

Microsoft.DP-100.vJul-2024.by.Isanco.262q

Number: DP-100
Passing Score: 800
Time Limit: 120
File Version: 21.0

Exam Code: DP-100
Exam Name: Designing and Implementing a Data Science Solution on Azure



01 - Run experiments and train models

QUESTION 1

You are solving a classification task.

You must evaluate your model on a limited data sample by using k-fold cross-validation. You start by configuring a k parameter as the number of splits.

You need to configure the k parameter for the cross-validation.

Which value should you use?

- A. k=1
- B. k=10
- C. k=0.5
- D. k=0.9

Correct Answer: B

Section:

Explanation:

Leave One Out (LOO) cross-validation

Setting $K = n$ (the number of observations) yields n-fold and is called leave-one out cross-validation (LOO), a special case of the K-fold approach.

LOO CV is sometimes useful but typically doesn't shake up the data enough. The estimates from each fold are highly correlated and hence their average can have high variance. This is why the usual choice is $K=5$ or 10 . It provides a good compromise for the bias-variance tradeoff.

QUESTION 2

You use the Azure Machine Learning service to create a tabular dataset named training_data. You plan to use this dataset in a training script.

You create a variable that references the dataset using the following code:

```
training_ds = workspace.datasets.get("training_data")
```

You define an estimator to run the script.

You need to set the correct property of the estimator to ensure that your script can access the training_data dataset.

Which property should you set?

- A. environment_definition = {"training_data":training_ds}
- B. inputs = [training_ds.as_named_input('training_ds')]
- C. script_params = {"--training_ds":training_ds}
- D. source_directory = training_ds

Correct Answer: B

Section:

Explanation:

Example:

```
# Get the training dataset diabetes_ds = ws.datasets.get("Diabetes Dataset") # Create an estimator that uses the remote compute hyper_estimator = SKLearn(source_directory=experiment_folder,
inputs=[diabetes_ds.as_named_input('diabetes')], # Pass the dataset as an input compute_target = cpu_cluster, conda_packages=['pandas','ipykernel','matplotlib'], pip_packages=['azureml-sdk','argparse','pyarrow'],
entry_script='diabetes_training.py')
```

Reference: <https://notebooks.azure.com/GraemeMalcolm/projects/azureml-primers/html/04%20-%20Optimizing%20Model%20Training.ipynb>

QUESTION 3

You register a file dataset named csv_folder that references a folder. The folder includes multiple comma-separated values (CSV) files in an Azure storage blob container.

You plan to use the following code to run a script that loads data from the file dataset. You create and instantiate the following variables:

Variable	Description
remote_cluster	References the Azure Machine Learning compute cluster
ws	References the Azure Machine Learning workspace

You have the following code:

```
from azureml.train.estimator import Estimator
file_dataset = ws.datasets.get('csv_folder')
estimator = Estimator(source_directory=script_folder,

compute_target = remote_cluster,
entry_script = 'script.py')
run = experiment.submit(config=estimator)
run.wait_for_completion(show_output=True)
```

You need to pass the dataset to ensure that the script can read the files it references.

Which code segment should you insert to replace the code comment?

- A. inputs=[file_dataset.as_named_input('training_files')],
- B. inputs=[file_dataset.as_named_input('training_files').as_mount()],
- C. inputs=[file_dataset.as_named_input('training_files').to_pandas_dataframe()],
- D. script_params={'--training_files': file_dataset},

Correct Answer: B

Section:

Explanation:

Example:

```
from azureml.train.estimator import Estimator
script_params = {
# to mount files referenced by mnist dataset
'--data-folder': mnist_file_dataset.as_named_input('mnist_opendataset').as_mount(),
'--regularization': 0.5
}
est = Estimator(source_directory=script_folder,
script_params=script_params,
compute_target=compute_target,
environment_definition=env,
entry_script='train.py')
```

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/tutorial-train-models-with-aml>

QUESTION 4

You are creating a new Azure Machine Learning pipeline using the designer.

The pipeline must train a model using data in a comma-separated values (CSV) file that is published on a website. You have not created a dataset for this file.

You need to ingest the data from the CSV file into the designer pipeline using the minimal administrative effort.

Which module should you add to the pipeline in Designer?

- A. Convert to CSV
- B. Enter Data Manually
- C. Import Data



D. Dataset

Correct Answer: D

Section:

Explanation:

QUESTION 5

You define a datastore named ml-data for an Azure Storage blob container. In the container, you have a folder named train that contains a file named data.csv. You plan to use the file to train a model by using the Azure Machine Learning SDK.

You plan to train the model by using the Azure Machine Learning SDK to run an experiment on local compute.

You define a DataReference object by running the following code:

```
from azureml.core import Workspace, Datastore, Environment
from azureml.train.estimator import Estimator
ws = Workspace.from_config()
ml_data = Datastore.get(ws, datastore_name='ml-data')
data_ref = ml_data.path('train').as_download(path_on_compute='train_data')
estimator = Estimator(source_directory='experiment_folder',
    script_params={'--data-folder': data_ref},
    compute_target = 'local',
    entry_script='training.py')
run = experiment.submit(config=estimator)
run.wait_for_completion(show_output=True)
```

You need to load the training data.

Which code segment should you use?

A.

```
import os
import argparse
import pandas as pd

parser = argparse.ArgumentParser()
parser.add_argument('--data-folder', type=str, dest='data_folder')
data_folder = args.data_folder
data = pd.read_csv(os.path.join(data_folder, 'ml-data', 'train_data', 'data.csv'))
```

B.

```
import os
import argparse
import pandas as pd

parser = argparse.ArgumentParser()
parser.add_argument('--data-folder', type=str, dest='data_folder')
data_folder = args.data_folder
data = pd.read_csv(os.path.join(data_folder, 'train', 'data.csv'))
```

C.




```
import pandas as pd
data = pd.read_csv('./data.csv')
```

D.

```
import os
import argparse
import pandas as pd

parser = argparse.ArgumentParser()
parser.add_argument('--data-folder', type=str, dest='data_folder')
data_folder = args.data_folder
data = pd.read_csv(os.path.join('ml_data', data_folder, 'data.csv'))
```

E.

```
import os
import argparse
import pandas as pd

parser = argparse.ArgumentParser()
parser.add_argument('--data-folder', type=str, dest='data_folder')
data_folder = args.data_folder
data = pd.read_csv(os.path.join(data_folder, 'data.csv'))
```

Correct Answer: E

Section:

Explanation:

Example:

```
data_folder = args.data_folder # Load Train and Test data
train_data = pd.read_csv(os.path.join(data_folder, 'data.csv'))
```

Reference:

<https://www.element61.be/en/resource/azure-machine-learning-services-complete-toolbox-ai>



QUESTION 6

Note: This question is part of a series of questions that present the same scenario. Each question in the series contains a unique solution that might meet the stated goals. Some question sets might have more than one correct solution, while others might not have a correct solution.

After you answer a question in this section, you will NOT be able to return to it. As a result, these questions will not appear in the review screen.

You create an Azure Machine Learning service datastore in a workspace. The datastore contains the following files:

```
/data/2018/Q1.csv
```

```
/data/2018/Q2.csv
```

```
/data/2018/Q3.csv
```

```
/data/2018/Q4.csv
```

```
/data/2019/Q1.csv
```

All files store data in the following format:

```
id,f1,f2,l
```

```
1,1,2,0
```

```
2,1,1,1
```

```
3,2,1,0
```

```
4,2,2,1
```

You run the following code:

```
data_store = Datastore.register_azure_blob_container(workspace=ws,
datastore_name= 'data_store',
container_name= 'quarterly_data',
account_name='companydata',
account_key='NRPxk8duxbM3...'
create_if_not_exists=False)
```

You need to create a dataset named training_data and load the data from all files into a single data frame by using the following code:

```
data_frame = training_data.to_pandas_dataframe()
```

Solution: Run the following code:

```
from azureml.core import Dataset
paths = (data_store, 'data/**/*.csv')
training_data = Dataset.Tabular.from_delimited_files(paths)
```

Does the solution meet the goal?

- A. Yes
- B. No

Correct Answer: B

Section:

Explanation:

Define paths with two file paths instead.

Use Dataset.Tabular_from_delimited as the data isn't cleansed.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/how-to-create-register-datasets>

QUESTION 7

Note: This question is part of a series of questions that present the same scenario. Each question in the series contains a unique solution that might meet the stated goals. Some question sets might have more than one correct solution, while others might not have a correct solution.

After you answer a question in this section, you will NOT be able to return to it. As a result, these questions will not appear in the review screen.

You create an Azure Machine Learning service datastore in a workspace. The datastore contains the following files:

```
/data/2018/Q1.csv
/data/2018/Q2.csv
/data/2018/Q3.csv
/data/2018/Q4.csv
/data/2019/Q1.csv
```

All files store data in the following format:

```
id,f1,f2,l
1,1,2,0
2,1,1,1
3,2,1,0
4,2,2,1
```

You run the following code:

```
data_store = Datastore.register_azure_blob_container(workspace=ws,
datastore_name= 'data_store',
container_name= 'quarterly_data',
account_name='companydata',
account_key='NRPxk8duxbM3...'
create_if_not_exists=False)
```

You need to create a dataset named training_data and load the data from all files into a single data frame by using the following code:

```
data_frame = training_data.to_pandas_dataframe()
```

Solution: Run the following code:

```
from azureml.core import Dataset
paths = [(data_store, 'data/2018/*.csv'), (data_store, 'data/2019/*.csv')]
training_data = Dataset.File.from_files(paths)
```

Does the solution meet the goal?

- A. Yes
- B. No

Correct Answer: B

Section:

Explanation:

Use two file paths.

Use Dataset.Tabular_from_delimited, instead of Dataset.File.from_files as the data isn't cleansed.

Note:

A FileDataset references single or multiple files in your datastores or public URLs. If your data is already cleansed, and ready to use in training experiments, you can download or mount the files to your compute as a FileDataset object.

A TabularDataset represents data in a tabular format by parsing the provided file or list of files. This provides you with the ability to materialize the data into a pandas or Spark DataFrame so you can work with familiar data preparation and training libraries without having to leave your notebook. You can create a TabularDataset object from .csv, .tsv, .parquet, .jsonl files, and from SQL query results.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/how-to-create-register-datasets>

QUESTION 8

Note: This question is part of a series of questions that present the same scenario. Each question in the series contains a unique solution that might meet the stated goals. Some question sets might have more than one correct solution, while others might not have a correct solution.

After you answer a question in this section, you will NOT be able to return to it. As a result, these questions will not appear in the review screen.

You create an Azure Machine Learning service datastore in a workspace. The datastore contains the following files:

/data/2018/Q1.csv

/data/2018/Q2.csv

/data/2018/Q3.csv

/data/2018/Q4.csv

/data/2019/Q1.csv

All files store data in the following format:

id,f1,f2,l

1,1,2,0

2,1,1,1

3,2,1,0

4,2,2,1

You run the following code:

```
data_store = Datastore.register_azure_blob_container(workspace=ws,
datastore_name= 'data_store',
container_name= 'quarterly_data',
account_name= 'companydata',
account_key= 'NRPxk8duxBM3...'
create_if_not_exists=False)
```

You need to create a dataset named training_data and load the data from all files into a single data frame by using the following code:

```
data_frame = training_data.to_pandas_dataframe()
```

Solution: Run the following code:

```
from azureml.core import Dataset
paths = [(data_store, 'data/2018/*.csv'), (data_store, 'data/2019/*.csv')]
training_data = Dataset.Tabular.from_delimited_files(paths)
```

Does the solution meet the goal?

- A. Yes
- B. No

Correct Answer: A

Section:

Explanation:

Use two file paths.

Use Dataset.Tabular_from_delimited as the data isn't cleansed.

Note:

A TabularDataset represents data in a tabular format by parsing the provided file or list of files. This provides you with the ability to materialize the data into a pandas or Spark DataFrame so you can work with familiar data preparation and training libraries without having to leave your notebook. You can create a TabularDataset object from .csv, .tsv, .parquet, .jsonl files, and from SQL query results.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/how-to-create-register-datasets>

QUESTION 9

You plan to use the Hyperdrive feature of Azure Machine Learning to determine the optimal hyperparameter values when training a model.

You must use Hyperdrive to try combinations of the following hyperparameter values:

learning_rate: any value between 0.001 and 0.1 batch_size: 16, 32, or 64

You need to configure the search space for the Hyperdrive experiment.

Which two parameter expressions should you use? Each correct answer presents part of the solution.

NOTE: Each correct selection is worth one point.

- A. a choice expression for learning_rate
- B. a uniform expression for learning_rate
- C. a normal expression for batch_size
- D. a choice expression for batch_size
- E. a uniform expression for batch_size

Correct Answer: B, D

Section:

Explanation:

B: Continuous hyperparameters are specified as a distribution over a continuous range of values. Supported distributions include: uniform(low, high) - Returns a value uniformly distributed between low and high

D: Discrete hyperparameters are specified as a choice among discrete values. choice can be:

one or more comma-separated values a range object any arbitrary list object

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/how-to-tune-hyperparameters>

QUESTION 10

You run an automated machine learning experiment in an Azure Machine Learning workspace. Information about the run is listed in the table below:

Experiment	Run ID	Status	Created on	Duration
auto_ml_clasification	AutoML_1234567890-123	Completed	11/11/2019 11:00:00 AM	00:27:11

You need to write a script that uses the Azure Machine Learning SDK to retrieve the best iteration of the experiment run.

Which Python code segment should you use?

- A.
- ```
from azureml.core import Workspace
from azureml.train.automl.run import AutoMLRun
ws = Workspace.from_config()
automl_ex = ws.experiments.get('auto_ml_classification')
best_iter = automl_ex.archived_time.find('11/11/2019 11:00:00 AM')
```
- B.
- ```
from azureml.core import Workspace
from azureml.train.automl.run import AutoMLRun
automl_ex = ws.experiments.get('auto_ml_classification')
automl_run = AutoMLRun(automl_ex, 'AutoML_1234567890-123')
best_iter = automl_run.current_run
```
- C.
- ```
from azureml.core import Workspace
from azureml.train.automl.run import AutoMLRun
ws = Workspace.from_config()
automl_ex = ws.experiments.get('auto_ml_classification')
best_iter = list(automl_ex.get_runs())[0]
```
- D.
- ```
from azureml.core import Workspace
from azureml.train.automl.run import AutoMLRun
ws = Workspace.from_config()
automl_ex = ws.experiments.get('auto_ml_classification')
automl_run = AutoMLRun(automl_ex, 'AutoML_1234567890-123')
best_iter = automl_run.get_output()[0]
```
- E.
- ```
from azureml.core import Workspace
from azureml.train.automl.run import AutoMLRun
ws = Workspace.from_config()
automl_ex = ws.experiments.get('auto_ml_classification')
best_iter = automl_ex.get_runs('AutoML_1234567890-123')
```

**Correct Answer: D**

**Section:**

**Explanation:**

The `get_output` method on `automl_classifier` returns the best run and the fitted model for the last invocation. Overloads on `get_output` allow you to retrieve the best run and fitted model for any logged metric or for a particular iteration.

In [ ]:

```
best_run, fitted_model = local_run.get_output()
```

Reference:



<https://notebooks.azure.com/azureml/projects/azureml-getting-started/html/how-to-use-azureml/automated-machine-learning/classification-with-deployment/auto-ml-classification-with-deployment.ipynb>

#### QUESTION 11

You have a comma-separated values (CSV) file containing data from which you want to train a classification model.

You are using the Automated Machine Learning interface in Azure Machine Learning studio to train the classification model. You set the task type to Classification.

You need to ensure that the Automated Machine Learning process evaluates only linear models.

What should you do?

- A. Add all algorithms other than linear ones to the blocked algorithms list.
- B. Set the Exit criterion option to a metric score threshold.
- C. Clear the option to perform automatic featurization.
- D. Clear the option to enable deep learning.
- E. Set the task type to Regression.

**Correct Answer: A**

**Section:**

**Explanation:**

Automatic featurization can fit non-linear models.

Reference: <https://econml.azurewebsites.net/spec/estimation/dml.html> <https://docs.microsoft.com/en-us/azure/machine-learning/how-to-use-automated-ml-for-ml-models>

#### QUESTION 12

You are analyzing a dataset containing historical data from a local taxi company. You are developing a regression model.

You must predict the fare of a taxi trip.

You need to select performance metrics to correctly evaluate the regression model.

Which two metrics can you use? Each correct answer presents a complete solution?

NOTE: Each correct selection is worth one point.

- A. a Root Mean Square Error value that is low
- B. an R-Squared value close to 0
- C. an F1 score that is low
- D. an R-Squared value close to 1
- E. an F1 score that is high
- F. a Root Mean Square Error value that is high

**Correct Answer: A, D**

**Section:**

**Explanation:**

RMSE and R2 are both metrics for regression models.

A: Root mean squared error (RMSE) creates a single value that summarizes the error in the model. By squaring the difference, the metric disregards the difference between over-prediction and under-prediction.

D: Coefficient of determination, often referred to as R2, represents the predictive power of the model as a value between 0 and 1. Zero means the model is random (explains nothing); 1 means there is a perfect fit. However, caution should be used in interpreting R2 values, as low values can be entirely normal and high values can be suspect.

Incorrect Answers:

C, E: F-score is used for classification models, not for regression models.

Reference: <https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/evaluate-model>

#### QUESTION 13

Note: This question is part of a series of questions that present the same scenario. Each question in the series contains a unique solution that might meet the stated goals. Some question sets might have more than one correct solution, while others might not have a correct solution.



After you answer a question in this section, you will NOT be able to return to it. As a result, these questions will not appear in the review screen.

You are using Azure Machine Learning to run an experiment that trains a classification model.

You want to use Hyperdrive to find parameters that optimize the AUC metric for the model. You configure a HyperDriveConfig for the experiment by running the following code:

```
hyperdrive = HyperDriveConfig(estimator=your_estimator,
 hyperparameter_sampling=your_params,
 policy=policy,
 primary_metric_name='AUC',
 primary_metric_goal=PrimaryMetricGoal.MAXIMIZE,
 max_total_runs=6,
 max_concurrent_runs=4)
```

You plan to use this configuration to run a script that trains a random forest model and then tests it with validation data. The label values for the validation data are stored in a variable named `y_test` variable, and the predicted probabilities from the model are stored in a variable named `y_predicted`.

You need to add logging to the script to allow Hyperdrive to optimize hyperparameters for the AUC metric.

Solution: Run the following code:

```
from sklearn.metrics import roc_auc_score
import logging
code to train model omitted
auc = roc_auc_score(y_test, y_predicted)
logging.info("AUC: " + str(auc))
```

Does the solution meet the goal?

- A. Yes
- B. No

**Correct Answer: A**

**Section:**

**Explanation:**

Python printing/logging example: `logging.info(message)`

Destination: Driver logs, Azure Machine Learning designer

Reference: <https://docs.microsoft.com/en-us/azure/machine-learning/how-to-debug-pipelines>

#### QUESTION 14

Note: This question is part of a series of questions that present the same scenario. Each question in the series contains a unique solution that might meet the stated goals. Some question sets might have more than one correct solution, while others might not have a correct solution.

After you answer a question in this section, you will NOT be able to return to it. As a result, these questions will not appear in the review screen.

You are using Azure Machine Learning to run an experiment that trains a classification model.

You want to use Hyperdrive to find parameters that optimize the AUC metric for the model. You configure a HyperDriveConfig for the experiment by running the following code:

```
hyperdrive = HyperDriveConfig(estimator=your_estimator,
 hyperparameter_sampling=your_params,
 policy=policy,
 primary_metric_name='AUC',
 primary_metric_goal=PrimaryMetricGoal.MAXIMIZE,
 max_total_runs=6,
 max_concurrent_runs=4)
```

You plan to use this configuration to run a script that trains a random forest model and then tests it with validation data. The label values for the validation data are stored in a variable named `y_test` variable, and the predicted probabilities from the model are stored in a variable named `y_predicted`.



You need to add logging to the script to allow Hyperdrive to optimize hyperparameters for the AUC metric.

Solution: Run the following code:

```
import json, os
from sklearn.metrics import roc_auc_score
code to train model omitted
auc = roc_auc_score(y_test, y_predicted)
os.makedirs("outputs", exist_ok = True)
with open("outputs/AUC.txt", "w") as file_cur:
 file_cur.write(auc)
```

Does the solution meet the goal?

- A. Yes
- B. No

**Correct Answer: B**

**Section:**

**Explanation:**

Use a solution with logging.info(message) instead.

Note: Python printing/logging example: logging.info(message)

Destination: Driver logs, Azure Machine Learning designer

Reference: <https://docs.microsoft.com/en-us/azure/machine-learning/how-to-debug-pipelines>

#### QUESTION 15

Note: This question is part of a series of questions that present the same scenario. Each question in the series contains a unique solution that might meet the stated goals. Some question sets might have more than one correct solution, while others might not have a correct solution.

After you answer a question in this section, you will NOT be able to return to it. As a result, these questions will not appear in the review screen.

You are using Azure Machine Learning to run an experiment that trains a classification model.

You want to use Hyperdrive to find parameters that optimize the AUC metric for the model. You configure a HyperDriveConfig for the experiment by running the following code:

```
hyperdrive = HyperDriveConfig(estimator=your_estimator,
 hyperparameter_sampling=your_params,
 policy=policy,
 primary_metric_name='AUC',
 primary_metric_goal=PrimaryMetricGoal.MAXIMIZE,
 max_total_runs=6,
 max_concurrent_runs=4)
```

You plan to use this configuration to run a script that trains a random forest model and then tests it with validation data. The label values for the validation data are stored in a variable named y\_test variable, and the predicted probabilities from the model are stored in a variable named y\_predicted.

You need to add logging to the script to allow Hyperdrive to optimize hyperparameters for the AUC metric.

Solution: Run the following code:

```
import numpy as np
from sklearn.metrics import roc_auc_score
code to train model omitted
auc = roc_auc_score(y_test, y_predicted)
print(np.float(auc))
```

Does the solution meet the goal?

- A. Yes
- B. No

**Correct Answer: B**

**Section:**

**Explanation:**

Use a solution with logging.info(message) instead.

Note: Python printing/logging example: logging.info(message)

Destination: Driver logs, Azure Machine Learning designer

Reference: <https://docs.microsoft.com/en-us/azure/machine-learning/how-to-debug-pipelines>

#### QUESTION 16

You use the following code to run a script as an experiment in Azure Machine Learning:

```
from azureml.core import Workspace, Experiment, Run
from azureml.core import RunConfig, ScriptRunConfig
ws = Workspace.from_config()
run_config = RunConfiguration()
run_config.target='local'
script_config = ScriptRunConfig(source_directory='./script', script='experiment.py', run_config=run_config)
experiment = Experiment(workspace=ws, name='script experiment')
run = experiment.submit(config=script_config)
run.wait_for_completion()
```

You must identify the output files that are generated by the experiment run.

You need to add code to retrieve the output file names.

Which code segment should you add to the script?

- A. files = run.get\_properties()
- B. files= run.get\_file\_names()
- C. files = run.get\_details\_with\_logs()
- D. files = run.get\_metrics()
- E. files = run.get\_details()

**Correct Answer: B**

**Section:**

**Explanation:**

You can list all of the files that are associated with this run record by called run.get\_file\_names()

Reference: <https://docs.microsoft.com/en-us/azure/machine-learning/how-to-track-experiments>

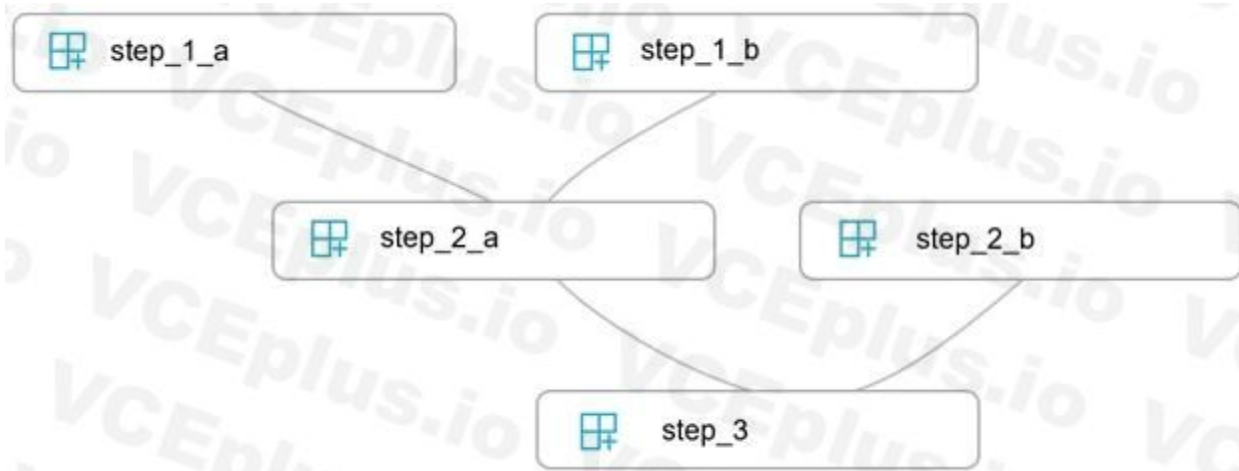
#### QUESTION 17

You write five Python scripts that must be processed in the order specified in Exhibit A – which allows the same modules to run in parallel, but will wait for modules with dependencies.

You must create an Azure Machine Learning pipeline using the Python SDK, because you want to script to create the pipeline to be tracked in your version control system. You have created five PythonScriptSteps and have named the variables to match the module names.







You need to create the pipeline shown. Assume all relevant imports have been done.  
Which Python code segment should you use?

- A.
- ```
p = Pipeline(ws, steps=[[[[step_1_a, step_1_b], step_2_a], step_2_b], step_3])
```
- B.
- ```
pipeline_steps = {
 "Pipeline": {
 "run": step_3,
 "run_after": [{
 {"run": step_2_a,
 "run_after": [
 {"run": step_1_a},
 {"run": step_1_b}
]},
 {"run": step_2_b}
]}
]}
}
p = Pipeline(ws, steps=pipeline_steps)
```
- C.
- ```
step_2_a.run_after(step_1_b)
step_2_a.run_after(step_1_a)
step_3.run_after(step_2_b)
step_3.run_after(step_2_a)
p = Pipeline(ws, steps=[step_3])
```
- D.
- ```
p = Pipeline(ws, steps=[step_1_a, step_1_b, step_2_a, step_2_b, step_3])
```

**Correct Answer: A**

**Section:**

**Explanation:**

The steps parameter is an array of steps. To build pipelines that have multiple steps, place the steps in order in this array.

Reference: <https://docs.microsoft.com/en-us/azure/machine-learning/how-to-use-parallel-run-step>





## QUESTION 18

You create a datastore named `training_data` that references a blob container in an Azure Storage account. The blob container contains a folder named `csv_files` in which multiple comma-separated values (CSV) files are stored. You have a script named `train.py` in a local folder named `./script` that you plan to run as an experiment using an estimator. The script includes the following code to read data from the `csv_files` folder:

```
import os
import argparse
import pandas as pd

from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression
from azureml.core import Run

run = Run.get_context()
parser = argparse.ArgumentParser()
parser.add_argument('--data-folder', type=str, dest='data_folder', help='data reference')
args = parser.parse_args()

data_folder = args.data_folder
csv_files = os.listdir(data_folder)
training_data = pd.concat((pd.read_csv(os.path.join(data_folder, csv_file)) for csv_file in csv_files))

Code goes on to split the training data and train a logistic regression model
```

You have the following script.

```
from azureml.core import Workspace, Datastore, Experiment
from azureml.train.sklearn import SKLearn
```

```
ws = Workspace.from_config()
exp = Experiment(workspace=ws, name='csv_training')
ds = Datastore.get(ws, datastore_name='training_data')
data_ref = ds.path('csv_files')
```

# Code to define estimator goes here

```
run = exp.submit(config=estimator)
run.wait_for_completion(show_output=True)
```

You need to configure the estimator for the experiment so that the script can read the data from a data reference named `data_ref` that references the `csv_files` folder in the `training_data` datastore. Which code should you use to configure the estimator?

A.

```
estimator = SKLearn(source_directory='./script',
 inputs=[data_ref.as_named_input('data-folder').to_pandas_dataframe()],
 compute_target='local',
 entry_script='train.py')
```

B.

```
script_params = {
 '--data-folder': data_ref.as_mount()
}
estimator = SKLearn(source_directory='./script',
 script_params=script_params,
 compute_target='local',
 entry_script='train.py')
```

C.



```
estimator = SKLearn(source_directory='./script',
 inputs=[data_ref.as_named_input('data-folder').as_mount()],
 compute_target='local',
 entry_script='train.py')
```

D.

```
script_params = {
 '--data-folder': data_ref.as_download(path_on_compute='csv_files')
}
estimator = SKLearn(source_directory='./script',
 script_params=script_params,
 compute_target='local',
 entry_script='train.py')
```

E.

```
estimator = SKLearn(source_directory='./script',
 inputs=[data_ref.as_named_input('data-folder').as_download(path_on_compute='csv_files')],
 compute_target='local',
 entry_script='train.py')
```

**Correct Answer: B**

**Section:**

**Explanation:**

Besides passing the dataset through the input parameters in the estimator, you can also pass the dataset through script\_params and get the data path (mounting point) in your training script via arguments. This way, you can keep your training script independent of azureml-sdk. In other words, you will be able use the same training script for local debugging and remote training on any cloud platform.

Example:

```
from azureml.train.sklearn import SKLearn
script_params = {
 # mount the dataset on the remote compute and pass the mounted path as an argument to the training script
 '--data-folder': mnist_ds.as_named_input('mnist').as_mount(),
 '--regularization': 0.5
}
est = SKLearn(source_directory=script_folder,
 script_params=script_params,
 compute_target=compute_target,
 environment_definition=env,
 entry_script='train_mnist.py')
Run the experiment
run = experiment.submit(est)
run.wait_for_completion(show_output=True)
```

Incorrect Answers:

A: Pandas DataFrame not used.

Reference:

<https://docs.microsoft.com/es-es/azure/machine-learning/how-to-train-with-datasets>

#### QUESTION 19

Note: This question is part of a series of questions that present the same scenario. Each question in the series contains a unique solution that might meet the stated goals. Some question sets might have more than one correct solution, while others might not have a correct solution.

After you answer a question in this section, you will NOT be able to return to it. As a result, these questions will not appear in the review screen.



An IT department creates the following Azure resource groups and resources:

| Resource group  | Resources                                                                                                                                                                                                                                                                                                                                                 |
|-----------------|-----------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|
| ml_resources    | <ul style="list-style-type: none"><li>• an Azure Machine Learning workspace named amlworkspace</li><li>• an Azure Storage account named amlworkspace12345</li><li>• an Application Insights instance named amlworkspace54321</li><li>• an Azure Key Vault named amlworkspace67890</li><li>• an Azure Container Registry named amlworkspace09876</li></ul> |
| general_compute | <p>A virtual machine named mlvm with the following configuration:</p> <ul style="list-style-type: none"><li>• Operating system: Ubuntu Linux</li><li>• Software installed: Python 3.6 and Jupyter Notebooks</li><li>• Size: NC6 (6 vCPUs, 1 vGPU, 56 Gb RAM)</li></ul>                                                                                    |

The IT department creates an Azure Kubernetes Service (AKS)-based inference compute target named aks-cluster in the Azure Machine Learning workspace.

You have a Microsoft Surface Book computer with a GPU. Python 3.6 and Visual Studio Code are installed.

You need to run a script that trains a deep neural network (DNN) model and logs the loss and accuracy metrics.

Solution: Attach the mlvm virtual machine as a compute target in the Azure Machine Learning workspace. Install the Azure ML SDK on the Surface Book and run Python code to connect to the workspace. Run the training script as an experiment on the mlvm remote compute resource.

Does the solution meet the goal?

A. Yes

B. No

**Correct Answer: A**

**Section:**

**Explanation:**

Use the VM as a compute target.

Note: A compute target is a designated compute resource/environment where you run your training script or host your service deployment. This location may be your local machine or a cloud-based compute resource.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/concept-compute-target>

## QUESTION 20

Note: This question is part of a series of questions that present the same scenario. Each question in the series contains a unique solution that might meet the stated goals. Some question sets might have more than one correct solution, while others might not have a correct solution.

After you answer a question in this section, you will NOT be able to return to it. As a result, these questions will not appear in the review screen.

An IT department creates the following Azure resource groups and resources:

| Resource group  | Resources                                                                                                                                                                                                                                                                                                                                                 |
|-----------------|-----------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|
| ml_resources    | <ul style="list-style-type: none"><li>• an Azure Machine Learning workspace named amlworkspace</li><li>• an Azure Storage account named amlworkspace12345</li><li>• an Application Insights instance named amlworkspace54321</li><li>• an Azure Key Vault named amlworkspace67890</li><li>• an Azure Container Registry named amlworkspace09876</li></ul> |
| general_compute | <p>A virtual machine named mlvm with the following configuration:</p> <ul style="list-style-type: none"><li>• Operating system: Ubuntu Linux</li><li>• Software installed: Python 3.6 and Jupyter Notebooks</li><li>• Size: NC6 (6 vCPUs, 1 vGPU, 56 Gb RAM)</li></ul>                                                                                    |

The IT department creates an Azure Kubernetes Service (AKS)-based inference compute target named aks-cluster in the Azure Machine Learning workspace.

You have a Microsoft Surface Book computer with a GPU. Python 3.6 and Visual Studio Code are installed.

You need to run a script that trains a deep neural network (DNN) model and logs the loss and accuracy metrics.

Solution: Install the Azure ML SDK on the Surface Book. Run Python code to connect to the workspace and then run the training script as an experiment on local compute.

Does the solution meet the goal?

A. Yes

B. No

**Correct Answer: B**

**Section:**

**Explanation:**

Need to attach the mlvm virtual machine as a compute target in the Azure Machine Learning workspace.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/concept-compute-target>

#### QUESTION 21

Note: This question is part of a series of questions that present the same scenario. Each question in the series contains a unique solution that might meet the stated goals. Some question sets might have more than one correct solution, while others might not have a correct solution.

After you answer a question in this section, you will NOT be able to return to it. As a result, these questions will not appear in the review screen.

An IT department creates the following Azure resource groups and resources:

| Resource group  | Resources                                                                                                                                                                                                                                                                                                                                                 |
|-----------------|-----------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|
| ml_resources    | <ul style="list-style-type: none"><li>• an Azure Machine Learning workspace named amlworkspace</li><li>• an Azure Storage account named amlworkspace12345</li><li>• an Application Insights instance named amlworkspace54321</li><li>• an Azure Key Vault named amlworkspace67890</li><li>• an Azure Container Registry named amlworkspace09876</li></ul> |
| general_compute | <p>A virtual machine named mlvm with the following configuration:</p> <ul style="list-style-type: none"><li>• Operating system: Ubuntu Linux</li><li>• Software installed: Python 3.6 and Jupyter Notebooks</li><li>• Size: NC6 (6 vCPUs, 1 vGPU, 56 Gb RAM)</li></ul>                                                                                    |

The IT department creates an Azure Kubernetes Service (AKS)-based inference compute target named aks-cluster in the Azure Machine Learning workspace.

You have a Microsoft Surface Book computer with a GPU. Python 3.6 and Visual Studio Code are installed.

You need to run a script that trains a deep neural network (DNN) model and logs the loss and accuracy metrics.

Solution: Install the Azure ML SDK on the Surface Book. Run Python code to connect to the workspace. Run the training script as an experiment on the aks-cluster compute target.

Does the solution meet the goal?

A. Yes

B. No

**Correct Answer: B**

**Section:**

**Explanation:**

Need to attach the mlvm virtual machine as a compute target in the Azure Machine Learning workspace.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/concept-compute-target>

#### QUESTION 22

You create a batch inference pipeline by using the Azure ML SDK. You configure the pipeline parameters by executing the following code:

```
from azureml.contrib.pipeline.steps import ParallelRunConfig
parallel_run_config = ParallelRunConfig(
 source_directory=scripts_folder,
 entry_script= "batch_pipeline.py",
 mini_batch_size= "5",
 error_threshold=10,
 output_action= "append_row",
 environment=batch_env,
 compute_target=compute_target,
 logging_level= "DEBUG",
 node_count=4)
```

You need to obtain the output from the pipeline execution.  
Where will you find the output?

- A. the digit\_identification.py script
- B. the debug log
- C. the Activity Log in the Azure portal for the Machine Learning workspace
- D. the Inference Clusters tab in Machine Learning studio
- E. a file named parallel\_run\_step.txt located in the output folder

**Correct Answer: E**

**Section:**

**Explanation:**

output\_action (str): How the output is to be organized. Currently supported values are 'append\_row' and 'summary\_only'.

'append\_row' - All values output by run() method invocations will be aggregated into one unique file named parallel\_run\_step.txt that is created in the output location. 'summary\_only'

Reference:

<https://docs.microsoft.com/en-us/python/api/azureml-contrib-pipeline-steps/azureml.contrib.pipeline.steps.parallelrunconfig>

### QUESTION 23

HOTSPOT

You create an experiment in Azure Machine Learning Studio. You add a training dataset that contains 10,000 rows. The first 9,000 rows represent class 0 (90 percent).

The remaining 1,000 rows represent class 1 (10 percent).

The training set is imbalances between two classes. You must increase the number of training examples for class 1 to 4,000 by using 5 data rows. You add the Synthetic Minority Oversampling Technique (SMOTE) module to the experiment.

You need to configure the module.

Which values should you use? To answer, select the appropriate options in the dialog box in the answer area.

NOTE: Each correct selection is worth one point.

**Hot Area:**



**Answer Area**

SMOTE

Label column

Selected columns:  
**All labels**

Launch column selector

SMOTE percentage

|      |
|------|
| 0    |
| 300  |
| 3000 |
| 4000 |

Number of nearest neighbors

|      |
|------|
| 0    |
| 1    |
| 5    |
| 4000 |

Random seed

|   |
|---|
| 0 |
|---|

Answer Area:



**Answer Area**

SMOTE

Label column

Selected columns:  
All labels

Launch column selector

SMOTE percentage

0  
300  
3000  
4000

Number of nearest neighbors

0  
1  
5  
4000

Random seed

0



**Section:**

**Explanation:**

Box 1: 300

You type 300 (%), the module triples the percentage of minority cases (3000) compared to the original dataset (1000).

Box 2: 5

We should use 5 data rows.

Use the Number of nearest neighbors option to determine the size of the feature space that the SMOTE algorithm uses when in building new cases. A nearest neighbor is a row of data (a case) that is very similar to some target case. The distance between any two cases is measured by combining the weighted vectors of all features.

By increasing the number of nearest neighbors, you get features from more cases.

By keeping the number of nearest neighbors low, you use features that are more like those in the original sample.

References:

<https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/smote>

**QUESTION 24**

**HOTSPOT**

You are running Python code interactively in a Conda environment. The environment includes all required Azure Machine Learning SDK and MLflow packages.

You must use MLflow to log metrics in an Azure Machine Learning experiment named mlflow-experiment.

How should you complete the code? To answer, select the appropriate options in the answer area.

NOTE: Each correct selection is worth one point.

**Hot Area:**

## Answer Area

```
import mlflow
from azureml.core import Workspace
ws = Workspace.from_config()
Set the MLflow logging target

mlflow.tracking.client = ws
mlflow.set_tracking_uri(ws.get_mlflow_tracking_uri())
mlflow.log_param('workspace', ws)

Configure experiment

mlflow-experiment = Run.get_context()
mlflow.get_run('mlflow-experiment')
mlflow.set_experiment('mlflow-experiment')

Begin the experiment run
with
 mlflow.active_run
 mlflow.start_run()
 Run.get_context()

Log my_metric with value 1.00
run.log()
mlflow.log_metric('my_metric', 1.00)
print

print("Finished!")
```

Answer Area:

## Answer Area

```
import mlflow
from azureml.core import Workspace
ws = Workspace.from_config()
Set the MLflow logging target

mlflow.tracking.client = ws
mlflow.set_tracking_uri(ws.get_mlflow_tracking_uri())
mlflow.log_param('workspace', ws)

Configure experiment

mlflow-experiment = Run.get_context()
mlflow.get_run('mlflow-experiment')
mlflow.set_experiment('mlflow-experiment')

Begin the experiment run
with
mlflow.active_run
mlflow.start_run()
Run.get_context()

Log my_metric with value 1.00
run.log() ('my_metric', 1.00)
mlflow.log_metric
print

print("Finished!")
```

### Section:

### Explanation:

Box 1: `mlflow.set_tracking_uri(ws.get_mlflow_tracking_uri())`

In the following code, the `get_mlflow_tracking_uri()` method assigns a unique tracking URI address to the workspace, `ws`, and `set_tracking_uri()` points the MLflow tracking URI to that address.

`mlflow.set_tracking_uri(ws.get_mlflow_tracking_uri())`

Box 2: `mlflow.set_experiment(experiment_name)`

Set the MLflow experiment name with `set_experiment()` and start your training run with `start_run()`.

Box 3: `mlflow.start_run()`

Box 4: `mlflow.log_metric`

Then use `log_metric()` to activate the MLflow logging API and begin logging your training run metrics.

Reference:



**QUESTION 25**

DRAG DROP

You are creating a machine learning model that can predict the species of a penguin from its measurements. You have a file that contains measurements for three species of penguin in comma-delimited format. The model must be optimized for area under the received operating characteristic curve performance metric, averaged for each class.

You need to use the Automated Machine Learning user interface in Azure Machine Learning studio to run an experiment and find the best performing model.

Which five actions should you perform in sequence? To answer, move the appropriate actions from the list of actions to the answer area and arrange them in the correct order.

Select and Place:

**Actions**

- Create and select a new dataset by uploading the comma-delimited file of penguin data.
- Configure the automated machine learning run by selecting the experiment name, target column, and compute target.
- Set the Primary metric configuration setting to **Accuracy**.
- Select the **Classification** task type.
- Select the **Regression** task type.
- Run the automated machine learning experiment and review the results.
- Set the Primary metric configuration setting to **AUC Weighted**.

**Answer Area**



Correct Answer:



## Actions

|                                                                       |
|-----------------------------------------------------------------------|
|                                                                       |
|                                                                       |
|                                                                       |
|                                                                       |
| Select the <b>Regression</b> task type.                               |
|                                                                       |
|                                                                       |
| Set the Primary metric configuration setting to <b>AUC Weighted</b> . |

## Answer Area

|   |                                                                                                                   |   |
|---|-------------------------------------------------------------------------------------------------------------------|---|
|   | Create and select a new dataset by uploading the comma-delimited file of penguin data.                            |   |
|   | Select the <b>Classification</b> task type.                                                                       |   |
|   | Set the Primary metric configuration setting to <b>Accuracy</b> .                                                 |   |
| ⏪ | Configure the automated machine learning run by selecting the experiment name, target column, and compute target. | ⏩ |
| ⏩ | Run the automated machine learning experiment and review the results.                                             | ⏪ |

### Section:

### Explanation:

Step 1: Create and select a new dataset by uploading the command-delimited file of penguin data.

Step 2: Select the Classification task type

Step 3: Set the Primary metric configuration setting to Accuracy.

The available metrics you can select is determined by the task type you choose.

Primary metrics for classification scenarios:

Post thresholded metrics, like accuracy, average\_precision\_score\_weighted, norm\_macro\_recall, and precision\_score\_weighted may not optimize as well for datasets which are very small, have very large class skew (class imbalance), or when the expected metric value is very close to 0.0 or 1.0. In those cases, AUC\_weighted can be a better choice for the primary metric.

Step 4: Configure the automated machine learning run by selecting the experiment name, target column, and compute target

Step 5: Run the automated machine learning experiment and review the results.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/how-to-configure-auto-train>

### QUESTION 26

#### HOTSPOT

You are tuning a hyperparameter for an algorithm. The following table shows a data set with different hyperparameter, training error, and validation errors.

| Hyperparameter (H) | Training error (TE) | Validation error (VE) |
|--------------------|---------------------|-----------------------|
| 1                  | 105                 | 95                    |
| 2                  | 200                 | 85                    |
| 3                  | 250                 | 100                   |
| 4                  | 105                 | 100                   |
| 5                  | 400                 | 50                    |

Use the drop-down menus to select the answer choice that answers each question based on the information presented in the graphic.

Hot Area:

| Question                                           | Answer Choice                                                                                                                  |
|----------------------------------------------------|--------------------------------------------------------------------------------------------------------------------------------|
| Which H value should you select based on the data? | <input type="text" value="1"/> <ul style="list-style-type: none"> <li>1</li> <li>2</li> <li>3</li> <li>4</li> <li>5</li> </ul> |
| What H value displays the poorest training result? | <input type="text" value="1"/> <ul style="list-style-type: none"> <li>1</li> <li>2</li> <li>3</li> <li>4</li> <li>5</li> </ul> |

Answer Area:

**Answer Area**

| Question                                           | Answer Choice                                                                                                                                                 |
|----------------------------------------------------|---------------------------------------------------------------------------------------------------------------------------------------------------------------|
| Which H value should you select based on the data? | <input type="checkbox"/> 1<br><input type="checkbox"/> 2<br><input type="checkbox"/> 3<br><input checked="" type="checkbox"/> 4<br><input type="checkbox"/> 5 |
| What H value displays the poorest training result? | <input type="checkbox"/> 1<br><input type="checkbox"/> 2<br><input type="checkbox"/> 3<br><input type="checkbox"/> 4<br><input checked="" type="checkbox"/> 5 |

**Section:**

**Explanation:**

Box 1: 4

Choose the one which has lower training and validation error and also the closest match.

Minimize variance (difference between validation error and train error).

Box 2: 5

Minimize variance (difference between validation error and train error).

Reference:

<https://medium.com/comet-ml/organizing-machine-learning-projects-project-management-guidelines-2d2b85651bbd>

**QUESTION 27**

**DRAG DROP**

You create machine learning models by using Azure Machine Learning.

You plan to train and score models by using a variety of compute contexts. You also plan to create a new compute resource in Azure Machine Learning studio.

You need to select the appropriate compute types.

Which compute types should you select? To answer, drag the appropriate compute types to the correct requirements. Each compute type may be used once, more than once, or not at all. You may need to drag the split bar between panes or scroll to view content.

NOTE: Each correct selection is worth one point.

**Select and Place:**



### Compute types

Attached compute

Inference cluster

Training cluster

### Answer Area

#### Requirement

Train models by using the Azure Machine Learning designer.

Score new data through a trained model published as a real-time web service.

Train models by using an Azure Databricks cluster.

Deploy models by using the Azure Machine Learning designer.

#### Compute type

Compute type

Compute type

Compute type

Compute type

Correct Answer:

### Compute types

Attached compute

Inference cluster

Training cluster

### Answer Area

#### Requirement

Train models by using the Azure Machine Learning designer.

Score new data through a trained model published as a real-time web service.

Train models by using an Azure Databricks cluster.

Deploy models by using the Azure Machine Learning designer.

#### Compute type

Attached compute

Inference cluster

Training cluster

Attached compute

Section:

Explanation:

Box 1: Attached compute

| Training targets                               | Automated ML                | ML pipelines | Azure Machine Learning designer |
|------------------------------------------------|-----------------------------|--------------|---------------------------------|
| Local computer                                 | yes                         |              |                                 |
| Azure Machine Learning compute cluster         | yes & hyperparameter tuning | yes          | yes                             |
| <b>Azure Machine Learning compute instance</b> | yes & hyperparameter tuning | yes          | <b>yes</b>                      |

Box 2: Inference cluster

Box 3: Training cluster

Box 4: Attached compute

**QUESTION 28**

DRAG DROP

You are building an experiment using the Azure Machine Learning designer.

You split a dataset into training and testing sets. You select the Two-Class Boosted Decision Tree as the algorithm.

You need to determine the Area Under the Curve (AUC) of the model.

Which three modules should you use in sequence? To answer, move the appropriate modules from the list of modules to the answer area and arrange them in the correct order.

Select and Place:

**Modules**

- Export Data
- Tune Model Hyperparameters
- Cross Validate Model
- Evaluate Model
- Score Model
- Train Model

**Answer Area**

Answer area for the question, currently empty. A large watermark 'VCEplus.io' is visible across the area.



Correct Answer:

**Modules**

- Export Data
- Tune Model Hyperparameters
- Cross Validate Model
- 
- 
- 

**Answer Area**

Answer area for the correct answer, containing three modules in sequence: Train Model, Score Model, and Evaluate Model. A large watermark 'VCEplus.io' is visible across the area.

**Section:**

**Explanation:**

Step 1: Train Model

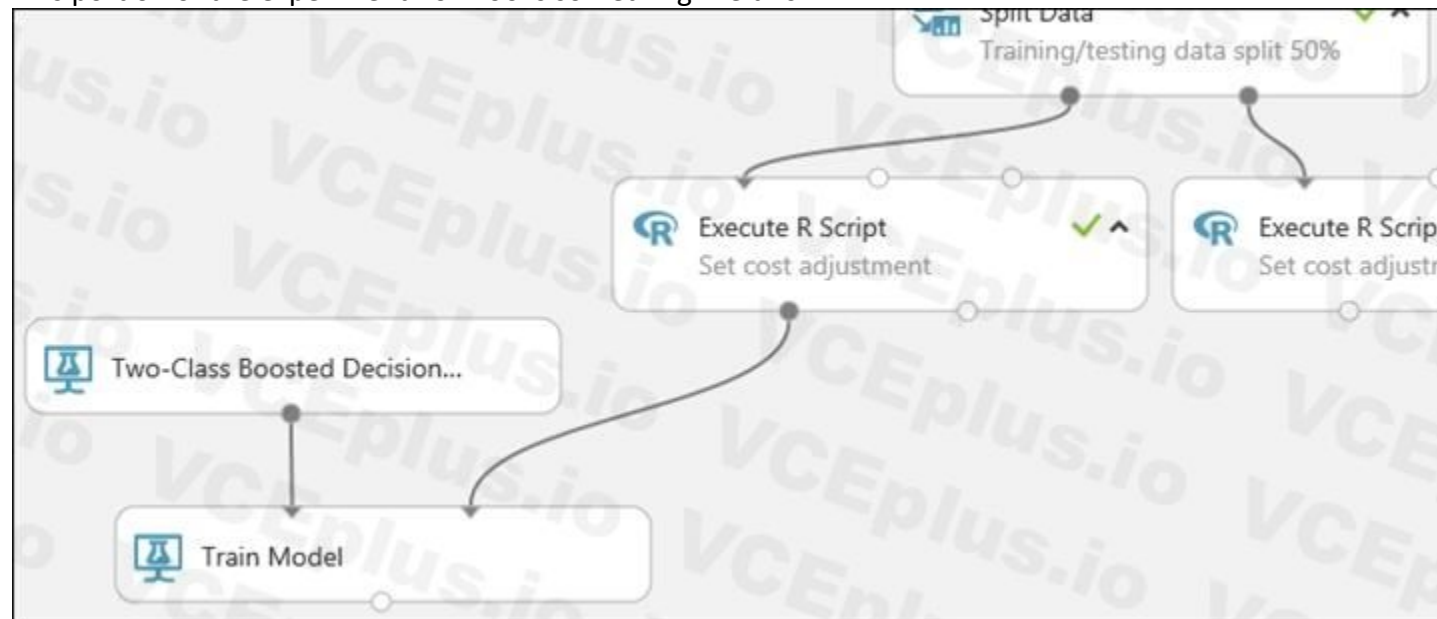
Two-Class Boosted Decision Tree

First, set up the boosted decision tree model.

1. Find the Two-Class Boosted Decision Tree module in the module palette and drag it onto the canvas.



- Find the Train Model module, drag it onto the canvas, and then connect the output of the Two-Class Boosted Decision Tree module to the left input port of the Train Model module. The Two-Class Boosted Decision Tree module initializes the generic model, and Train Model uses training data to train the model.
- Connect the left output of the left Execute R Script module to the right input port of the Train Model module (in this tutorial you used the data coming from the left side of the Split Data module for training). This portion of the experiment now looks something like this:

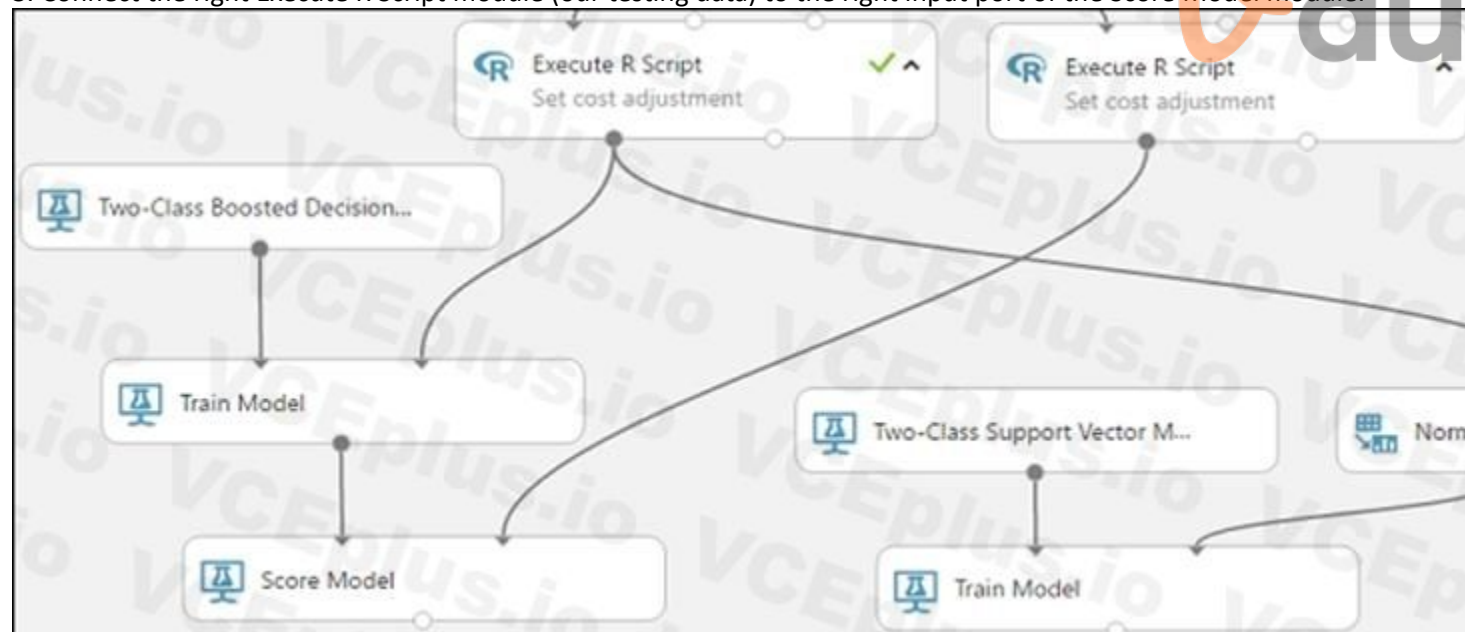


#### Step 2: Score Model

Score and evaluate the models You use the testing data that was separated out by the Split Data module to score our trained models. You can then compare the results of the two models to see which generated better results.

Add the Score Model modules

- Find the Score Model module and drag it onto the canvas.
- Connect the Train Model module that's connected to the Two-Class Boosted Decision Tree module to the left input port of the Score Model module.
- Connect the right Execute R Script module (our testing data) to the right input port of the Score Model module.

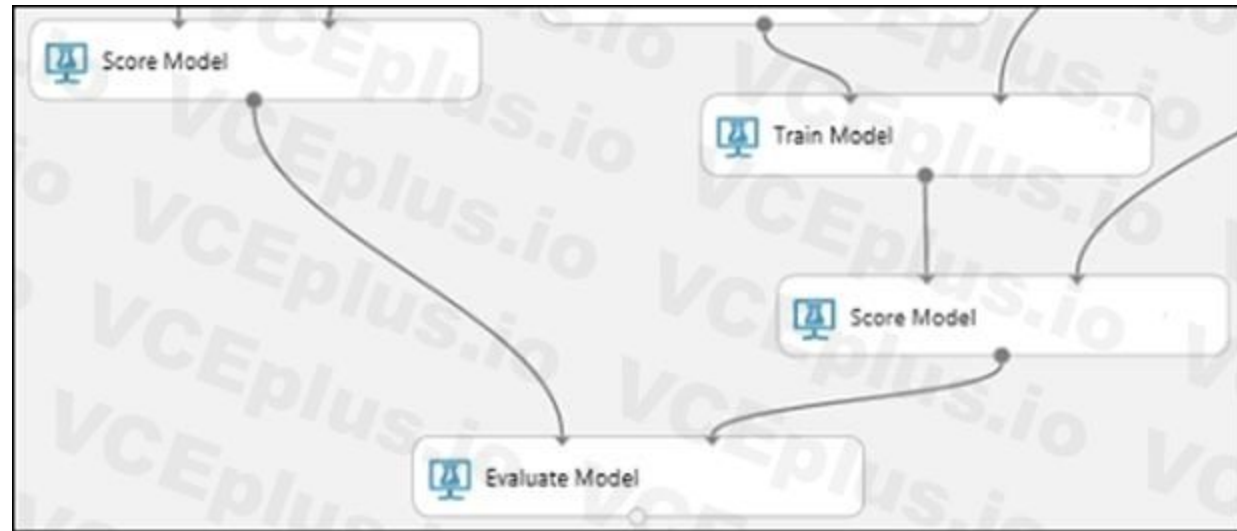


#### Step 3: Evaluate Model

To evaluate the two scoring results and compare them, you use an Evaluate Model module.

- Find the Evaluate Model module and drag it onto the canvas.
- Connect the output port of the Score Model module associated with the boosted decision tree model to the left input port of the Evaluate Model module.
- Connect the other Score Model module to the right input port.





### QUESTION 29

You plan to run a script as an experiment using a Script Run Configuration. The script uses modules from the scipy library as well as several Python packages that are not typically installed in a default conda environment. You plan to run the experiment on your local workstation for small datasets and scale out the experiment by running it on more powerful remote compute clusters for larger datasets.

You need to ensure that the experiment runs successfully on local and remote compute with the least administrative effort.

What should you do?

- A. Do not specify an environment in the run configuration for the experiment. Run the experiment by using the default environment.
- B. Create a virtual machine (VM) with the required Python configuration and attach the VM as a compute target. Use this compute target for all experiment runs.
- C. Create and register an Environment that includes the required packages. Use this Environment for all experiment runs.
- D. Create a config.yaml file defining the conda packages that are required and save the file in the experiment folder.
- E. Always run the experiment with an Estimator by using the default packages.

**Correct Answer: C**

**Section:**

**Explanation:**

If you have an existing Conda environment on your local computer, then you can use the service to create an environment object. By using this strategy, you can reuse your local interactive environment on remote runs.

Reference: <https://docs.microsoft.com/en-us/azure/machine-learning/how-to-use-environments>

### QUESTION 30

You write a Python script that processes data in a comma-separated values (CSV) file.

You plan to run this script as an Azure Machine Learning experiment.

The script loads the data and determines the number of rows it contains using the following code:

```

from azureml.core import Run
import pandas as pd

run = Run.get_context()
data = pd.read_csv('./data.csv')
rows = (len(data))
record row_count metric here
...

```

You need to record the row count as a metric named row\_count that can be returned using the get\_metrics method of the Run object after the experiment run completes.

Which code should you use?

- A. run.upload\_file('row\_count', './data.csv')

- B. `run.log('row_count', rows)`
- C. `run.tag('row_count', rows)`
- D. `run.log_table('row_count', rows)`
- E. `run.log_row('row_count', rows)`

**Correct Answer: B**

**Section:**

**Explanation:**

Log a numerical or string value to the run with the given name using `log(name, value, description=)`. Logging a metric to a run causes that metric to be stored in the run record in the experiment. You can log the same metric multiple times within a run, the result being considered a vector of that metric.

Example: `run.log("accuracy", 0.95)`

Incorrect Answers:

E: Using `log_row(name, description=None, **kwargs)` creates a metric with multiple columns as described in `kwargs`. Each named parameter generates a column with the value specified. `log_row` can be called once to log an arbitrary tuple, or multiple times in a loop to generate a complete table.

Example: `run.log_row("Y over X", x=1, y=0.4)`

Reference: <https://docs.microsoft.com/en-us/python/api/azureml-core/azureml.core.run>

### QUESTION 31

Note: This question-is part of a series of questions that present the same scenario. Each question-in the series contains a unique solution that might meet the stated goals. Some question-sets might have more than one correct solution, while others might not have a correct solution.

After you answer a question-in this section, you will NOT be able to return to it. As a result, these questions will not appear in the review screen.

You are creating a new experiment in Azure Machine Learning Studio.

One class has a much smaller number of observations than the other classes in the training set.

You need to select an appropriate data sampling strategy to compensate for the class imbalance.

Solution: You use the Synthetic Minority Oversampling Technique (SMOTE) sampling mode.

Does the solution meet the goal?

- A. Yes
- B. No

**Correct Answer: A**

**Section:**

**Explanation:**

SMOTE is used to increase the number of underrepresented cases in a dataset used for machine learning. SMOTE is a better way of increasing the number of rare cases than simply duplicating existing cases.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/smote>

### QUESTION 32

Note: This question-is part of a series of questions that present the same scenario. Each question-in the series contains a unique solution that might meet the stated goals. Some question-sets might have more than one correct solution, while others might not have a correct solution.

After you answer a question-in this section, you will NOT be able to return to it. As a result, these questions will not appear in the review screen.

You are creating a new experiment in Azure Machine Learning Studio.

One class has a much smaller number of observations than the other classes in the training set.

You need to select an appropriate data sampling strategy to compensate for the class imbalance.

Solution: You use the Stratified split for the sampling mode.

Does the solution meet the goal?

- A. Yes
- B. No

**Correct Answer: B**

**Section:**

**Explanation:**

Instead use the Synthetic Minority Oversampling Technique (SMOTE) sampling mode.

Note: SMOTE is used to increase the number of underrepresented cases in a dataset used for machine learning. SMOTE is a better way of increasing the number of rare cases than simply duplicating existing cases.

Reference: <https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/smote>

### QUESTION 33

You are creating a machine learning model.

You need to identify outliers in the data.

Which two visualizations can you use? Each correct answer presents a complete solution.

NOTE: Each correct selection is worth one point.

- A. Venn diagram
- B. Box plot
- C. ROC curve
- D. Random forest diagram
- E. Scatter plot

**Correct Answer: B, E**

**Section:**

**Explanation:**

The box-plot algorithm can be used to display outliers.

One other way to quickly identify Outliers visually is to create scatter plots.

Reference:

<https://blogs.msdn.microsoft.com/azuredev/2017/05/27/data-cleansing-tools-in-azure-machine-learning/>

The logo for Vdumps.com, featuring a stylized orange 'V' followed by the word 'dumps' in a grey, lowercase, sans-serif font.

### QUESTION 34

You are evaluating a completed binary classification machine learning model.

You need to use the precision as the evaluation metric.

Which visualization should you use?

- A. Violin plot
- B. Gradient descent
- C. Box plot
- D. Binary classification confusion matrix

**Correct Answer: D**

**Section:**

**Explanation:**

Incorrect Answers:

A: A violin plot is a visual that traditionally combines a box plot and a kernel density plot.

B: Gradient descent is a first-order iterative optimization algorithm for finding the minimum of a function. To find a local minimum of a function using gradient descent, one takes steps proportional to the negative of the gradient (or approximate gradient) of the function at the current point.

C: A box plot lets you see basic distribution information about your data, such as median, mean, range and quartiles but doesn't show you how your data looks throughout its range.

Reference:

<https://machinelearningknowledge.ai/confusion-matrix-and-performance-metrics-machine-learning/>



### QUESTION 35

You create a multi-class image classification deep learning model that uses the PyTorch deep learning framework.

You must configure Azure Machine Learning Hyperdrive to optimize the hyperparameters for the classification model.

You need to define a primary metric to determine the hyperparameter values that result in the model with the best accuracy score.

Which three actions must you perform? Each correct answer presents part of the solution.

NOTE: Each correct selection is worth one point.

- A. Set the `primary_metric_goal` of the estimator used to run the `bird_classifier_train.py` script to maximize.
- B. Add code to the `bird_classifier_train.py` script to calculate the validation loss of the model and log it as a float value with the key `loss`.
- C. Set the `primary_metric_goal` of the estimator used to run the `bird_classifier_train.py` script to minimize.
- D. Set the `primary_metric_name` of the estimator used to run the `bird_classifier_train.py` script to `accuracy`.
- E. Set the `primary_metric_name` of the estimator used to run the `bird_classifier_train.py` script to `loss`.
- F. Add code to the `bird_classifier_train.py` script to calculate the validation accuracy of the model and log it as a float value with the key `accuracy`.

**Correct Answer: A, D, F**

**Section:**

**Explanation:**

AD:

`primary_metric_name="accuracy"`, `primary_metric_goal=PrimaryMetricGoal.MAXIMIZE` Optimize the runs to maximize "accuracy". Make sure to log this value in your training script. Note: `primary_metric_name`: The name of the primary metric to optimize. The name of the primary metric needs to exactly match the name of the metric logged by the training script. `primary_metric_goal`: It can be either `PrimaryMetricGoal.MAXIMIZE` or `PrimaryMetricGoal.MINIMIZE` and determines whether the primary metric will be maximized or minimized when evaluating the runs.

F: The training script calculates the `val_accuracy` and logs it as "accuracy", which is used as the primary metric.

### QUESTION 36

You are performing a filter-based feature selection for a dataset to build a multi-class classifier by using Azure Machine Learning Studio.

The dataset contains categorical features that are highly correlated to the output label column.

You need to select the appropriate feature scoring statistical method to identify the key predictors.

Which method should you use?

- A. Kendall correlation
- B. Spearman correlation
- C. Chi-squared
- D. Pearson correlation

**Correct Answer: D**

**Section:**

**Explanation:**

Pearson's correlation statistic, or Pearson's correlation coefficient, is also known in statistical models as the  $r$  value. For any two variables, it returns a value that indicates the strength of the correlation. Pearson's correlation coefficient is the test statistics that measures the statistical relationship, or association, between two continuous variables. It is known as the best method of measuring the association between variables of interest because it is based on the method of covariance. It gives information about the magnitude of the association, or correlation, as well as the direction of the relationship.

Incorrect Answers:

C: The two-way chi-squared test is a statistical method that measures how close expected values are to actual results.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/filter-based-feature-selection> <https://www.statisticssolutions.com/pearsons-correlation-coefficient/>

### QUESTION 37

You plan to use automated machine learning to train a regression model. You have data that has features which have missing values, and categorical features with few distinct values.

You need to configure automated machine learning to automatically impute missing values and encode categorical features as part of the training task.

Which parameter and value pair should you use in the AutoMLConfig class?

- A. featurization = 'auto'
- B. enable\_voting\_ensemble = True
- C. task = 'classification'
- D. exclude\_nan\_labels = True
- E. enable\_tf = True

**Correct Answer: A**

**Section:**

**Explanation:**

Featurization str or FeaturizationConfig

Values: 'auto' / 'off' / FeaturizationConfig

Indicator for whether featurization step should be done automatically or not, or whether customized featurization should be used.

Column type is automatically detected. Based on the detected column type preprocessing/featurization is done as follows:

Categorical: Target encoding, one hot encoding, drop high cardinality categories, impute missing values.

Numeric: Impute missing values, cluster distance, weight of evidence.

DateTime: Several features such as day, seconds, minutes, hours etc.

Text: Bag of words, pre-trained Word embedding, text target encoding.

Reference:

<https://docs.microsoft.com/en-us/python/api/azureml-train-automl-client/azureml.train.automl.automlconfig.automlconfig>

#### QUESTION 38

You are building a regression model for estimating the number of calls during an event.

You need to determine whether the feature values achieve the conditions to build a Poisson regression model.

Which two conditions must the feature set contain? Each correct answer presents part of the solution.

NOTE: Each correct selection is worth one point.

- A. The label data must be a negative value.
- B. The label data must be whole numbers.
- C. The label data must be non-discrete.
- D. The label data must be a positive value.
- E. The label data can be positive or negative.

**Correct Answer: B, D**

**Section:**

**Explanation:**

Poisson regression is intended for use in regression models that are used to predict numeric values, typically counts. Therefore, you should use this module to create your regression model only if the values you are trying to predict fit the following conditions:

The response variable has a Poisson distribution.

Counts cannot be negative. The method will fail outright if you attempt to use it with negative labels.

A Poisson distribution is a discrete distribution; therefore, it is not meaningful to use this method with non-whole numbers.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/poisson-regression>

#### QUESTION 39

Note: This question-is part of a series of questions that present the same scenario. Each question-in the series contains a unique solution that might meet the stated goals. Some question-sets might have more than one correct solution, while others might not have a correct solution.

After you answer a question-in this section, you will NOT be able to return to it. As a result, these questions will not appear in the review screen.

You are creating a new experiment in Azure Machine Learning Studio.

One class has a much smaller number of observations than the other classes in the training set.

You need to select an appropriate data sampling strategy to compensate for the class imbalance.

Solution: You use the Principal Components Analysis (PCA) sampling mode.

Does the solution meet the goal?

- A. Yes
- B. No

**Correct Answer: B**

**Section:**

**Explanation:**

Instead use the Synthetic Minority Oversampling Technique (SMOTE) sampling mode.

Note: SMOTE is used to increase the number of underrepresented cases in a dataset used for machine learning. SMOTE is a better way of increasing the number of rare cases than simply duplicating existing cases.

Incorrect Answers:

The Principal Component Analysis module in Azure Machine Learning Studio (classic) is used to reduce the dimensionality of your training data. The module analyzes your data and creates a reduced feature set that captures all the information contained in the dataset, but in a smaller number of features.

Reference: <https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/smote> <https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/principal-component-analysis>

#### QUESTION 40

You are performing feature engineering on a dataset.

You must add a feature named CityName and populate the column value with the text London.

You need to add the new feature to the dataset.

Which Azure Machine Learning Studio module should you use?

- A. Edit Metadata
- B. Filter Based Feature Selection
- C. Execute Python Script
- D. Latent Dirichlet Allocation

**Correct Answer: A**

**Section:**

**Explanation:**

Typical metadata changes might include marking columns as features.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/edit-metadata>

#### QUESTION 41

You are evaluating a completed binary classification machine learning model.

You need to use the precision as the evaluation metric.

Which visualization should you use?

- A. violin plot
- B. Gradient descent
- C. Scatter plot
- D. Receiver Operating Characteristic (ROC) curve





**Correct Answer: D**

**Section:**

**Explanation:**

Receiver operating characteristic (or ROC) is a plot of the correctly classified labels vs. the incorrectly classified labels for a particular model.

Incorrect Answers:

A: A violin plot is a visual that traditionally combines a box plot and a kernel density plot.

B: Gradient descent is a first-order iterative optimization algorithm for finding the minimum of a function. To find a local minimum of a function using gradient descent, one takes steps proportional to the negative of the gradient (or approximate gradient) of the function at the current point.

C: A scatter plot graphs the actual values in your data against the values predicted by the model. The scatter plot displays the actual values along the X-axis, and displays the predicted values along the Y-axis. It also displays a line that illustrates the perfect prediction, where the predicted value exactly matches the actual value.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/how-to-understand-automated-ml#confusion-matrix>

#### QUESTION 42

Note: This question is part of a series of questions that present the same scenario. Each question in the series contains a unique solution that might meet the stated goals. Some question sets might have more than one correct solution, while others might not have a correct solution.

After you answer a question in this section, you will NOT be able to return to it. As a result, these questions will not appear in the review screen.

You plan to use a Python script to run an Azure Machine Learning experiment. The script creates a reference to the experiment run context, loads data from a file, identifies the set of unique values for the label column, and completes the experiment run:

```
from azureml.core import Run
import pandas as pd
run = Run.get_context()
data = pd.read_csv('data.csv')
label_vals = data['label'].unique()
Add code to record metrics here
run.complete()
```

The experiment must record the unique labels in the data as metrics for the run that can be reviewed later.

You must add code to the script to record the unique label values as run metrics at the point indicated by the comment.

Solution: Replace the comment with the following code:

```
run.upload_file('outputs/labels.csv', './data.csv')
```

Does the solution meet the goal?

A. Yes

B. No

**Correct Answer: B**

**Section:**

**Explanation:**

label\_vals has the unique labels (from the statement label\_vals = data['label'].unique()), and it has to be logged.

Note:

Instead use the run\_log function to log the contents in label\_vals:

```
for label_val in label_vals: run.log('Label Values', label_val)
```

Reference:

<https://www.element61.be/en/resource/azure-machine-learning-services-complete-toolbox-ai>

#### QUESTION 43

Note: This question is part of a series of questions that present the same scenario. Each question in the series contains a unique solution that might meet the stated goals. Some question sets might have more than one correct solution, while others might not have a correct solution.

After you answer a question in this section, you will NOT be able to return to it. As a result, these questions will not appear in the review screen.

You plan to use a Python script to run an Azure Machine Learning experiment. The script creates a reference to the experiment run context, loads data from a file, identifies the set of unique values for the label column, and

```
completes the experiment run:
from azureml.core import Run
import pandas as pd
run = Run.get_context()
data = pd.read_csv('data.csv')
label_vals = data['label'].unique()
Add code to record metrics here
run.complete()
```

The experiment must record the unique labels in the data as metrics for the run that can be reviewed later.

You must add code to the script to record the unique label values as run metrics at the point indicated by the comment.

Solution: Replace the comment with the following code:

```
run.log_table('Label Values', label_vals)
```

Does the solution meet the goal?

A. Yes

B. No

**Correct Answer: B**

**Section:**

**Explanation:**

Instead use the run\_log function to log the contents in label\_vals:

```
for label_val in label_vals: run.log('Label Values', label_val)
```

Reference:

<https://www.element61.be/en/resource/azure-machine-learning-services-complete-toolbox-ai>

#### QUESTION 44

Note: This question-is part of a series of questions that present the same scenario. Each question-in the series contains a unique solution that might meet the stated goals. Some question-sets might have more than one correct solution, while others might not have a correct solution.

After you answer a question-in this section, you will NOT be able to return to it. As a result, these questions will not appear in the review screen.

You plan to use a Python script to run an Azure Machine Learning experiment. The script creates a reference to the experiment run context, loads data from a file, identifies the set of unique values for the label column, and completes the experiment run:

```
from azureml.core import Run
import pandas as pd
run = Run.get_context()
data = pd.read_csv('data.csv')
label_vals = data['label'].unique()
Add code to record metrics here
run.complete()
```

The experiment must record the unique labels in the data as metrics for the run that can be reviewed later.

You must add code to the script to record the unique label values as run metrics at the point indicated by the comment.

Solution: Replace the comment with the following code:

```
for label_val in label_vals:
run.log('Label Values', label_val)
```

Does the solution meet the goal?

A. Yes

B. No

**Correct Answer: A**

**Section:**

**Explanation:**

The run\_log function is used to log the contents in label\_vals:

```
for label_val in label_vals: run.log('Label Values', label_val)
```

Reference: <https://www.element61.be/en/resource/azure-machine-learning-services-complete-toolbox-ai>

#### QUESTION 45

You are solving a classification task.

You must evaluate your model on a limited data sample by using k-fold cross-validation. You start by configuring a k parameter as the number of splits.

You need to configure the k parameter for the cross-validation.

Which value should you use?

- A. k=0.5
- B. k=0.01
- C. k=5
- D. k=1

**Correct Answer: C**

**Section:**

**Explanation:**

Leave One Out (LOO) cross-validation

Setting  $K = n$  (the number of observations) yields n-fold and is called leave-one out cross-validation (LOO), a special case of the K-fold approach.

LOO CV is sometimes useful but typically doesn't shake up the data enough. The estimates from each fold are highly correlated and hence their average can have high variance. This is why the usual choice is  $K=5$  or  $10$ . It provides a good compromise for the bias-variance tradeoff.

#### QUESTION 46

Note: This question is part of a series of questions that present the same scenario. Each question in the series contains a unique solution that might meet the stated goals. Some question sets might have more than one correct solution, while others might not have a correct solution.

After you answer a question in this section, you will NOT be able to return to it. As a result, these questions will not appear in the review screen.

You create a model to forecast weather conditions based on historical data.

You need to create a pipeline that runs a processing script to load data from a datastore and pass the processed data to a machine learning model training script.

Solution: Run the following code:

```
datastore = ws.get_default_datastore()
data_output = pd.read_csv("traindata.csv")
process_step = PythonScriptStep(script_name="process.py",
 arguments=["--data_for_train", data_output],
 outputs=[data_output], compute_target=aml_compute,
 source_directory=process_directory)
train_step = PythonScriptStep(script_name="train.py",
 arguments=["--data_for_train", data_output],
 inputs=[data_output], compute_target=aml_compute,
 source_directory=train_directory)
pipeline = Pipeline(workspace=ws, steps=[process_step, train_step])
```

Does the solution meet the goal?

- A. Yes
- B. No

**Correct Answer: B**

**Section:**

**Explanation:**

The two steps are present: process\_step and train\_step



The training data input is not setup correctly.

Note:

Data used in pipeline can be produced by one step and consumed in another step by providing a PipelineData object as an output of one step and an input of one or more subsequent steps.

PipelineData objects are also used when constructing Pipelines to describe step dependencies. To specify that a step requires the output of another step as input, use a PipelineData object in the constructor of both steps.

For example, the pipeline train step depends on the process\_step\_output output of the pipeline process step:

```
from azureml.pipeline.core import Pipeline, PipelineData
from azureml.pipeline.steps import PythonScriptStep
datastore = ws.get_default_datastore()
process_step_output = PipelineData("processed_data", datastore=datastore)
process_step = PythonScriptStep(script_name="process.py",
arguments=["--data_for_train", process_step_output],
outputs=[process_step_output],
compute_target=aml_compute,
source_directory=process_directory)
train_step = PythonScriptStep(script_name="train.py",
arguments=["--data_for_train", process_step_output],
inputs=[process_step_output],
compute_target=aml_compute,
source_directory=train_directory)
pipeline = Pipeline(workspace=ws, steps=[process_step, train_step])
```

Reference:

<https://docs.microsoft.com/en-us/python/api/azureml-pipeline-core/azureml.pipeline.core.pipelinedata?view=azure-ml-py>

#### QUESTION 47

Note: This question is part of a series of questions that present the same scenario. Each question in the series contains a unique solution that might meet the stated goals. Some question sets might have more than one correct solution, while others might not have a correct solution.

After you answer a question in this section, you will NOT be able to return to it. As a result, these questions will not appear in the review screen.

You create a model to forecast weather conditions based on historical data.

You need to create a pipeline that runs a processing script to load data from a datastore and pass the processed data to a machine learning model training script.

Solution: Run the following code:

```
datastore = ws.get_default_datastore()
data_output = PipelineData("processed_data", datastore=datastore)
process_step = PythonScriptStep(script_name="process.py",
arguments=["--data_for_train", data_output],
outputs=[data_output], compute_target=aml_compute,
source_directory=process_directory)
pipeline = Pipeline(workspace=ws, steps=[process_step])
```

Does the solution meet the goal?

- A. Yes
- B. No

**Correct Answer: B**

**Section:**

**Explanation:**

train\_step is missing.

Reference:

<https://docs.microsoft.com/en-us/python/api/azureml-pipeline-core/azureml.pipeline.core.pipelinedata?view=azure-ml-py>

#### QUESTION 48

Note: This question is part of a series of questions that present the same scenario. Each question in the series contains a unique solution that might meet the stated goals. Some question sets might have more than one correct solution, while others might not have a correct solution.

After you answer a question in this section, you will NOT be able to return to it. As a result, these questions will not appear in the review screen.

You create a model to forecast weather conditions based on historical data.

You need to create a pipeline that runs a processing script to load data from a datastore and pass the processed data to a machine learning model training script.

Solution: Run the following code:

```
datastore = ws.get_default_datastore()
data_input = PipelineData("raw_data", datastore=rawdatastore)
data_output = PipelineData("processed_data", datastore=datastore)
process_step = PythonScriptStep(script_name="process.py",
 arguments=["--data_for_train", data_input],
 outputs=[data_output], compute_target=aml_compute,
 source_directory=process_directory)
train_step = PythonScriptStep(script_name="train.py",
 arguments=["--data_for_train", data_input], inputs=[data_output],
 compute_target=aml_compute, source_directory=train_directory)
pipeline = Pipeline(workspace=ws, steps=[process_step, train_step])
```

Does the solution meet the goal?

- A. Yes
- B. No

**Correct Answer: B**

**Section:**

**Explanation:**

Note: Data used in pipeline can be produced by one step and consumed in another step by providing a PipelineData object as an output of one step and an input of one or more subsequent steps.

Compare with this example, the pipeline train step depends on the process\_step\_output output of the pipeline process step:

```
from azureml.pipeline.core import Pipeline, PipelineData
from azureml.pipeline.steps import PythonScriptStep
datastore = ws.get_default_datastore()
process_step_output = PipelineData("processed_data", datastore=datastore)
process_step = PythonScriptStep(script_name="process.py",
 arguments=["--data_for_train", process_step_output],
 outputs=[process_step_output],
 compute_target=aml_compute,
 source_directory=process_directory)
train_step = PythonScriptStep(script_name="train.py",
 arguments=["--data_for_train", process_step_output],
 inputs=[process_step_output],
 compute_target=aml_compute,
 source_directory=train_directory)
pipeline = Pipeline(workspace=ws, steps=[process_step, train_step])
```

Reference:

<https://docs.microsoft.com/en-us/python/api/azureml-pipeline-core/azureml.pipeline.core.pipelinedata?view=azure-ml-py>

#### QUESTION 49



Note: This question is part of a series of questions that present the same scenario. Each question in the series contains a unique solution that might meet the stated goals. Some question sets might have more than one correct solution, while others might not have a correct solution.

After you answer a question in this section, you will NOT be able to return to it. As a result, these questions will not appear in the review screen.

You have a Python script named train.py in a local folder named scripts. The script trains a regression model by using scikit-learn. The script includes code to load a training data file which is also located in the scripts folder.

You must run the script as an Azure ML experiment on a compute cluster named aml-compute.

You need to configure the run to ensure that the environment includes the required packages for model training. You have instantiated a variable named aml-compute that references the target compute cluster.

Solution: Run the following code:

```
from azureml.train.sklearn import SKLearn
sk_est = SKLearn(source_directory='./scripts',
 compute_target=aml-compute,
 entry_script='train.py')
```

Does the solution meet the goal?

A. Yes

B. No

**Correct Answer: A**

**Section:**

**Explanation:**

The scikit-learn estimator provides a simple way of launching a scikit-learn training job on a compute target. It is implemented through the SKLearn class, which can be used to support single-node CPU training.

Example:

```
from azureml.train.sklearn import SKLearn
}
estimator = SKLearn(source_directory=project_folder,
 compute_target=compute_target,
 entry_script='train_iris.py'
)
```

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/how-to-train-scikit-learn>



#### QUESTION 50

Note: This question is part of a series of questions that present the same scenario. Each question in the series contains a unique solution that might meet the stated goals. Some question sets might have more than one correct solution, while others might not have a correct solution.

After you answer a question in this section, you will NOT be able to return to it. As a result, these questions will not appear in the review screen.

You have a Python script named train.py in a local folder named scripts. The script trains a regression model by using scikit-learn. The script includes code to load a training data file which is also located in the scripts folder.

You must run the script as an Azure ML experiment on a compute cluster named aml-compute.

You need to configure the run to ensure that the environment includes the required packages for model training. You have instantiated a variable named aml-compute that references the target compute cluster.

Solution: Run the following code:

```
from azureml.train.dnn import TensorFlow
sk_est = TensorFlow(source_directory='./scripts',
 compute_target=aml-compute,
 entry_script='train.py')
```

Does the solution meet the goal?

A. Yes

B. No



**Correct Answer: B**

**Section:**

**Explanation:**

The scikit-learn estimator provides a simple way of launching a scikit-learn training job on a compute target. It is implemented through the SKLearn class, which can be used to support single-node CPU training.

Example:

```
from azureml.train.sklearn import SKLearn
}
estimator = SKLearn(source_directory=project_folder, compute_target=compute_target,
entry_script='train_iris.py')
```

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/how-to-train-scikit-learn>

#### QUESTION 51

Note: This question is part of a series of questions that present the same scenario. Each question in the series contains a unique solution that might meet the stated goals. Some question sets might have more than one correct solution, while others might not have a correct solution.

After you answer a question in this section, you will NOT be able to return to it. As a result, these questions will not appear in the review screen.

You have a Python script named train.py in a local folder named scripts. The script trains a regression model by using scikit-learn. The script includes code to load a training data file which is also located in the scripts folder.

You must run the script as an Azure ML experiment on a compute cluster named aml-compute.

You need to configure the run to ensure that the environment includes the required packages for model training. You have instantiated a variable named aml-compute that references the target compute cluster.

Solution: Run the following code:

```
from azureml.train.estimator import Estimator
sk_est = Estimator(source_directory='./scripts',
compute_target=aml-compute,
entry_script='train.py',
conda_packages=['scikit-learn'])
```

Does the solution meet the goal?

A. Yes

B. No

**Correct Answer: B**

**Section:**

**Explanation:**

The scikit-learn estimator provides a simple way of launching a scikit-learn training job on a compute target. It is implemented through the SKLearn class, which can be used to support single-node CPU training.

Example:

```
from azureml.train.sklearn import SKLearn
}
estimator = SKLearn(source_directory=project_folder, compute_target=compute_target,
entry_script='train_iris.py')
```

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/how-to-train-scikit-learn>

#### QUESTION 52

You create a multi-class image classification deep learning model that uses a set of labeled images. You create a script file named train.py that uses the PyTorch 1.3 framework to train the model.

You must run the script by using an estimator. The code must not require any additional Python libraries to be installed in the environment for the estimator. The time required for model training must be minimized.

You need to define the estimator that will be used to run the script.

Which estimator type should you use?



- A. TensorFlow
- B. PyTorch
- C. SKLearn
- D. Estimator

**Correct Answer: B**

**Section:**

**Explanation:**

For PyTorch, TensorFlow and Chainer tasks, Azure Machine Learning provides respective PyTorch, TensorFlow, and Chainer estimators to simplify using these frameworks.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/how-to-train-ml-models>

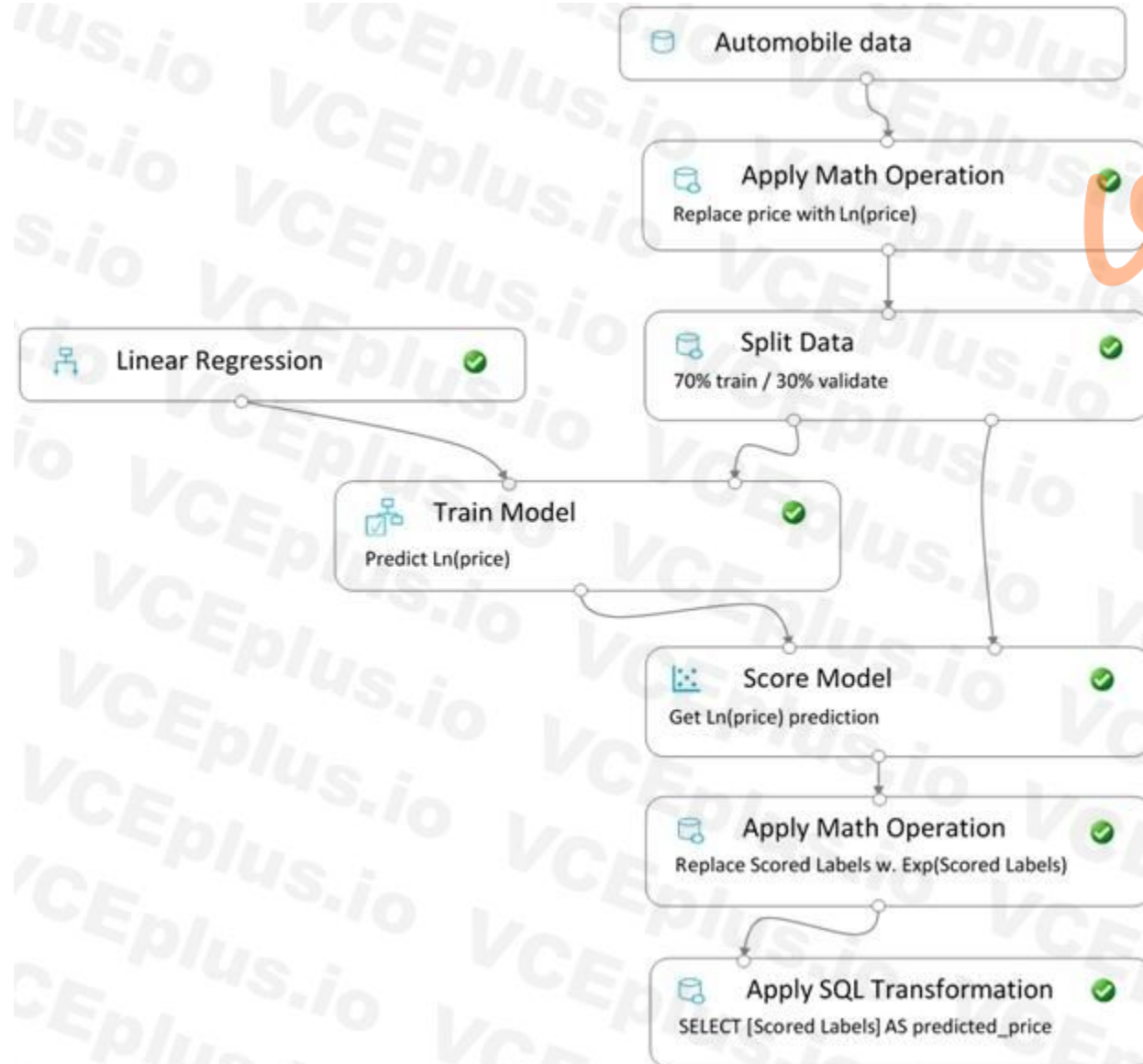
**QUESTION 53**

You create a pipeline in designer to train a model that predicts automobile prices.

Because of non-linear relationships in the data, the pipeline calculates the natural log (Ln) of the prices in the training data, trains a model to predict this natural log of price value, and then calculates the exponential of the scored label to get the predicted price.

The training pipeline is shown in the exhibit. (Click the Training pipeline tab.)

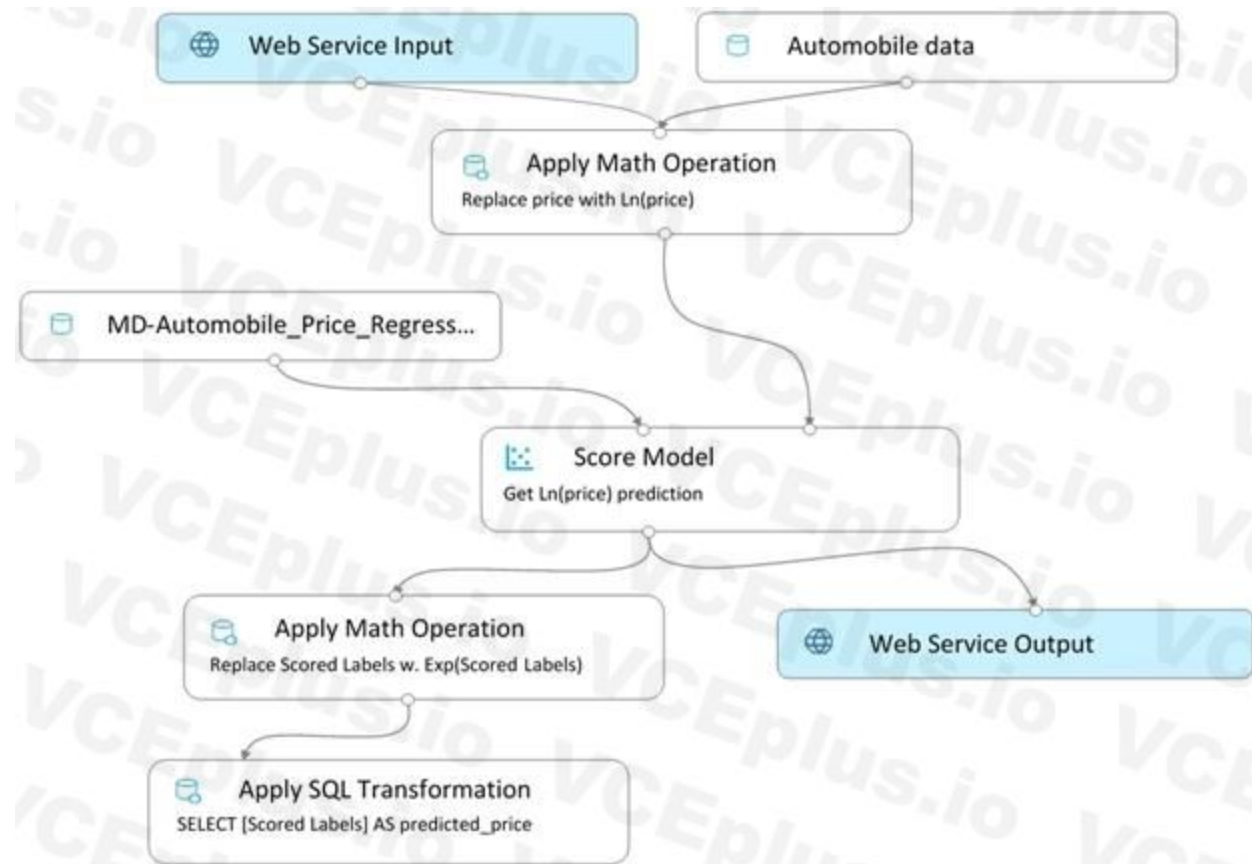
Training pipeline



You create a real-time inference pipeline from the training pipeline, as shown in the exhibit. (Click the Real-time pipeline tab.)

Real-time pipeline

**vdumps**



You need to modify the inference pipeline to ensure that the web service returns the exponential of the scored label as the predicted automobile price and that client applications are not required to include a price value in the input values.

Which three modifications must you make to the inference pipeline? Each correct answer presents part of the solution.

NOTE: Each correct selection is worth one point.

- A. Connect the output of the Apply SQL Transformation to the Web Service Output module.
- B. Replace the Web Service Input module with a data input that does not include the price column.
- C. Add a Select Columns module before the Score Model module to select all columns other than price.
- D. Replace the training dataset module with a data input that does not include the price column.
- E. Remove the Apply Math Operation module that replaces price with its natural log from the data flow.
- F. Remove the Apply SQL Transformation module from the data flow.

**Correct Answer: A, C, E**

**Section:**

#### QUESTION 54

You are creating a classification model for a banking company to identify possible instances of credit card fraud. You plan to create the model in Azure Machine Learning by using automated machine learning.

The training dataset that you are using is highly unbalanced.

You need to evaluate the classification model.

Which primary metric should you use?

- A. normalized\_mean\_absolute\_error
- B. AUC\_weighted
- C. accuracy
- D. normalized\_root\_mean\_squared\_error
- E. spearman\_correlation

**Correct Answer: B**

**Section:**

**Explanation:**

AUC\_weighted is a Classification metric.

Note: AUC is the Area under the Receiver Operating Characteristic Curve. Weighted is the arithmetic mean of the score for each class, weighted by the number of true instances in each class.

Incorrect Answers:

A: normalized\_mean\_absolute\_error is a regression metric, not a classification metric.

C: When comparing approaches to imbalanced classification problems, consider using metrics beyond accuracy such as recall, precision, and AUROC. It may be that switching the metric you optimize for during parameter selection or model selection is enough to provide desirable performance detecting the minority class.

D: normalized\_root\_mean\_squared\_error is a regression metric, not a classification metric.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/how-to-understand-automated-ml>

#### QUESTION 55

You create a machine learning model by using the Azure Machine Learning designer. You publish the model as a real-time service on an Azure Kubernetes Service (AKS) inference compute cluster. You make no change to the deployed endpoint configuration.

You need to provide application developers with the information they need to consume the endpoint.

Which two values should you provide to application developers? Each correct answer presents part of the solution.

NOTE: Each correct selection is worth one point.

- A. The name of the AKS cluster where the endpoint is hosted.
- B. The name of the inference pipeline for the endpoint.
- C. The URL of the endpoint.
- D. The run ID of the inference pipeline experiment for the endpoint.
- E. The key for the endpoint.



**Correct Answer: C, E**

**Section:**

**Explanation:**

Deploying an Azure Machine Learning model as a web service creates a REST API endpoint. You can send data to this endpoint and receive the prediction returned by the model.

You create a web service when you deploy a model to your local environment, Azure Container Instances, Azure Kubernetes Service, or field-programmable gate arrays (FPGA). You retrieve the URI used to access the web service by using the Azure Machine Learning SDK. If authentication is enabled, you can also use the SDK to get the authentication keys or tokens.

Example:

```
URL for the web service
```

```
scoring_uri = '<your web service URI>'
```

```
If the service is authenticated, set the key or token key = '<your key or token>'
```

Reference: <https://docs.microsoft.com/en-us/azure/machine-learning/how-to-consume-web-service>

#### QUESTION 56

Note: This question is part of a series of questions that present the same scenario. Each question in the series contains a unique solution that might meet the stated goals. Some question sets might have more than one correct solution, while others might not have a correct solution.

After you answer a question in this section, you will NOT be able to return to it. As a result, these questions will not appear in the review screen.

You create a model to forecast weather conditions based on historical data.

You need to create a pipeline that runs a processing script to load data from a datastore and pass the processed data to a machine learning model training script.

Solution: Run the following code:



```

data_store = Datastore.get(ws, "ml-data")
data_input = DataReference(
 datastore = data_store,
 data_reference_name = "training_data",
 path_on_datastore = "train/data.txt")
data_output = PipelineData("processed_data", datastore=datastore)
process_step = PythonScriptStep(script_name="process.py",
 arguments=["- -data", data_input], outputs=[data_output],
 compute_target=aml_compute, source_directory=process_directory)
train_step = PythonScriptStep(script_name="train.py",
 arguments=["- -data", data_output], inputs=[data_output],
 compute_target=aml_compute, source_directory=train_directory)
pipeline = Pipeline(workspace=ws, steps = [process_step, train_step])

```

Does the solution meet the goal?

- A. Yes
- B. No

**Correct Answer: A**

**Section:**

**Explanation:**

The two steps are present: process\_step and train\_step Data\_input correctly references the data in the data store.

Note:

Data used in pipeline can be produced by one step and consumed in another step by providing a PipelineData object as an output of one step and an input of one or more subsequent steps.

PipelineData objects are also used when constructing Pipelines to describe dependencies. To specify that a step requires the output of another step as input, use a PipelineData object in the constructor of both steps.

For example, the pipeline train step depends on the process\_step\_output output of the pipeline process step:

```
from azureml.pipeline.core import Pipeline, PipelineData
from azureml.pipeline.steps import PythonScriptStep
```

```
datastore = ws.get_default_datastore()
```

```
process_step_output = PipelineData("processed_data", datastore=datastore)
process_step = PythonScriptStep(script_name="process.py", arguments=["--data_for_train", process_step_output], outputs=[process_step_output], compute_target=aml_compute, source_directory=process_directory)
```

```
train_step = PythonScriptStep(script_name="train.py", arguments=["--data_for_train", process_step_output], inputs=[process_step_output], compute_target=aml_compute, source_directory=train_directory)
```

```
pipeline = Pipeline(workspace=ws, steps=[process_step, train_step])
```

Reference:

<https://docs.microsoft.com/en-us/python/api/azureml-pipeline-core/azureml.pipeline.core.pipelinedata?view=azure-ml-py>

#### QUESTION 57

You run an experiment that uses an AutoMLConfig class to define an automated machine learning task with a maximum of ten model training iterations. The task will attempt to find the best performing model based on a metric named accuracy.

You submit the experiment with the following code:

```

from azureml.core.experiment import Experiment
automl_experiment = Experiment(ws, 'automl_experiment')
automl_run = automl_experiment.submit(automl_config, show_output=True)

```

You need to create Python code that returns the best model that is generated by the automated machine learning task.

Which code segment should you use?

- A. best\_model = automl\_run.get\_details()
- B. best\_model = automl\_run.get\_metrics()
- C. best\_model = automl\_run.get\_file\_names()[1]

D. `best_model = automl_run.get_output()[1]`

**Correct Answer: D**

**Section:**

**Explanation:**

The `get_output` method returns the best run and the fitted model.

Reference:

<https://notebooks.azure.com/azureml/projects/azureml-getting-started/html/how-to-use-azureml/automated-machine-learning/classification/auto-ml-classification.ipynb>

#### QUESTION 58

You plan to use the Hyperdrive feature of Azure Machine Learning to determine the optimal hyperparameter values when training a model.

You must use Hyperdrive to try combinations of the following hyperparameter values. You must not apply an early termination policy.

`learning_rate`: any value between 0.001 and 0.1

`batch_size`: 16, 32, or 64

You need to configure the sampling method for the Hyperdrive experiment.

Which two sampling methods can you use? Each correct answer is a complete solution.

NOTE: Each correct selection is worth one point.

- A. No sampling
- B. Grid sampling
- C. Bayesian sampling
- D. Random sampling

**Correct Answer: C, D**

**Section:**

**Explanation:**

C: Bayesian sampling is based on the Bayesian optimization algorithm and makes intelligent choices on the hyperparameter values to sample next. It picks the sample based on how the previous samples performed, such that the new sample improves the reported primary metric.

Bayesian sampling does not support any early termination policy

Example:

```
from azureml.train.hyperdrive import BayesianParameterSampling
from azureml.train.hyperdrive import uniform, choice
param_sampling = BayesianParameterSampling({
"learning_rate": uniform(0.05, 0.1),
"batch_size": choice(16, 32, 64, 128)
}
)
```

D: In random sampling, hyperparameter values are randomly selected from the defined search space. Random sampling allows the search space to include both discrete and continuous hyperparameters.

Incorrect Answers:

B: Grid sampling can be used if your hyperparameter space can be defined as a choice among discrete values and if you have sufficient budget to exhaustively search over all values in the defined search space. Additionally, one can use automated early termination of poorly performing runs, which reduces wastage of resources.

Example, the following space has a total of six samples:

```
from azureml.train.hyperdrive import GridParameterSampling
from azureml.train.hyperdrive import choice
param_sampling = GridParameterSampling({
"num_hidden_layers": choice(1, 2, 3),
"batch_size": choice(16, 32)
}
)
```



Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/how-to-tune-hyperparameters>

#### QUESTION 59

You are training machine learning models in Azure Machine Learning. You use Hyperdrive to tune the hyperparameter.

In previous model training and tuning runs, many models showed similar performance.

You need to select an early termination policy that meets the following requirements:

accounts for the performance of all previous runs when evaluating the current run

avoids comparing the current run with only the best performing run to date

Which two early termination policies should you use? Each correct answer presents part of the solution.

NOTE: Each correct selection is worth one point.

- A. Median stopping
- B. Bandit
- C. Default
- D. Truncation selection

**Correct Answer: A, D**

**Section:**

**Explanation:**

The Median Stopping policy computes running averages across all runs and cancels runs whose best performance is worse than the median of the running averages. If no policy is specified, the hyperparameter tuning service will let all training runs execute to completion.

Reference: <https://docs.microsoft.com/en-us/python/api/azureml-train-core/azureml.train.hyperdrive.medianstoppingpolicy>

<https://docs.microsoft.com/en-us/python/api/azureml-train-core/azureml.train.hyperdrive.truncationselectionpolicy>

<https://docs.microsoft.com/en-us/python/api/azureml-train-core/azureml.train.hyperdrive.banditpolicy>

#### QUESTION 60

You use the Azure Machine Learning SDK in a notebook to run an experiment using a script file in an experiment folder.

The experiment fails.

You need to troubleshoot the failed experiment.

What are two possible ways to achieve this goal? Each correct answer presents a complete solution.

- A. Use the `get_metrics()` method of the run object to retrieve the experiment run logs.
- B. Use the `get_details_with_logs()` method of the run object to display the experiment run logs.
- C. View the log files for the experiment run in the experiment folder.
- D. View the logs for the experiment run in Azure Machine Learning studio.
- E. Use the `get_output()` method of the run object to retrieve the experiment run logs.

**Correct Answer: B, D**

**Section:**

**Explanation:**

Use `get_details_with_logs()` to fetch the run details and logs created by the run.

You can monitor Azure Machine Learning runs and view their logs with the Azure Machine Learning studio.

Incorrect Answers:

A: You can view the metrics of a trained model using `run.get_metrics()`. E: `get_output()` gets the output of the step as `PipelineData`.

Reference:

<https://docs.microsoft.com/en-us/python/api/azureml-pipeline-core/azureml.pipeline.core.steprun> <https://docs.microsoft.com/en-us/azure/machine-learning/how-to-monitor-view-training-logs>

#### QUESTION 61

You use the Two-Class Neural Network module in Azure Machine Learning Studio to build a binary classification model. You use the Tune Model Hyperparameters module to tune accuracy for the model. You need to configure the Tune Model Hyperparameters module. Which two values should you use? Each correct answer presents part of the solution. NOTE: Each correct selection is worth one point.

- A. Number of hidden nodes
- B. Learning Rate
- C. The type of the normalizer
- D. Number of learning iterations
- E. Hidden layer specification

**Correct Answer: D, E**

**Section:**

**Explanation:**

D: For Number of learning iterations, specify the maximum number of times the algorithm should process the training cases.

E: For Hidden layer specification, select the type of network architecture to create.

Between the input and output layers you can insert multiple hidden layers. Most predictive tasks can be accomplished easily with only one or a few hidden layers.

Reference: <https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/two-class-neural-network>

#### QUESTION 62

You create a binary classification model by using Azure Machine Learning Studio.

You must tune hyperparameters by performing a parameter sweep of the model. The parameter sweep must meet the following requirements:

- iterate all possible combinations of hyperparameters
  - minimize computing resources required to perform the sweep
- You need to perform a parameter sweep of the model. Which parameter sweep mode should you use?

- A. Random sweep
- B. Sweep clustering
- C. Entire grid
- D. Random grid

**Correct Answer: D**

**Section:**

**Explanation:**

Maximum number of runs on random grid: This option also controls the number of iterations over a random sampling of parameter values, but the values are not generated randomly from the specified range; instead, a matrix is created of all possible combinations of parameter values and a random sampling is taken over the matrix. This method is more efficient and less prone to regional oversampling or undersampling.

If you are training a model that supports an integrated parameter sweep, you can also set a range of seed values to use and iterate over the random seeds as well. This is optional, but can be useful for avoiding bias introduced by seed selection.

Incorrect Answers:

B: If you are building a clustering model, use Sweep Clustering to automatically determine the optimum number of clusters and other parameters.

C: Entire grid: When you select this option, the module loops over a grid predefined by the system, to try different combinations and identify the best learner. This option is useful for cases where you don't know what the best parameter settings might be and want to try all possible combination of values.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/tune-model-hyperparameters>

#### QUESTION 63

You are building a recurrent neural network to perform a binary classification.





You review the training loss, validation loss, training accuracy, and validation accuracy for each training epoch.

You need to analyze model performance.

You need to identify whether the classification model is overfitted.

Which of the following is correct?

- A. The training loss stays constant and the validation loss stays on a constant value and close to the training loss value when training the model.
- B. The training loss decreases while the validation loss increases when training the model.
- C. The training loss stays constant and the validation loss decreases when training the model.
- D. The training loss increases while the validation loss decreases when training the model.

**Correct Answer: B**

**Section:**

**Explanation:**

An overfit model is one where performance on the train set is good and continues to improve, whereas performance on the validation set improves to a point and then begins to degrade.

Reference:

<https://machinelearningmastery.com/diagnose-overfitting-underfitting-lstm-models/>

#### QUESTION 64

Note: This question is part of a series of questions that present the same scenario. Each question in the series contains a unique solution that might meet the stated goals. Some question sets might have more than one correct solution, while others might not have a correct solution.

After you answer a question in this section, you will NOT be able to return to it. As a result, these questions will not appear in the review screen.

You have a Python script named train.py in a local folder named scripts. The script trains a regression model by using scikit-learn. The script includes code to load a training data file which is also located in the scripts folder.

You must run the script as an Azure ML experiment on a compute cluster named aml-compute.

You need to configure the run to ensure that the environment includes the required packages for model training. You have instantiated a variable named aml-compute that references the target compute cluster.

Solution: Run the following code:

```
from azureml.train.estimator import Estimator
sk_est = Estimator(source_directory='./scripts',
compute_target=aml_compute,
entry_script='train.py')
```

Does the solution meet the goal?

- A. Yes
- B. No

**Correct Answer: B**

**Section:**

**Explanation:**

There is a missing line: `conda_packages=['scikit-learn']`, which is needed.

Correct example:

```
sk_est = Estimator(source_directory='./my-sklearn-proj',
script_params=script_params,
compute_target=compute_target,
entry_script='train.py',
conda_packages=['scikit-learn'])
```

Note:

The Estimator class represents a generic estimator to train data using any supplied framework.

This class is designed for use with machine learning frameworks that do not already have an Azure Machine Learning pre-configured estimator. Pre-configured estimators exist for Chainer, PyTorch, TensorFlow, and SKLearn.

Example:

```
from azureml.train.estimator import Estimator
script_params = {
to mount files referenced by mnist dataset
'--data-folder': ds.as_named_input('mnist').as_mount(),
'--regularization': 0.8
}
Reference:
https://docs.microsoft.com/en-us/python/api/azureml-train-core/azureml.train.estimator.estimator
```

#### QUESTION 65

You are performing clustering by using the K-means algorithm.  
You need to define the possible termination conditions.  
Which three conditions can you use? Each correct answer presents a complete solution.  
NOTE: Each correct selection is worth one point.

- A. Centroids do not change between iterations.
- B. The residual sum of squares (RSS) rises above a threshold.
- C. The residual sum of squares (RSS) falls below a threshold.
- D. A fixed number of iterations is executed.
- E. The sum of distances between centroids reaches a maximum.

**Correct Answer: A, C, D**

**Section:**

**Explanation:**

AD: The algorithm terminates when the centroids stabilize or when a specified number of iterations are completed.

C: A measure of how well the centroids represent the members of their clusters is the residual sum of squares or RSS, the squared distance of each vector from its centroid summed over all vectors. RSS is the objective function and our goal is to minimize it.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/k-means-clustering> <https://nlp.stanford.edu/IR-book/html/htmledition/k-means-1.html>

#### QUESTION 66

You are building a machine learning model for translating English language textual content into French language textual content.  
You need to build and train the machine learning model to learn the sequence of the textual content.  
Which type of neural network should you use?

- A. Multilayer Perceptions (MLPs)
- B. Convolutional Neural Networks (CNNs)
- C. Recurrent Neural Networks (RNNs)
- D. Generative Adversarial Networks (GANs)

**Correct Answer: C**

**Section:**

**Explanation:**

To translate a corpus of English text to French, we need to build a recurrent neural network (RNN).

Note: RNNs are designed to take sequences of text as inputs or return sequences of text as outputs, or both. They're called recurrent because the network's hidden layers have a loop in which the output and cell state from each time step become inputs at the next time step. This recurrence serves as a form of memory. It allows contextual information to flow through the network so that relevant outputs from previous time steps can be applied to network operations at the current time step.

Reference: <https://towardsdatascience.com/language-translation-with-rnns-d84d43b40571>

**QUESTION 67**

You create a binary classification model.

You need to evaluate the model performance.

Which two metrics can you use? Each correct answer presents a complete solution.

NOTE: Each correct selection is worth one point.

- A. relative absolute error
- B. precision
- C. accuracy
- D. mean absolute error
- E. coefficient of determination

**Correct Answer: B, C**

**Section:**

**Explanation:**

The evaluation metrics available for binary classification models are: Accuracy, Precision, Recall, F1 Score, and AUC.

Note: A very natural question-is: 'Out of the individuals whom the model, how many were classified correctly (TP)?'

This question-can be answered by looking at the Precision of the model, which is the proportion of positives that are classified correctly.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/studio/evaluate-model-performance>

**QUESTION 68**

You create a script that trains a convolutional neural network model over multiple epochs and logs the validation loss after each epoch. The script includes arguments for batch size and learning rate.

You identify a set of batch size and learning rate values that you want to try.

You need to use Azure Machine Learning to find the combination of batch size and learning rate that results in the model with the lowest validation loss.

What should you do?

- A. Run the script in an experiment based on an AutoMLConfig object
- B. Create a PythonScriptStep object for the script and run it in a pipeline
- C. Use the Automated Machine Learning interface in Azure Machine Learning studio
- D. Run the script in an experiment based on a ScriptRunConfig object
- E. Run the script in an experiment based on a HyperDriveConfig object

**Correct Answer: E**

**Section:**

**Explanation:**

Reference: <https://docs.microsoft.com/en-us/azure/machine-learning/how-to-tune-hyperparameters>

**QUESTION 69**

You use the Azure Machine Learning Python SDK to define a pipeline to train a model.

The data used to train the model is read from a folder in a datastore.

You need to ensure the pipeline runs automatically whenever the data in the folder changes.

What should you do?

- A. Set the regenerate\_outputs property of the pipeline to True
- B. Create a ScheduleRecurrence object with a Frequency of auto. Use the object to create a Schedule for the pipeline
- C. Create a PipelineParameter with a default value that references the location where the training data is stored

D. Create a Schedule for the pipeline. Specify the datastore in the datastore property, and the folder containing the training data in the path\_on\_datastore property

**Correct Answer: D**

**Section:**

**Explanation:**

Reference: <https://docs.microsoft.com/en-us/azure/machine-learning/how-to-trigger-published-pipeline>

#### QUESTION 70

You plan to run a Python script as an Azure Machine Learning experiment.

The script must read files from a hierarchy of folders. The files will be passed to the script as a dataset argument.

You must specify an appropriate mode for the dataset argument.

Which two modes can you use? Each correct answer presents a complete solution.

NOTE: Each correct selection is worth one point.

A. to\_pandas\_dataframe()

B. as\_download()

C. as\_upload()

D. as\_mount()

**Correct Answer: B**

**Section:**

**Explanation:**

Reference: <https://docs.microsoft.com/en-us/python/api/azureml-core/azureml.data.filedataset?view=azure-ml-py>

#### QUESTION 71

DRAG DROP

You create a multi-class image classification deep learning experiment by using the PyTorch framework. You plan to run the experiment on an Azure Compute cluster that has nodes with GPU's.

You need to define an Azure Machine Learning service pipeline to perform the monthly retraining of the image classification model. The pipeline must run with minimal cost and minimize the time required to train the model.

Which three pipeline steps should you run in sequence? To answer, move the appropriate actions from the list of actions to the answer area and arrange them in the correct order.

**Select and Place:**



| Actions                                                                                                                                          |
|--------------------------------------------------------------------------------------------------------------------------------------------------|
| Configure a DataTransferStep() to fetch new image data from public web portal, running on the cpu-compute compute target.                        |
| Configure an EstimatorStep() to run an estimator that runs the bird_classifier_train.py model training script on the gpu_compute compute target. |
| Configure a PythonScriptStep() to run both image_fetcher.py and image_resize.py on the cpu-compute compute target.                               |
| Configure an EstimatorStep() to run an estimator that runs the bird_classifier_train.py model training script on the cpu_compute compute target. |
| Configure a PythonScriptStep() to run image_fetcher.py on the cpu-compute compute target.                                                        |
| Configure a PythonScriptStep() to run image_resize.py on the cpu-compute compute target.                                                         |
| Configure a PythonScriptStep() to run bird_classifier_train.py on the cpu-compute compute target.                                                |
| Configure a PythonScriptStep() to run bird_classifier_train.py on the gpu-compute compute target.                                                |

| Answer Area |
|-------------|
|             |
|             |
|             |
|             |
|             |
|             |
|             |
|             |

**Correct Answer:**

| Actions                                                                                                                                          |
|--------------------------------------------------------------------------------------------------------------------------------------------------|
|                                                                                                                                                  |
|                                                                                                                                                  |
| Configure a PythonScriptStep() to run both image_fetcher.py and image_resize.py on the cpu-compute compute target.                               |
| Configure an EstimatorStep() to run an estimator that runs the bird_classifier_train.py model training script on the cpu_compute compute target. |
| Configure a PythonScriptStep() to run image_fetcher.py on the cpu-compute compute target.                                                        |
|                                                                                                                                                  |
| Configure a PythonScriptStep() to run bird_classifier_train.py on the cpu-compute compute target.                                                |
| Configure a PythonScriptStep() to run bird_classifier_train.py on the gpu-compute compute target.                                                |

| Answer Area                                                                                                                                      |
|--------------------------------------------------------------------------------------------------------------------------------------------------|
| Configure a DataTransferStep() to fetch new image data from public web portal, running on the cpu-compute compute target.                        |
| Configure a PythonScriptStep() to run image_resize.py on the cpu-compute compute target.                                                         |
| Configure an EstimatorStep() to run an estimator that runs the bird_classifier_train.py model training script on the gpu_compute compute target. |



**Section:**

**Explanation:**

Step 1: Configure a DataTransferStep() to fetch new image data...

Step 2: Configure a PythonScriptStep() to run image\_resize.y on the cpu-compute compute target.

Step 3: Configure the EstimatorStep() to run training script on the gpu\_compute computer target.

The PyTorch estimator provides a simple way of launching a PyTorch training job on a compute target.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/how-to-train-pytorch>

#### QUESTION 72

##### HOTSPOT

You plan to use Hyperdrive to optimize the hyperparameters selected when training a model. You create the following code to define options for the hyperparameter experiment:

```
import azureml.train.hyperdrive.parameter_expressions as pe
from azureml.train.hyperdrive import GridParameterSampling, HyperDriveConfig

param_sampling = GridParameterSampling({
 "max_depth" : pe.choice(6, 7, 8, 9),
 "learning_rate" : pe.choice(0.05, 0.1, 0.15)
})
hyperdrive_run_config = HyperDriveConfig(
 estimator = estimator,
 hyperparameter_sampling = param_sampling,
 policy = None,
 primary_metric_name = "auc",
 primary_metric_goal = PrimaryMetricGoal.MAXIMIZE,
 max_total_runs = 50,
 max_concurrent_runs = 4)
```

For each of the following statements, select Yes if the statement is true. Otherwise, select No.

NOTE: Each correct selection is worth one point.



Hot Area:

#### Answer Area

|                                                                                                                  | Yes                   | No                    |
|------------------------------------------------------------------------------------------------------------------|-----------------------|-----------------------|
| There will be 50 runs for this hyperparameter tuning experiment.                                                 | <input type="radio"/> | <input type="radio"/> |
| You can use the policy parameter in the HyperDriveConfig class to specify a security policy.                     | <input type="radio"/> | <input type="radio"/> |
| The experiment will create a run for every possible value for the learning rate parameter between 0.05 and 0.15. | <input type="radio"/> | <input type="radio"/> |

Answer Area:

## Answer Area

There will be 50 runs for this hyperparameter tuning experiment.

Yes

No

You can use the policy parameter in the HyperDriveConfig class to specify a security policy.

The experiment will create a run for every possible value for the learning rate parameter between 0.05 and 0.15.

### Section:

#### Explanation:

Box 1: No max\_total\_runs (50 here)

The maximum total number of runs to create. This is the upper bound; there may be fewer runs when the sample space is smaller than this value.

Box 2: Yes

Policy EarlyTerminationPolicy

The early termination policy to use. If None - the default, no early termination policy will be used.

Box 3: No

Discrete hyperparameters are specified as a choice among discrete values. choice can be:

one or more comma-separated values

a range object

any arbitrary list object

Reference:

<https://docs.microsoft.com/en-us/python/api/azureml-train-core/azureml.train.hyperdrive.hyperdriveconfig>

<https://docs.microsoft.com/en-us/azure/machine-learning/how-to-tune-hyperparameters>



### QUESTION 73

#### HOTSPOT

You are using Azure Machine Learning to train machine learning models. You need to compute target on which to remotely run the training script.

You run the following Python code:

```
from azureml.core.compute import ComputeTarget, AmlCompute
from azureml.core.compute_target import ComputeTargetException
the_cluster_name = "NewCompute"
config = AmlCompute.provisioning_configuration(vm_size= 'STANDARD_D2', max_nodes=3)
the_cluster = ComputeTarget.create(ws, the_cluster_name, config)
```

For each of the following statements, select Yes if the statement is true. Otherwise, select No.

NOTE: Each correct selection is worth one point.

#### Hot Area:



## Answer Area

|                                                                                                              | Yes                   | No                    |
|--------------------------------------------------------------------------------------------------------------|-----------------------|-----------------------|
| The compute is created in the same region as the Machine Learning service workspace.                         | <input type="radio"/> | <input type="radio"/> |
| The compute resource created by the code is displayed as a compute cluster in Azure Machine Learning studio. | <input type="radio"/> | <input type="radio"/> |
| The minimum number of nodes will be zero.                                                                    | <input type="radio"/> | <input type="radio"/> |

Answer Area:

## Answer Area

|                                                                                                              | Yes                              | No                    |
|--------------------------------------------------------------------------------------------------------------|----------------------------------|-----------------------|
| The compute is created in the same region as the Machine Learning service workspace.                         | <input checked="" type="radio"/> | <input type="radio"/> |
| The compute resource created by the code is displayed as a compute cluster in Azure Machine Learning studio. | <input checked="" type="radio"/> | <input type="radio"/> |
| The minimum number of nodes will be zero.                                                                    | <input checked="" type="radio"/> | <input type="radio"/> |

**Section:**

**Explanation:**

Box 1: Yes

The compute is created within your workspace region as a resource that can be shared with other users.

Box 2: Yes

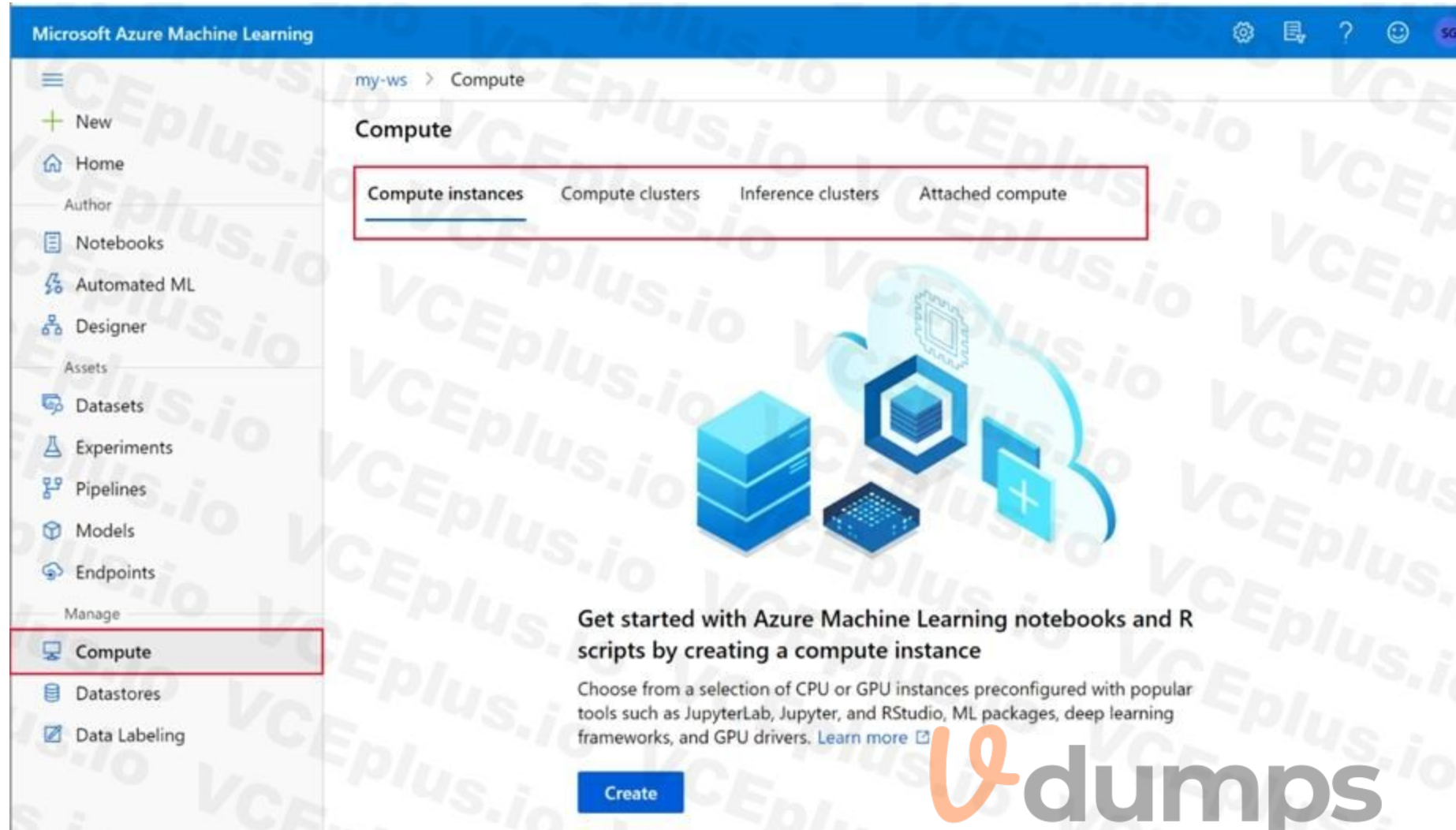
It is displayed as a compute cluster.

View compute targets

1. To see all compute targets for your workspace, use the following steps:
2. Navigate to Azure Machine Learning studio.
3. Under Manage, select Compute.
4. Select tabs at the top to show each type of compute target.

Vdumps





Box 3: Yes  
min\_nodes is not specified, so it defaults to 0.

Reference:

<https://docs.microsoft.com/en-us/python/api/azureml-core/azureml.core.compute.azuremlcompute.azuremlcompute.provisioningconfiguration>

<https://docs.microsoft.com/en-us/azure/machine-learning/how-to-create-attach-compute-studio>

#### QUESTION 74

##### HOTSPOT

You have an Azure blob container that contains a set of TSV files. The Azure blob container is registered as a datastore for an Azure Machine Learning service workspace. Each TSV file uses the same data schema.

You plan to aggregate data for all of the TSV files together and then register the aggregated data as a dataset in an Azure Machine Learning workspace by using the Azure Machine Learning SDK for Python.

You run the following code.

```
from azureml.core.workspace import Workspace
from azureml.core.datastore import Datastore
from azureml.core.dataset import Dataset
import pandas as pd
datastore_paths = (datastore, './data/*.tsv')
myDataset_1 = Dataset.File.from_files(path=datastore_paths)
myDataset_2 = Dataset.Tabular.from_delimited_files(path=datastore_paths, separator='\t')
```

For each of the following statements, select Yes if the statement is true. Otherwise, select No.

NOTE: Each correct selection is worth one point.

Hot Area:

## Answer Area

Yes

No

The myDataset\_1 dataset can be converted into a pandas dataframe by using the following method:

```
using myDataset_1.to_pandas_dataframe()
```

The myDataset\_1.to\_path() method returns an array of file paths for all of the TSV files in the dataset.

The myDataset\_2 dataset can be converted into a pandas dataframe by using the following method:

```
myDataset_2.to_pandas_dataframe()
```

Answer Area:

## Answer Area

Yes

No

The myDataset\_1 dataset can be converted into a pandas dataframe by using the following method:

```
using myDataset_1.to_pandas_dataframe()
```

The myDataset\_1.to\_path() method returns an array of file paths for all of the TSV files in the dataset.

The myDataset\_2 dataset can be converted into a pandas dataframe by using the following method:

```
myDataset_2.to_pandas_dataframe()
```

Section:

Explanation:

Box 1: No

FileDataset references single or multiple files in datastores or from public URLs. The TSV files need to be parsed.

Box 2: Yes

to\_path() gets a list of file paths for each file stream defined by the dataset.

Box 3: Yes

TabularDataset.to\_pandas\_dataframe loads all records from the dataset into a pandas DataFrame.

TabularDataset represents data in a tabular format created by parsing the provided file or list of files.

Note: TSV is a file extension for a tab-delimited file used with spreadsheet software. TSV stands for Tab Separated Values. TSV files are used for raw data and can be imported into and exported from spreadsheet software. TSV files are essentially text files, and the raw data can be viewed by text editors, though they are often used when moving raw data between spreadsheets.

Reference:



<https://docs.microsoft.com/en-us/python/api/azureml-core/azureml.data.tabulardataset>

### QUESTION 75

DRAG DROP

You create a multi-class image classification deep learning model.

The model must be retrained monthly with the new image data fetched from a public web portal. You create an Azure Machine Learning pipeline to fetch new data, standardize the size of images, and retrain the model.

You need to use the Azure Machine Learning SDK to configure the schedule for the pipeline.

Which four actions should you perform in sequence? To answer, move the appropriate actions from the list of actions to the answer area and arrange them in the correct order.

Select and Place:

#### Actions

Publish the pipeline.

Retrieve the pipeline ID.

Create a ScheduleRecurrence(frequency= 'Month', interval=1, start\_time='2019-01-01T00:00:00') object.

Define a pipeline parameter named **RunDate**.

Define a new Azure Machine Learning pipeline StepRun object with the step ID of the first step in the pipeline.

Define an Azure Machine Learning pipeline schedule using the schedule.create method with the defined recurrence specification.

#### Answer Area



Vdumps

Correct Answer:

| Actions                                                                                                         | Answer Area                                                                                                                    |
|-----------------------------------------------------------------------------------------------------------------|--------------------------------------------------------------------------------------------------------------------------------|
|                                                                                                                 | Publish the pipeline.                                                                                                          |
|                                                                                                                 | Retrieve the pipeline ID.                                                                                                      |
|                                                                                                                 | Create a ScheduleRecurrence(frequency= 'Month', interval=1, start_time='2019-01-01T00:00:00') object                           |
| Define a pipeline parameter named <b>RunDate</b> .                                                              | Define an Azure Machine Learning pipeline schedule using the schedule.create method with the defined recurrence specification. |
| Define a new Azure Machine Learning pipeline StepRun object with the step ID of the first step in the pipeline. |                                                                                                                                |
|                                                                                                                 |                                                                                                                                |

**Section:**

**Explanation:**

Step 1: Publish the pipeline.

To schedule a pipeline, you'll need a reference to your workspace, the identifier of your published pipeline, and the name of the experiment in which you wish to create the schedule.

Step 2: Retrieve the pipeline ID.

Needed for the schedule.

Step 3: Create a ScheduleRecurrence..

To run a pipeline on a recurring basis, you'll create a schedule. A Schedule associates a pipeline, an experiment, and a trigger.

First create a schedule. Example: Create a Schedule that begins a run every 15 minutes:

```
recurrence = ScheduleRecurrence(frequency="Minute", interval=15)
```

Step 4: Define an Azure Machine Learning pipeline schedule..

Example, continued:

```
recurring_schedule = Schedule.create(ws, name="MyRecurringSchedule",
description="Based on time",
pipeline_id=pipeline_id,
experiment_name=experiment_name,
recurrence=recurrence)
```

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/how-to-schedule-pipelines>

**QUESTION 76**

**HOTSPOT**

You create a script for training a machine learning model in Azure Machine Learning service.

You create an estimator by running the following code:



```
from azureml.core import Workspace, Datastore
from azureml.core.compute import ComputeTarget
from azureml.train.estimator import Estimator
work_space = Workspace.from_config()
data_source = work_space.get_default_datastore()
train_cluster = ComputeTarget(workspace=work_space, name='train-cluster')
estimator = Estimator(source_directory =
 'training-experiment',
 script_params = { '--data-folder' : data_source.as_mount(), '--regularization':0.8},
 compute_target = train_cluster,
 entry_script = 'train.py',
 conda_packages = ['scikit-learn'])
```

For each of the following statements, select Yes if the statement is true. Otherwise, select No.

NOTE: Each correct selection is worth one point.

Hot Area:

### Answer Area

Yes

No

The estimator will look for the files it needs to run an experiment in the training-experiment directory of the local compute environment.

The estimator will mount the local data-folder folder and make it available to the script through a parameter.

The train.py script file will be created if it does not exist.

The estimator can run Scikit-learn experiments.

Answer Area:

## Answer Area

Yes

No

The estimator will look for the files it needs to run an experiment in the training-experiment directory of the local compute environment.  Yes  No

The estimator will mount the local data-folder folder and make it available to the script through a parameter.  Yes  No

The train.py script file will be created if it does not exist.  Yes  No

The estimator can run Scikit-learn experiments.  Yes  No

### Section:

### Explanation:

Box 1: Yes

Parameter `source_directory` is a local directory containing experiment configuration and code files needed for a training job.

Box 2: Yes

`script_params` is a dictionary of command-line arguments to pass to the training script specified in `entry_script`.

Box 3: No

Box 4: Yes

The `conda_packages` parameter is a list of strings representing conda packages to be added to the Python environment for the experiment.

### QUESTION 77

#### HOTSPOT

You have a Python data frame named `salesData` in the following format:

|   | shop   | 2017 | 2018 |
|---|--------|------|------|
| 0 | Shop X | 34   | 25   |
| 1 | Shop Y | 65   | 76   |
| 2 | Shop Z | 48   | 55   |

The data frame must be unpivoted to a long data format as follows:

|   | shop   | year | value |
|---|--------|------|-------|
| 0 | Shop X | 2017 | 34    |
| 1 | Shop Y | 2017 | 65    |
| 2 | Shop Z | 2017 | 48    |
| 3 | Shop X | 2018 | 25    |
| 4 | Shop Y | 2018 | 76    |
| 5 | Shop Z | 2018 | 55    |

You need to use the `pandas.melt()` function in Python to perform the transformation.

How should you complete the code segment? To answer, select the appropriate options in the answer area.

NOTE: Each correct selection is worth one point.

### Hot Area:

Answer Area

```
import pandas as pd
salesData = pd.melt(
 ,id_vars='
 ',value_vars=
)
```

|           |
|-----------|
| dataFrame |
| pandas    |
| salesData |
| year      |

|                        |
|------------------------|
| shop                   |
| year                   |
| value                  |
| Shop X, Shop Y, Shop Z |

|                  |
|------------------|
| 'shop'           |
| 'year'           |
| ['year']         |
| ['2017', '2018'] |

Answer Area:

Answer Area

```
import pandas as pd
salesData = pd.melt(
 ,id_vars='
 ',value_vars=
)
```

|           |
|-----------|
| dataFrame |
| pandas    |
| salesData |
| year      |

|                        |
|------------------------|
| shop                   |
| year                   |
| value                  |
| Shop X, Shop Y, Shop Z |

|                  |
|------------------|
| 'shop'           |
| 'year'           |
| ['year']         |
| ['2017', '2018'] |

Section:

Explanation:

Box 1: dataFrame

Syntax: pandas.melt(frame, id\_vars=None, value\_vars=None, var\_name=None, value\_name='value', col\_level=None)[source]

Where frame is a DataFrame

Box 2: shop

Parameter id\_vars id\_vars : tuple, list, or ndarray, optional

Column(s) to use as identifier variables.

Box 3: ['2017','2018']

value\_vars : tuple, list, or ndarray, optional

Column(s) to unpivot. If not specified, uses all columns that are not set as id\_vars.

Example:

```
df = pd.DataFrame({'A': {0: 'a', 1: 'b', 2: 'c'},
```

```
... 'B': {0: 1, 1: 3, 2: 5},
```

```
... 'C': {0: 2, 1: 4, 2: 6}})
```

```
pd.melt(df, id_vars=['A'], value_vars=['B', 'C'])
```

A variable value

0 a B 1

1 b B 3

2 c B 5

3 a C 2

4 b C 4

5 c C 6

References:



<https://pandas.pydata.org/pandas-docs/stable/reference/api/pandas.melt.html>

**QUESTION 78**

**HOTSPOT**

You are working on a classification task. You have a dataset indicating whether a student would like to play soccer and associated attributes. The dataset includes the following columns:

| Name          | Description                  |
|---------------|------------------------------|
| IsPlaySoccer  | Values can be 1 and 0.       |
| Gender        | Values can be M or F.        |
| PrevExamMarks | Stores values from 0 to 100  |
| Height        | Stores values in centimeters |
| Weight        | Stores values in kilograms   |

You need to classify variables by type.

Which variable should you add to each category? To answer, select the appropriate options in the answer area.

NOTE: Each correct selection is worth one point.

**Hot Area:**

**Answer Area**

| Category              | Variables                                                                                                                                                                                                                     |
|-----------------------|-------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|
| Categorical variables | <input type="checkbox"/> Gender, IsPlaySoccer<br><input checked="" type="checkbox"/> Gender, PrevExamMarks, Height, Weight<br><input type="checkbox"/> PrevExamMarks, Height, Weight<br><input type="checkbox"/> IsPlaySoccer |
| Continuous variables  | <input type="checkbox"/> Gender, IsPlaySoccer<br><input type="checkbox"/> Gender, PrevExamMarks, Height, Weight<br><input type="checkbox"/> PrevExamMarks, Height, Weight<br><input type="checkbox"/> IsPlaySoccer            |

**Answer Area:**





**Answer Area**

| Category              | Variables                                                                                                                                                                                                                     |
|-----------------------|-------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|
| Categorical variables | <input type="checkbox"/> Gender, IsPlaySoccer<br><input checked="" type="checkbox"/> Gender, PrevExamMarks, Height, Weight<br><input type="checkbox"/> PrevExamMarks, Height, Weight<br><input type="checkbox"/> IsPlaySoccer |
| Continuous variables  | <input type="checkbox"/> Gender, IsPlaySoccer<br><input type="checkbox"/> Gender, PrevExamMarks, Height, Weight<br><input checked="" type="checkbox"/> PrevExamMarks, Height, Weight<br><input type="checkbox"/> IsPlaySoccer |

**Section:**

**Explanation:**

References:

<https://www.edureka.co/blog/classification-algorithms/>

**QUESTION 79**

HOTSPOT

You plan to preprocess text from CSV files. You load the Azure Machine Learning Studio default stop words list.

You need to configure the Preprocess Text module to meet the following requirements:

Ensure that multiple related words from a single canonical form.

Remove pipe characters from text.

Remove words to optimize information retrieval.

Which three options should you select? To answer, select the appropriate options in the answer area.

NOTE: Each correct selection is worth one point.

**Hot Area:**



**Answer Area**

Preprocess Text

Language  
English

Remove by part of speech  
False

Text column to clean

**Selected columns:**  
Column names: **String, Feature**

Launch column selector

- Remove stop words
- Lemmatization
- Detect sentences
- Normalize case to lowercase
- Remove numbers
- Remove special characters
- Remove duplicate characters
- Remove email addresses
- Remove URLs
- Expand verb contractions
- Normalize backslashes to slashes
- Split tokens on special characters

Answer Area:



**Answer Area**

Preprocess Text

Language  
English

Remove by part of speech  
False

Text column to clean  
**Selected columns:**  
 Column names: **String, Feature**

Launch column selector

Remove stop words

Lemmatization

Detect sentences

Normalize case to lowercase

Remove numbers

Remove special characters

Remove duplicate characters

Remove email addresses

Remove URLs

Expand verb contractions

Normalize backslashes to slashes

Split tokens on special characters



**Section:**

**Explanation:**

Box 1: Remove stop words

Remove words to optimize information retrieval.

Remove stop words: Select this option if you want to apply a predefined stopword list to the text column. Stop word removal is performed before any other processes.

Box 2: Lemmatization

Ensure that multiple related words from a single canonical form.

Lemmatization converts multiple related words to a single canonical form

Box 3: Remove special characters

Remove special characters: Use this option to replace any non-alphanumeric special characters with the pipe | character.

References:

<https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/preprocess-text>

**QUESTION 80**

DRAG DROP

You have a dataset that contains over 150 features. You use the dataset to train a Support Vector Machine (SVM) binary classifier.

You need to use the Permutation Feature Importance module in Azure Machine Learning Studio to compute a set of feature importance scores for the dataset.

In which order should you perform the actions? To answer, move all actions from the list of actions to the answer area and arrange them in the correct order.

Select and Place:

**Actions**

Add a Two-Class Support Vector Machine module to initialize the SVM classifier.

Set the Metric for measuring performance property to **Classification - Accuracy** and then run the experiment.

Add a Permutation Feature Importance module and connect the trained model and test dataset.

Add a dataset to the experiment.

Add a Split Data module to create training and test datasets.

**Answer Area**



Correct Answer:



## Actions

|  |
|--|
|  |
|  |
|  |
|  |
|  |

## Answer Area

Add a Two-Class Support Vector Machine module to initialize the SVM classifier.

Add a dataset to the experiment.

Add a Split Data module to create training and test datasets.

Add a Permutation Feature Importance module and connect the trained model and test dataset.

Set the Metric for measuring performance property to **Classification - Accuracy** and then run the experiment.



### Section:

### Explanation:

Step 1: Add a Two-Class Support Vector Machine module to initialize the SVM classifier.

Step 2: Add a dataset to the experiment

Step 3: Add a Split Data module to create training and test dataset.

To generate a set of feature scores requires that you have an already trained model, as well as a test dataset.

Step 4: Add a Permutation Feature Importance module and connect to the trained model and test dataset.

Step 5: Set the Metric for measuring performance property to Classification - Accuracy and then run the experiment.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/two-class-support-vector-machine>

<https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/permutation-feature-importance>

### QUESTION 81

#### HOTSPOT

You are using the Hyperdrive feature in Azure Machine Learning to train a model.

You configure the Hyperdrive experiment by running the following code:

```
from azureml.train.hyperdrive import RandomParameterSampling
param_sampling = RandomParameterSampling({
 "learning_rate": normal(10, 3),
 "keep_probability": uniform(0.05, 0.1),
 "batch_size": choice(16, 32, 64, 128)
 "number_of_hidden_layers": choice(range(3,5))
})
```

For each of the following statements, select Yes if the statement is true. Otherwise, select No.

NOTE: Each correct selection is worth one point.



**Hot Area:**

By defining sampling in this manner, every possible combination of the parameters will be tested.

| Yes                   | No                    |
|-----------------------|-----------------------|
| <input type="radio"/> | <input type="radio"/> |

Random values of the learning\_rate parameter will be selected from a normal distribution with a mean of 10 and a standard deviation of 3.

|                       |                       |
|-----------------------|-----------------------|
| <input type="radio"/> | <input type="radio"/> |
|-----------------------|-----------------------|

The keep\_probability parameter value will always be either **0.05** or **0.1**.

|                       |                       |
|-----------------------|-----------------------|
| <input type="radio"/> | <input type="radio"/> |
|-----------------------|-----------------------|

Random values for the number\_of\_hidden\_layers parameter will be selected from a normal distribution with a mean of 3 and a standard deviation of 5.

|                       |                       |
|-----------------------|-----------------------|
| <input type="radio"/> | <input type="radio"/> |
|-----------------------|-----------------------|

**Answer Area:**

By defining sampling in this manner, every possible combination of the parameters will be tested.

| Yes                              | No                    |
|----------------------------------|-----------------------|
| <input checked="" type="radio"/> | <input type="radio"/> |

Random values of the learning\_rate parameter will be selected from a normal distribution with a mean of 10 and a standard deviation of 3.

|                                  |                       |
|----------------------------------|-----------------------|
| <input checked="" type="radio"/> | <input type="radio"/> |
|----------------------------------|-----------------------|

The keep\_probability parameter value will always be either **0.05** or **0.1**.

|                       |                                  |
|-----------------------|----------------------------------|
| <input type="radio"/> | <input checked="" type="radio"/> |
|-----------------------|----------------------------------|

Random values for the number\_of\_hidden\_layers parameter will be selected from a normal distribution with a mean of 3 and a standard deviation of 5.

|                       |                                  |
|-----------------------|----------------------------------|
| <input type="radio"/> | <input checked="" type="radio"/> |
|-----------------------|----------------------------------|

**Section:**

**Explanation:**

Box 1: Yes  
In random sampling, hyperparameter values are randomly selected from the defined search space. Random sampling allows the search space to include both discrete and continuous hyperparameters.

Box 2: Yes  
learning\_rate has a normal distribution with mean value 10 and a standard deviation of 3.

Box 3: No  
keep\_probability has a uniform distribution with a minimum value of 0.05 and a maximum value of 0.1.

Box 4: No  
number\_of\_hidden\_layers takes on one of the values [3, 4, 5].

Reference:  
<https://docs.microsoft.com/en-us/azure/machine-learning/how-to-tune-hyperparameters>

**QUESTION 82**

**HOTSPOT**

You create a binary classification model to predict whether a person has a disease. You need to detect possible classification errors.



Which error type should you choose for each description? To answer, select the appropriate options in the answer area.

NOTE: Each correct selection is worth one point.

Hot Area:

Answer Area

| Description                                                                           | Error type                                                                                                                                                                                                                                                                                                                  |                |                |                 |                 |
|---------------------------------------------------------------------------------------|-----------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|----------------|----------------|-----------------|-----------------|
| A person has a disease. The model classifies the case as having a disease.            | <div style="border: 1px solid black; padding: 2px;"><div style="background-color: #cccccc; padding: 2px; text-align: right;">▼</div><table border="1" style="width: 100%;"><tr><td>True Positives</td></tr><tr><td>True Negatives</td></tr><tr><td>False Positives</td></tr><tr><td>False Negatives</td></tr></table></div> | True Positives | True Negatives | False Positives | False Negatives |
| True Positives                                                                        |                                                                                                                                                                                                                                                                                                                             |                |                |                 |                 |
| True Negatives                                                                        |                                                                                                                                                                                                                                                                                                                             |                |                |                 |                 |
| False Positives                                                                       |                                                                                                                                                                                                                                                                                                                             |                |                |                 |                 |
| False Negatives                                                                       |                                                                                                                                                                                                                                                                                                                             |                |                |                 |                 |
| A person does not have a disease. The model classifies the case as having no disease. | <div style="border: 1px solid black; padding: 2px;"><div style="background-color: #cccccc; padding: 2px; text-align: right;">▼</div><table border="1" style="width: 100%;"><tr><td>True Positives</td></tr><tr><td>True Negatives</td></tr><tr><td>False Positives</td></tr><tr><td>False Negatives</td></tr></table></div> | True Positives | True Negatives | False Positives | False Negatives |
| True Positives                                                                        |                                                                                                                                                                                                                                                                                                                             |                |                |                 |                 |
| True Negatives                                                                        |                                                                                                                                                                                                                                                                                                                             |                |                |                 |                 |
| False Positives                                                                       |                                                                                                                                                                                                                                                                                                                             |                |                |                 |                 |
| False Negatives                                                                       |                                                                                                                                                                                                                                                                                                                             |                |                |                 |                 |
| A person does not have a disease. The model classifies the case as having a disease.  | <div style="border: 1px solid black; padding: 2px;"><div style="background-color: #cccccc; padding: 2px; text-align: right;">▼</div><table border="1" style="width: 100%;"><tr><td>True Positives</td></tr><tr><td>True Negatives</td></tr><tr><td>False Positives</td></tr><tr><td>False Negatives</td></tr></table></div> | True Positives | True Negatives | False Positives | False Negatives |
| True Positives                                                                        |                                                                                                                                                                                                                                                                                                                             |                |                |                 |                 |
| True Negatives                                                                        |                                                                                                                                                                                                                                                                                                                             |                |                |                 |                 |
| False Positives                                                                       |                                                                                                                                                                                                                                                                                                                             |                |                |                 |                 |
| False Negatives                                                                       |                                                                                                                                                                                                                                                                                                                             |                |                |                 |                 |
| A person has a disease. The model classifies the case as having no disease.           | <div style="border: 1px solid black; padding: 2px;"><div style="background-color: #cccccc; padding: 2px; text-align: right;">▼</div><table border="1" style="width: 100%;"><tr><td>True Positives</td></tr><tr><td>True Negatives</td></tr><tr><td>False Positives</td></tr><tr><td>False Negatives</td></tr></table></div> | True Positives | True Negatives | False Positives | False Negatives |
| True Positives                                                                        |                                                                                                                                                                                                                                                                                                                             |                |                |                 |                 |
| True Negatives                                                                        |                                                                                                                                                                                                                                                                                                                             |                |                |                 |                 |
| False Positives                                                                       |                                                                                                                                                                                                                                                                                                                             |                |                |                 |                 |
| False Negatives                                                                       |                                                                                                                                                                                                                                                                                                                             |                |                |                 |                 |

Answer Area:

| Answer Area | Description                                                                           | Error type                                                                                                                                                                                                                                                                                                                                     |
|-------------|---------------------------------------------------------------------------------------|------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|
|             | A person has a disease. The model classifies the case as having a disease.            | <div style="border: 1px solid black; padding: 2px;"> <div style="background-color: #cccccc; text-align: right; padding: 2px;">▼</div> <div style="padding: 2px;">True Positives</div> <div style="padding: 2px;">True Negatives</div> <div style="padding: 2px;">False Positives</div> <div style="padding: 2px;">False Negatives</div> </div> |
|             | A person does not have a disease. The model classifies the case as having no disease. | <div style="border: 1px solid black; padding: 2px;"> <div style="background-color: #cccccc; text-align: right; padding: 2px;">▼</div> <div style="padding: 2px;">True Positives</div> <div style="padding: 2px;">True Negatives</div> <div style="padding: 2px;">False Positives</div> <div style="padding: 2px;">False Negatives</div> </div> |
|             | A person does not have a disease. The model classifies the case as having a disease.  | <div style="border: 1px solid black; padding: 2px;"> <div style="background-color: #cccccc; text-align: right; padding: 2px;">▼</div> <div style="padding: 2px;">True Positives</div> <div style="padding: 2px;">True Negatives</div> <div style="padding: 2px;">False Positives</div> <div style="padding: 2px;">False Negatives</div> </div> |
|             | A person has a disease. The model classifies the case as having no disease.           | <div style="border: 1px solid black; padding: 2px;"> <div style="background-color: #cccccc; text-align: right; padding: 2px;">▼</div> <div style="padding: 2px;">True Positives</div> <div style="padding: 2px;">True Negatives</div> <div style="padding: 2px;">False Positives</div> <div style="padding: 2px;">False Negatives</div> </div> |

**Section:**

**Explanation:**

Box 1: True Positive

A true positive is an outcome where the model correctly predicts the positive class

Box 2: True Negative

A true negative is an outcome where the model correctly predicts the negative class.

Box 3: False Positive

A false positive is an outcome where the model incorrectly predicts the positive class.

Box 4: False Negative

A false negative is an outcome where the model incorrectly predicts the negative class.

Note: Let's make the following definitions:

"Wolf" is a positive class.

"No wolf" is a negative class.

We can summarize our "wolf-prediction" model using a 2x2 confusion matrix that depicts all four possible outcomes:



Reference:  
<https://developers.google.com/machine-learning/crash-course/classification/true-false-positive-negative>

### QUESTION 83

#### HOTSPOT

You are using the Azure Machine Learning Service to automate hyperparameter exploration of your neural network classification model. You must define the hyperparameter space to automatically tune hyperparameters using random sampling according to following requirements: The learning rate must be selected from a normal distribution with a mean value of 10 and a standard deviation of 3. Batch size must be 16, 32 and 64. Keep probability must be a value selected from a uniform distribution between the range of 0.05 and 0.1. You need to use the param\_sampling method of the Python API for the Azure Machine Learning Service. How should you complete the code segment? To answer, select the appropriate options in the answer area. NOTE: Each correct selection is worth one point.

#### Hot Area:


**Answer Area**

```
from azureml.train.hyperdrive import RandomParameterSampling
param_sampling = RandomParameterSampling({
 "learning_rate" :
 "batch_size":
 "keep_probability" :
}
```

|                  |
|------------------|
| uniform(10,3)    |
| normal(10,3)     |
| choice(10,3)     |
| Loguniform(10,3) |

|                      |
|----------------------|
| choice(16,32,64)     |
| choice(range(16,64)) |
| normal(16,32,64)     |
| normal(range(16,64)) |

|                          |
|--------------------------|
| choice(range(0.05, 0.1)) |
| uniform(0.05, 0.1)       |
| normal(0.05, 0.1)        |
| lognormal(0.05, 0.1)     |



#### Answer Area:

## Answer Area

```
from azureml.train.hyperdrive import RandomParameterSampling
param_sampling = RandomParameterSampling({
 "learning_rate" :
 uniform(10,3)
 normal(10,3)
 choice(10,3)
 Loguniform(10,3)
 "batch_size":
 choice(16,32,64)
 choice(range(16,64))
 normal(16,32,64)
 normal(range(16,64))
 "keep_probability" :
 choice(range(0.05, 0.1))
 uniform(0.05, 0.1)
 normal(0.05, 0.1)
 lognormal(0.05, 0.1)
})
```



### Section:

### Explanation:

In random sampling, hyperparameter values are randomly selected from the defined search space. Random sampling allows the search space to include both discrete and continuous hyperparameters.

Example:

```
from azureml.train.hyperdrive import RandomParameterSampling
param_sampling = RandomParameterSampling({
 "learning_rate": normal(10, 3),
 "keep_probability": uniform(0.05, 0.1),
 "batch_size": choice(16, 32, 64)
})
```

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/service/how-to-tune-hyperparameters>

### QUESTION 84

DRAG DROP

You create a training pipeline using the Azure Machine Learning designer. You upload a CSV file that contains the data from which you want to train your model.

You need to use the designer to create a pipeline that includes steps to perform the following tasks:

Select the training features using the pandas filter method.

Train a model based on the naive\_bayes.GaussianNB algorithm.

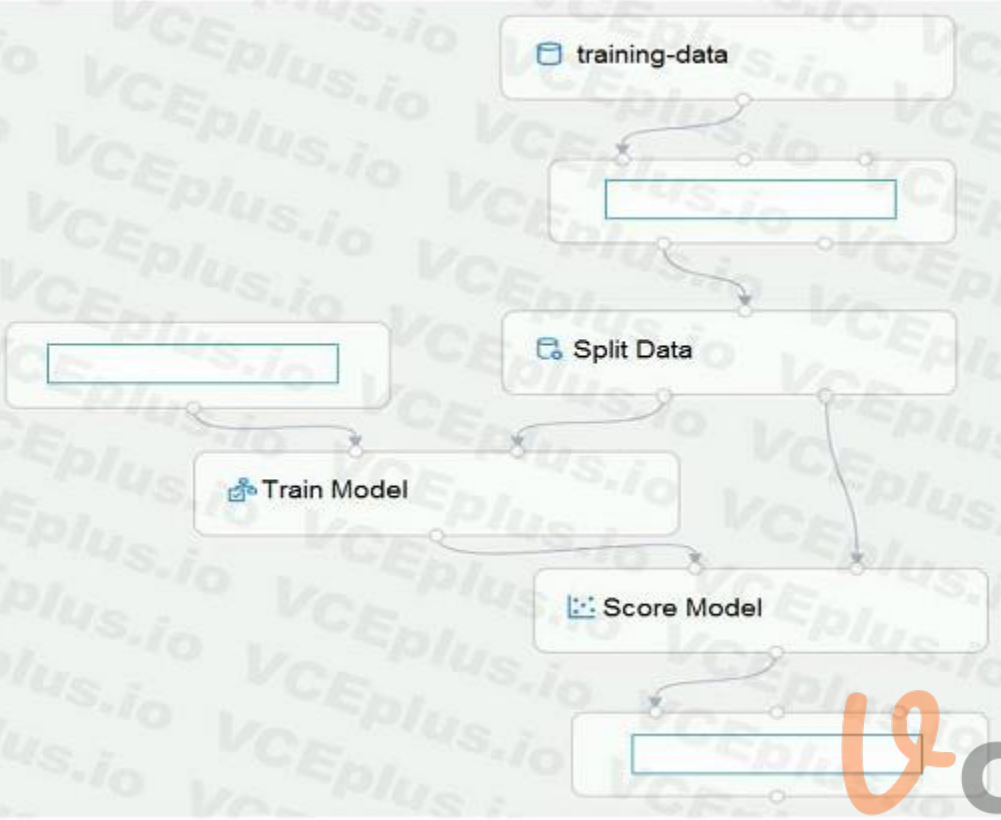
Return only the Scored Labels column by using the query SELECT [Scored Labels] FROM t1;

Which modules should you use? To answer, drag the appropriate modules to the appropriate locations. Each module name may be used once, more than once, or not at all. You may need to drag the split bar between panes

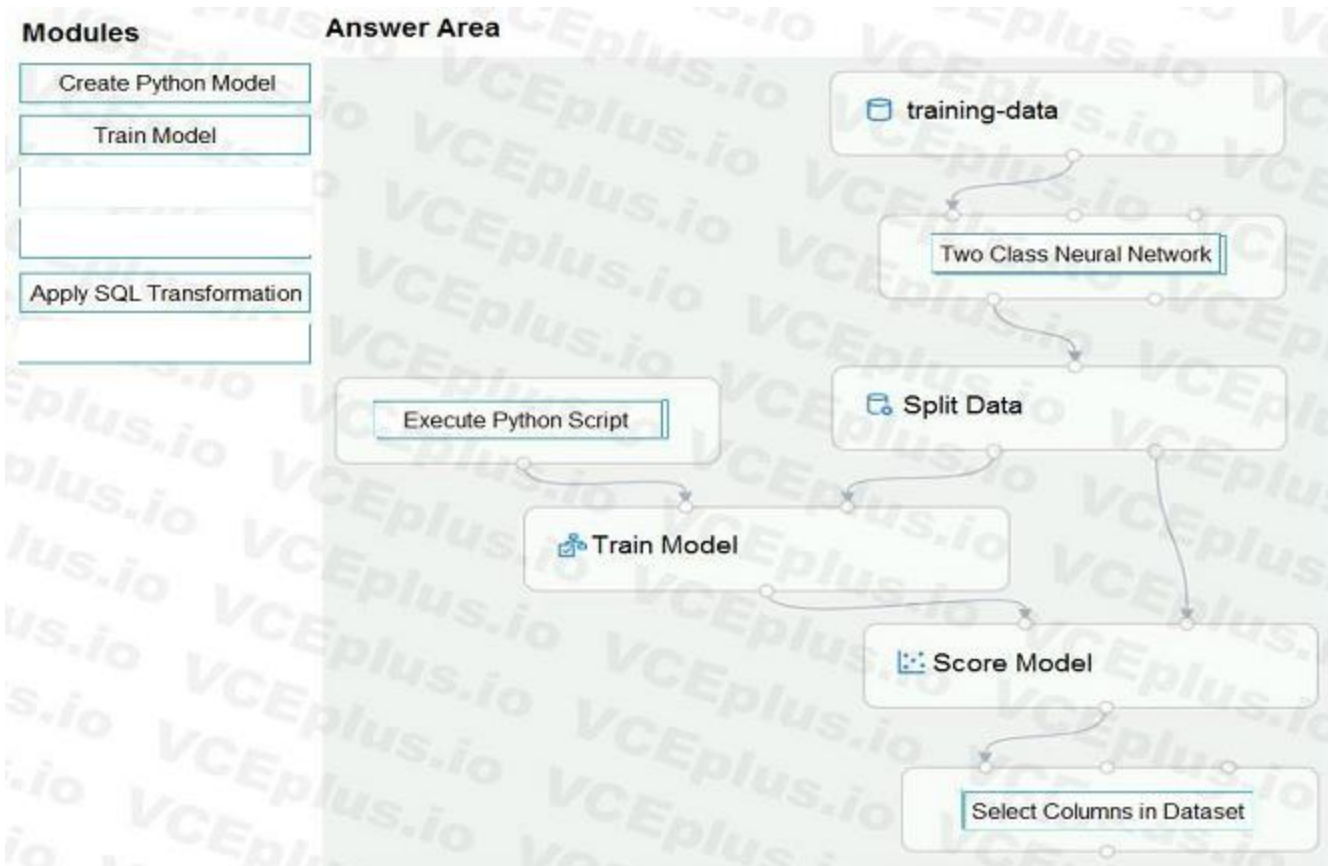
or scroll to view content.

NOTE: Each correct selection is worth one point.

**Select and Place:**

| Modules                   | Answer Area                                                                         |
|---------------------------|-------------------------------------------------------------------------------------|
| Create Python Model       |  |
| Train Model               |                                                                                     |
| Two Class Neural Network  |                                                                                     |
| Execute Python Script     |                                                                                     |
| Apply SQL Transformation  |                                                                                     |
| Select Columns in Dataset |                                                                                     |

**Correct Answer:**



**Section:**

**Explanation:**

Box 1: Two-Class Neural Network

The Two-Class Neural Network creates a binary classifier using a neural network algorithm.

Train a model based on the naive\_bayes.GaussianNB algorithm.

Box 2: Execute python script

Select the training features using the pandas filter method

Box 3: Select Columns in DataSet

Return only the Scored Labels column by using the query `SELECT [Scored Labels] FROM t1;`

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/two-class-neural-network>



**QUESTION 85**

**HOTSPOT**

You have a dataset created for multiclass classification tasks that contains a normalized numerical feature set with 10,000 data points and 150 features.

You use 75 percent of the data points for training and 25 percent for testing. You are using the scikit-learn machine learning library in Python. You use X to denote the feature set and Y to denote class labels.

You create the following Python data frames:

| Name    | Description           |
|---------|-----------------------|
| X_train | training feature set  |
| Y_train | training class labels |
| x_train | testing feature set   |
| y_train | testing class labels  |

You need to apply the Principal Component Analysis (PCA) method to reduce the dimensionality of the feature set to 10 features in both training and testing sets.

How should you complete the code segment? To answer, select the appropriate options in the answer area.

NOTE: Each correct selection is worth one point.

**Hot Area:**



Answer Area

```
from sklearn.decomposition import PCA
pca = PCA(n_components=10)
X_train = pca.fit_transform(X_train)
x_test = pca.transform(x_test)
```

The image shows a code editor with three dropdown menus. The first dropdown, for 'pca', has options: PCA(), PCA(n\_components = 150), PCA(n\_components = 10), and PCA(n\_components = 10000). The second dropdown, for 'X\_train=', has options: pca, model, and sklearn.decomposition. The third dropdown, for 'x\_test = pca.', has options: x\_test, X\_train, fit(x\_test), and transform(x\_test).

Answer Area:

Answer Area

```
from sklearn.decomposition import PCA
pca = PCA(n_components=10)
X_train = pca.fit_transform(X_train)
x_test = pca.transform(x_test)
```

The image shows the same code editor as above, but with the correct options highlighted in green: 'PCA(n\_components = 10)' in the first dropdown, 'pca' in the second dropdown, and 'transform(x\_test)' in the third dropdown.



Section:

Explanation:

Box 1: PCA(n\_components = 10)

Need to reduce the dimensionality of the feature set to 10 features in both training and testing sets.

Example:

```
from sklearn.decomposition import PCA
pca = PCA(n_components=2) ;2 dimensions
principalComponents = pca.fit_transform(x)
Box 2: pca
fit_transform(X[, y])fits the model with X and apply the dimensionality reduction on X.
Box 3: transform(x_test)
transform(X) applies dimensionality reduction to X.
References:
https://scikit-learn.org/stable/modules/generated/sklearn.decomposition.PCA.html
```

**QUESTION 86**

HOTSPOT

You have a feature set containing the following numerical features: X, Y, and Z.  
The Poisson correlation coefficient (r-value) of X, Y, and Z features is shown in the following image:

|   | X         | Y        | Z         |
|---|-----------|----------|-----------|
| X | 1         | 0.149676 | -0.106276 |
| Y | 0.149676  | 1        | 0.859122  |
| Z | -0.106276 | 0.859122 | 1         |

Use the drop-down menus to select the answer choice that answers each question based on the information presented in the graphic.

NOTE: Each correct selection is worth one point.

Hot Area:

**Answer Area**

What is the r-value for the correlation of Y to Z?

- 0.106276
- 0.149676
- 0.859122
- 1

Which type of relationship exists between Z and Y in the feature set?

- a positive linear relationship
- a negative linear relationship
- no linear relationship

Answer Area:

**Answer Area**

What is the r-value for the correlation of Y to Z?

|           |
|-----------|
| -0.106276 |
| 0.149676  |
| 0.859122  |
| 1         |

Which type of relationship exists between Z and Y in the feature set?

|                                |
|--------------------------------|
| a positive linear relationship |
| a negative linear relationship |
| no linear relationship         |

**Section:**

**Explanation:**

Box 1: 0.859122

Box 2: a positively linear relationship +1 indicates a strong positive linear relationship

-1 indicates a strong negative linear correlation

0 denotes no linear relationship between the two variables.

References:

<https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/compute-linear-correlation>

**QUESTION 87**

DRAG DROP

You plan to explore demographic data for home ownership in various cities. The data is in a CSV file with the following format:

age,city,income,home\_owner

21,Chicago,50000,0

35,Seattle,120000,1

23,Seattle,65000,0

45,Seattle,130000,1

18,Chicago,48000,0

You need to run an experiment in your Azure Machine Learning workspace to explore the data and log the results. The experiment must log the following information:

the number of observations in the dataset

a box plot of income by home\_owner

a dictionary containing the city names and the average income for each city

You need to use the appropriate logging methods of the experiment's run object to log the required information.

How should you complete the code? To answer, drag the appropriate code segments to the correct locations. Each code segment may be used once, more than once, or not at all. You may need to drag the split bar between panes or scroll to view content.

NOTE: Each correct selection is worth one point.

**Select and Place:**



**Code segments**

- log
- log\_list
- log\_row
- log\_table
- log\_image

**Answer Area**

```

from azureml.core import Experiment, Run
import pandas as pd
import matplotlib.pyplot as plt
Create an Azure ML experiment in workspace
experiment = Experiment(workspace = ws, name = "demo-experiment")
Start logging data from the experiment
run = experiment.start_logging()
load the dataset
data = pd.read_csv('research/demographics.csv')
Log the number of observations
row_count = (len(data))
run.log("observations", row_count)
Log box plot for income by home_owner
fig = plt.figure(figsize=(9, 6))
ax = fig.gca()
data.boxplot(column = 'income', by = "home_owner", ax = ax)
ax.set_title('income by home_owner')
ax.set_ylabel('income')
run.log(name = 'income_by_home_owner', plot = fig)
Create a dataframe of mean income per city
mean_inc_df = data.groupby('city')['income'].agg(np.mean).to_frame().reset_index()
Convert to a dictionary
mean_inc_dict = mean_inc_df.to_dict('dict')
Log city names and average income dictionary
run.log(name="mean_income_by_city", value= mean_inc_dict)
Complete tracking and get link to details
run.complete()

```

**Correct Answer:****Code segments**

- 
- log\_list
- log\_row
- 
- 

**Answer Area**

```

from azureml.core import Experiment, Run
import pandas as pd
import matplotlib.pyplot as plt
Create an Azure ML experiment in workspace
experiment = Experiment(workspace = ws, name = "demo-experiment")
Start logging data from the experiment
run = experiment.start_logging()
load the dataset
data = pd.read_csv('research/demographics.csv')
Log the number of observations
row_count = (len(data))
run.log("observations", row_count)
Log box plot for income by home_owner
fig = plt.figure(figsize=(9, 6))
ax = fig.gca()
data.boxplot(column = 'income', by = "home_owner", ax = ax)
ax.set_title('income by home_owner')
ax.set_ylabel('income')
run.log_image(name = 'income_by_home_owner', plot = fig)
Create a dataframe of mean income per city
mean_inc_df = data.groupby('city')['income'].agg(np.mean).to_frame().reset_index()
Convert to a dictionary
mean_inc_dict = mean_inc_df.to_dict('dict')
Log city names and average income dictionary
run.log_table(name="mean_income_by_city", value= mean_inc_dict)
Complete tracking and get link to details
run.complete()

```

**Section:****Explanation:**

Box 1: log

The number of observations in the dataset.

run.log(name, value, description="")

Scalar values: Log a numerical or string value to the run with the given name. Logging a metric to a run causes that metric to be stored in the run record in the experiment. You can log the same metric multiple times within a run, the result being considered a vector of that metric.



Example: `run.log("accuracy", 0.95)`

Box 2: `log_image`

A box plot of income by `home_owner`.

`log_image` Log an image to the run record. Use `log_image` to log a .PNG image file or a matplotlib plot to the run. These images will be visible and comparable in the run record.

Example: `run.log_image("ROC", plot=plt)`

Box 3: `log_table`

A dictionary containing the city names and the average income for each city.

`log_table`: Log a dictionary object to the run with the given name.

### QUESTION 88

#### HOTSPOT

Your Azure Machine Learning workspace has a dataset named `real_estate_data`. A sample of the data in the dataset follows.

| <code>postal_code</code> | <code>num_bedrooms</code> | <code>sq_feet</code> | <code>garage</code> | <code>price</code> |
|--------------------------|---------------------------|----------------------|---------------------|--------------------|
| 12345                    | 3                         | 1300                 | 0                   | 23,9000            |
| 54321                    | 1                         | 950                  | 0                   | 11,0000            |
| 12346                    | 2                         | 1200                 | 1                   | 15,0000            |

You want to use automated machine learning to find the best regression model for predicting the price column.

You need to configure an automated machine learning experiment using the Azure Machine Learning SDK.

How should you complete the code? To answer, select the appropriate options in the answer area.

NOTE: Each correct selection is worth one point.

Hot Area:



## Answer Area

```
from azureml.core import Workspace
from azureml.core.compute import ComputeTarget
from azureml.core.runconfig import RunConfiguration
from azureml.train.automl import AutoMLConfig

ws = Workspace.from_config()
training_cluster = ComputeTarget(workspace=ws, name= 'aml-cluster1')
real_estate_ds = ws.datasets.get('real_estate_data')
split1_ds, split2_ds = real_estate_ds.random_split(percentage=0.7, seed=123)
automl_run_config = RunConfiguration(
 framework="python")
automl_config = AutoMLConfig(
 task='regression',
 compute_target= training_cluster,
 run_configuration=automl_run_config,
 primary_metric='r2_score',
```

|                    |               |
|--------------------|---------------|
|                    | ▼ =split1_ds, |
| X                  |               |
| Y                  |               |
| X_valid            |               |
| Y_valid            |               |
| training_data      |               |
|                    | ▼ =split2_ds  |
| X                  |               |
| Y                  |               |
| X_valid            |               |
| Y_valid            |               |
| validation_data    |               |
| training_data      |               |
|                    | ▼ ='price')   |
| y                  |               |
| y_valid            |               |
| y_max              |               |
| label_column_name  |               |
| exclude_nan_labels |               |

 **Vdumps**

Answer Area:

## Answer Area

```
from azureml.core import Workspace
from azureml.core.compute import ComputeTarget
from azureml.core.runconfig import RunConfiguration
from azureml.train.automl import AutoMLConfig

ws = Workspace.from_config()
training_cluster = ComputeTarget(workspace=ws, name='aml-cluster1')
real_estate_ds = ws.datasets.get('real_estate_data')
split1_ds, split2_ds = real_estate_ds.random_split(percentage=0.7, seed=123)
automl_run_config = RunConfiguration(framework="python")
automl_config = AutoMLConfig(
 task='regression',
 compute_target=training_cluster,
 run_configuration=automl_run_config,
 primary_metric='r2_score',
```

|                    |             |
|--------------------|-------------|
| ▼                  | =split1_ds, |
| X                  |             |
| Y                  |             |
| X_valid            |             |
| Y_valid            |             |
| training_data      |             |
| ▼                  | =split2_ds  |
| X                  |             |
| Y                  |             |
| X_valid            |             |
| Y_valid            |             |
| validation_data    |             |
| training_data      |             |
| ▼                  | ='price')   |
| y                  |             |
| y_valid            |             |
| y_max              |             |
| label_column_name  |             |
| exclude_nan_labels |             |

### Section:

#### Explanation:

Box 1: training\_data The training data to be used within the experiment. It should contain both training features and a label column (optionally a sample weights column). If training\_data is specified, then the label\_column\_name parameter must also be specified.

Box 2: validation\_data Provide validation data: In this case, you can either start with a single data file and split it into training and validation sets or you can provide a separate data file for the validation set. Either way, the validation\_data parameter in your

AutoMLConfig object assigns which data to use as your validation set.

Example, the following code example explicitly defines which portion of the provided data in dataset to use for training and validation.

```
dataset = Dataset.Tabular.from_delimited_files(data)
training_data, validation_data = dataset.random_split(percentage=0.8, seed=1)
automl_config = AutoMLConfig(compute_target = aml_remote_compute,
task = 'classification',
```



```
primary_metric = 'AUC_weighted',
training_data = training_data,
validation_data = validation_data,
label_column_name = 'Class'
)
```

Box 3: label\_column\_name

label\_column\_name:

The name of the label column. If the input data is from a pandas.DataFrame which doesn't have column names, column indices can be used instead, expressed as integers.

This parameter is applicable to training\_data and validation\_data parameters.

Incorrect Answers:

X: The training features to use when fitting pipelines during an experiment. This setting is being deprecated. Please use training\_data and label\_column\_name instead.

Y: The training labels to use when fitting pipelines during an experiment. This is the value your model will predict. This setting is being deprecated. Please use training\_data and label\_column\_name instead.

X\_valid: Validation features to use when fitting pipelines during an experiment.

If specified, then y\_valid or sample\_weight\_valid must also be specified.

Y\_valid: Validation labels to use when fitting pipelines during an experiment.

Both X\_valid and y\_valid must be specified together.

exclude\_nan\_labels: Whether to exclude rows with NaN values in the label. The default is True.

y\_max: y\_max (float)

Maximum value of y for a regression experiment. The combination of y\_min and y\_max are used to normalize test set metrics based on the input data range. If not specified, the maximum value is inferred from the data.

Reference:

<https://docs.microsoft.com/en-us/python/api/azureml-train-automl-client/azureml.train.automl.automlconfig.automlconfig?view=azure-ml-py>

#### QUESTION 89

HOTSPOT

You have a multi-class image classification deep learning model that uses a set of labeled photographs. You create the following code to select hyperparameter values when training the model.

```
from azureml.train.hyperdrive import BayesianParameterSampling
param_sampling = BayesianParametersSampling ({
 "learning_rate": uniform(0.01, 0.1),
 "batch_size": choice(16, 32, 64, 128)}
)
```

For each of the following statements, select Yes if the statement is true. Otherwise, select No.

NOTE: Each correct selection is worth one point.

Hot Area:

|                                                                                                                               | Yes                   | No                    |
|-------------------------------------------------------------------------------------------------------------------------------|-----------------------|-----------------------|
| Hyperparameter combinations for the runs are selected based on how previous samples performed in the previous experiment run. | <input type="radio"/> | <input type="radio"/> |
| The learning rate value 0.09 might be used during model training.                                                             | <input type="radio"/> | <input type="radio"/> |
| You can define an early termination policy for this hyperparameter tuning run.                                                | <input type="radio"/> | <input type="radio"/> |



**Answer Area:**

|                                                                                                                               | Yes                                 | No                                  |
|-------------------------------------------------------------------------------------------------------------------------------|-------------------------------------|-------------------------------------|
| Hyperparameter combinations for the runs are selected based on how previous samples performed in the previous experiment run. | <input checked="" type="checkbox"/> | <input type="checkbox"/>            |
| The learning rate value 0.09 might be used during model training.                                                             | <input checked="" type="checkbox"/> | <input type="checkbox"/>            |
| You can define an early termination policy for this hyperparameter tuning run.                                                | <input type="checkbox"/>            | <input checked="" type="checkbox"/> |

**Section:**

**Explanation:**

Box 1: Yes

Hyperparameters are adjustable parameters you choose to train a model that govern the training process itself. Azure Machine Learning allows you to automate hyperparameter exploration in an efficient manner, saving you significant time and resources. You specify the range of hyperparameter values and a maximum number of training runs. The system then automatically launches multiple simultaneous runs with different parameter configurations and finds the configuration that results in the best performance, measured by the metric you choose. Poorly performing training runs are automatically early terminated, reducing wastage of compute resources. These resources are instead used to explore other hyperparameter configurations.

Box 2: Yes

uniform(low, high) - Returns a value uniformly distributed between low and high

Box 3: No

Bayesian sampling does not currently support any early termination policy.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/how-to-tune-hyperparameters>



**QUESTION 90**

**HOTSPOT**

You publish a batch inferencing pipeline that will be used by a business application.

The application developers need to know which information should be submitted to and returned by the REST interface for the published pipeline.

You need to identify the information required in the REST request and returned as a response from the published pipeline.

Which values should you use in the REST request and to expect in the response? To answer, select the appropriate options in the answer area.

NOTE: Each correct selection is worth one point.

**Hot Area:**

**Answer Area**

**REST Request**

**Value**

Request Header

|                                       |
|---------------------------------------|
| JSON containing the run ID            |
| JSON containing the pipeline ID       |
| JSON containing the experiment name   |
| JSON containing an OAuth bearer token |

Request Body

|                                       |
|---------------------------------------|
| JSON containing the run ID            |
| JSON containing the pipeline ID       |
| JSON containing the experiment name   |
| JSON containing an OAuth bearer token |

Response

|                                                                 |
|-----------------------------------------------------------------|
| JSON containing the run ID                                      |
| JSON containing a list of predictions                           |
| JSON containing the experiment name                             |
| JSON containing a path to the parallel_run_step.txt output file |



Answer Area:

## Answer Area

### REST Request

### Value

#### Request Header

- JSON containing the run ID
- JSON containing the pipeline ID
- JSON containing the experiment name
- JSON containing an OAuth bearer token

#### Request Body

- JSON containing the run ID
- JSON containing the pipeline ID
- JSON containing the experiment name
- JSON containing an OAuth bearer token

#### Response

- JSON containing the run ID
- JSON containing a list of predictions
- JSON containing the experiment name
- JSON containing a path to the parallel\_run\_step.txt output file



### Section:

### Explanation:

Box 1: JSON containing an OAuth bearer token

Specify your authentication header in the request.

To run the pipeline from the REST endpoint, you need an OAuth2 Bearer-type authentication header.

Box 2: JSON containing the experiment name

Add a JSON payload object that has the experiment name.

Example:

```
rest_endpoint = published_pipeline.endpoint
response = requests.post(rest_endpoint,
headers=auth_header,
json={"ExperimentName": "batch_scoring",
"ParameterAssignments": {"process_count_per_node": 6}})
run_id = response.json()["Id"]
```

Box 3: JSON containing the run ID

Make the request to trigger the run. Include code to access the Id key from the response dictionary to get the value of the run ID.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/tutorial-pipeline-batch-scoring-classification>

### QUESTION 91

#### HOTSPOT

You register the following versions of a model.

| Model name       | Model version | Tags                              | Properties  |
|------------------|---------------|-----------------------------------|-------------|
| healthcare_model | 3             | 'Training context': 'CPU Compute' | value:87.43 |
| healthcare_model | 2             | 'Training context': 'CPU Compute' | value:54.98 |
| healthcare_model | 1             | 'Training context': 'CPU Compute' | value:23.56 |

You use the Azure ML Python SDK to run a training experiment. You use a variable named run to reference the experiment run.

After the run has been submitted and completed, you run the following code:

```
run.register_model(model_path='outputs/model.pkl',
 model_name='healthcare_model',
 tags={'Training context': 'CPU Compute'})
```

For each of the following statements, select Yes if the statement is true. Otherwise, select No.

NOTE: Each correct selection is worth one point.

Hot Area:

Yes No

The code will cause a previous version of the saved model to be overwritten.

The version number will now be 4.

The latest version of the stored model will have a property of value: 87.43.

Answer Area:

Yes No

The code will cause a previous version of the saved model to be overwritten.

The version number will now be 4.

The latest version of the stored model will have a property of value: 87.43.

Section:

Explanation:

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/how-to-deploy-and-where>

QUESTION 92

HOTSPOT

You collect data from a nearby weather station. You have a pandas dataframe named weather\_df that includes the following data:



| Temperature | Observation_time | Humidity | Pressure | Visibility | Days_since_last_observation |
|-------------|------------------|----------|----------|------------|-----------------------------|
| 74          | 2019/10/2 00:00  | 0.62     | 29.87    | 3          | 0.5                         |
| 89          | 2019/10/2 12:00  | 0.70     | 28.88    | 10         | 0.5                         |
| 72          | 2019/10/3 00:00  | 0.64     | 30.00    | 8          | 0.5                         |
| 80          | 2019/10/3 12:00  | 0.66     | 29.75    | 7          | 0.5                         |

The data is collected every 12 hours: noon and midnight.

You plan to use automated machine learning to create a time-series model that predicts temperature over the next seven days. For the initial round of training, you want to train a maximum of 50 different models.

You must use the Azure Machine Learning SDK to run an automated machine learning experiment to train these models.

You need to configure the automated machine learning run.

How should you complete the AutoMLConfig definition? To answer, select the appropriate options in the answer area.

NOTE: Each correct selection is worth one point.

**Hot Area:**



```
automl_config = AutoMLConfig(task="
```

|                |
|----------------|
| ▼              |
| regression     |
| forecasting    |
| classification |
| deep learning  |

```
training_data=weather_df,
```

```
label_column_name="
```

|                  |
|------------------|
| ▼                |
| humidity         |
| pressure         |
| visibility       |
| temperature      |
| days_since_last  |
| observation_time |

```
time_column_name="
```

|                  |
|------------------|
| ▼                |
| humidity         |
| pressure         |
| visibility       |
| temperature      |
| days_since_last  |
| observation_time |

```
max_horizon=
```

|    |
|----|
| ▼  |
| 2  |
| 6  |
| 7  |
| 12 |
| 14 |
| 50 |

```
iterations=
```

|    |
|----|
| ▼  |
| 2  |
| 6  |
| 7  |
| 12 |
| 14 |
| 50 |

```
iteration_timeout_minutes=5,
```

```
primary_metric="r2_score")
```

Answer Area:

```

automl_config = AutoMLConfig(task="
 regression
 forecasting
 classification
 deep learning
 ",
 training_data=weather_df,
 label_column_name="
 humidity
 pressure
 visibility
 temperature
 days_since_last
 observation_time
 ",
 time_column_name="
 humidity
 pressure
 visibility
 temperature
 days_since_last
 observation_time
 ",
 max_horizon=
 2
 6
 7
 12
 14
 50
 ",
 iterations=
 2
 6
 7
 12
 14
 50
 ",
 iteration_timeout_minutes=5,
 primary_metric="r2_score")

```



**Section:**

**Explanation:**

Box 1: forecasting

Task: The type of task to run. Values can be 'classification', 'regression', or 'forecasting' depending on the type of automated ML problem to solve.

Box 2: temperature

The training data to be used within the experiment. It should contain both training features and a label column (optionally a sample weights column).

Box 3: observation\_time

time\_column\_name: The name of the time column. This parameter is required when forecasting to specify the datetime column in the input data used for building the time series and inferring its frequency. This setting is being deprecated.

Please use forecasting\_parameters instead.

Box 4: 7

"predicts temperature over the next seven days"

max\_horizon: The desired maximum forecast horizon in units of time-series frequency. The default value is 1.

Units are based on the time interval of your training data, e.g., monthly, weekly that the forecaster should predict out. When task type is forecasting, this parameter is required.

Box 5: 50

"For the initial round of training, you want to train a maximum of 50 different models."

Iterations: The total number of different algorithm and parameter combinations to test during an automated ML experiment.

Reference:

<https://docs.microsoft.com/en-us/python/api/azureml-train-automl-client/azureml.train.automl.automlconfig.automlconfig>

### QUESTION 93

You plan to run a Python script as an Azure Machine Learning experiment.

The script contains the following code:

```
import os, argparse, glob
from azureml.core import Run
parser = argparse.ArgumentParser()
parser.add_argument('--input-data',
 type=str, dest='data_folder')
args = parser.parse_args()
data_path = args.data_folder
file_paths = glob.glob(data_path + "/*.jpg")
```

You must specify a file dataset as an input to the script. The dataset consists of multiple large image files and must be streamed directly from its source.

You need to write code to define a ScriptRunConfig object for the experiment and pass the ds dataset as an argument.

Which code segment should you use?

- A. arguments = ['--input-data', ds.to\_pandas\_dataframe()]
- B. arguments = ['--input-data', ds.as\_mount()]
- C. arguments = ['--data-data', ds]
- D. arguments = ['--input-data', ds.as\_download()]



**Correct Answer: A**

**Section:**

**Explanation:**

If you have structured data not yet registered as a dataset, create a TabularDataset and use it directly in your training script for your local or remote experiment.

To load the TabularDataset to pandas DataFrame

```
df = dataset.to_pandas_dataframe()
```

Note: TabularDataset represents data in a tabular format created by parsing the provided file or list of files.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/how-to-train-with-datasets>

### QUESTION 94

HOTSPOT

You are hired as a data scientist at a winery. The previous data scientist used Azure Machine Learning.

You need to review the models and explain how each model makes decisions.

Which explainer modules should you use? To answer, select the appropriate options in the answer area.

NOTE: Each correct selection is worth one point.

**Hot Area:**



**Answer Area**

**Model type**

**Explainer**

A random forest model for predicting the alcohol content in wine given a set of covariates

|         |
|---------|
| ▼       |
| Tabular |
| HAN     |
| Text    |
| Image   |

A natural language processing model for analyzing field reports

|       |
|-------|
| ▼     |
| Tree  |
| HAN   |
| Text  |
| Image |

An image classifier that determines the quality of the grape based upon its physical characteristics.

|        |
|--------|
| ▼      |
| Kernel |
| HAN    |
| Text   |
| Image  |

**Answer Area:**

**Answer Area**

**Model type**

**Explainer**

A random forest model for predicting the alcohol content in wine given a set of covariates

|         |
|---------|
| ▼       |
| Tabular |
| HAN     |
| Text    |
| Image   |

A natural language processing model for analyzing field reports

|       |
|-------|
| ▼     |
| Tree  |
| HAN   |
| Text  |
| Image |

An image classifier that determines the quality of the grape based upon its physical characteristics.

|        |
|--------|
| ▼      |
| Kernel |
| HAN    |
| Text   |
| Image  |

**Section:**

**Explanation:**

Meta explainers automatically select a suitable direct explainer and generate the best explanation info based on the given model and data sets. The meta explainers leverage all the libraries (SHAP, LIME, Mimic, etc.) that we have integrated or developed. The following are the meta explainers available in the SDK:



Tabular Explainer: Used with tabular datasets.

Text Explainer: Used with text datasets.

Image Explainer: Used with image datasets.

Box 1: Tabular

Box 2: Text

Box 3: Image

Incorrect Answers:

Hierarchical Attention Network (HAN)

HAN was proposed by Yang et al. in 2016. Key features of HAN that differentiates itself from existing approaches to document classification are (1) it exploits the hierarchical nature of text data and (2) attention mechanism is adapted for document classification.

Reference:

<https://medium.com/microsoftazure/automated-and-interpretable-machine-learning-d07975741298>

### QUESTION 95

HOTSPOT

You have a dataset that includes home sales data for a city. The dataset includes the following columns.

| Name      | Description                                                                  |
|-----------|------------------------------------------------------------------------------|
| Price     | The sales price for the house.                                               |
| Bedrooms  | The number of bedrooms in the house.                                         |
| Size      | The size of the house in square feet.                                        |
| HasGarage | A binary value indicating whether or not the house has a garage.             |
| HomeType  | The category of home, for example, apartment, townhouse, single-family home. |

Each row in the dataset corresponds to an individual home sales transaction.

You need to use automated machine learning to generate the best model for predicting the sales price based on the features of the house.

Which values should you use? To answer, select the appropriate options in the answer area.

NOTE: Each correct selection is worth one point.



Hot Area:

Answer Area

| Setting         | Value                                                                                                                     |
|-----------------|---------------------------------------------------------------------------------------------------------------------------|
| Prediction task | <div style="border: 1px solid black; padding: 2px;">▼<br/>Classification<br/>Forecasting<br/>Regression<br/>Outlier</div> |
| Target column   | <div style="border: 1px solid black; padding: 2px;">▼<br/>Price<br/>Bedrooms<br/>Size<br/>HasGarage<br/>HomeType</div>    |

Answer Area:

## Answer Area

| Setting         | Value                                                                                                                     |
|-----------------|---------------------------------------------------------------------------------------------------------------------------|
| Prediction task | <div style="border: 1px solid black; padding: 2px;">▼<br/>Classification<br/>Forecasting<br/>Regression<br/>Outlier</div> |
| Target column   | <div style="border: 1px solid black; padding: 2px;">▼<br/>Price<br/>Bedrooms<br/>Size<br/>HasGarage<br/>HomeType</div>    |

### Section:

### Explanation:

Box 1: Regression

Regression is a supervised machine learning technique used to predict numeric values.

Box 2: Price

Reference:

<https://docs.microsoft.com/en-us/learn/modules/create-regression-model-azure-machine-learning-designer>

### QUESTION 96

#### DRAG DROP

You have an Azure Machine Learning workspace that contains a CPU-based compute cluster and an Azure Kubernetes Services (AKS) inference cluster. You create a tabular dataset containing data that you plan to use to create a classification model.

You need to use the Azure Machine Learning designer to create a web service through which client applications can consume the classification model by submitting new data and getting an immediate prediction as a response.

Which three actions should you perform in sequence? To answer, move the appropriate actions from the list of actions to the answer area and arrange them in the correct order.

### Select and Place:



### Actions

Create and run a batch inference pipeline on the compute cluster.

Deploy a real-time endpoint on the inference cluster.

Create and run a real-time inference pipeline on the compute cluster.

Create and run a training pipeline that prepares the data and trains a classification model on the compute cluster.

Use the automated ML user interface to train a classification model on the compute cluster.

Create and start a Compute Instance.

### Answer Area



### Correct Answer:

### Actions

Create and run a batch inference pipeline on the compute cluster.

Deploy a real-time endpoint on the inference cluster.

Use the automated ML user interface to train a classification model on the compute cluster.

### Answer Area

Create and start a Compute Instance.

Create and run a training pipeline that prepares the data and trains a classification model on the compute cluster.

Create and run a real-time inference pipeline on the compute cluster.



### Section:

### Explanation:

Step 1: Create and start a Compute Instance To train and deploy models using Azure Machine Learning designer, you need compute on which to run the training process, test the model, and host the model in a deployed service.

There are four kinds of compute resource you can create:

Compute Instances: Development workstations that data scientists can use to work with data and models.

Compute Clusters: Scalable clusters of virtual machines for on-demand processing of experiment code.

Inference Clusters: Deployment targets for predictive services that use your trained models.

Attached Compute: Links to existing Azure compute resources, such as Virtual Machines or Azure Databricks clusters.

Step 2: Create and run a training pipeline..

After you've used data transformations to prepare the data, you can use it to train a machine learning model. Create and run a training pipeline



### Step 3: Create and run a real-time inference pipeline

After creating and running a pipeline to train the model, you need a second pipeline that performs the same data transformations for new data, and then uses the trained model to inference (in other words, predict) label values based on its features. This pipeline will form the basis for a predictive service that you can publish for applications to use.

Reference:

<https://docs.microsoft.com/en-us/learn/modules/create-classification-model-azure-machine-learning-designer/>

### QUESTION 97

#### HOTSPOT

You are running a training experiment on remote compute in Azure Machine Learning.

The experiment is configured to use a conda environment that includes the mlflow and azureml-contrib-run packages.

You must use MLflow as the logging package for tracking metrics generated in the experiment.

You need to complete the script for the experiment.

How should you complete the code? To answer, select the appropriate options in the answer area.

NOTE: Each correct selection is worth one point.

#### Hot Area:

**Answer Area**

```
import numpy as np
Import library to log metrics
```

▼

```
from azureml.core import Run
import mlflow
import logging
```

# Start logging for this run

▼

```
run = Run.get_context()
mlflow.start_run()
logger = logging.getLogger('Run')
reg_rate = 0.01
Log the reg_rate metric
```

▼

```
run.log('reg_rate', np.float(reg_rate))
mlflow.log_metric('reg_rate', np.float(reg_rate))
logger.info(np.float(reg_rate))
```

# Stop logging for this run

▼

```
run.complete()
mlflow.end_run()
logger.setLevel(logging.INFO)
```

#### Answer Area:



## Answer Area

```
import numpy as np
Import library to log metrics
```

```
from azureml.core import Run
```

```
import mlflow
```

```
import logging
```

```
Start logging for this run
```

```
run = Run.get_context()
```

```
mlflow.start_run()
```

```
logger = logging.getLogger('Run')
```

```
reg_rate = 0.01
```

```
Log the reg_rate metric
```

```
run.log('reg_rate', np.float(reg_rate))
```

```
mlflow.log_metric('reg_rate', np.float(reg_rate))
```

```
logger.info(np.float(reg_rate))
```

```
Stop logging for this run
```

```
run.complete()
```

```
mlflow.end_run()
```

```
logger.setLevel(logging.INFO)
```



### Section:

#### Explanation:

Box 1: import mlflow

Import the mlflow and Workspace classes to access MLflow's tracking URI and configure your workspace.

Box 2: mlflow.start\_run()

Set the MLflow experiment name with set\_experiment() and start your training run with start\_run().

Box 3: mlflow.log\_metric('..')

Use log\_metric() to activate the MLflow logging API and begin logging your training run metrics.

Box 4: mlflow.end\_run()

Close the run:

```
run.endRun()
```

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/how-to-use-mlflow>

### QUESTION 98

#### HOTSPOT

You are using C-Support Vector classification to do a multi-class classification with an unbalanced training dataset. The C-Support Vector classification using Python code shown below:

```
from sklearn.svm import svc
import numpy as np
svc = SVC(kernel= 'linear', class_weight= 'balanced', C=1.0, random_state=0)
modell = svc.fit(X_train, y)
```

You need to evaluate the C-Support Vector classification code.  
Which evaluation statement should you use? To answer, select the appropriate options in the answer area.  
NOTE: Each correct selection is worth one point.

Hot Area:

**Answer Area**

| Code Segment          | Evaluation Statement                                                                                                                                                                                                                                                                                                                                                                                                                                                                           |
|-----------------------|------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|
| class_weight=balanced | <div style="border: 1px solid black; background-color: #f0f0f0; padding: 2px;">▼</div> <div style="border: 1px solid black; padding: 2px;"> <p>Automatically select the performance metrics for the classification.</p> <p>Automatically adjust weights directly proportional to class frequencies in the input data.</p> <p>Automatically adjust weights inversely proportional to class frequencies in the input data.</p> </div> |
| C parameter           | <div style="border: 1px solid black; background-color: #f0f0f0; padding: 2px;">▼</div> <div style="border: 1px solid black; padding: 2px;"> <p>Penalty parameter</p> <p>Degree of polynomial kernel function</p> <p>Size of the kernel cache</p> </div>                                                                                                                                                                          |

Answer Area:



## Answer Area



### Code Segment

```
class_weight=balanced
```

Automatically select the performance metrics for the classification.  
Automatically adjust weights directly proportional to class frequencies in the input data.  
Automatically adjust weights inversely proportional to class frequencies in the input data.

```
C parameter
```

Penalty parameter  
Degree of polynomial kernel function  
Size of the kernel cache

#### Section:

#### Explanation:

Box 1: Automatically adjust weights inversely proportional to class frequencies in the input data

The "balanced" mode uses the values of y to automatically adjust weights inversely proportional to class frequencies in the input data as  $n\_samples / (n\_classes * np.bincount(y))$ .

Box 2: Penalty parameter

Parameter: C : float, optional (default=1.0)

Penalty parameter C of the error term.

References:

<https://scikit-learn.org/stable/modules/generated/sklearn.svm.SVC.html>

#### QUESTION 99

You create a Python script that runs a training experiment in Azure Machine Learning. The script uses the Azure Machine Learning SDK for Python.

You must add a statement that retrieves the names of the logs and outputs generated by the script.

You need to reference a Python class object from the SDK for the statement.

Which class object should you use?

- A. Run
- B. ScriptRunConfig
- C. Workspace
- D. Experiment

**Correct Answer: A**

#### Section:

#### Explanation:

A run represents a single trial of an experiment. Runs are used to monitor the asynchronous execution of a trial, log metrics and store output of the trial, and to analyze results and access artifacts generated by the trial.

The run Class get\_all\_logs method downloads all logs for the run to a directory.



Incorrect Answers:

A: A run represents a single trial of an experiment. Runs are used to monitor the asynchronous execution of a trial, log metrics and store output of the trial, and to analyze results and access artifacts generated by the trial.

B: A ScriptRunConfig packages together the configuration information needed to submit a run in Azure ML, including the script, compute target, environment, and any distributed job-specific configs.

Reference:

[https://docs.microsoft.com/en-us/python/api/azureml-core/azureml.core.run\(class\)](https://docs.microsoft.com/en-us/python/api/azureml-core/azureml.core.run(class))

#### QUESTION 100

You run a script as an experiment in Azure Machine Learning.

You have a Run object named run that references the experiment run. You must review the log files that were generated during the experiment run.

You need to download the log files to a local folder for review.

Which two code segments can you run to achieve this goal? Each correct answer presents a complete solution.

NOTE: Each correct selection is worth one point.

- A. run.get\_details()
- B. run.get\_file\_names()
- C. run.get\_metrics()
- D. run.download\_files(output\_directory='./runfiles')
- E. run.get\_all\_logs(destination='./runlogs')

**Correct Answer: A, E**

**Section:**

**Explanation:**

The run Class get\_all\_logs method downloads all logs for the run to a directory.

The run Class get\_details gets the definition, status information, current log files, and other details of the run.

Incorrect Answers:

B: The run get\_file\_names list the files that are stored in association with the run.

Reference:

[https://docs.microsoft.com/en-us/python/api/azureml-core/azureml.core.run\(class\)](https://docs.microsoft.com/en-us/python/api/azureml-core/azureml.core.run(class))

#### QUESTION 101

You have the following code. The code prepares an experiment to run a script:

```
from azureml.core import Workspace, Experiment, Run, ScriptRunConfig

ws = Workspace.from_config()
script_config = ScriptRunConfig(source_directory='experiment_files',
 script='experiment.py')

script_experiment = Experiment(workspace=ws, name='script-experiment')
```

The experiment must be run on local computer using the default environment.

You need to add code to start the experiment and run the script.

Which code segment should you use?

- A. run = script\_experiment.start\_logging()
- B. run = Run(experiment=script\_experiment)
- C. ws.get\_run(run\_id=experiment.id)
- D. run = script\_experiment.submit(config=script\_config)

**Correct Answer: D**

**Section:****Explanation:**

The experiment class submit method submits an experiment and return the active created run.

Syntax: submit(config, tags=None, \*\*kwargs)

Reference:

<https://docs.microsoft.com/en-us/python/api/azureml-core/azureml.core.experiment.experiment>

**QUESTION 102**

You use the following code to define the steps for a pipeline:

```
from azureml.core import Workspace, Experiment, Run
from azureml.pipeline.core import Pipeline
from azureml.pipeline.steps import PythonScriptStep
ws = Workspace.from_config()
```

```
...
```

```
step1 = PythonScriptStep(name="step1", ...)
step2 = PythonScriptStep(name="step2", ...)
pipeline_steps = [step1, step2]
```

You need to add code to run the steps.

Which two code segments can you use to achieve this goal? Each correct answer presents a complete solution.

NOTE: Each correct selection is worth one point.

- A. 

```
experiment = Experiment(workspace=ws,
name='pipeline-experiment')
run = experiment.submit(config=pipeline_steps)
```
- B. 

```
run = Run(pipeline_steps)
```
- C. 

```
pipeline = Pipeline(workspace=ws, steps=pipeline_steps)
experiment = Experiment(workspace=ws,
name='pipeline-experiment')
run = experiment.submit(pipeline)
```
- D. 

```
pipeline = Pipeline(workspace=ws, steps=pipeline_steps)
run = pipeline.submit(experiment_name='pipeline-experiment')
```



**Correct Answer: C, D**

**Section:****Explanation:**

After you define your steps, you build the pipeline by using some or all of those steps.

# Build the pipeline. Example:

```
pipeline1 = Pipeline(workspace=ws, steps=[compare_models])
```

# Submit the pipeline to be run

```
pipeline_run1 = Experiment(ws, 'Compare_Models_Exp').submit(pipeline1)
```

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/how-to-create-machine-learning-pipelines>

**QUESTION 103****HOTSPOT**

You create an Azure Databricks workspace and a linked Azure Machine Learning workspace.

You have the following Python code segment in the Azure Machine Learning workspace:

```
import mlflow
import mlflow.azureml
```

```
import azureml.mlflow
import azureml.core
from azureml.core import Workspace
subscription_id = 'subscription_id'
resource_group = 'resource_group_name'
workspace_name = 'workspace_name'
ws = Workspace.get(name=workspace_name,
subscription_id=subscription_id,
resource_group=resource_group)
experimentName = "/Users/{user_name}/{experiment_folder}/{experiment_name}"
mlflow.set_experiment(experimentName)
uri = ws.get_mlflow_tracking_uri()
mlflow.set_tracking_uri(uri)
```

Instructions: For each of the following statements, select Yes if the statement is true. Otherwise, select No.

NOTE: Each correct selection is worth one point.

Hot Area:

**Answer Area**

|                                                                                              | Yes                   | No                    |
|----------------------------------------------------------------------------------------------|-----------------------|-----------------------|
| A resource group and Azure Machine Learning workspace will be created.                       | <input type="radio"/> | <input type="radio"/> |
| An Azure Databricks experiment will be tracked only in the Azure Machine Learning workspace. | <input type="radio"/> | <input type="radio"/> |
| The epoch loss metric is set to be tracked.                                                  | <input type="radio"/> | <input type="radio"/> |

Answer Area:

**Answer Area**

|                                                                                              | Yes                              | No                               |
|----------------------------------------------------------------------------------------------|----------------------------------|----------------------------------|
| A resource group and Azure Machine Learning workspace will be created.                       | <input type="radio"/>            | <input checked="" type="radio"/> |
| An Azure Databricks experiment will be tracked only in the Azure Machine Learning workspace. | <input checked="" type="radio"/> | <input type="radio"/>            |
| The epoch loss metric is set to be tracked.                                                  | <input checked="" type="radio"/> | <input type="radio"/>            |

**Section:**

**Explanation:**

Box 1: No

The Workspace.get method loads an existing workspace without using configuration files.

```
ws = Workspace.get(name="myworkspace",
subscription_id='<azure-subscription-id>',
resource_group='myresourcegroup')
```

Box 2: Yes

MLflow Tracking with Azure Machine Learning lets you store the logged metrics and artifacts from your local runs into your Azure Machine Learning workspace.

The get\_mlflow\_tracking\_uri() method assigns a unique tracking URI address to the workspace, ws, and set\_tracking\_uri() points the MLflow tracking URI to that address.

Box 3: Yes

Note: In Deep Learning, epoch means the total dataset is passed forward and backward in a neural network once.

Reference:

<https://docs.microsoft.com/en-us/python/api/azureml-core/azureml.core.workspace.workspace>

<https://docs.microsoft.com/en-us/azure/machine-learning/how-to-use-mlflow>



**QUESTION 104**

You create and register a model in an Azure Machine Learning workspace.

You must use the Azure Machine Learning SDK to implement a batch inference pipeline that uses a ParallelRunStep to score input data using the model. You must specify a value for the ParallelRunConfig compute\_target setting of the pipeline step.

You need to create the compute target.

Which class should you use?

- A. BatchCompute
- B. AdlaCompute
- C. AmlCompute
- D. AksCompute

**Correct Answer: C**

**Section:**



**Explanation:**

Compute target to use for ParallelRunStep. This parameter may be specified as a compute target object or the string name of a compute target in the workspace.

The compute\_target target is of AmlCompute or string.

Note: An Azure Machine Learning Compute (AmlCompute) is a managed-compute infrastructure that allows you to easily create a single or multi-node compute. The compute is created within your workspace region as a resource that can be shared with other users

Reference:

<https://docs.microsoft.com/en-us/python/api/azureml-contrib-pipeline-steps/azureml.contrib.pipeline.steps.parallelrunconfig>

[https://docs.microsoft.com/en-us/python/api/azureml-core/azureml.core.compute.amlcompute\(class\)](https://docs.microsoft.com/en-us/python/api/azureml-core/azureml.core.compute.amlcompute(class))

**QUESTION 105**

DRAG DROP

You previously deployed a model that was trained using a tabular dataset named training-dataset, which is based on a folder of CSV files.

Over time, you have collected the features and predicted labels generated by the model in a folder containing a CSV file for each month. You have created two tabular datasets based on the folder containing the inference data: one named predictions-dataset with a schema that matches the training data exactly, including the predicted label; and another named features-dataset with a schema containing all of the feature columns and a timestamp column based on the filename, which includes the day, month, and year.

You need to create a data drift monitor to identify any changing trends in the feature data since the model was trained. To accomplish this, you must define the required datasets for the data drift monitor.

Which datasets should you use to configure the data drift monitor? To answer, drag the appropriate datasets to the correct data drift monitor options. Each source may be used once, more than once, or not at all. You may need to drag the split bar between panes or scroll to view content.

NOTE: Each correct selection is worth one point.

Select and Place:

**Target datasets**

- training-dataset
- predictions-dataset
- features-dataset

**Answer Area**

- Baseline dataset: training-dataset
- Target dataset: predictions-dataset

Correct Answer:

**Target datasets**

- 
- 
- features-dataset

**Answer Area**

- Baseline dataset: training-dataset
- Target dataset: predictions-dataset

**Section:**

**Explanation:**

Box 1: training-dataset

Baseline dataset - usually the training dataset for a model.

Box 2: predictions-dataset

Target dataset - usually model input data - is compared over time to your baseline dataset. This comparison means that your target dataset must have a timestamp column specified.

The monitor will compare the baseline and target datasets.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/how-to-monitor-datasets>

## 02 - Run experiments and train models

Case study

Overview

You are a data scientist in a company that provides data science for professional sporting events. Models will use global and local market data to meet the following business goals:

Understand sentiment of mobile device users at sporting events based on audio from crowd reactions.

Assess a user's tendency to respond to an advertisement.

Customize styles of ads served on mobile devices.

Use video to detect penalty events

Current environment

Media used for penalty event detection will be provided by consumer devices. Media may include images and videos captured during the sporting event and shared using social media. The images and videos will have varying sizes and formats.

The data available for model building comprises of seven years of sporting event media. The sporting event media includes; recorded video transcripts or radio commentary, and logs from related social media feeds captured during the sporting events.

Crowd sentiment will include audio recordings submitted by event attendees in both mono and stereo formats.

Penalty detection and sentiment

Data scientists must build an intelligent solution by using multiple machine learning models for penalty event detection.

Data scientists must build notebooks in a local environment using automatic feature engineering and model building in machine learning pipelines.

Notebooks must be deployed to retrain by using Spark instances with dynamic worker allocation.

Notebooks must execute with the same code on new Spark instances to recode only the source of the data.

Global penalty detection models must be trained by using dynamic runtime graph computation during training.

Local penalty detection models must be written by using BrainScript.

Experiments for local crowd sentiment models must combine local penalty detection data.

Crowd sentiment models must identify known sounds such as cheers and known catch phrases. Individual crowd sentiment models will detect similar sounds.

All shared features for local models are continuous variables.

Shared features must use double precision. Subsequent layers must have aggregate running mean and standard deviation metrics available.

Advertisements

During the initial weeks in production, the following was observed:

Ad response rated declined.

Drops were not consistent across ad styles.

The distribution of features across training and production data are not consistent

Analysis shows that, of the 100 numeric features on user location and behavior, the 47 features that come from location sources are being used as raw features. A suggested experiment to remedy the bias and variance issue is to engineer 10 linearly uncorrelated features.

Initial data discovery shows a wide range of densities of target states in training data used for crowd sentiment models.

All penalty detection models show inference phases using a Stochastic Gradient Descent (SGD) are running too slow.

Audio samples show that the length of a catch phrase varies between 25%-47% depending on region. The performance of the global penalty detection models shows lower variance but higher bias when comparing training and validation sets. Before implementing any feature changes, you must confirm the bias and variance using all training and validation cases.

Ad response models must be trained at the beginning of each event and applied during the sporting event.

Market segmentation models must optimize for similar ad response history.

Sampling must guarantee mutual and collective exclusivity between local and global segmentation models that share the same features.

Local market segmentation models will be applied before determining a user's propensity to respond to an advertisement.

Ad response models must support non-linear boundaries of features.

The ad propensity model uses a cut threshold is 0.45 and retrains occur if weighted Kappa deviated from 0.1 +/- 5%.

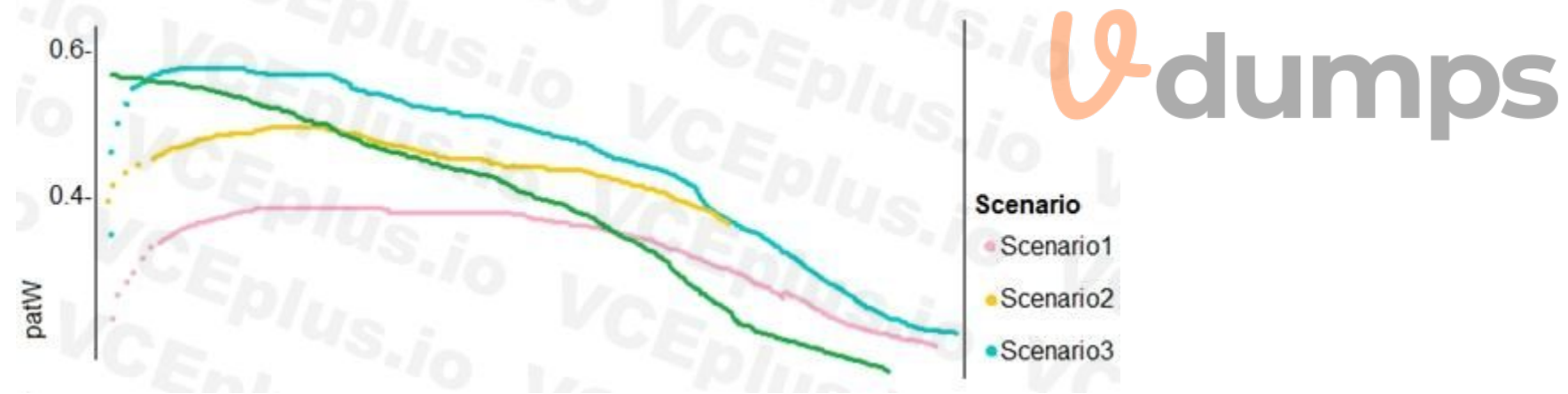
The ad propensity model uses cost factors shown in the following diagram:

|           |   | Actual |   |
|-----------|---|--------|---|
|           |   | 1      | 0 |
| Predicted | 0 | 1      | 2 |
|           | 1 | 2      | 1 |

The ad propensity model uses proposed cost factors shown in the following diagram:

|           |   | Actual |   |
|-----------|---|--------|---|
|           |   | 1      | 0 |
| Predicted | 0 | 1      | 5 |
|           | 1 | 5      | 1 |

Performance curves of current and proposed cost factor scenarios are shown in the following diagram:



**QUESTION 1**

**HOTSPOT**

You need to use the Python language to build a sampling strategy for the global penalty detection models.

How should you complete the code segment? To answer, select the appropriate options in the answer area.

NOTE: Each correct selection is worth one point.

**Hot Area:**

## Answer Area

```
import torch as deeplearninglib
import tensorflow as deeplearninglib
import cntk as deeplearninglib
```

```
train_sampler = deeplearninglib.DistributedSampler(penalty_video_dataset)
train_sampler = deeplearninglib.log_uniform_candidate_sampler(penalty_video_dataset)
train_sampler = deeplearninglib.WeightedRandomSampler(penalty_video_dataset)
train_sampler = deeplearninglib.all_candidate_sampler(penalty_video_dataset)
```

```
...
train_loader =
```

```
...
(train_sampler, penalty_video_dataset)
```

```
optimizer = deeplearninglib.optim.SGD(model.parameters(), lr=0.01)
optimizer = deeplearninglib.train.GradientDescentOptimizer(learning_rate=0.10)
```

```
model = deeplearninglib.parallel.Distributed(DataParallel(model))
model = deeplearninglib.nn.parallel.DistributedDataParallelCPU(model)
model = deeplearninglib.keras.Model([
model = deeplearninglib.keras.Sequential([
```

```
...
train_sampler.set_epoch(epoch)
for data, target in train_loader:
 data, target = data.to(device), target.to(device)
..
```

Answer Area:

 Vdumps



## Answer Area

```
import torch as deeplearninglib
import tensorflow as deeplearninglib
import cntk as deeplearninglib
```

```
train_sampler = deeplearninglib.DistributedSampler(penalty_video_dataset)
train_sampler = deeplearninglib.log_uniform_candidate_sampler(penalty_video_dataset)
train_sampler = deeplearninglib.WeightedRandomSampler(penalty_video_dataset)
train_sampler = deeplearninglib.all_candidate_sampler(penalty_video_dataset)
```

```
...
train_loader =
```

```
...
(train_sampler, penalty_video_dataset)
```

```
optimizer = deeplearninglib.optim.SGD(model.parameters(), lr=0.01)
optimizer = deeplearninglib.train.GradientDescentOptimizer(learning_rate=0.10)
```

```
model = deeplearninglib.parallel.Distributed(DataParallel(model))
model = deeplearninglib.nn.parallel.DistributedDataParallelCPU(model)
model = deeplearninglib.keras.Model([
model = deeplearninglib.keras.Sequential([
```

```
...
train_sampler.set_epoch(epoch)
for data, target in train_loader:
 data, target = data.to(device), target.to(device)
..
```

### Section:

#### Explanation:

Box 1: import torch as deeplearninglib

Box 2: ..DistributedSampler(Sampler)..

DistributedSampler(Sampler):

Sampler that restricts data loading to a subset of the dataset.

It is especially useful in conjunction with class: `torch.nn.parallel.DistributedDataParallel`. In such case, each process can pass a DistributedSampler instance as a DataLoader sampler, and load a subset of the original dataset that is exclusive to it.

Scenario: Sampling must guarantee mutual and collective exclusivity between local and global segmentation models that share the same features.

Box 3: optimizer = deeplearninglib.train.GradientDescentOptimizer(learning\_rate=0.10)

Incorrect Answers: ..SGD..

Scenario: All penalty detection models show inference phases using a Stochastic Gradient Descent (SGD) are running too slow.

Box 4: .. nn.parallel.DistributedDataParallel..

DistributedSampler(Sampler): The sampler that restricts data loading to a subset of the dataset.

It is especially useful in conjunction with :class:`torch.nn.parallel.DistributedDataParallel`.

References:

<https://github.com/pytorch/pytorch/blob/master/torch/utils/data/distributed.py>



**QUESTION 2**


DRAG DROP

You need to define an evaluation strategy for the crowd sentiment models.

Which three actions should you perform in sequence? To answer, move the appropriate actions from the list of actions to the answer area and arrange them in the correct order.

Select and Place:

| Actions                                                                            | Answer Area |
|------------------------------------------------------------------------------------|-------------|
| Add new features for retraining supervised models.                                 |             |
| Filter labeled cases for retraining using the shortest distance from centroids.    |             |
| Evaluate the changes in correlation between model error rate and centroid distance | ⬅️          |
| Impute unavailable features with centroid aligned models                           | ➡️          |
| Filter labeled cases for retraining using the longest distance from centroids.     |             |
| Remove features before retraining supervised models.                               |             |



Correct Answer:

**Actions**

Filter labeled cases for retraining using the shortest distance from centroids.

Impute unavailable features with centroid aligned models

Remove features before retraining supervised models.

**Answer Area**

Add new features for retraining supervised models.

Evaluate the changes in correlation between model error rate and centroid distance

Filter labeled cases for retraining using the longest distance from centroids.

**Section:**

**Explanation:**

Scenario:

Experiments for local crowd sentiment models must combine local penalty detection data.

Crowd sentiment models must identify known sounds such as cheers and known catch phrases. Individual crowd sentiment models will detect similar sounds.

Note: Evaluate the changed in correlation between model error rate and centroid distance

In machine learning, a nearest centroid classifier or nearest prototype classifier is a classification model that assigns to observations the label of the class of training samples whose mean (centroid) is closest to the observation.

References:

[https://en.wikipedia.org/wiki/Nearest\\_centroid\\_classifier](https://en.wikipedia.org/wiki/Nearest_centroid_classifier)

<https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/sweep-clustering>

**QUESTION 3**

DRAG DROP

You need to define a modeling strategy for ad response.

Which three actions should you perform in sequence? To answer, move the appropriate actions from the list of actions to the answer area and arrange them in the correct order.

**Select and Place:**



**Action**

- Implement a K-Means Clustering model.
- Use the raw score as a feature in a Score Matchbox Recommender model.
- Use the cluster as a feature in a Decision Jungle model.
- Use the raw score as a feature in a Logistic Regression model.
- Implement a Sweep Clustering model.

**Answer area**



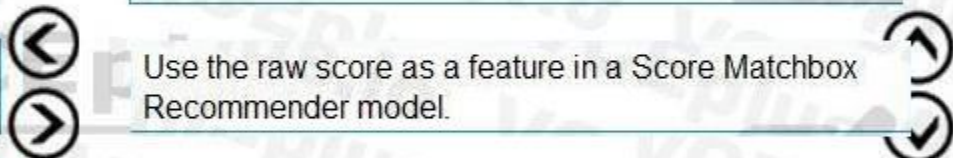
**Correct Answer:**

**Action**

- 
- 
- 
- Use the raw score as a feature in a Logistic Regression model.
- Implement a Sweep Clustering model.

**Answer area**

- Implement a K-Means Clustering model.
- Use the cluster as a feature in a Decision Jungle model.
- Use the raw score as a feature in a Score Matchbox Recommender model.



**Section:**

**Explanation:**

Step 1: Implement a K-Means Clustering model

Step 2: Use the cluster as a feature in a Decision jungle model.

Decision jungles are non-parametric models, which can represent non-linear decision boundaries.

Step 3: Use the raw score as a feature in a Score Matchbox Recommender model

The goal of creating a recommendation system is to recommend one or more "items" to "users" of the system. Examples of an item could be a movie, restaurant, book, or song. A user could be a person, group of persons, or other entity with item preferences.

Scenario:

Ad response rated declined.

Ad response models must be trained at the beginning of each event and applied during the sporting event.

Market segmentation models must optimize for similar ad response history.



Ad response models must support non-linear boundaries of features.

References:

<https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/multiclass-decision-jungle>

<https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/score-matchbox-recommender>

#### QUESTION 4

DRAG DROP

You need to define an evaluation strategy for the crowd sentiment models.

Which three actions should you perform in sequence? To answer, move the appropriate actions from the list of actions to the answer area and arrange them in the correct order.

Select and Place:

| Actions                                       | Answer Area |
|-----------------------------------------------|-------------|
| Define a cross-entropy function activation.   |             |
| Add cost functions for each target state.     |             |
| Evaluate the classification error metric.     |             |
| Evaluate the distance error metric.           |             |
| Add cost functions for each component metric. |             |
| Define a sigmoid loss function activation.    |             |

**Correct Answer:**

| Actions                                       | Answer Area                                 |
|-----------------------------------------------|---------------------------------------------|
|                                               | Define a cross-entropy function activation. |
|                                               | Add cost functions for each target state.   |
| Evaluate the classification error metric.     | Evaluate the distance error metric.         |
|                                               |                                             |
| Add cost functions for each component metric. |                                             |
| Define a sigmoid loss function activation.    |                                             |

Correct Answer:

| Actions                                       | Answer Area                                 |
|-----------------------------------------------|---------------------------------------------|
|                                               | Define a cross-entropy function activation. |
|                                               | Add cost functions for each target state.   |
| Evaluate the classification error metric.     | Evaluate the distance error metric.         |
|                                               |                                             |
| Add cost functions for each component metric. |                                             |
| Define a sigmoid loss function activation.    |                                             |

Section:

Explanation:

Step 1: Define a cross-entropy function activation

When using a neural network to perform classification and prediction, it is usually better to use cross-entropy error than classification error, and somewhat better to use cross-entropy error than mean squared error to evaluate the quality of the neural network.

Step 2: Add cost functions for each target state.

Step 3: Evaluated the distance error metric.

References:

<https://www.analyticsvidhya.com/blog/2018/04/fundamentals-deep-learning-regularization-techniques/>

#### QUESTION 5

You need to implement a scaling strategy for the local penalty detection data.

Which normalization type should you use?

- A. Streaming
- B. Weight
- C. Batch
- D. Cosine

**Correct Answer: C**

**Section:**

**Explanation:**

Post batch normalization statistics (PBN) is the Microsoft Cognitive Toolkit (CNTK) version of how to evaluate the population mean and variance of Batch Normalization which could be used in inference Original Paper.

In CNTK, custom networks are defined using the BrainScriptNetworkBuilder and described in the CNTK network description language "BrainScript." Scenario:

Local penalty detection models must be written by using BrainScript.

Reference:

<https://docs.microsoft.com/en-us/cognitive-toolkit/post-batch-normalization-statistics>

#### QUESTION 6

You need to implement a feature engineering strategy for the crowd sentiment local models.

What should you do?

- A. Apply an analysis of variance (ANOVA).
- B. Apply a Pearson correlation coefficient.
- C. Apply a Spearman correlation coefficient.
- D. Apply a linear discriminant analysis.

**Correct Answer: D**

**Section:**

**Explanation:**

The linear discriminant analysis method works only on continuous variables, not categorical or ordinal variables.

Linear discriminant analysis is similar to analysis of variance (ANOVA) in that it works by comparing the means of the variables.

Scenario:

Data scientists must build notebooks in a local environment using automatic feature engineering and model building in machine learning pipelines. Experiments for local crowd sentiment models must combine local penalty detection data. All shared features for local models are continuous variables.

Incorrect Answers:

B: The Pearson correlation coefficient, sometimes called Pearson's R test, is a statistical value that measures the linear relationship between two variables. By examining the coefficient values, you can infer something about the strength of the relationship between the two variables, and whether they are positively correlated or negatively correlated.

C: Spearman's correlation coefficient is designed for use with non-parametric and non-normally distributed data. Spearman's coefficient is a nonparametric measure of statistical dependence between two variables, and is sometimes denoted by the Greek letter rho. The Spearman's coefficient expresses the degree to which two variables are monotonically related. It is also called Spearman rank correlation, because it can be used with ordinal variables.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/fisher-linear-discriminant-analysis> <https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/compute-linear-correlation>

#### QUESTION 7

You need to implement a model development strategy to determine a user's tendency to respond to an ad.

Which technique should you use?

- A. Use a Relative Expression Split module to partition the data based on centroid distance.
- B. Use a Relative Expression Split module to partition the data based on distance travelled to the event.
- C. Use a Split Rows module to partition the data based on distance travelled to the event.
- D. Use a Split Rows module to partition the data based on centroid distance.

**Correct Answer: A**

**Section:**

**Explanation:**

Split Data partitions the rows of a dataset into two distinct sets.

The Relative Expression Split option in the Split Data module of Azure Machine Learning Studio is helpful when you need to divide a dataset into training and testing datasets using a numerical expression.

Relative Expression Split: Use this option whenever you want to apply a condition to a number column. The number could be a date/time field, a column containing age or dollar amounts, or even a percentage. For example, you might want to divide your data set depending on the cost of the items, group people by age ranges, or separate data by a calendar date.

Scenario:

Local market segmentation models will be applied before determining a user's propensity to respond to an advertisement. The distribution of features across training and production data are not consistent

Reference: <https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/split-data>

### QUESTION 8

You need to implement a new cost factor scenario for the ad response models as illustrated in the performance curve exhibit.

Which technique should you use?

- A. Set the threshold to 0.5 and retrain if weighted Kappa deviates +/- 5% from 0.45.
- B. Set the threshold to 0.05 and retrain if weighted Kappa deviates +/- 5% from 0.5.
- C. Set the threshold to 0.2 and retrain if weighted Kappa deviates +/- 5% from 0.6.
- D. Set the threshold to 0.75 and retrain if weighted Kappa deviates +/- 5% from 0.15.



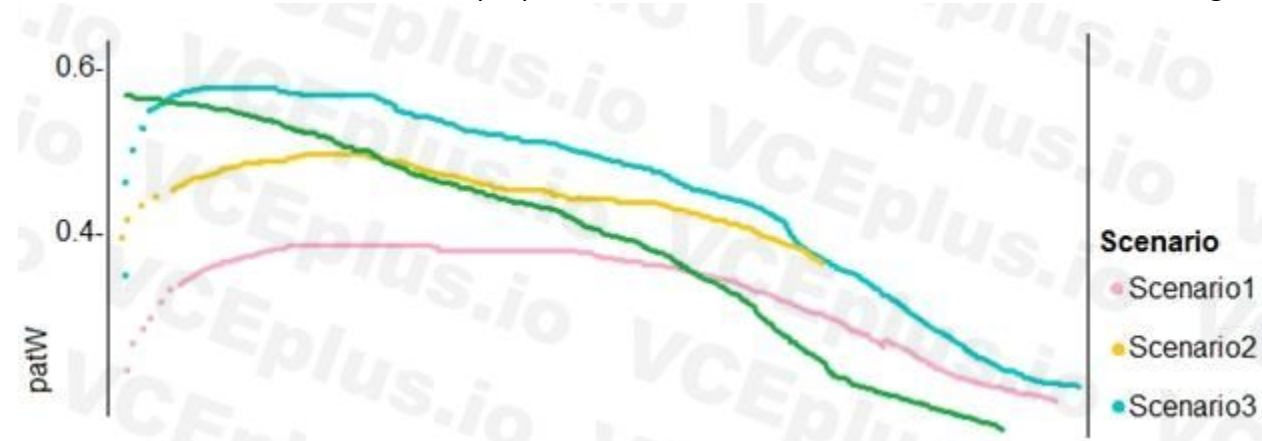
**Correct Answer: A**

**Section:**

**Explanation:**

Scenario:

Performance curves of current and proposed cost factor scenarios are shown in the following diagram:



The ad propensity model uses a cut threshold is 0.45 and retrains occur if weighted Kappa deviated from 0.1 +/- 5%.

### 03 - Run experiments and train models

Case study

This is a case study. Case studies are not timed separately. You can use as much exam time as you would like to complete each case. However, there may be additional case studies and sections on this exam. You must manage



your time to ensure that you are able to complete all questions included on this exam in the time provided.

To answer the questions included in a case study, you will need to reference information that is provided in the case study. Case studies might contain exhibits and other resources that provide more information about the scenario that is described in the case study. Each question is independent of the other questions in this case study.

At the end of this case study, a review screen will appear. This screen allows you to review your answers and to make changes before you move to the next section of the exam. After you begin a new section, you cannot return to this section.

To start the case study To display the first question in this case study, click the Next button. Use the buttons in the left pane to explore the content of the case study before you answer the questions. Clicking these buttons displays information such as business requirements, existing environment, and problem statements. If the case study has an All Information tab, note that the information displayed is identical to the information displayed on the subsequent tabs. When you are ready to answer a question, click the Question button to return to the question.

#### Overview

You are a data scientist for Fabrikam Residences, a company specializing in quality private and commercial property in the United States. Fabrikam Residences is considering expanding into Europe and has asked you to investigate prices for private residences in major European cities.

You use Azure Machine Learning Studio to measure the median value of properties. You produce a regression model to predict property prices by using the Linear Regression and Bayesian Linear Regression modules.

#### Datasets

There are two datasets in CSV format that contain property details for two cities, London and Paris. You add both files to Azure Machine Learning Studio as separate datasets to the starting point for an experiment. Both datasets contain the following columns:

| Column heading             | Description                                                                |
|----------------------------|----------------------------------------------------------------------------|
| CapitaCrimeRate            | per capita crime rate by town                                              |
| Zoned                      | proportion of residential land zoned for lots over 25,000 square feet      |
| NonRetailAcres             | proportion of retail business acres per town                               |
| NextToRiver                | proximity of a property to the river                                       |
| NitrogenOxideConcentration | nitric oxides concentration (parts per 10 million)                         |
| AvgRoomsPerHouse           | average number of rooms per dwelling                                       |
| Age                        | proportion of owner-occupied units built prior to 1940                     |
| DistanceToEmploymentCenter | weighted distances to employment centers                                   |
| AccessibilityToHighway     | index of accessibility to radial highways to a value of two decimal places |
| Tax                        | full value property tax rate per \$10,000                                  |
| PupilTeacherRatio          | pupil to teacher ratio by town                                             |
| ProfessionalClass          | professional class percentage                                              |
| LowerStatus                | percentage lower status of the population                                  |
| MedianValue                | median value of owner-occupied homes in \$1000s                            |

An initial investigation shows that the datasets are identical in structure apart from the MedianValue column. The smaller Paris dataset contains the MedianValue in text format, whereas the larger London dataset contains the MedianValue in numerical format.

#### Data issues

##### Missing values

The AccessibilityToHighway column in both datasets contains missing values. The missing data must be replaced with new data so that it is modeled conditionally using the other variables in the data before filling in the missing values.

Columns in each dataset contain missing and null values. The datasets also contain many outliers. The Age column has a high proportion of outliers. You need to remove the rows that have outliers in the Age column. The MedianValue and AvgRoomsInHouse columns both hold data in numeric format. You need to select a feature selection algorithm to analyze the relationship between the two columns in more detail.

#### Model fit

The model shows signs of overfitting. You need to produce a more refined regression model that reduces the overfitting.

#### Experiment requirements

You must set up the experiment to cross-validate the Linear Regression and Bayesian Linear Regression modules to evaluate performance. In each case, the predictor of the dataset is the column named MedianValue. You must ensure that the datatype of the MedianValue column of the Paris dataset matches the structure of the London dataset.

You must prioritize the columns of data for predicting the outcome. You must use non-parametric statistics to measure relationships.

You must use a feature selection algorithm to analyze the relationship between the MedianValue and AvgRoomsInHouse columns.

#### Model training

##### Permutation Feature Importance

Given a trained model and a test dataset, you must compute the Permutation Feature Importance scores of feature variables. You must be determined the absolute fit for the model.

#### Hyperparameters

You must configure hyperparameters in the model learning process to speed the learning phase. In addition, this configuration should cancel the lowest performing runs at each evaluation interval, thereby directing effort and resources towards models that are more likely to be successful.

You are concerned that the model might not efficiently use compute resources in hyperparameter tuning. You also are concerned that the model might prevent an increase in the overall tuning time. Therefore, must implement an early stopping



criterion on models that provides savings without terminating promising jobs.

Testing

You must produce multiple partitions of a dataset based on sampling using the Partition and Sample module in Azure Machine Learning Studio.

Cross-validation

You must create three equal partitions for cross-validation. You must also configure the cross-validation process so that the rows in the test and training datasets are divided evenly by properties that are near each city's main river. You must complete this task before the data goes through the sampling process.

Linear regression module

When you train a Linear Regression module, you must determine the best features to use in a model. You can choose standard metrics provided to measure performance before and after the feature importance process completes. The distribution of features across multiple training models must be consistent.

Data visualization

You need to provide the test results to the Fabrikam Residences team. You create data visualizations to aid in presenting the results.

You must produce a Receiver Operating Characteristic (ROC) curve to conduct a diagnostic test evaluation of the model. You need to select appropriate methods for producing the ROC curve in Azure Machine Learning Studio to compare the Two-Class Decision Forest and the Two-Class Decision Jungle modules with one another.

## QUESTION 1

HOTSPOT

You need to configure the Permutation Feature Importance module for the model training requirements.

What should you do? To answer, select the appropriate options in the dialog box in the answer area.

NOTE: Each correct selection is worth one point.

Hot Area:

**Answer Area**

Permutation Feature importance

Random seed

|     |   |
|-----|---|
|     | ▼ |
| 0   |   |
| 500 |   |

Regression – Root Mean Square Error

Regression – R-squared

Regression – Mean Zero One Error

Regression – Mean Absolute Error

Answer Area:

**Answer Area**

Permutation Feature importance

Random seed

|     |   |
|-----|---|
|     | ▼ |
| 0   |   |
| 500 |   |

Regression – Root Mean Square Error

Regression – R-squared

Regression – Mean Zero One Error

Regression – Mean Absolute Error

**vdumps**

**Section:**

**Explanation:**

Box 1: 500

For Random seed, type a value to use as seed for randomization. If you specify 0 (the default), a number is generated based on the system clock.

A seed value is optional, but you should provide a value if you want reproducibility across runs of the same experiment.

Here we must replicate the findings.

Box 2: Mean Absolute Error

Scenario: Given a trained model and a test dataset, you must compute the Permutation Feature Importance scores of feature variables. You need to set up the Permutation Feature Importance module to select the correct metric to investigate the model's accuracy and replicate the findings.

Regression. Choose one of the following: Precision, Recall, Mean Absolute Error , Root Mean Squared Error, Relative Absolute Error, Relative Squared Error, Coefficient of Determination

References:

<https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/permutation-feature-importance>

**QUESTION 2**

**HOTSPOT**

You need to set up the Permutation Feature Importance module according to the model training requirements.

Which properties should you select? To answer, select the appropriate options in the answer area.

NOTE: Each correct selection is worth one point.

Hot Area:

Answer Area

**Tune Model Hyperparameters**

Specify parameter sweeping mode  
Random sweep

Maximum number of runs on random sweep  
5

Random seed  
0

Label column  
Selected columns:  
Column names: MedianValue  
Launch column selector

Metric for measuring performance for classification

|           |
|-----------|
| ▼         |
| F-score   |
| Precision |
| Recall    |
| Accuracy  |

Metric for measuring performance for regression

|                            |
|----------------------------|
| ▼                          |
| Root of mean squared error |
| R-squared                  |
| Mean zero one error        |
| Mean absolute error        |

 **Vdumps**

Answer Area:

## Answer Area

### Tune Model Hyperparameters

Specify parameter sweeping mode

Random sweep

Maximum number of runs on random sweep

5

Random seed

0

Label column

Selected columns:  
Column names: MedianValue

Launch column selector

Metric for measuring performance for classification

F-score  
Precision  
Recall  
Accuracy

Metric for measuring performance for regression

Root of mean squared error  
R-squared  
Mean zero one error  
Mean absolute error

### Section:

#### Explanation:

Box 1: Accuracy

Scenario: You want to configure hyperparameters in the model learning process to speed the learning phase by using hyperparameters. In addition, this configuration should cancel the lowest performing runs at each evaluation interval, thereby directing effort and resources towards models that are more likely to be successful.

Box 2: R-Squared

### QUESTION 3

#### HOTSPOT

You need to configure the Feature Based Feature Selection module based on the experiment requirements and datasets.

How should you configure the module properties? To answer, select the appropriate options in the dialog box in the answer area.

NOTE: Each correct selection is worth one point.

### Hot Area:



**Answer Area**

Filter Based Feature Selection

Feature scoring method

|                    |
|--------------------|
| ▼                  |
| Fisher Score       |
| Chi-squared        |
| Mutual information |
| Counts             |

Operate on feature columns only

Target column

|                 |
|-----------------|
| ▼               |
| MedianValue     |
| AvgRooms/nHouse |

Launch column selector

Number of desired features

1

Answer Area:

**Answer Area**

Filter Based Feature Selection

Feature scoring method

|                    |
|--------------------|
| ▼                  |
| Fisher Score       |
| Chi-squared        |
| Mutual information |
| Counts             |

Operate on feature columns only

Target column

|                 |
|-----------------|
| ▼               |
| MedianValue     |
| AvgRooms/nHouse |

Launch column selector

Number of desired features

1

Section:



**Explanation:**

Box 1: Mutual Information.

The mutual information score is particularly useful in feature selection because it maximizes the mutual information between the joint distribution and target variables in datasets with many dimensions.

Box 2: MedianValue

MedianValue is the feature column, it is the predictor of the dataset.

Scenario: The MedianValue and AvgRoomsInHouse columns both hold data in numeric format. You need to select a feature selection algorithm to analyze the relationship between the two columns in more detail.

References:

<https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/filter-based-feature-selection>

**QUESTION 4**



DRAG DROP

You need to implement an early stopping criteria policy for model training.

Which three code segments should you use to develop the solution? To answer, move the appropriate code segments from the list of code segments to the answer area and arrange them in the correct order.

NOTE: More than one order of answer choices is correct. You will receive credit for any of the correct orders you select.

Select and Place:

| Code segments                                                                                                                        | Answer Area                                                                                                                                                                               |
|--------------------------------------------------------------------------------------------------------------------------------------|-------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|
| <pre>early_termination_policy = TruncationSelectionPolicy(evaluation_interval=1, truncation_percentage=20, delay_evaluation=5)</pre> | <br><br>plus.com |
| <pre>import TruncationSelectionPolicy</pre>                                                                                          |                                                                                                                                                                                           |
| <pre>from azureml.train.hyperdrive</pre>                                                                                             |                                                                                                                                                                                           |
| <pre>import BanditPolicy</pre>                                                                                                       |                                                                                                                                                                                           |
| <pre>early_termination_policy = BanditPolicy (slack_factor = 0.1, evaluation_interval=1, delay_evaluation=5)</pre>                   |                                                                                                                                                                                           |

Correct Answer:

**Code segments**

import BanditPolicy

early\_termination\_policy = BanditPolicy  
(slack\_factor = 0.1, evaluation\_interval=1,  
delay\_evaluation=5)

**Answer Area**

from azureml.train.hyperdrive

import TruncationSelectionPolicy

⏪
⏩
 early\_termination\_policy =  
 TruncationSelectionPolicy(evaluation\_interval=1,  
 truncation\_percentage=20, delay\_evaluation=5)

**Section:**

**Explanation:**

You need to implement an early stopping criterion on models that provides savings without terminating promising jobs.

Truncation selection cancels a given percentage of lowest performing runs at each evaluation interval. Runs are compared based on their performance on the primary metric and the lowest X% are terminated.

Example:

```
from azureml.train.hyperdrive import TruncationSelectionPolicy
early_termination_policy = TruncationSelectionPolicy(evaluation_interval=1, truncation_percentage=20, delay_evaluation=5)
```

Incorrect Answers:

Bandit is a termination policy based on slack factor/slack amount and evaluation interval. The policy early terminates any runs where the primary metric is not within the specified slack factor / slack amount with respect to the best performing training run.

Example:

```
from azureml.train.hyperdrive import BanditPolicy
early_termination_policy = BanditPolicy(slack_factor = 0.1, evaluation_interval=1, delay_evaluation=5)
```

References:

<https://docs.microsoft.com/en-us/azure/machine-learning/service/how-to-tune-hyperparameters>

**QUESTION 5**

**DRAG DROP**

You need to implement early stopping criteria as stated in the model training requirements.

Which three code segments should you use to develop the solution? To answer, move the appropriate code segments from the list of code segments to the answer area and arrange them in the correct order.

NOTE: More than one order of answer choices is correct. You will receive the credit for any of the correct orders you select.

**Select and Place:**

 **Code segments**

**Answer Area**

```
early_termination_policy = TruncationSelectionPolicy
(evaluation_interval=1, truncation_percentage=20,
delay_evaluation = 5)
```

```
import BanditPolicy
```

```
import TruncationSelectionPolicy
```

```
early_termination_policy= BanditPolicy (slack_factor =
0.1, evaluation_interval = 1, delay_evaluation = 5)
```

```
from azureml.train.hyperdrive
```

```
early_termination_policy = MedianStoppingPolicy
(evaluation_interval = 1, delay_evaluation=5)
```

```
import MedianStoppingPolicy
```



 **vdumps**



Correct Answer:



**Code segments**

```

import BanditPolicy

early_termination_policy= BanditPolicy (slack_factor = 0.1, evaluation_interval = 1, delay_evaluation = 5)

early_termination_policy = MedianStoppingPolicy (evaluation_interval = 1, delay_evaluation=5)

import MedianStoppingPolicy

```

**Answer Area**

```

from azureml.train.hyperdrive

import TruncationSelectionPolicy

early_termination_policy = TruncationSelectionPolicy (evaluation_interval=1, truncation_percentage=20, delay_evaluation = 5)

```

Navigation icons: Right arrow, Left arrow, Checkmark.

**Section:**

**Explanation:**

Step 1: from azureml.train.hyperdrive

Step 2: Import TruncationCelectionPolicy

Truncation selection cancels a given percentage of lowest performing runs at each evaluation interval. Runs are compared based on their performance on the primary metric and the lowest X% are terminated.

Scenario: You must configure hyperparameters in the model learning process to speed the learning phase. In addition, this configuration should cancel the lowest performing runs at each evaluation interval, thereby directing effort and resources towards models that are more likely to be successful.

Step 3: early\_termination\_policy = TruncationSelectionPolicy..

Example:

```
from azureml.train.hyperdrive import TruncationSelectionPolicy
```

```
early_termination_policy = TruncationSelectionPolicy(evaluation_interval=1, truncation_percentage=20, delay_evaluation=5)
```

In this example, the early termination policy is applied at every interval starting at evaluation interval 5. A run will be terminated at interval 5 if its performance at interval 5 is in the lowest 20% of performance of all runs at interval 5.

Incorrect Answers:

Median:

Median stopping is an early termination policy based on running averages of primary metrics reported by the runs. This policy computes running averages across all training runs and terminates runs whose performance is worse than the median of the running averages.

Slack:

Bandit is a termination policy based on slack factor/slack amount and evaluation interval. The policy early terminates any runs where the primary metric is not within the specified slack factor / slack amount with respect to the best performing training run.

References:

<https://docs.microsoft.com/en-us/azure/machine-learning/service/how-to-tune-hyperparameters>

#### QUESTION 6

You need to visually identify whether outliers exist in the Age column and quantify the outliers before the outliers are removed.

Which three Azure Machine Learning Studio modules should you use? Each correct answer presents part of the solution.

NOTE: Each correct selection is worth one point.

- A. Create Scatterplot
- B. Summarize Data
- C. Clip Values
- D. Replace Discrete Values
- E. Build Counting Transform

**Correct Answer: A, B, C**

**Section:**

**Explanation:**

B: To have a global view, the summarize data module can be used. Add the module and connect it to the data set that needs to be visualized. A: One way to quickly identify Outliers visually is to create scatter plots.

C: The easiest way to treat the outliers in Azure ML is to use the Clip Values module. It can identify and optionally replace data values that are above or below a specified threshold.

You can use the Clip Values module in Azure Machine Learning Studio, to identify and optionally replace data values that are above or below a specified threshold. This is useful when you want to remove outliers or replace them with a mean, a constant, or other substitute value.

Reference:

<https://blogs.msdn.microsoft.com/azuredev/2017/05/27/data-cleansing-tools-in-azure-machine-learning/> <https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/clip-values>

#### QUESTION 7

You need to select a feature extraction method.

Which method should you use?

- A. Mutual information
- B. Pearson's correlation
- C. Spearman correlation
- D. Fisher Linear Discriminant Analysis

**Correct Answer: C**

**Section:**

**Explanation:**

Spearman's rank correlation coefficient assesses how well the relationship between two variables can be described using a monotonic function.

Note: Both Spearman's and Kendall's can be formulated as special cases of a more general correlation coefficient, and they are both appropriate in this scenario.

Scenario: The MedianValue and AvgRoomsInHouse columns both hold data in numeric format. You need to select a feature selection algorithm to analyze the relationship between the two columns in more detail.

Incorrect Answers:

B: The Spearman correlation between two variables is equal to the Pearson correlation between the rank values of those two variables; while Pearson's correlation assesses linear relationships, Spearman's correlation assesses monotonic relationships (whether linear or not).

Reference: <https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/feature-selection-modules>

#### QUESTION 8

You need to select a feature extraction method.

Which method should you use?

- A. Mutual information
- B. Mood's median test
- C. Kendall correlation
- D. Permutation Feature Importance

**Correct Answer: C**

**Section:**

**Explanation:**

In statistics, the Kendall rank correlation coefficient, commonly referred to as Kendall's tau coefficient (after the Greek letter  $\tau$ ), is a statistic used to measure the ordinal association between two measured quantities. It is a supported method of the Azure Machine Learning Feature selection.

Note: Both Spearman's and Kendall's can be formulated as special cases of a more general correlation coefficient, and they are both appropriate in this scenario.

Scenario: The MedianValue and AvgRoomsInHouse columns both hold data in numeric format. You need to select a feature selection algorithm to analyze the relationship between the two columns in more detail.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/feature-selection-modules>

#### QUESTION 9

HOTSPOT

You need to replace the missing data in the AccessibilityToHighway columns.

How should you configure the Clean Missing Data module? To answer, select the appropriate options in the answer area.

NOTE: Each correct selection is worth one point.

**Hot Area:**



Answer Area

Properties Project

Clean Missing Data

Columns to be cleaned

Selected columns:

Column names: AccessibilityToHighway

Launch column selector

Minimum missing value ratio

0

Maximum missing value ratio

1

Cleaning mode

- Replace using MICE
- Replace with Mean
- Replace with Median
- Replace with Mode

Cols with all missing values.

- Propagate
- Remove

Generate missing value indicator column

Number of iterations

5



Answer Area:



Answer Area

Properties Project

Clean Missing Data

Columns to be cleaned

Selected columns:

Column names: AccessibilityToHighway

Launch column selector

Minimum missing value ratio

0

Maximum missing value ratio

1

Cleaning mode

Replace using MICE

Replace with Mean

Replace with Median

Replace with Mode

Cols with all missing values.

Propagate

Remove

Generate missing value indicator column

Number of iterations

5



Section:

Explanation:

Box 1: Replace using MICE Replace using MICE: For each missing value, this option assigns a new value, which is calculated by using a method described in the statistical literature as "Multivariate Imputation using Chained Equations" or "Multiple Imputation by Chained Equations". With a multiple imputation method, each variable with missing data is modeled conditionally using the other variables in the data before filling in the missing values.

Scenario: The AccessibilityToHighway column in both datasets contains missing values. The missing data must be replaced with new data so that it is modeled conditionally using the other variables in the data before filling in the missing values.

Box 2: Propagate

Cols with all missing values indicate if columns of all missing values should be preserved in the output.

References:

<https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/clean-missing-data>

#### QUESTION 10

DRAG DROP

You need to produce a visualization for the diagnostic test evaluation according to the data visualization requirements.

Which three modules should you recommend be used in sequence? To answer, move the appropriate modules from the list of modules to the answer area and arrange them in the correct order.

Select and Place:

**Modules**

- Score Matchbox Recommender
- Apply Transformation
- Evaluate Recommender
- Evaluate Model
- Train Model
- Sweep Clustering
- Score Model
- Load Trained Model

**Answer Area**

Correct Answer:

The screenshot shows the Azure Machine Learning Studio interface. On the left, under the heading "Modules", there is a list of available modules: "Score Matchbox Recommender", "Apply Transformation", "Evaluate Recommender", an empty box, another empty box, "Score Model", and "Load Trained Model". On the right, under the heading "Answer Area", there is a list of selected modules: "Sweep Clustering", "Train Model", and "Evaluate Model". The "Evaluate Model" module has a checkmark icon next to it. Between the two columns, there are four circular navigation arrows: a left arrow, a right arrow, an up arrow, and a down arrow. A watermark "VCEplus.io" is visible across the image, and a logo for "Vdumps" is in the bottom right corner of the screenshot area.

**Section:**

**Explanation:**

Step 1: Sweep Clustering

Start by using the "Tune Model Hyperparameters" module to select the best sets of parameters for each of the models we're considering.

One of the interesting things about the "Tune Model Hyperparameters" module is that it not only outputs the results from the Tuning, it also outputs the Trained Model.

Step 2: Train Model

Step 3: Evaluate Model

Scenario: You need to provide the test results to the Fabrikam Residences team. You create data visualizations to aid in presenting the results.

You must produce a Receiver Operating Characteristic (ROC) curve to conduct a diagnostic test evaluation of the model. You need to select appropriate methods for producing the ROC curve in Azure Machine Learning Studio to compare the Two-Class Decision Forest and the Two-Class Decision Jungle modules with one another.

References:

<http://breaking-bi.blogspot.com/2017/01/azure-machine-learning-model-evaluation.html>

**QUESTION 11**

**HOTSPOT**

You need to identify the methods for dividing the data according to the testing requirements.

Which properties should you select? To answer, select the appropriate options in the answer area.

NOTE: Each correct selection is worth one point.

**Hot Area:**



Properties Project

Partition and Sample

|                 |
|-----------------|
| Assign to Folds |
| Sampling        |
| Head            |

Partition or sample mode

Use replacement in the partitioning

Randomized split

Random seed

|                                  |
|----------------------------------|
| True                             |
| False                            |
| Partition evenly                 |
| Partition with custom partitions |

Specify the partitioner method

Specify number of folds to split evenly into

Stratified split

Stratification key column

**Selected columns:**  
**Column names:** NextToRiver

Answer Area:





Properties Project

Partition and Sample

|                 |
|-----------------|
| ▼               |
| Assign to Folds |
| Sampling        |
| Head            |

Partition or sample mode

Use replacement in the partitioning

Randomized split

Random seed

|                                  |
|----------------------------------|
| ▼                                |
| True                             |
| False                            |
| Partition evenly                 |
| Partition with custom partitions |

Specify the partitioner method

Specify number of folds to split evenly into

Stratified split

Stratification key column

**Selected columns:**  
**Column names:** NextToRiver

**Section:**

**Explanation:**

Scenario: Testing

You must produce multiple partitions of a dataset based on sampling using the Partition and Sample module in Azure Machine Learning Studio.

Box 1: Assign to folds



Use Assign to folds option when you want to divide the dataset into subsets of the data. This option is also useful when you want to create a custom number of folds for cross-validation, or to split rows into several groups.

Not Head: Use Head mode to get only the first n rows. This option is useful if you want to test a pipeline on a small number of rows, and don't need the data to be balanced or sampled in any way.

Not Sampling: The Sampling option supports simple random sampling or stratified random sampling. This is useful if you want to create a smaller representative sample dataset for testing.

Box 2: Partition evenly

Specify the partitioner method: Indicate how you want data to be apportioned to each partition, using these options:

Partition evenly: Use this option to place an equal number of rows in each partition. To specify the number of output partitions, type a whole number in the Specify number of folds to split evenly into text box.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/algorithm-module-reference/partition-and-sample>

## QUESTION 12

HOTSPOT

You need to configure the Edit Metadata module so that the structure of the datasets match.

Which configuration options should you select? To answer, select the appropriate options in the answer area.

NOTE: Each correct selection is worth one point.

**Hot Area:**



Answer Area

Properties

Project

▲ Edit Metadata

Column

**Selected columns:**

Column names: MedianValue

Launch column selector

- Floating point
- DateTime
- TimeSpan
- Integer

- Unchanged
- Make Categorical
- Make Uncategorical

Fields

5

Answer Area:



## Answer Area

Properties

Project

### ▲ Edit Metadata

Column

**Selected columns:**

Column names: MedianValue

Launch column selector

|                |
|----------------|
| ▼              |
| Floating point |
| DateTime       |
| TimeSpan       |
| Integer        |

|                    |
|--------------------|
| ▼                  |
| Unchanged          |
| Make Categorical   |
| Make Uncategorical |

Fields

5

 **vdumps**

#### Section:

#### Explanation:

Box 1: Floating point

Need floating point for Median values.

Scenario: An initial investigation shows that the datasets are identical in structure apart from the MedianValue column. The smaller Paris dataset contains the MedianValue in text format, whereas the larger London dataset contains the

MedianValue in numerical format.

Box 2: Unchanged

Note: Select the Categorical option to specify that the values in the selected columns should be treated as categories.

For example, you might have a column that contains the numbers 0,1 and 2, but know that the numbers actually mean "Smoker", "Non smoker" and "Unknown". In that case, by flagging the column as categorical you can ensure that the values are not used in numeric calculations, only to group data.

01 - Deploy and operationalize machine learning solutions



### QUESTION 1

You create a batch inference pipeline by using the Azure ML SDK. You run the pipeline by using the following code:

```
from azureml.pipeline.core import Pipeline
from azureml.core.experiment import Experiment
pipeline = Pipeline(workspace=ws, steps=[parallelrun_step])
pipeline_run = Experiment(ws, 'batch_pipeline').submit(pipeline)
```

You need to monitor the progress of the pipeline execution.

What are two possible ways to achieve this goal? Each correct answer presents a complete solution.

NOTE: Each correct selection is worth one point.

A. Run the following code in a notebook:

```
from azureml.contrib.interpret.explanation.explanation_client import ExplanationClient
client = ExplanationClient.from_run(pipeline_run)
explanation = client.download_model_explanation()
explanation = client.download_model_explanation(top_k=4)
global_importance_values = explanation.get_ranked_global_values()
global_importance_names = explanation.get_ranked_global_names()
print('global importance values: {}'.format(global_importance_values))
print('global importance names: {}'.format(global_importance_names))
```

B. Use the Inference Clusters tab in Machine Learning Studio.

C. Use the Activity log in the Azure portal for the Machine Learning workspace.

D. Run the following code in a notebook:

```
from azureml.widgets import RunDetails
RunDetails(pipeline_run).show()
```

E. Run the following code and monitor the console output from the PipelineRun object:

```
pipeline_run.wait_for_completion(show_output=True)
```

**Correct Answer: D, E**

**Section:**

**Explanation:**

A batch inference job can take a long time to finish. This example monitors progress by using a Jupyter widget. You can also manage the job's progress by using: Azure Machine Learning Studio.

Console output from the PipelineRun object.

```
from azureml.widgets import RunDetails
RunDetails(pipeline_run).show()
pipeline_run.wait_for_completion(show_output=True)
```

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/how-to-use-parallel-run-step#monitor-the-parallel-run-job>

### QUESTION 2

You create a deep learning model for image recognition on Azure Machine Learning service using GPU-based training.

You must deploy the model to a context that allows for real-time GPU-based inferencing.

You need to configure compute resources for model inferencing.

Which compute type should you use?

A. Azure Container Instance

- B. Azure Kubernetes Service
- C. Field Programmable Gate Array
- D. Machine Learning Compute

**Correct Answer: B**

**Section:**

**Explanation:**

You can use Azure Machine Learning to deploy a GPU-enabled model as a web service. Deploying a model on Azure Kubernetes Service (AKS) is one option. The AKS cluster provides a GPU resource that is used by the model for inference.

Inference, or model scoring, is the phase where the deployed model is used to make predictions. Using GPUs instead of CPUs offers performance advantages on highly parallelizable computation.

Reference: <https://docs.microsoft.com/en-us/azure/machine-learning/how-to-deploy-inferencing-gpus>

### QUESTION 3

Note: This question-is part of a series of questions that present the same scenario. Each question-in the series contains a unique solution that might meet the stated goals. Some question-sets might have more than one correct solution, while others might not have a correct solution.

After you answer a question-in this section, you will NOT be able to return to it. As a result, these questions will not appear in the review screen.

You train and register a machine learning model.

You plan to deploy the model as a real-time web service. Applications must use key-based authentication to use the model.

You need to deploy the web service.

Solution:

Create an AciWebservice instance.

Set the value of the ssl\_enabled property to True. Deploy the model to the service.

Does the solution meet the goal?

- A. Yes
- B. No



**Correct Answer: B**

**Section:**

**Explanation:**

Instead use only auth\_enabled = TRUE

Note: Key-based authentication.

Web services deployed on AKS have key-based auth enabled by default. ACI-deployed services have key-based auth disabled by default, but you can enable it by setting auth\_enabled = TRUE when creating the ACI web service. The following is an example of creating an ACI deployment configuration with key-based auth enabled.

```
deployment_config <- aci_webservice_deployment_config(cpu_cores = 1,
memory_gb = 1,
auth_enabled = TRUE)
```

Reference:

<https://azure.github.io/azureml-sdk-for-r/articles/deploying-models.html>

### QUESTION 4

Note: This question-is part of a series of questions that present the same scenario. Each question-in the series contains a unique solution that might meet the stated goals. Some question-sets might have more than one correct solution, while others might not have a correct solution.

After you answer a question-in this section, you will NOT be able to return to it. As a result, these questions will not appear in the review screen.

You train and register a machine learning model.

You plan to deploy the model as a real-time web service. Applications must use key-based authentication to use the model.

You need to deploy the web service.

Solution:

Create an AksWebservice instance.

Set the value of the `auth_enabled` property to `True`. Deploy the model to the service.  
Does the solution meet the goal?

- A. Yes
- B. No

**Correct Answer: A**

**Section:**

**Explanation:**

Key-based authentication.

Web services deployed on AKS have key-based auth enabled by default. ACI-deployed services have key-based auth disabled by default, but you can enable it by setting `auth_enabled = TRUE` when creating the ACI web service. The following is an example of creating an ACI deployment configuration with key-based auth enabled.

```
deployment_config <- aci_webservice_deployment_config(cpu_cores = 1, memory_gb = 1, auth_enabled = TRUE)
```

Reference:

<https://azure.github.io/azureml-sdk-for-r/articles/deploying-models.html>

#### QUESTION 5

Note: This question is part of a series of questions that present the same scenario. Each question in the series contains a unique solution that might meet the stated goals. Some question sets might have more than one correct solution, while others might not have a correct solution.

After you answer a question in this section, you will NOT be able to return to it. As a result, these questions will not appear in the review screen.

You train and register a machine learning model.

You plan to deploy the model as a real-time web service. Applications must use key-based authentication to use the model.

You need to deploy the web service.

Solution:

Create an `AksWebservice` instance.

Set the value of the `auth_enabled` property to `False`.

Set the value of the `token_auth_enabled` property to `True`. Deploy the model to the service.

Does the solution meet the goal?

- A. Yes
- B. No

**Correct Answer: B**

**Section:**

**Explanation:**

Instead use only `auth_enabled = TRUE`

Note: Key-based authentication.

Web services deployed on AKS have key-based auth enabled by default. ACI-deployed services have key-based auth disabled by default, but you can enable it by setting `auth_enabled = TRUE` when creating the ACI web service. The following is an example of creating an ACI deployment configuration with key-based auth enabled.

```
deployment_config <- aci_webservice_deployment_config(cpu_cores = 1, memory_gb = 1, auth_enabled = TRUE)
```

Reference:

<https://azure.github.io/azureml-sdk-for-r/articles/deploying-models.html>

#### QUESTION 6

You use the following Python code in a notebook to deploy a model as a web service:

```
from azureml.core.webservice import AciWebservice
```

```
from azureml.core.model import InferenceConfig
```

```
inference_config = InferenceConfig(runtime='python', source_directory='model_files', entry_script='score.py', conda_file='env.yml')
```

```
deployment_config = AciWebservice.deploy_configuration(cpu_cores=1, memory_gb=1)
```



```
service = Model.deploy(ws, 'my-service', [model], inference_config, deployment_config)
```

```
service.wait_for_deployment(True)
```

The deployment fails.

You need to use the Python SDK in the notebook to determine the events that occurred during service deployment an initialization.

Which code segment should you use?

- A. `service.state`
- B. `service.get_logs()`
- C. `service.serialize()`
- D. `service.environment`

**Correct Answer: B**

**Section:**

**Explanation:**

The first step in debugging errors is to get your deployment logs. In Python: `service.get_logs()`

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/how-to-troubleshoot-deployment>

#### QUESTION 7

You use the Azure Machine Learning Python SDK to define a pipeline that consists of multiple steps.

When you run the pipeline, you observe that some steps do not run. The cached output from a previous run is used instead.

You need to ensure that every step in the pipeline is run, even if the parameters and contents of the source directory have not changed since the previous run.

What are two possible ways to achieve this goal? Each correct answer presents a complete solution.

NOTE: Each correct selection is worth one point.

- A. Use a `PipelineData` object that references a datastore other than the default datastore.
- B. Set the `regenerate_outputs` property of the pipeline to `True`.
- C. Set the `allow_reuse` property of each step in the pipeline to `False`.
- D. Restart the compute cluster where the pipeline experiment is configured to run.
- E. Set the `outputs` property of each step in the pipeline to `True`.

**Correct Answer: B, C**

**Section:**

**Explanation:**

B: If `regenerate_outputs` is set to `True`, a new submit will always force generation of all step outputs, and disallow data reuse for any step of this run. Once this run is complete, however, subsequent runs may reuse the results of this run.

C: Keep the following in mind when working with pipeline steps, input/output data, and step reuse.

If data used in a step is in a datastore and `allow_reuse` is `True`, then changes to the data change won't be detected. If the data is uploaded as part of the snapshot (under the step's `source_directory`), though this is not recommended, then the hash will change and will trigger a rerun.

Reference: <https://docs.microsoft.com/en-us/python/api/azureml-pipeline-core/azureml.pipeline.core.pipelinestep> <https://github.com/Azure/MachineLearningNotebooks/blob/master/how-to-use-azureml/machine-learning-pipelines/intro-to-pipelines/aml-pipelines-getting-started.ipynb>

#### QUESTION 8

You train and register a model in your Azure Machine Learning workspace.

You must publish a pipeline that enables client applications to use the model for batch inferencing. You must use a pipeline with a single `ParallelRunStep` step that runs a Python inferencing script to get predictions from the input data.

You need to create the inferencing script for the `ParallelRunStep` pipeline step.

Which two functions should you include? Each correct answer presents part of the solution.



NOTE: Each correct selection is worth one point.

- A. run(mini\_batch)
- B. main()
- C. batch()
- D. init()
- E. score(mini\_batch)

**Correct Answer: A, D**

**Section:**

**Explanation:**

Reference: <https://github.com/Azure/MachineLearningNotebooks/tree/master/how-to-use-azureml/machine-learning-pipelines/parallel-run>

#### QUESTION 9

You deploy a model as an Azure Machine Learning real-time web service using the following code.

```
ws, model, inference_config, and deployment_config defined previously
service = Model.deploy(ws, 'classification-service', [model], inference_config, deployment_config)
service.wait_for_deployment(True)
```

The deployment fails.

You need to troubleshoot the deployment failure by determining the actions that were performed during deployment and identifying the specific action that failed.

Which code segment should you run?

- A. service.get\_logs()
- B. service.state
- C. service.serialize()
- D. service.update\_deployment\_state()



**Correct Answer: A**

**Section:**

**Explanation:**

You can print out detailed Docker engine log messages from the service object. You can view the log for ACI, AKS, and Local deployments. The following example demonstrates how to print the logs.

```
if you already have the service object handy print(service.get_logs())
```

```
if you only know the name of the service (note there might be multiple services with the same name but different version number) print(ws.webservices['mysvc'].get_logs())
```

Reference: <https://docs.microsoft.com/en-us/azure/machine-learning/how-to-troubleshoot-deployment>

#### QUESTION 10

You create a multi-class image classification deep learning model.

You train the model by using PyTorch version 1.2.

You need to ensure that the correct version of PyTorch can be identified for the inferencing environment when the model is deployed.

What should you do?

- A. Save the model locally as a .pt file, and deploy the model as a local web service.
- B. Deploy the model on computer that is configured to use the default Azure Machine Learning conda environment.
- C. Register the model with a .pt file extension and the default version property.
- D. Register the model, specifying the model\_framework and model\_framework\_version properties.

**Correct Answer: D**

**Section:**

**Explanation:**

framework\_version: The PyTorch version to be used for executing training code.

Reference: <https://docs.microsoft.com/en-us/python/api/azureml-train-core/azureml.train.dnn.pytorch?view=azure-ml-py>

**QUESTION 11**

You train a machine learning model.

You must deploy the model as a real-time inference service for testing. The service requires low CPU utilization and less than 48 MB of RAM. The compute target for the deployed service must initialize automatically while minimizing cost and administrative overhead.

Which compute target should you use?

- A. Azure Container Instance (ACI)
- B. attached Azure Databricks cluster
- C. Azure Kubernetes Service (AKS) inference cluster
- D. Azure Machine Learning compute cluster

**Correct Answer: A**

**Section:**

**Explanation:**

Azure Container Instances (ACI) are suitable only for small models less than 1 GB in size. Use it for low-scale CPU-based workloads that require less than 48 GB of RAM.

Note: Microsoft recommends using single-node Azure Kubernetes Service (AKS) clusters for dev-test of larger models.

Reference: <https://docs.microsoft.com/id-id/azure/machine-learning/how-to-deploy-and-where>

**QUESTION 12**

You register a model that you plan to use in a batch inference pipeline.

The batch inference pipeline must use a ParallelRunStep step to process files in a file dataset. The script has the ParallelRunStep step runs must process six input files each time the inferencing function is called.

You need to configure the pipeline.

Which configuration setting should you specify in the ParallelRunConfig object for the ParallelRunStep step?

- A. process\_count\_per\_node= "6"
- B. node\_count= "6"
- C. mini\_batch\_size= "6"
- D. error\_threshold= "6"

**Correct Answer: B**

**Section:**

**Explanation:**

node\_count is the number of nodes in the compute target used for running the ParallelRunStep.

Incorrect Answers:

A: process\_count\_per\_node

Number of processes executed on each node. (optional, default value is number of cores on node.)

C: mini\_batch\_size

For FileDataset input, this field is the number of files user script can process in one run() call. For TabularDataset input, this field is the approximate size of data the user script can process in one run() call. Example values are 1024, 1024KB, 10MB, and 1GB.

D: error\_threshold

The number of record failures for TabularDataset and file failures for FileDataset that should be ignored during processing. If the error count goes above this value, then the job will be aborted.

Reference:

<https://docs.microsoft.com/en-us/python/api/azureml-contrib-pipeline-steps/azureml.contrib.pipeline.steps.parallelrunconfig?view=azure-ml-py>

**QUESTION 13**

You deploy a real-time inference service for a trained model.

The deployed model supports a business-critical application, and it is important to be able to monitor the data submitted to the web service and the predictions the data generates.

You need to implement a monitoring solution for the deployed model using minimal administrative effort.

What should you do?

- A. View the explanations for the registered model in Azure ML studio.
- B. Enable Azure Application Insights for the service endpoint and view logged data in the Azure portal.
- C. View the log files generated by the experiment used to train the model.
- D.

**Correct Answer: B**

**Section:**

**Explanation:**

B. Enable Azure Application Insights for the service endpoint and view logged data in the Azure portal.

C. View the log files generated by the experiment used to train the model.

D. Create an ML Flow tracking URI that references the endpoint, and view the data logged by ML Flow.

Answer: B

Explanation:

Configure logging with Azure Machine Learning studio

You can also enable Azure Application Insights from Azure Machine Learning studio. When you're ready to deploy your model as a web service, use the following steps to enable Application Insights:

1. Sign in to the studio at <https://ml.azure.com>.
2. Go to Models and select the model you want to deploy.
3. Select +Deploy.
4. Populate the Deploy model form.
5. Expand the Advanced menu.
6. Select Enable Application Insights diagnostics and data collection.



Advanced

Enable Application Insights diagnostics and data collection

Enable SSL

Enable SSL

Max concurrent requests per container

1

CPU reserve capacity ⓘ

0.1

Memory reserve capacity ⓘ

0.5

Deploy Cancel

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/how-to-enable-app-insights>

#### QUESTION 14

An organization creates and deploys a multi-class image classification deep learning model that uses a set of labeled photographs.

The software engineering team reports there is a heavy inferencing load for the prediction web services during the summer. The production web service for the model fails to meet demand despite having a fully-utilized compute cluster where the web service is deployed.

You need to improve performance of the image classification web service with minimal downtime and minimal administrative effort.

What should you advise the IT Operations team to do?

- A. Create a new compute cluster by using larger VM sizes for the nodes, redeploy the web service to that cluster, and update the DNS registration for the service endpoint to point to the new cluster.
- B. Increase the node count of the compute cluster where the web service is deployed.
- C. Increase the minimum node count of the compute cluster where the web service is deployed.
- D. Increase the VM size of nodes in the compute cluster where the web service is deployed.

**Correct Answer: B**

**Section:**

**Explanation:**

The Azure Machine Learning SDK does not provide support scaling an AKS cluster. To scale the nodes in the cluster, use the UI for your AKS cluster in the Azure Machine Learning studio. You can only change the node count, not the VM size of the cluster.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/how-to-create-attach-kubernetes>

#### QUESTION 15



You use Azure Machine Learning designer to create a real-time service endpoint. You have a single Azure Machine Learning service compute resource. You train the model and prepare the real-time pipeline for deployment. You need to publish the inference pipeline as a web service. Which compute type should you use?

- A. a new Machine Learning Compute resource
- B. Azure Kubernetes Services
- C. HDInsight
- D. the existing Machine Learning Compute resource
- E. Azure Databricks

**Correct Answer: B**

**Section:**

**Explanation:**

Azure Kubernetes Service (AKS) can be used real-time inference.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/concept-compute-target>

#### QUESTION 16

You train a model and register it in your Azure Machine Learning workspace. You are ready to deploy the model as a real-time web service. You deploy the model to an Azure Kubernetes Service (AKS) inference cluster, but the deployment fails because an error occurs when the service runs the entry script that is associated with the model deployment. You need to debug the error by iteratively modifying the code and reloading the service, without requiring a re-deployment of the service for each code update. What should you do?

- A. Modify the AKS service deployment configuration to enable application insights and re-deploy to AKS.
- B. Create an Azure Container Instances (ACI) web service deployment configuration and deploy the model on ACI.
- C. Add a breakpoint to the first line of the entry script and redeploy the service to AKS.
- D. Create a local web service deployment configuration and deploy the model to a local Docker container.
- E. Register a new version of the model and update the entry script to load the new version of the model from its registered path.

**Correct Answer: B**

**Section:**

**Explanation:**

How to work around or solve common Docker deployment errors with Azure Container Instances (ACI) and Azure Kubernetes Service (AKS) using Azure Machine Learning.

The recommended and the most up to date approach for model deployment is via the Model.deploy() API using an Environment object as an input parameter. In this case our service will create a base docker image for you during deployment stage and mount the required models all in one call. The basic deployment tasks are:

1. Register the model in the workspace model registry.
2. Define Inference Configuration:
  - a) Create an Environment object based on the dependencies you specify in the environment yaml file or use one of our procured environments.
  - b) Create an inference configuration (InferenceConfig object) based on the environment and the scoring script.
3. Deploy the model to Azure Container Instance (ACI) service or to Azure Kubernetes Service (AKS).

#### QUESTION 17

You use Azure Machine Learning designer to create a training pipeline for a regression model. You need to prepare the pipeline for deployment as an endpoint that generates predictions asynchronously for a dataset of input data values. What should you do?

- A. Clone the training pipeline.

- B. Create a batch inference pipeline from the training pipeline.
- C. Create a real-time inference pipeline from the training pipeline.
- D. Replace the dataset in the training pipeline with an Enter Data Manually module.

**Correct Answer: C**

**Section:**

**Explanation:**

You must first convert the training pipeline into a real-time inference pipeline. This process removes training modules and adds web service inputs and outputs to handle requests.

Incorrect Answers:

A: Use the Enter Data Manually module to create a small dataset by typing values.

Reference: <https://docs.microsoft.com/en-us/azure/machine-learning/tutorial-designer-automobile-price-deploy> <https://docs.microsoft.com/en-us/azure/machine-learning/algorithm-module-reference/enter-data-manually>

#### QUESTION 18

You retrain an existing model.

You need to register the new version of a model while keeping the current version of the model in the registry.

What should you do?

- A. Register a model with a different name from the existing model and a custom property named version with the value 2.
- B. Register the model with the same name as the existing model.
- C. Save the new model in the default datastore with the same name as the existing model. Do not register the new model.
- D. Delete the existing model and register the new one with the same name.

**Correct Answer: B**

**Section:**

**Explanation:**

Model version: A version of a registered model. When a new model is added to the Model Registry, it is added as Version 1. Each model registered to the same model name increments the version number.

Reference:

<https://docs.microsoft.com/en-us/azure/databricks/applications/mlflow/model-registry>

#### QUESTION 19

You use the Azure Machine Learning SDK to run a training experiment that trains a classification model and calculates its accuracy metric.

The model will be retrained each month as new data is available.

You must register the model for use in a batch inference pipeline.

You need to register the model and ensure that the models created by subsequent retraining experiments are registered only if their accuracy is higher than the currently registered model.

What are two possible ways to achieve this goal? Each correct answer presents a complete solution.

NOTE: Each correct selection is worth one point.

- A. Specify a different name for the model each time you register it.
- B. Register the model with the same name each time regardless of accuracy, and always use the latest version of the model in the batch inferencing pipeline.
- C. Specify the model framework version when registering the model, and only register subsequent models if this value is higher.
- D. Specify a property named accuracy with the accuracy metric as a value when registering the model, and only register subsequent models if their accuracy is higher than the accuracy property value of the currently registered model.
- E. Specify a tag named accuracy with the accuracy metric as a value when registering the model, and only register subsequent models if their accuracy is higher than the accuracy tag value of the currently registered model.

**Correct Answer: C, E**

**Section:**

**Explanation:**

E: Using tags, you can track useful information such as the name and version of the machine learning library used to train the model. Note that tags must be alphanumeric.

Reference:

<https://notebooks.azure.com/xavierheriat/projects/azureml-getting-started/html/how-to-use-azureml/deployment/register-model-create-image-deploy-service/register-model-create-image-deploy-service.ipynb>

#### QUESTION 20

You are a data scientist working for a hotel booking website company. You use the Azure Machine Learning service to train a model that identifies fraudulent transactions.

You must deploy the model as an Azure Machine Learning real-time web service using the Model.deploy method in the Azure Machine Learning SDK. The deployed web service must return real-time predictions of fraud based on transaction data input.

You need to create the script that is specified as the entry\_script parameter for the InferenceConfig class used to deploy the model.

What should the entry script do?

- A. Register the model with appropriate tags and properties.
- B. Create a Conda environment for the web service compute and install the necessary Python packages.
- C. Load the model and use it to predict labels from input data.
- D. Start a node on the inference cluster where the web service is deployed.
- E. Specify the number of cores and the amount of memory required for the inference compute.

**Correct Answer: C**

**Section:**

**Explanation:**

The entry script receives data submitted to a deployed web service and passes it to the model. It then takes the response returned by the model and returns that to the client. The script is specific to your model. It must understand the data that the model expects and returns.

The two things you need to accomplish in your entry script are:

Loading your model (using a function called init())

Running your model on input data (using a function called run())

Reference: <https://docs.microsoft.com/en-us/azure/machine-learning/how-to-deploy-and-where>



#### QUESTION 21

You develop and train a machine learning model to predict fraudulent transactions for a hotel booking website.

Traffic to the site varies considerably. The site experiences heavy traffic on Monday and Friday and much lower traffic on other days. Holidays are also high web traffic days.

You need to deploy the model as an Azure Machine Learning real-time web service endpoint on compute that can dynamically scale up and down to support demand.

Which deployment compute option should you use?

- A. attached Azure Databricks cluster
- B. Azure Container Instance (ACI)
- C. Azure Kubernetes Service (AKS) inference cluster
- D. Azure Machine Learning Compute Instance
- E. attached virtual machine in a different region

**Correct Answer: E**

**Section:**

**Explanation:**

Azure Machine Learning compute cluster is a managed-compute infrastructure that allows you to easily create a single or multi-node compute. The compute is created within your workspace region as a resource that can be shared with other users in your workspace. The compute scales up automatically when a job is submitted, and can be put in an Azure Virtual Network.

Reference: <https://docs.microsoft.com/en-us/azure/machine-learning/how-to-create-attach-compute-sdk> Question Set 1

#### QUESTION 22

HOTSPOT

You are a lead data scientist for a project that tracks the health and migration of birds. You create a multi-image classification deep learning model that uses a set of labeled bird photos collected by experts. You plan to use

the model to develop a cross-platform mobile app that predicts the species of bird captured by app users.

You must test and deploy the trained model as a web service. The deployed model must meet the following requirements:

An authenticated connection must not be required for testing.

The deployed model must perform with low latency during inferencing.

The REST endpoints must be scalable and should have a capacity to handle large number of requests when multiple end users are using the mobile application.

You need to verify that the web service returns predictions in the expected JSON format when a valid REST request is submitted.

Which compute resources should you use? To answer, select the appropriate options in the answer area.

NOTE: Each correct selection is worth one point.

Hot Area:

| Context    | Resource                                                                                                                                                                                            |
|------------|-----------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|
| Test       | <input type="checkbox"/> ds-workstation notebook VM<br><input type="checkbox"/> aks-compute cluster<br><input type="checkbox"/> cpu-compute cluster<br><input type="checkbox"/> gpu-compute cluster |
| Production | <input type="checkbox"/> ds-workstation notebook VM<br><input type="checkbox"/> aks-compute cluster<br><input type="checkbox"/> cpu-compute cluster<br><input type="checkbox"/> gpu-compute cluster |

Answer Area:

| Context    | Resource                                                                                                                                                                                                       |
|------------|----------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|
| Test       | <input checked="" type="checkbox"/> ds-workstation notebook VM<br><input type="checkbox"/> aks-compute cluster<br><input type="checkbox"/> cpu-compute cluster<br><input type="checkbox"/> gpu-compute cluster |
| Production | <input type="checkbox"/> ds-workstation notebook VM<br><input type="checkbox"/> aks-compute cluster<br><input type="checkbox"/> cpu-compute cluster<br><input checked="" type="checkbox"/> gpu-compute cluster |

Section:

Explanation:

Box 1: ds-workstation notebook VM

An authenticated connection must not be required for testing.

On a Microsoft Azure virtual machine (VM), including a Data Science Virtual Machine (DSVM), you create local user accounts while provisioning the VM. Users then authenticate to the VM by using these credentials.



Box 2: gpu-compute cluster

Image classification is well suited for GPU compute clusters

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/data-science-virtual-machine/dsvm-common-identity>

<https://docs.microsoft.com/en-us/azure/architecture/reference-architectures/ai/training-deep-learning>

### QUESTION 23

HOTSPOT

You deploy a model in Azure Container Instance.

You must use the Azure Machine Learning SDK to call the model API.

You need to invoke the deployed model using native SDK classes and methods.

How should you complete the command? To answer, select the appropriate options in the answer areas.

NOTE: Each correct selection is worth one point.

Hot Area:

```
from azureml.core import Workspace
```

```
from azureml.core.webservice import requests
```

```
from azureml.core.webservice import Webservice
```

```
from azureml.core.webservice import LocalWebservice
```

```
import json
```

```
ws = Workspace.from_config()
```

```
service_name = "mlmodel1-service"
```

```
service = Webservice(name=service_name, workspace=ws)
```

```
x_new = [[2,101.5,1,24,21], [1,89.7,4,41,21]]
```

```
input_json = json.dumps({"data": x_new})
```

```
predictions = service.run(input_json)
```

```
predictions = requests.post(service.scoring_uri, input_json)
```

```
predictions = service.deserialize(ws, input_json)
```

Answer Area:

```
from azureml.core import Workspace
```

```
from azureml.core.webservice import requests
```

```
from azureml.core.webservice import Webservice
```

```
from azureml.core.webservice import LocalWebservice
```

```
import json
```

```
ws = Workspace.from_config()
```

```
service_name = "mlmodel1-service"
```

```
service = Webservice(name=service_name, workspace=ws)
```

```
x_new = [[2,101.5,1,24,21], [1,89.7,4,41,21]]
```

```
input_json = json.dumps({"data": x_new})
```

```
predictions = service.run(input_json)
```

```
predictions = requests.post(service.scoring_uri, input_json)
```

```
predictions = service.deserialize(ws, input_json)
```

#### Section:

#### Explanation:

Box 1: from azureml.core.webservice import Webservice

The following code shows how to use the SDK to update the model, environment, and entry script for a web service to Azure Container Instances:

```
from azureml.core import Environment
```

```
from azureml.core.webservice import Webservice
```

```
from azureml.core.model import Model, InferenceConfig
```

Box 2: predictions = service.run(input\_json)

Example: The following code demonstrates sending data to the service:

```
import json
```

```
test_sample = json.dumps({'data': [
```

```
[1, 2, 3, 4, 5, 6, 7, 8, 9, 10],
```

```
[10, 9, 8, 7, 6, 5, 4, 3, 2, 1]
```

```
]])
```

```
test_sample = bytes(test_sample, encoding='utf8')
```

```
prediction = service.run(input_data=test_sample)
```

```
print(prediction)
```

Reference:

<https://docs.microsoft.com/bs-latn-ba/azure/machine-learning/how-to-deploy-azure-container-instance>

<https://docs.microsoft.com/en-us/azure/machine-learning/how-to-troubleshoot-deployment>

#### QUESTION 24

##### HOTSPOT

You create an Azure Machine Learning workspace.

You need to detect data drift between a baseline dataset and a subsequent target dataset by using the DataDriftDetector class.

How should you complete the code segment? To answer, select the appropriate options in the answer area.

NOTE: Each correct selection is worth one point.

#### Hot Area:

### Answer Area

```
from azureml.core import Workspace, Dataset
from datetime import datetime

ws = Workspace.from_config()
dset = Dataset.get_by_name(ws, 'target')
baseline = target.time_before(datetime(2021, 2, 1))
features = ['windAngle', 'windSpeed', 'temperature', 'stationName']

monitor = DataDriftDetector. (ws, 'drift-monitor', baseline,
 backfill
 create_from_datasets
 create_from_model
 target, compute_target='cpu-cluster', frequency='Week', feature_list=None,
 drift_threshold=.6, latency=24)

monitor = DataDriftDetector.get_by_name(ws, 'drift-monitor')
monitor = monitor.update(feature_list=features)
complete = monitor. (datetime(2021, 1, 1), datetime.today())
```

|          |
|----------|
| backfill |
| list     |
| update   |

 **vdumps**

Answer Area:



### Answer Area

```
from azureml.core import Workspace, Dataset
from datetime import datetime

ws = Workspace.from_config()
dset = Dataset.get_by_name(ws, 'target')
baseline = target.time_before(datetime(2021, 2, 1))
features = ['windAngle', 'windSpeed', 'temperature', 'stationName']

monitor = DataDriftDetector. (ws, 'drift-monitor', baseline,
 backfill
 create_from_datasets
 create_from_model
 target, compute_target='cpu-cluster', frequency='Week', feature_list=None,
 drift_threshold=.6, latency=24)

monitor = DataDriftDetector.get_by_name(ws, 'drift-monitor')
monitor = monitor.update(feature_list=features)
complete = monitor. (datetime(2021, 1, 1), datetime.today())
 backfill
 list
 update
```

### Section:

### Explanation:

Box 1: create\_from\_datasets

The create\_from\_datasets method creates a new DataDriftDetector object from a baseline tabular dataset and a target time series dataset.

Box 2: backfill

The backfill method runs a backfill job over a given specified start and end date.

Syntax: backfill(start\_date, end\_date, compute\_target=None, create\_compute\_target=False)

Incorrect Answers:

List and update do not have datetime parameters.

Reference:

[https://docs.microsoft.com/en-us/python/api/azureml-datadrift/azureml.datadrift.datadriftdetector\(class\)](https://docs.microsoft.com/en-us/python/api/azureml-datadrift/azureml.datadrift.datadriftdetector(class))

### QUESTION 25

You are planning to register a trained model in an Azure Machine Learning workspace.

You must store additional metadata about the model in a key-value format. You must be able to add new metadata and modify or delete metadata after creation.

You need to register the model.

Which parameter should you use?

- A. description
- B. model\_framework



- C. tags
- D. properties

**Correct Answer: D**

**Section:**

**Explanation:**

azureml.core.Model.properties:

Dictionary of key value properties for the Model. These properties cannot be changed after registration, however new key value pairs can be added.

Reference:

<https://docs.microsoft.com/en-us/python/api/azureml-core/azureml.core.model.model>

#### QUESTION 26

You have a Python script that executes a pipeline. The script includes the following code:

```
from azureml.core import Experiment pipeline_run = Experiment(ws, 'pipeline_test').submit(pipeline)
```

You want to test the pipeline before deploying the script.

You need to display the pipeline run details written to the STDOUT output when the pipeline completes.

Which code segment should you add to the test script?

- A. `pipeline_run.get.metrics()`
- B. `pipeline_run.wait_for_completion(show_output=True)`
- C. `pipeline_param = PipelineParameter(name="stdout", default_value="console")`
- D. `pipeline_run.get_status()`

**Correct Answer: B**

**Section:**

**Explanation:**

`wait_for_completion`: Wait for the completion of this run. Returns the status object after the wait.

Syntax: `wait_for_completion(show_output=False, wait_post_processing=False, raise_on_error=True)`

Parameter: `show_output`

Indicates whether to show the run output on `sys.stdout`.

#### QUESTION 27

HOTSPOT

You use Azure Machine Learning to train and register a model.

You must deploy the model into production as a real-time web service to an inference cluster named `service-compute` that the IT department has created in the Azure Machine Learning workspace.

Client applications consuming the deployed web service must be authenticated based on their Azure Active Directory service principal.

You need to write a script that uses the Azure Machine Learning SDK to deploy the model. The necessary modules have been imported.

How should you complete the code? To answer, select the appropriate options in the answer area.

NOTE: Each correct selection is worth one point.

**Hot Area:**



## Answer Area

```
Assume the necessary modules have been imported
deploy_target = (ws, "service-compute")

AksCompute
AmlCompute
RemoteCompute
BatchCompute

deployment_config = .deploy_configuration(cpu_cores=1, memory_gb=1,

AksWebservice
AciWebservice
LocalWebService

)

token_auth_enabled=True
token_auth_enabled=False
auth_enabled=True
auth_enabled=False

service = Model.deploy(ws, "ml-service",
 [model], inference_config, deployment_config, deploy_target)
service.wait_for_deployment(show_output = True)
```

Answer Area:

## Answer Area

```
Assume the necessary modules have been imported
deploy_target = (ws, "service-compute")

AksCompute
AmlCompute
RemoteCompute
BatchCompute

deployment_config = .deploy_configuration(cpu_cores=1, memory_gb=1,

AksWebservice
AciWebservice
LocalWebService

)

token_auth_enabled=True
token_auth_enabled=False
auth_enabled=True
auth_enabled=False

service = Model.deploy(ws, "ml-service",
 [model], inference_config, deployment_config, deploy_target)
service.wait_for_deployment(show_output = True)
```

Section:

Explanation:



Box 1: AksCompute

Example:

```
aks_target = AksCompute(ws,"myaks")
If deploying to a cluster configured for dev/test, ensure that it was created with enough
cores and memory to handle this deployment configuration. Note that memory is also used by
things such as dependencies and AML components.
deployment_config = AksWebservice.deploy_configuration(cpu_cores = 1, memory_gb = 1)
service = Model.deploy(ws, "myservice", [model], inference_config, deployment_config, aks_target)
```

Box 2: AksWebservice

Box 3: token\_auth\_enabled=Yes

Whether or not token auth is enabled for the Webservice.

Note: A Service principal defined in Azure Active Directory (Azure AD) can act as a principal on which authentication and authorization policies can be enforced in Azure Databricks.

The Azure Active Directory Authentication Library (ADAL) can be used to programmatically get an Azure AD access token for a user.

Incorrect Answers:

auth\_enabled (bool): Whether or not to enable key auth for this Webservice. Defaults to True.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/how-to-deploy-azure-kubernetes-service>

<https://docs.microsoft.com/en-us/azure/databricks/dev-tools/api/latest/aad/service-prin-aad-token>

## QUESTION 28

DRAG DROP

You use Azure Machine Learning to deploy a model as a real-time web service.

You need to create an entry script for the service that ensures that the model is loaded when the service starts and is used to score new data as it is received.

Which functions should you include in the script? To answer, drag the appropriate functions to the correct actions. Each function may be used once, more than once, or not at all. You may need to drag the split bar between panes or scroll to view content.

NOTE: Each correct selection is worth one point.

Select and Place:

### Answer Area

| Functions                              | Action                                  | Function             |
|----------------------------------------|-----------------------------------------|----------------------|
| <input type="text" value="main()"/>    |                                         |                      |
| <input type="text" value="score()"/>   | Load the model when the service starts. | <input type="text"/> |
| <input type="text" value="run()"/>     | Use the model to score new data.        | <input type="text"/> |
| <input type="text" value="init()"/>    |                                         |                      |
| <input type="text" value="predict()"/> |                                         |                      |

Correct Answer:

| Functions | Action                                  | Function |
|-----------|-----------------------------------------|----------|
| main()    |                                         |          |
| score()   | Load the model when the service starts. | init()   |
|           | Use the model to score new data.        | run()    |
|           |                                         |          |
| predict() |                                         |          |

**Section:**

**Explanation:**

Box 1: init()

The entry script has only two required functions, init() and run(data). These functions are used to initialize the service at startup and run the model using request data passed in by a client. The rest of the script handles loading and running the model(s).

Box 2: run()

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/how-to-deploy-existing-model>



#### QUESTION 29

You use the designer to create a training pipeline for a classification model. The pipeline uses a dataset that includes the features and labels required for model training.

You create a real-time inference pipeline from the training pipeline. You observe that the schema for the generated web service input is based on the dataset and includes the label column that the model predicts. Client applications that use the service must not be required to submit this value.

You need to modify the inference pipeline to meet the requirement.

What should you do?

- A. Add a Select Columns in Dataset module to the inference pipeline after the dataset and use it to select all columns other than the label.
- B. Delete the dataset from the training pipeline and recreate the real-time inference pipeline.
- C. Delete the Web Service Input module from the inference pipeline.
- D. Replace the dataset in the inference pipeline with an Enter Data Manually module that includes data for the feature columns but not the label column.

**Correct Answer: A**

**Section:**

**Explanation:**

By default, the Web Service Input will expect the same data schema as the module output data which connects to the same downstream port as it. You can remove the target variable column in the inference pipeline using Select Columns in Dataset module. Make sure that the output of Select Columns in Dataset removing target variable column is connected to the same port as the output of the Web Service Input module.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/tutorial-designer-automobile-price-deploy>

#### QUESTION 30

You use the Azure Machine Learning designer to create and run a training pipeline. You then create a real-time inference pipeline.

You must deploy the real-time inference pipeline as a web service.



What must you do before you deploy the real-time inference pipeline?

- A. Run the real-time inference pipeline.
- B. Create a batch inference pipeline.
- C. Clone the training pipeline.
- D. Create an Azure Machine Learning compute cluster.

**Correct Answer: D**

**Section:**

**Explanation:**

You need to create an inferencing cluster.

Deploy the real-time endpoint

After your AKS service has finished provisioning, return to the real-time inferencing pipeline to complete deployment.

1. Select Deploy above the canvas.
2. Select Deploy new real-time endpoint.
3. Select the AKS cluster you created.
4. Select Deploy.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/tutorial-designer-automobile-price-deploy>

#### QUESTION 31

You create an Azure Machine Learning workspace named ML-workspace. You also create an Azure Databricks workspace named DB-workspace. DB-workspace contains a cluster named DB-cluster.

You must use DB-cluster to run experiments from notebooks that you import into DB-workspace.

You need to use ML-workspace to track MLflow metrics and artifacts generated by experiments running on DB-cluster. The solution must minimize the need for custom code.

What should you do?

- A. From DB-cluster, configure the Advanced Logging option.
- B. From DB-workspace, configure the Link Azure ML workspace option.
- C. From ML-workspace, create an attached compute.
- D. From ML-workspace, create a compute cluster.

**Correct Answer: B**

**Section:**

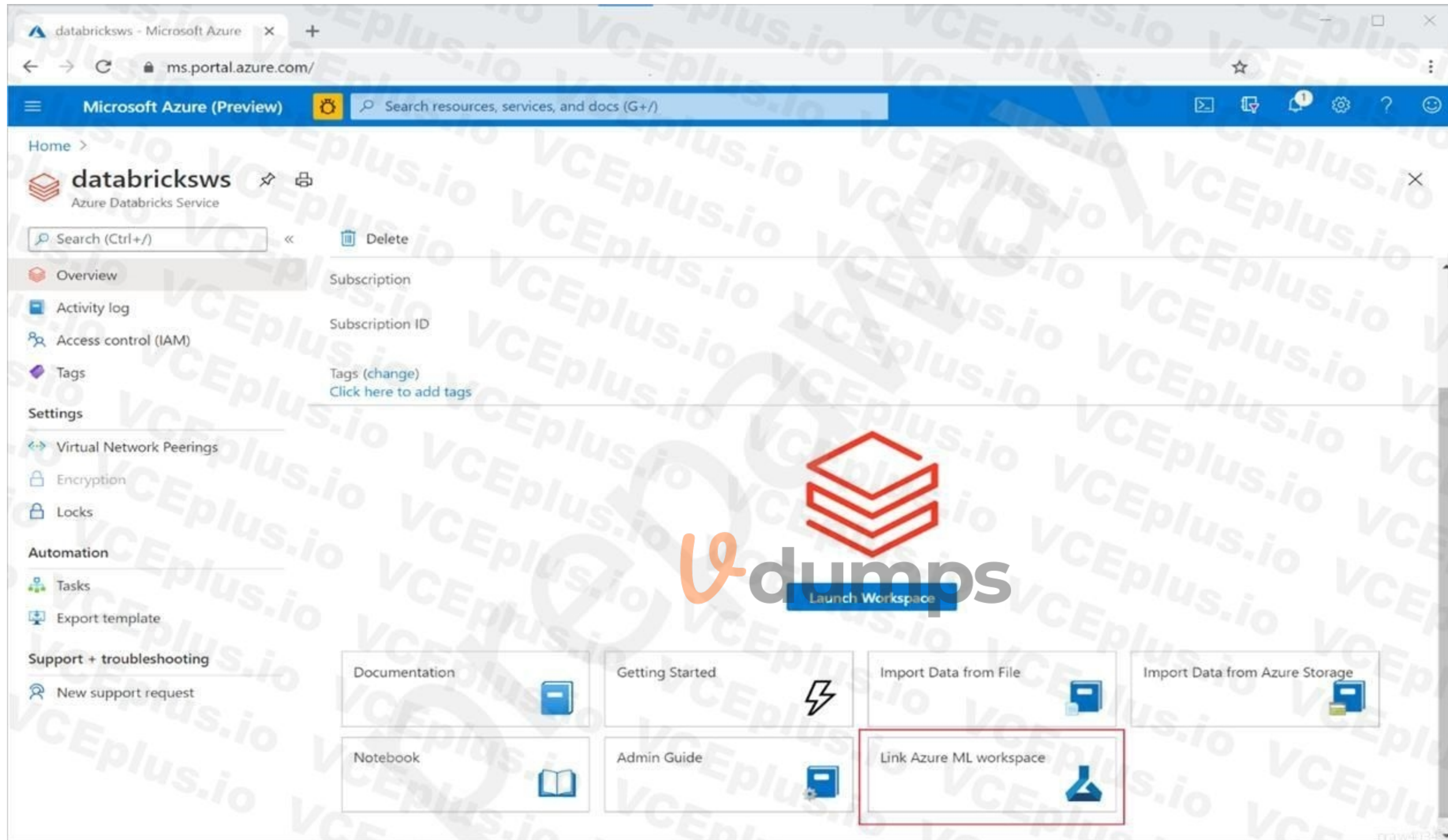
**Explanation:**

Connect your Azure Databricks and Azure Machine Learning workspaces:

Linking your ADB workspace to your Azure Machine Learning workspace enables you to track your experiment data in the Azure Machine Learning workspace.

To link your ADB workspace to a new or existing Azure Machine Learning workspace

1. Sign in to Azure portal.
2. Navigate to your ADB workspace's Overview page.
3. Select the Link Azure Machine Learning workspace button on the bottom right.



Reference:  
<https://docs.microsoft.com/en-us/azure/machine-learning/how-to-use-mlflow-azure-databricks>

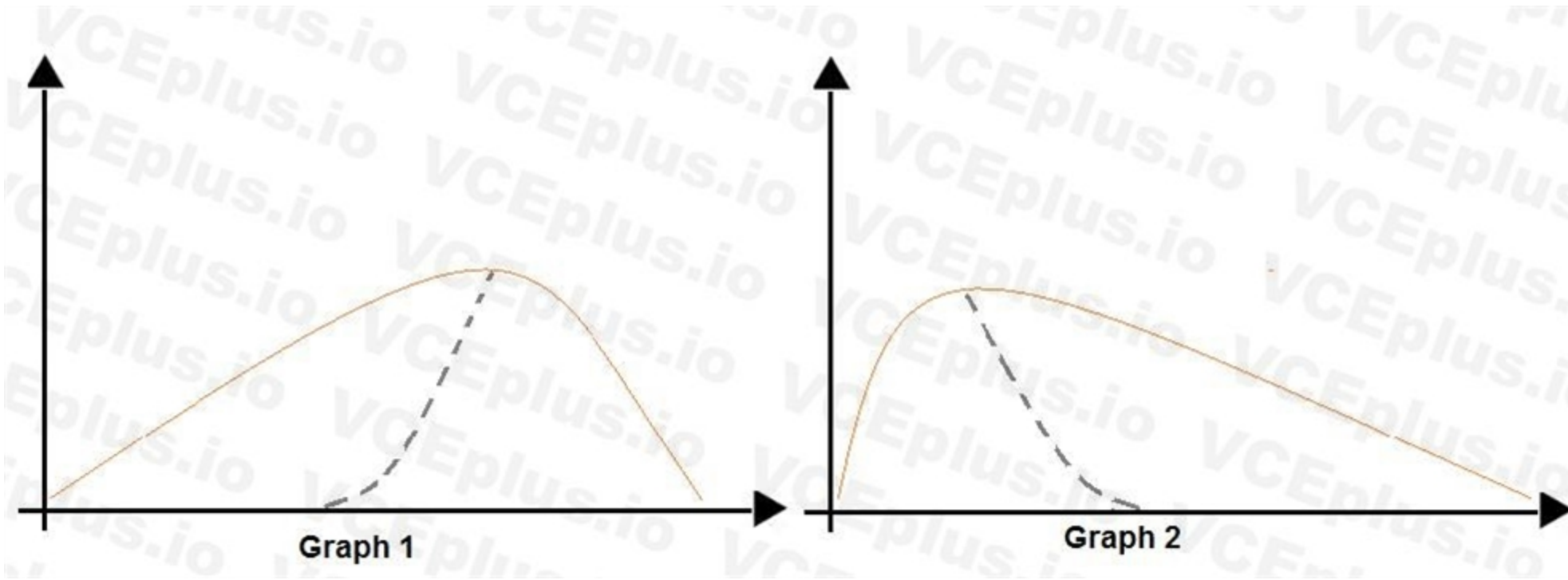
## 01 - Implement Responsible ML

### QUESTION 1

#### HOTSPOT

You are analyzing the asymmetry in a statistical distribution.

The following image contains two density curves that show the probability distribution of two datasets.



Use the drop-down menus to select the answer choice that answers each question based on the information presented in the graphic.  
NOTE: Each correct selection is worth one point.

Hot Area:

**Answer Area**

**Question**

Which type of distribution is shown for the dataset density curve of Graph 1?

**Answer choice**

- Negative skew
- Positive skew
- Normal distribution
- Bimodal distribution

Which type of distribution is shown for the dataset density curve of Graph 2?

- Negative skew
- Positive skew
- Normal distribution
- Bimodal distribution

Answer Area:

| Question                                                                      | Answer choice                                                                                                                                                                                |
|-------------------------------------------------------------------------------|----------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|
| Which type of distribution is shown for the dataset density curve of Graph 1? | <input type="checkbox"/> Negative skew<br><input checked="" type="checkbox"/> Positive skew<br><input type="checkbox"/> Normal distribution<br><input type="checkbox"/> Bimodal distribution |
| Which type of distribution is shown for the dataset density curve of Graph 2? | <input checked="" type="checkbox"/> Negative skew<br><input type="checkbox"/> Positive skew<br><input type="checkbox"/> Normal distribution<br><input type="checkbox"/> Bimodal distribution |

**Section:**

**Explanation:**

Box 1: Positive skew

Positive skew values means the distribution is skewed to the right.

Box 2: Negative skew

Negative skewness values mean the distribution is skewed to the left.

References:

<https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/compute-elementary-statistics>



**QUESTION 2**

**HOTSPOT**

You train a classification model by using a decision tree algorithm.

You create an estimator by running the following Python code. The variable `feature_names` is a list of all feature names, and `class_names` is a list of all class names.

```
from interpret.ext.blackbox import TabularExplainer
explainer = TabularExplainer(model,
 x_train,
 features=feature_names,
 classes=class_names)
```

You need to explain the predictions made by the model for all classes by determining the importance of all features.

For each of the following statements, select Yes if the statement is true. Otherwise, select No.

NOTE: Each correct selection is worth one point.

**Hot Area:**



### Answer Area

Yes

No

The SHAP TreeExplainer will be used to interpret the model.

If you omit the features and classes parameters in the TabularExplainer instantiation, the explainer still works as expected.

You could interpret the model by using a MimicExplainer instead of a TabularExplainer.

Answer Area:

### Answer Area

Yes

No

The SHAP TreeExplainer will be used to interpret the model.

If you omit the features and classes parameters in the TabularExplainer instantiation, the explainer still works as expected.

You could interpret the model by using a MimicExplainer instead of a TabularExplainer.

Section:

Explanation:

Box 1: Yes

TabularExplainer calls one of the three SHAP explainers underneath (TreeExplainer, DeepExplainer, or KernelExplainer).

Box 2: Yes

To make your explanations and visualizations more informative, you can choose to pass in feature names and output class names if doing classification.

Box 3: No TabularExplainer automatically selects the most appropriate one for your use case, but you can call each of its three underlying explainers underneath (TreeExplainer, DeepExplainer, or KernelExplainer) directly.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/how-to-machine-learning-interpretability-aml>

### QUESTION 3

DRAG DROP

You have several machine learning models registered in an Azure Machine Learning workspace.

You must use the Fairlearn dashboard to assess fairness in a selected model.

Which three actions should you perform in sequence? To answer, move the appropriate actions from the list of actions to the answer area and arrange them in the correct order.

Select and Place:

**Actions**

Select a binary classification or regression model.

Select a metric to be measured.

Select a multiclass classification model.

Select a model feature to be evaluated.

Select a clustering model.

**Answer Area**

Select a model feature to be evaluated.

Select a binary classification or regression model.

Select a metric to be measured.

Correct Answer:

**Actions**

Select a multiclass classification model.

Select a clustering model.

**Answer Area**

Select a model feature to be evaluated.

Select a binary classification or regression model.

Select a metric to be measured.



**Section:**

**Explanation:**

Step 1: Select a model feature to be evaluated.

Step 2: Select a binary classification or regression model.

Register your models within Azure Machine Learning. For convenience, store the results in a dictionary, which maps the id of the registered model (a string in name:version format) to the predictor itself.

Example:

```
model_dict = {}
lr_reg_id = register_model("fairness_logistic_regression", lr_predictor)
model_dict[lr_reg_id] = lr_predictor
svm_reg_id = register_model("fairness_svm", svm_predictor)
model_dict[svm_reg_id] = svm_predictor
```

Step 3: Select a metric to be measured

Precompute fairness metrics.

Create a dashboard dictionary using Fairlearn's metrics package.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/how-to-machine-learning-fairness-aml>

**QUESTION 4**

Note: This question-is part of a series of questions that present the same scenario. Each question-in the series contains a unique solution that might meet the stated goals. Some question-sets might have more than one correct solution, while others might not have a correct solution.

After you answer a question-in this section, you will NOT be able to return to it. As a result, these questions will not appear in the review screen.

You are creating a model to predict the price of a student's artwork depending on the following variables: the student's length of education, degree type, and art form.

You start by creating a linear regression model.

You need to evaluate the linear regression model.

Solution: Use the following metrics: Relative Squared Error, Coefficient of Determination, Accuracy, Precision, Recall, F1 score, and AUC.

Does the solution meet the goal?

- A. Yes
- B. No

**Correct Answer: B**

**Section:**

**Explanation:**

Relative Squared Error, Coefficient of Determination are good metrics to evaluate the linear regression model, but the others are metrics for classification models.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/evaluate-model>

#### QUESTION 5

You are a data scientist creating a linear regression model.

You need to determine how closely the data fits the regression line.

Which metric should you review?

- A. Root Mean Square Error
- B. Coefficient of determination
- C. Recall
- D. Precision
- E. Mean absolute error

**Correct Answer: B**

**Section:**

**Explanation:**

Coefficient of determination, often referred to as R<sup>2</sup>, represents the predictive power of the model as a value between 0 and 1. Zero means the model is random (explains nothing); 1 means there is a perfect fit. However, caution should be used in interpreting R<sup>2</sup> values, as low values can be entirely normal and high values can be suspect.

Incorrect Answers:

A: Root mean squared error (RMSE) creates a single value that summarizes the error in the model. By squaring the difference, the metric disregards the difference between over-prediction and under-prediction.

C: Recall is the fraction of all correct results returned by the model.

D: Precision is the proportion of true results over all positive results.

E: Mean absolute error (MAE) measures how close the predictions are to the actual outcomes; thus, a lower score is better.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/evaluate-model>

#### QUESTION 6

You are creating a binary classification by using a two-class logistic regression model.

You need to evaluate the model results for imbalance.

Which evaluation metric should you use?

- A. Relative Absolute Error
- B. AUC Curve
- C. Mean Absolute Error
- D. Relative Squared Error
- E. Accuracy



F. Root Mean Square Error

**Correct Answer: B**

**Section:**

**Explanation:**

One can inspect the true positive rate vs. the false positive rate in the Receiver Operating Characteristic (ROC) curve and the corresponding Area Under the Curve (AUC) value. The closer this curve is to the upper left corner; the better the classifier's performance is (that is maximizing the true positive rate while minimizing the false positive rate). Curves that are close to the diagonal of the plot, result from classifiers that tend to make predictions that are close to random guessing.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/studio/evaluate-model-performance#evaluating-a-binary-classification-model>

**QUESTION 7**

You are a data scientist working for a bank and have used Azure ML to train and register a machine learning model that predicts whether a customer is likely to repay a loan.

You want to understand how your model is making selections and must be sure that the model does not violate government regulations such as denying loans based on where an applicant lives.

You need to determine the extent to which each feature in the customer data is influencing predictions.

What should you do?

- A. Enable data drift monitoring for the model and its training dataset.
- B. Score the model against some test data with known label values and use the results to calculate a confusion matrix.
- C. Use the Hyperdrive library to test the model with multiple hyperparameter values.
- D. Use the interpretability package to generate an explainer for the model.
- E. Add tags to the model registration indicating the names of the features in the training dataset.

**Correct Answer: D**

**Section:**

**Explanation:**

When you compute model explanations and visualize them, you're not limited to an existing model explanation for an automated ML model. You can also get an explanation for your model with different test data. The steps in this section show you how to compute and visualize engineered feature importance based on your test data.

Incorrect Answers:

A: In the context of machine learning, data drift is the change in model input data that leads to model performance degradation. It is one of the top reasons where model accuracy degrades over time, thus monitoring data drift helps detect model performance issues.

B: A confusion matrix is used to describe the performance of a classification model. Each row displays the instances of the true, or actual class in your dataset, and each column represents the instances of the class that was predicted by the model.

C: Hyperparameters are adjustable parameters you choose for model training that guide the training process. The HyperDrive package helps you automate choosing these parameters.

Reference: <https://docs.microsoft.com/en-us/azure/machine-learning/how-to-machine-learning-interpretability-automl>

**QUESTION 8**

Note: This question-is part of a series of questions that present the same scenario. Each question-in the series contains a unique solution that might meet the stated goals. Some question-sets might have more than one correct solution, while others might not have a correct solution.

After you answer a question-in this section, you will NOT be able to return to it. As a result, these questions will not appear in the review screen.

You train a classification model by using a logistic regression algorithm.

You must be able to explain the model's predictions by calculating the importance of each feature, both as an overall global relative importance value and as a measure of local importance for a specific set of predictions.

You need to create an explainer that you can use to retrieve the required global and local feature importance values.

Solution: Create a MimicExplainer.

Does the solution meet the goal?

- A. Yes
- B. No





**Correct Answer: B**

**Section:**

**Explanation:**

Instead use Permutation Feature Importance Explainer (PFI).

Note 1: Mimic explainer is based on the idea of training global surrogate models to mimic blackbox models. A global surrogate model is an intrinsically interpretable model that is trained to approximate the predictions of any black box model as accurately as possible. Data scientists can interpret the surrogate model to draw conclusions about the black box model.

Note 2: Permutation Feature Importance Explainer (PFI): Permutation Feature Importance is a technique used to explain classification and regression models. At a high level, the way it works is by randomly shuffling data one feature at a time for the entire dataset and calculating how much the performance metric of interest changes. The larger the change, the more important that feature is. PFI can explain the overall behavior of any underlying model but does not explain individual predictions.

Reference: <https://docs.microsoft.com/en-us/azure/machine-learning/how-to-machine-learning-interpretability>

#### **QUESTION 9**

Note: This question-is part of a series of questions that present the same scenario. Each question-in the series contains a unique solution that might meet the stated goals. Some question-sets might have more than one correct solution, while others might not have a correct solution.

After you answer a question-in this section, you will NOT be able to return to it. As a result, these questions will not appear in the review screen.

You train a classification model by using a logistic regression algorithm.

You must be able to explain the model's predictions by calculating the importance of each feature, both as an overall global relative importance value and as a measure of local importance for a specific set of predictions.

You need to create an explainer that you can use to retrieve the required global and local feature importance values.

Solution: Create a TabularExplainer.

Does the solution meet the goal?

A. Yes

B. No

**Correct Answer: B**

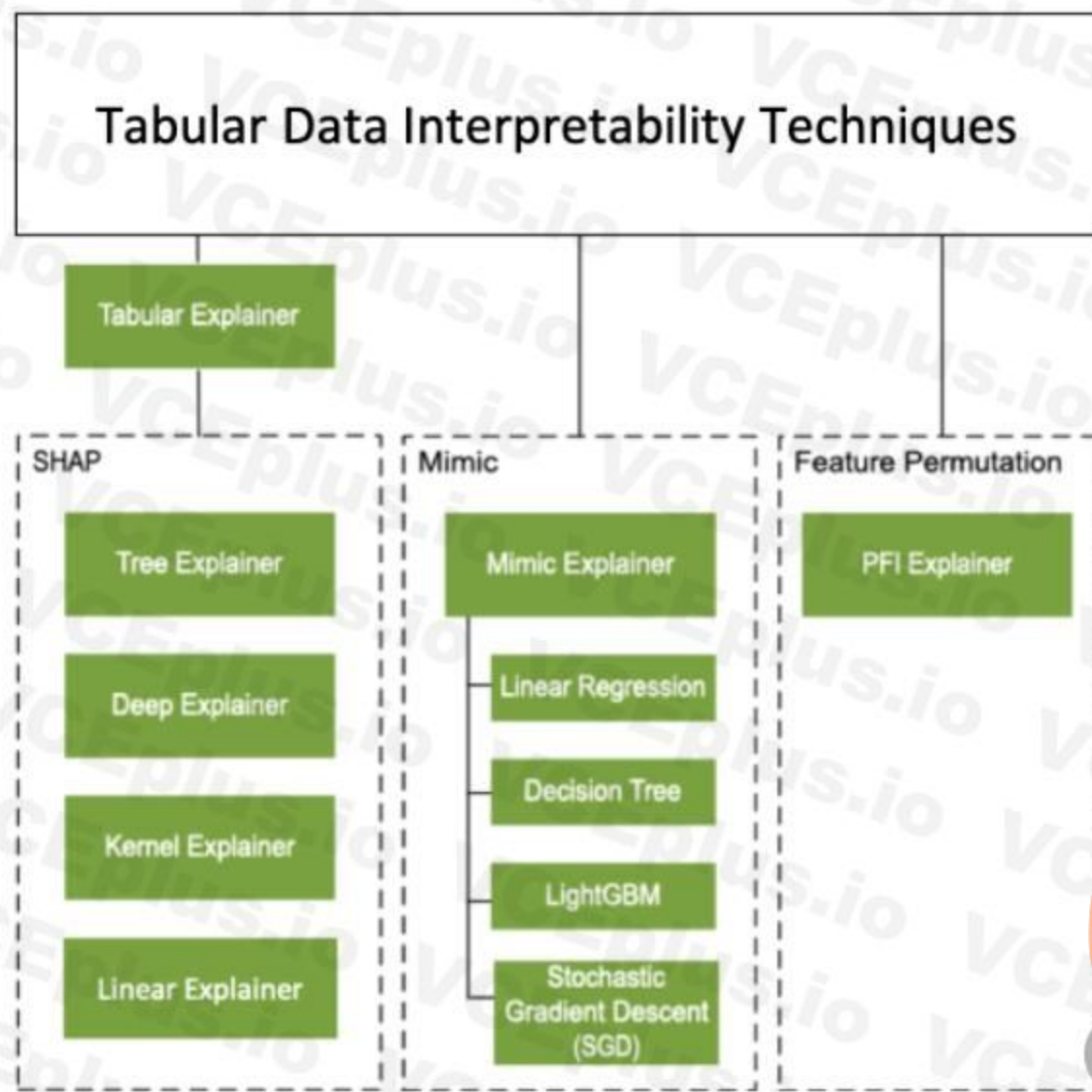
**Section:**

**Explanation:**

Instead use Permutation Feature Importance Explainer (PFI).

Note 1:





Note 2: Permutation Feature Importance Explainer (PFI): Permutation Feature Importance is a technique used to explain classification and regression models. At a high level, the way it works is by randomly shuffling data one feature at a time for the entire dataset and calculating how much the performance metric of interest changes. The larger the change, the more important that feature is. PFI can explain the overall behavior of any underlying model but does not explain individual predictions.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/how-to-machine-learning-interpretability>

#### QUESTION 10

Note: This question-is part of a series of questions that present the same scenario. Each question-in the series contains a unique solution that might meet the stated goals. Some question-sets might have more than one correct solution, while others might not have a correct solution.

After you answer a question-in this section, you will NOT be able to return to it. As a result, these questions will not appear in the review screen.

You train a classification model by using a logistic regression algorithm.

You must be able to explain the model's predictions by calculating the importance of each feature, both as an overall global relative importance value and as a measure of local importance for a specific set of predictions.

You need to create an explainer that you can use to retrieve the required global and local feature importance values.

Solution: Create a PFIExplainer.

Does the solution meet the goal?

A. Yes

B. No

**Correct Answer: A**

**Section:**

**Explanation:**

Permutation Feature Importance Explainer (PFI): Permutation Feature Importance is a technique used to explain classification and regression models. At a high level, the way it works is by randomly shuffling data one feature at a time for the entire dataset and calculating how much the performance metric of interest changes. The larger the change, the more important that feature is. PFI can explain the overall behavior of any underlying model but does not explain individual predictions.

Reference: <https://docs.microsoft.com/en-us/azure/machine-learning/how-to-machine-learning-interpretability>

**QUESTION 11**

You are determining if two sets of data are significantly different from one another by using Azure Machine Learning Studio.

Estimated values in one set of data may be more than or less than reference values in the other set of data. You must produce a distribution that has a constant Type I error as a function of the correlation.

You need to produce the distribution.

Which type of distribution should you produce?

- A. Unpaired t-test with a two-tail option
- B. Unpaired t-test with a one-tail option
- C. Paired t-test with a one-tail option
- D. Paired t-test with a two-tail option

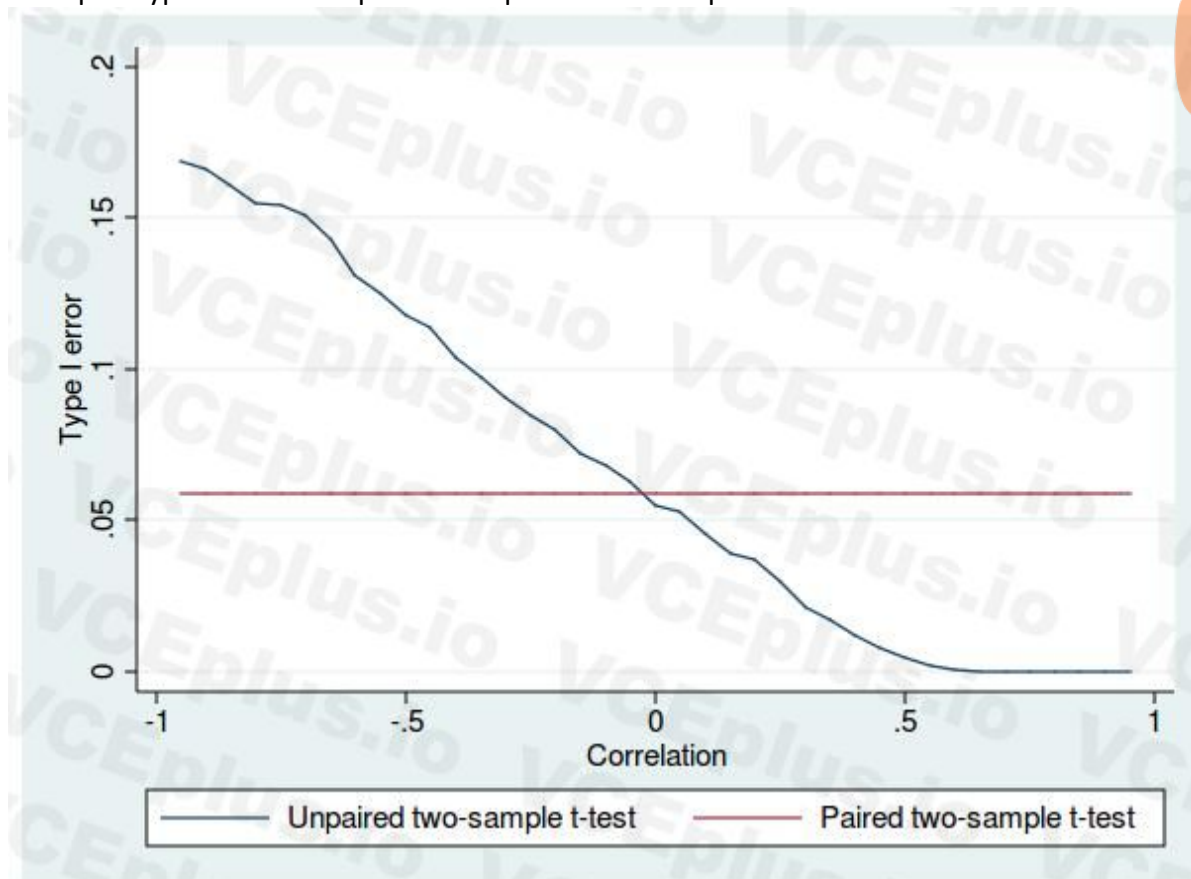
**Correct Answer: D**

**Section:**

**Explanation:**

Choose a one-tail or two-tail test. The default is a two-tailed test. This is the most common type of test, in which the expected distribution is symmetric around zero.

Example: Type I error of unpaired and paired two-sample t-tests as a function of the correlation. The simulated random numbers originate from a bivariate normal distribution with a variance of 1.



Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/test-hypothesis-using-t-test>

[https://en.wikipedia.org/wiki/Student%27s\\_t-test](https://en.wikipedia.org/wiki/Student%27s_t-test)

### QUESTION 12

You are performing feature engineering on a dataset.

You must add a feature named CityName and populate the column value with the text London.

You need to add the new feature to the dataset.

Which Azure Machine Learning Studio module should you use?

- A. Extract N-Gram Features from Text
- B. Edit Metadata
- C. Preprocess Text
- D. Apply SQL Transformation

**Correct Answer: B**

**Section:**

**Explanation:**

Typical metadata changes might include marking columns as features.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/edit-metadata>

### QUESTION 13

Note: This question-is part of a series of questions that present the same scenario. Each question-in the series contains a unique solution that might meet the stated goals. Some question-sets might have more than one correct solution, while others might not have a correct solution.

After you answer a question-in this section, you will NOT be able to return to it. As a result, these questions will not appear in the review screen.

You are creating a model to predict the price of a student's artwork depending on the following variables: the student's length of education, degree type, and art form.

You start by creating a linear regression model.

You need to evaluate the linear regression model.

Solution: Use the following metrics: Mean Absolute Error, Root Mean Absolute Error, Relative Absolute Error, Relative Squared Error, and the Coefficient of Determination.

Does the solution meet the goal?

- A. Yes
- B. No

**Correct Answer: A**

**Section:**

**Explanation:**

The following metrics are reported for evaluating regression models. When you compare models, they are ranked by the metric you select for evaluation.

Mean absolute error (MAE) measures how close the predictions are to the actual outcomes; thus, a lower score is better.

Root mean squared error (RMSE) creates a single value that summarizes the error in the model. By squaring the difference, the metric disregards the difference between over-prediction and under-prediction.

Relative absolute error (RAE) is the relative absolute difference between expected and actual values; relative because the mean difference is divided by the arithmetic mean.

Relative squared error (RSE) similarly normalizes the total squared error of the predicted values by dividing by the total squared error of the actual values.

Mean Zero One Error (MZOE) indicates whether the prediction was correct or not. In other words:  $\text{ZeroOneLoss}(x,y) = 1$  when  $x \neq y$ ; otherwise 0.

Coefficient of determination, often referred to as R<sup>2</sup>, represents the predictive power of the model as a value between 0 and 1. Zero means the model is random (explains nothing); 1 means there is a perfect fit. However, caution should be used in interpreting R<sup>2</sup> values, as low values can be entirely normal and high values can be suspect.

AUC.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/evaluate-model>

### QUESTION 14

Note: This question-is part of a series of questions that present the same scenario. Each question-in the series contains a unique solution that might meet the stated goals. Some question-sets might have more than one correct solution, while others might not have a correct solution.



After you answer a question-in this section, you will NOT be able to return to it. As a result, these questions will not appear in the review screen.

You are creating a model to predict the price of a student's artwork depending on the following variables: the student's length of education, degree type, and art form.

You start by creating a linear regression model.

You need to evaluate the linear regression model.

Solution: Use the following metrics: Accuracy, Precision, Recall, F1 score, and AUC.

Does the solution meet the goal?

- A. Yes
- B. No

**Correct Answer: B**

**Section:**

**Explanation:**

Those are metrics for evaluating classification models, instead use: Mean Absolute Error, Root Mean Absolute Error, Relative Absolute Error, Relative Squared Error, and the Coefficient of Determination.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/evaluate-model>

#### QUESTION 15

You are a data scientist building a deep convolutional neural network (CNN) for image classification.

The CNN model you build shows signs of overfitting.

You need to reduce overfitting and converge the model to an optimal fit.

Which two actions should you perform? Each correct answer presents a complete solution.

NOTE: Each correct selection is worth one point.

- A. Add an additional dense layer with 512 input units.
- B. Add L1/L2 regularization.
- C. Use training data augmentation.
- D. Reduce the amount of training data.
- E. Add an additional dense layer with 64 input units.



**Correct Answer: B, D**

**Section:**

**Explanation:**

B: Weight regularization provides an approach to reduce the overfitting of a deep learning neural network model on the training data and improve the performance of the model on new data, such as the holdout test set.

Keras provides a weight regularization API that allows you to add a penalty for weight size to the loss function.

Three different regularizer instances are provided; they are:

L1: Sum of the absolute weights.

L2: Sum of the squared weights.

L1L2: Sum of the absolute and the squared weights.

D: Because a fully connected layer occupies most of the parameters, it is prone to overfitting. One method to reduce overfitting is dropout. At each training stage, individual nodes are either "dropped out" of the net with probability  $1-p$  or kept with probability  $p$ , so that a reduced network is left; incoming and outgoing edges to a dropped-out node are also removed.

By avoiding training all nodes on all training data, dropout decreases overfitting.

Reference:

<https://machinelearningmastery.com/how-to-reduce-overfitting-in-deep-learning-with-weight-regularization/>

[https://en.wikipedia.org/wiki/Convolutional\\_neural\\_network](https://en.wikipedia.org/wiki/Convolutional_neural_network)

#### QUESTION 16

Note: This question-is part of a series of questions that present the same scenario. Each question-in the series contains a unique solution that might meet the stated goals. Some question-sets might have more than one correct solution, while others might not have a correct solution.

After you answer a question-in this section, you will NOT be able to return to it. As a result, these questions will not appear in the review screen.

You are creating a model to predict the price of a student's artwork depending on the following variables: the student's length of education, degree type, and art form.

You start by creating a linear regression model.

You need to evaluate the linear regression model.

Solution: Use the following metrics: Mean Absolute Error, Root Mean Absolute Error, Relative Absolute Error, Accuracy, Precision, Recall, F1 score, and AUC.

Does the solution meet the goal?

- A. Yes
- B. No

**Correct Answer: B**

**Section:**

**Explanation:**

Accuracy, Precision, Recall, F1 score, and AUC are metrics for evaluating classification models.

Note: Mean Absolute Error, Root Mean Absolute Error, Relative Absolute Error are OK for the linear regression model.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/evaluate-model>

#### QUESTION 17

You are building a binary classification model by using a supplied training set.

The training set is imbalanced between two classes.

You need to resolve the data imbalance.

What are three possible ways to achieve this goal? Each correct answer presents a complete solution.

NOTE: Each correct selection is worth one point.

- A. Penalize the classification
- B. Resample the dataset using undersampling or oversampling
- C. Normalize the training feature set
- D. Generate synthetic samples in the minority class
- E. Use accuracy as the evaluation metric of the model

**Correct Answer: A, B, D**

**Section:**

**Explanation:**

A: Try Penalized Models

You can use the same algorithms but give them a different perspective on the problem.

Penalized classification imposes an additional cost on the model for making classification mistakes on the minority class during training. These penalties can bias the model to pay more attention to the minority class.

B: You can change the dataset that you use to build your predictive model to have more balanced data.

This change is called sampling your dataset and there are two main methods that you can use to even-up the classes:

Consider testing under-sampling when you have an a lot data (tens- or hundreds of thousands of instances or more)

Consider testing over-sampling when you don't have a lot of data (tens of thousands of records or less)

D: Try Generate Synthetic Samples

A simple way to generate synthetic samples is to randomly sample the attributes from instances in the minority class.

Reference:

<https://machinelearningmastery.com/tactics-to-combat-imbalanced-classes-in-your-machine-learning-dataset/>

#### QUESTION 18

HOTSPOT

You write code to retrieve an experiment that is run from your Azure Machine Learning workspace.



Hot Area:

```
Assume required modules are imported
```

```
ws = Workspace.from_config()
```

```
feature_importances = explanation.
```

|                            |   |
|----------------------------|---|
|                            | ▼ |
| from_run                   |   |
| list_model_explanations    |   |
| from_run_id                |   |
| download_model_explanation |   |

```
(workspace = ws,
 experiment_name='train_and_explain',
 run_id='train_and_explain_12345')
```

```
explanation = client.
```

|                            |   |
|----------------------------|---|
|                            | ▼ |
| upload_model_explanation   |   |
| list_model_explanations    |   |
| run                        |   |
| download_model_explanation |   |

```
()
```

```
feature_importances = explanation.
```

|                            |   |
|----------------------------|---|
|                            | ▼ |
| explanation                |   |
| explanation_client         |   |
| get_feature_important_dict |   |
| download_model_explanation |   |

```
()
```

```
for key, value in feature_importances.items():
 print(key, "\t", value)
```



Answer Area:

```
Assume required modules are imported
```

```
ws = Workspace.from_config()
```

```
feature_importances = explanation.
```

|                            |
|----------------------------|
| from_run                   |
| list_model_explanations    |
| from_run_id                |
| download_model_explanation |

```
(workspace = ws,
 experiment_name='train_and_explain',
 run_id='train_and_explain_12345')
```

```
explanation = client.
```

|                            |
|----------------------------|
| upload_model_explanation   |
| list_model_explanations    |
| run                        |
| download_model_explanation |

```
()
```

```
feature_importances = explanation.
```

|                            |
|----------------------------|
| explanation                |
| explanation_client         |
| get_feature_important_dict |
| download_model_explanation |

```
()
```

```
for key, value in feature_importances.items():
 print(key, "\t", value)
```



**Section:**

**Explanation:**

Business managers in your organization want to see the importance of the features in the model.

You need to print out the model features and their relative importance in an output that looks similar to the following.

| Feature | Importance          |
|---------|---------------------|
| 0       | 1.5627435610083558  |
| 2       | 0.6077689312583112  |
| 4       | 0.5574002432900718  |
| 3       | 0.42858759955671777 |
| 1       | 0.3501361539771977  |

How should you complete the code? To answer, select the appropriate options in the answer area.

NOTE: Each correct selection is worth one point.

**Explanation:**

Box 1: from\_run\_id from\_run\_id(workspace, experiment\_name, run\_id)

Create the client with factory method given a run ID.

Returns an instance of the ExplanationClient.

Parameters

Workspace Workspace An object that represents a workspace.

experiment\_name str The name of an experiment.

run\_id str A GUID that represents a run.

Box 2: list\_model\_explanations



list\_model\_explanations returns a dictionary of metadata for all model explanations available.

Returns

A dictionary of explanation metadata such as id, data type, explanation method, model type, and upload time, sorted by upload time

Box 3: explanation

Reference:

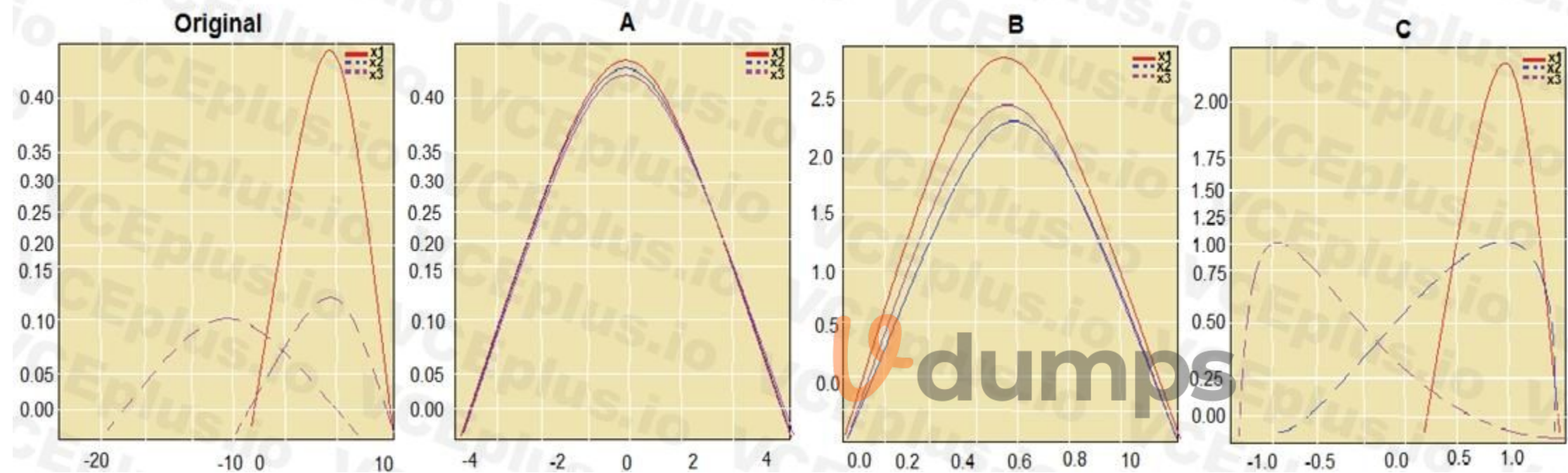
[https://docs.microsoft.com/en-us/python/api/azureml-contrib-interpret/azureml.contrib.interpret.explanation.explanation\\_client.explanationclient?view=azure-ml-py](https://docs.microsoft.com/en-us/python/api/azureml-contrib-interpret/azureml.contrib.interpret.explanation.explanation_client.explanationclient?view=azure-ml-py)

### QUESTION 19

HOTSPOT

You are performing feature scaling by using the scikit-learn Python library for x1, x2, and x3 features.

Original and scaled data is shown in the following image.



Use the drop-down menus to select the answer choice that answers each question based on the information presented in the graphic.

NOTE: Each correct selection is worth one point.

Hot Area:

**Answer Area**

| Question                         | Answer choice                                                            |
|----------------------------------|--------------------------------------------------------------------------|
| Which scaler is used in graph A? | <input type="text"/> ▼<br>Standard Scaler<br>Min Max Scale<br>Normalizer |
| Which scaler is used in graph B? | <input type="text"/> ▼<br>Standard Scaler<br>Min Max Scale<br>Normalizer |
| Which scaler is used in graph C? | <input type="text"/> ▼<br>Standard Scaler<br>Min Max Scale<br>Normalizer |

Answer Area:

**Answer Area**

| Question                         | Answer choice                                                                                                            |
|----------------------------------|--------------------------------------------------------------------------------------------------------------------------|
| Which scaler is used in graph A? | <input type="text"/> ▼<br><span style="background-color: #e0ffe0;">Standard Scaler</span><br>Min Max Scale<br>Normalizer |
| Which scaler is used in graph B? | <input type="text"/> ▼<br>Standard Scaler<br><span style="background-color: #e0ffe0;">Min Max Scale</span><br>Normalizer |
| Which scaler is used in graph C? | <input type="text"/> ▼<br>Standard Scaler<br>Min Max Scale<br><span style="background-color: #e0ffe0;">Normalizer</span> |

Section:

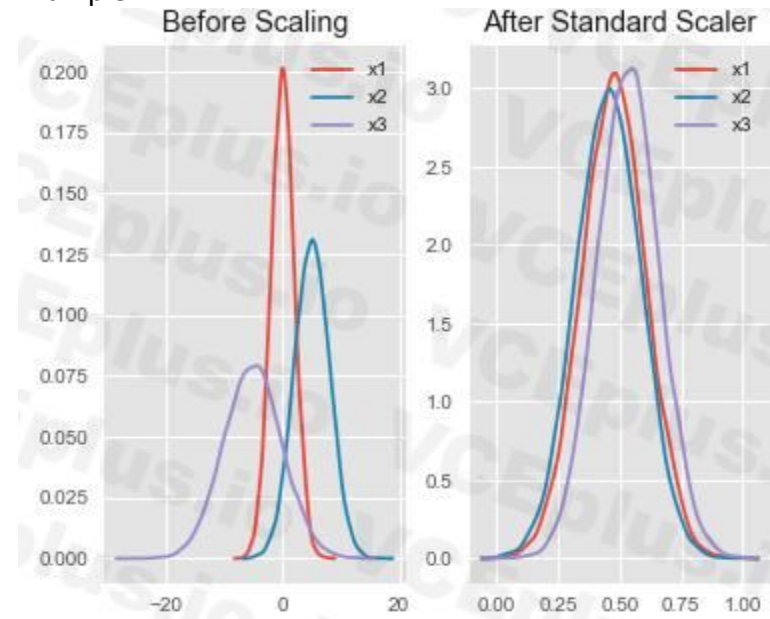


**Explanation:**

**Box 1: StandardScaler**

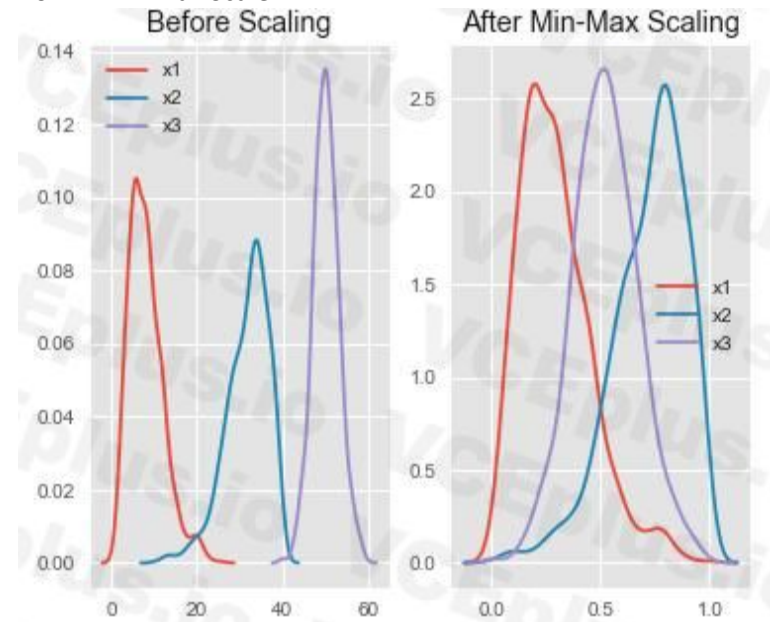
The StandardScaler assumes your data is normally distributed within each feature and will scale them such that the distribution is now centred around 0, with a standard deviation of 1.

Example:



All features are now on the same scale relative to one another.

**Box 2: Min Max Scaler**



Notice that the skewness of the distribution is maintained but the 3 distributions are brought into the same scale so that they overlap.

**Box 3: Normalizer**

References:

<http://benalexkeen.com/feature-scaling-with-scikit-learn/>

**QUESTION 20**

DRAG DROP

You are producing a multiple linear regression model in Azure Machine Learning Studio.

Several independent variables are highly correlated.

You need to select appropriate methods for conducting effective feature engineering on all the data.

Which three actions should you perform in sequence? To answer, move the appropriate actions from the list of actions to the answer area and arrange them in the correct order.





Select and Place:

| Action                                        | Answer area |
|-----------------------------------------------|-------------|
| Evaluate the probability function             |             |
| Remove duplicate rows                         |             |
| Use the Filter Based Feature Selection module |             |
| Test the hypothesis using t-Test              |             |
| Compute linear correlation                    |             |
| Build a counting transform                    |             |

Correct Answer:

| Action                            | Answer area                                   |
|-----------------------------------|-----------------------------------------------|
| Evaluate the probability function | Use the Filter Based Feature Selection module |
| Remove duplicate rows             | Build a counting transform                    |
|                                   | Test the hypothesis using t-Test              |
|                                   |                                               |
| Compute linear correlation        |                                               |
|                                   |                                               |

Section:

Explanation:

Step 1: Use the Filter Based Feature Selection module

Filter Based Feature Selection identifies the features in a dataset with the greatest predictive power.

The module outputs a dataset that contains the best feature columns, as ranked by predictive power. It also outputs the names of the features and their scores from the selected metric.

Step 2: Build a counting transform

A counting transform creates a transformation that turns count tables into features, so that you can apply the transformation to multiple datasets.

Step 3: Test the hypothesis using t-Test



References:

<https://docs.microsoft.com/bs-latn-ba/azure/machine-learning/studio-module-reference/filter-based-feature-selection>

<https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/build-counting-transform>

**QUESTION 21**

HOTSPOT

You are developing a linear regression model in Azure Machine Learning Studio. You run an experiment to compare different algorithms.


The following image displays the results dataset output:

| Algorithm             | Mean Absolute Error | Root Mean Squared Error | Relative Absolute Error | Relative Squared Error |
|-----------------------|---------------------|-------------------------|-------------------------|------------------------|
| Bayesian Liner        | 3.276025            | 4.655442                | 0.511436                | 0.282138               |
| Neural Network        | 2.676538            | 3.621476                | 0.417847                | 0.17073                |
| Boosted Decision Tree | 2.168847            | 2.878077                | 0.338589                | 0.107831               |
| Linear                | 6.350005            | 8.720718                | 0.99133                 | 0.99002                |
| Decision Forest       | 2.390206            | 3.315164                | 0.373146                | 0.14307                |

Use the drop-down menus to select the answer choice that answers each question based on the information presented in the image.

NOTE: Each correct selection is worth one point.

**Hot Area:**

 **Answer Area**

Which algorithm minimizes differences between actual and predicted values?

**Linear Regression**

Bayesian Linear Regression  
Neutral Network Regression  
Boosted Decision Tree Regression  
Linear Regression  
Decision Forest Regression

Which approach should you use to find the best parameters for a Linear Regression model for the Online Gradient Descent method?

**Set the Create trainer mode option to Parameter Range.**

Set the Decrease learning rate option to True.  
Set the Decrease learning rate option to False.  
Increase the number of epochs.  
Decrease the number of epochs.

Answer Area:



**CEplus**  
Answer Area

Which algorithm minimizes differences between actual and predicted values?

Bayesian Linear Regression  
 Neutral Network Regression  
 Boosted Decision Tree Regression  
 Linear Regression  
 Decision Forest Regression

Which approach should you use to find the best parameters for a Linear Regression model for the Online Gradient Descent method?

Set the Decrease learning rate option to True.  
 Set the Decrease learning rate option to False.  
 Set the Create trainer mode option to Parameter Range.  
 Increase the number of epochs.  
 Decrease the number of epochs.

**Section:**

**Explanation:**

Box 1: Boosted Decision Tree Regression

Mean absolute error (MAE) measures how close the predictions are to the actual outcomes; thus, a lower score is better.

Box 2:

Online Gradient Descent: If you want the algorithm to find the best parameters for you, set Create trainer mode option to Parameter Range. You can then specify multiple values for the algorithm to try.

References:

<https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/evaluate-model>

<https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/linear-regression>

**QUESTION 22**

**HOTSPOT**

You are using a decision tree algorithm. You have trained a model that generalizes well at a tree depth equal to 10.

You need to select the bias and variance properties of the model with varying tree depth values.

Which properties should you select for each tree depth? To answer, select the appropriate options in the answer area.

**Hot Area:**

**Answer Area**

| Tree Depth | Bias                                                                                                | Variance                                                                                            |
|------------|-----------------------------------------------------------------------------------------------------|-----------------------------------------------------------------------------------------------------|
| 5          | <input type="checkbox"/> High<br><input type="checkbox"/> Low<br><input type="checkbox"/> Identical | <input type="checkbox"/> High<br><input type="checkbox"/> Low<br><input type="checkbox"/> Identical |
| 15         | <input type="checkbox"/> High<br><input type="checkbox"/> Low<br><input type="checkbox"/> Identical | <input type="checkbox"/> High<br><input type="checkbox"/> Low<br><input type="checkbox"/> Identical |

Answer Area:

**Answer Area**

| Tree Depth | Bias                                                                                                           | Variance                                                                                                       |
|------------|----------------------------------------------------------------------------------------------------------------|----------------------------------------------------------------------------------------------------------------|
| 5          | <input checked="" type="checkbox"/> High<br><input type="checkbox"/> Low<br><input type="checkbox"/> Identical | <input type="checkbox"/> High<br><input checked="" type="checkbox"/> Low<br><input type="checkbox"/> Identical |
| 15         | <input type="checkbox"/> High<br><input checked="" type="checkbox"/> Low<br><input type="checkbox"/> Identical | <input checked="" type="checkbox"/> High<br><input type="checkbox"/> Low<br><input type="checkbox"/> Identical |

**vdumps**

**Section:**

**Explanation:**

In decision trees, the depth of the tree determines the variance. A complicated decision tree (e.g. deep) has low bias and high variance.

Note: In statistics and machine learning, the bias–variance tradeoff is the property of a set of predictive models whereby models with a lower bias in parameter estimation have a higher variance of the parameter estimates across samples, and vice versa. Increasing the bias will decrease the variance. Increasing the variance will decrease the bias.

References:

<https://machinelearningmastery.com/gentle-introduction-to-the-bias-variance-trade-off-in-machine-learning/>

**QUESTION 23**

DRAG DROP

You have a model with a large difference between the training and validation error values.

You must create a new model and perform cross-validation.

You need to identify a parameter set for the new model using Azure Machine Learning Studio.

Which module you should use for each step? To answer, drag the appropriate modules to the correct steps. Each module may be used once or more than once, or not at all. You may need to drag the split bar between panes or scroll to view content.



NOTE: Each correct selection is worth one point.

Select and Place:

**Answer Area**

| Modules                         | Step                                 | Module |
|---------------------------------|--------------------------------------|--------|
| Two-Class Boosted Decision Tree | Define the parameter scope           |        |
| Partition and Sample            | Define the cross-validation settings |        |
| Tune Model Hyperparameters      | Define the metric                    |        |
| Split Data                      | Train, evaluate, and compare         |        |

Correct Answer:

**Answer Area**

| Modules | Step                                 | Module                          |
|---------|--------------------------------------|---------------------------------|
|         | Define the parameter scope           | Split Data                      |
|         | Define the cross-validation settings | Partition and Sample            |
|         | Define the metric                    | Two-Class Boosted Decision Tree |
|         | Train, evaluate, and compare         | Tune Model Hyperparameters      |

Section:

**Explanation:**

Box 1: Split data

Box 2: Partition and Sample

Box 3: Two-Class Boosted Decision Tree

Box 4: Tune Model Hyperparameters Integrated train and tune: You configure a set of parameters to use, and then let the module iterate over multiple combinations, measuring accuracy until it finds a "best" model. With most learner modules, you can choose which parameters should be changed during the training process, and which should remain fixed.

We recommend that you use Cross-Validate Model to establish the goodness of the model given the specified parameters. Use Tune Model Hyperparameters to identify the optimal parameters.

References:

<https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/partition-and-sample>

**QUESTION 24**

HOTSPOT

A biomedical research company plans to enroll people in an experimental medical treatment trial.

You create and train a binary classification model to support selection and admission of patients to the trial. The model includes the following features: Age, Gender, and Ethnicity.

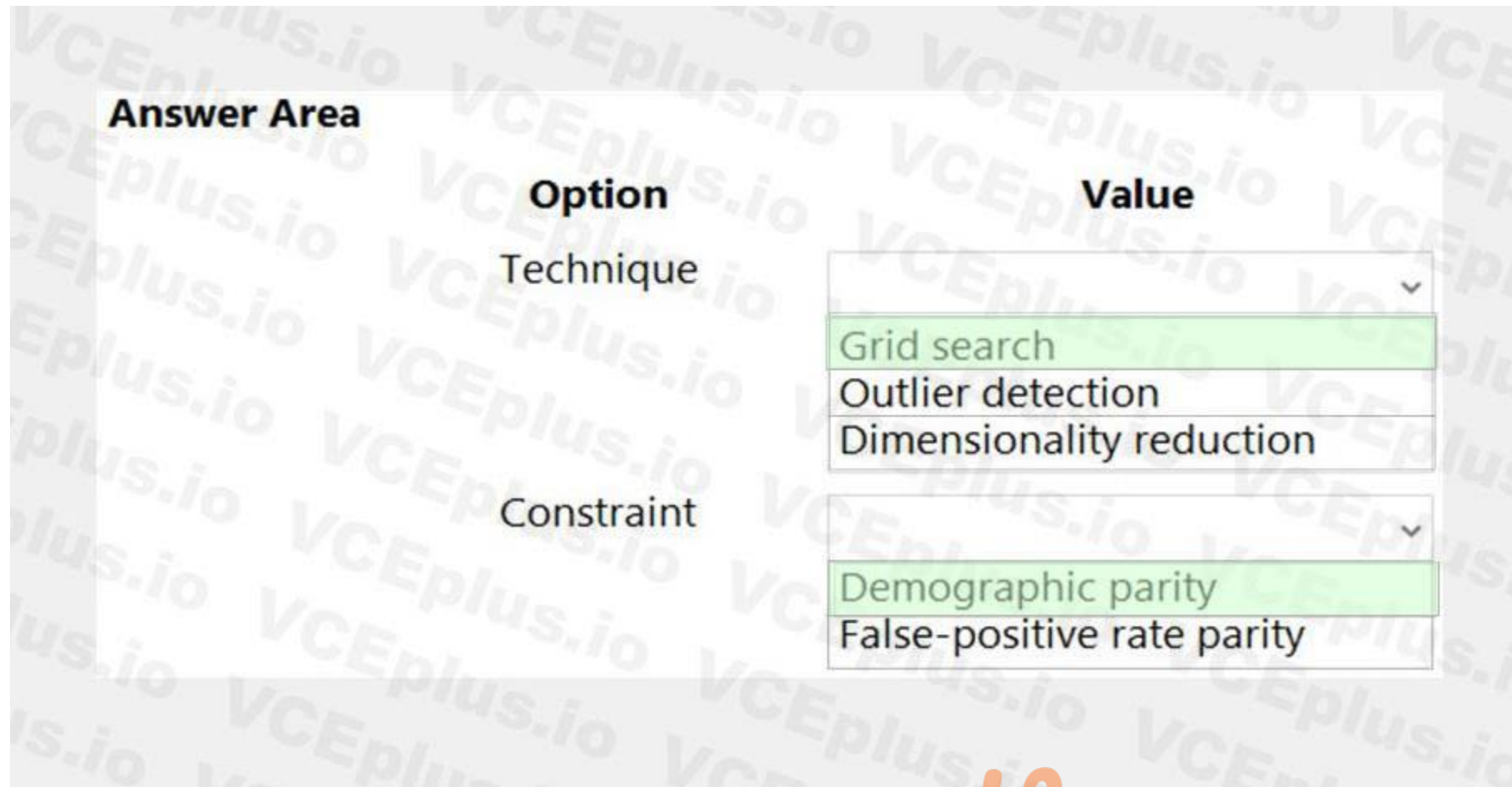
The model returns different performance metrics for people from different ethnic groups.

You need to use Fairlearn to mitigate and minimize disparities for each category in the Ethnicity feature.  
Which technique and constraint should you use? To answer, select the appropriate options in the answer area.  
NOTE: Each correct selection is worth one point.

Hot Area:

| Answer Area |                                                                                          |
|-------------|------------------------------------------------------------------------------------------|
| Option      | Value                                                                                    |
| Technique   | <input type="checkbox"/><br>Grid search<br>Outlier detection<br>Dimensionality reduction |
| Constraint  | <input type="checkbox"/><br>Demographic parity<br>False-positive rate parity             |

Answer Area:



**Section:**

**Explanation:**

Box 1: Grid Search

Fairlearn open-source package provides postprocessing and reduction unfairness mitigation algorithms: ExponentiatedGradient, GridSearch, and ThresholdOptimizer.

Note: The Fairlearn open-source package provides postprocessing and reduction unfairness mitigation algorithms types:

Reduction: These algorithms take a standard black-box machine learning estimator (e.g., a LightGBM model) and generate a set of retrained models using a sequence of re-weighted training datasets.

Post-processing: These algorithms take an existing classifier and the sensitive feature as input.

Box 2: Demographic parity

The Fairlearn open-source package supports the following types of parity constraints: Demographic parity, Equalized odds, Equal opportunity, and Bounded group loss.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/concept-fairness-ml>

**Exam F**

**QUESTION 1**

You create a training pipeline by using the Azure Machine Learning designer. You need to load data into a machine learning pipeline by using the Import Data component. Which two data sources could you use? Each correct answer presents a complete solution.

NOTE: Each correct selection is worth one point

- A. Azure Blob storage container through a registered datastore
- B. Azure SQL Database
- C. URL via HTTP
- D. Azure Data Lake Storage Gen2
- E. Registered dataset



**Correct Answer: A, C**

**Section:**

**QUESTION 2**

You create a multi-class image classification model with automated machine learning in Azure Machine Learning.

You need to prepare labeled image data as input for model training in the form of an Azure Machine Learning tabular dataset.

Which data format should you use?

- A. COCO
- B. JSONL
- C. JSON
- D. Pascal VOC

**Correct Answer: C**

**Section:**

**QUESTION 3**

DRAG DROP

You use a training pipeline in the Azure Machine Learning designer. You register a datastore named ds1. The datastore contains multiple training data files. You use the Import Data module with the configured datastore.

You need to retrain a model on a different set of data files.

Which four actions should you perform in sequence? To answer, move the appropriate actions from the list of actions to the answer area and arrange them in the correct order.

**Select and Place:**

**Actions**

- Add a new parameter in the module indicating the path to the training file.
- Register each training file as a new datastore.
- Specify a new path to the training file as a parameter value.
- Run the training pipeline by using the studio portal.
- Publish a training pipeline.

**Answer area**

**Correct Answer:**



| Actions                                                                     | Answer area                                                   |
|-----------------------------------------------------------------------------|---------------------------------------------------------------|
| Add a new parameter in the module indicating the path to the training file. | Register each training file as a new datastore.               |
|                                                                             | Specify a new path to the training file as a parameter value. |
|                                                                             | Run the training pipeline by using the studio portal.         |
|                                                                             | Publish a training pipeline.                                  |
|                                                                             |                                                               |

**Section:**  
**Explanation:**

**QUESTION 4**

HOTSPOT

You are developing code to analyse a dataset that includes age information for a large group of diabetes patients. You create an Azure Machine Learning workspace and install all required libraries. You set the privacy budget to 1.0).

You must analyze the dataset and preserve data privacy. The code must run twice before the privacy budget is depleted.

You need to complete the code.

Which values should you use? To answer, select the appropriate options in the answer area.

NOTE: Each correct selection is worth one point.



**Hot Area:**

```

Answer Area
import pandas as pd
data_path = "data/diabetes.csv"
diabetes = pd.read_csv(data_path)
import as lib
cols =
azureml.datadrift
with lib.Analysis() as analysis:
 data = dp.Dataset(path=data_path, column_names=cols)
 age_mean = lib.dp_mean(data = lib.cast(data['age'], type="FLOAT"),
 privacy_usage = (, .50),
 data_lower = 0.,
 data_upper = 100.,
)

```

**Answer Area:**

Answer Area

```
import pandas as pd
data_path = 'data/diabetes.csv'
diabetes = pd.read_csv(data_path)
import opendp.smartnoise.core as lib
cols = opendp.smartnoise.core
azureml.datadrift
with lib.Analysis() as analysis:
 data = dp.Dataset(path=data_path, column_names=cols)
 age_mean = lib.dp_mean(data = lib.cast(data['age'], type="FLOAT"),
 privacy_usage = ('epsilon', .50),
 data_lower = 0.,
 data_upper = 100..)
```

Section:

Explanation:

Answer Area

```
import pandas as pd
data_path = 'data/diabetes.csv'
diabetes = pd.read_csv(data_path)
import opendp.smartnoise.core as lib
cols = list(diabetes.columns)

with lib.Analysis() as analysis:
 data = dp.Dataset(path=data_path, column_names=cols)
 age_mean = lib.dp_mean(data = lib.cast(data['age'], type="FLOAT"),
 privacy_usage = ('epsilon', .50),
 data_lower = 0.,
 data_upper = 100..)
```

### QUESTION 5

HOTSPOT

You train a machine learning model by using Aunt Machine Learning.

You use the following training script in Python to log an accuracy value.

```
from azureml.core.run import Run
run_logger = Run.get_context()
run_logger.log("accuracy", float(val_accuracy))
```

You must use a Python script to define a sweep job.

You need to provide the primary metric and goal you want hyper parameter tuning to optimize.

How should you complete the Python script? To answer select the appropriate options in the answer area NOTE: Each correct selection is worth one point.

Hot Area:



Answer Area:



Section:

Explanation:

#### QUESTION 6

You create an Azure Machine Learning workspace. You use Azure Machine Learning designer to create a pipeline within the workspace. You need to submit a pipeline run from the designer. What should you do first?

- A. Create a compute cluster.
- B. Create an attached compute resource.
- C. Select a model.
- D. Create an experiment.

Correct Answer: B

Section:

#### QUESTION 7

DRAG DROP

You have an Azure Machine Learning workspace. You are running an experiment on your local computer.

You need to ensure that you can use MLflow Tracking with Azure Machine Learning Python SDK v2 to store metrics and artifacts from your local experiment runs in the workspace.

In which order should you perform the actions? To answer, move all actions from the list of actions to the answer area and arrange them in the correct order.

Select and Place:



**Actions**

- Go to the workspace in the Azure portal.
- Retrieve the tracking URI of the workspace.
- Import MLflow and MLClient classes.
- Set the MLflow tracking URI and the experiment name.



**Answer area**

Empty answer area



**Correct Answer:**

**Actions**

- 
- 
- 
- 



**Answer area**

- Go to the workspace in the Azure portal.
- Retrieve the tracking URI of the workspace.
- Import MLflow and MLClient classes.
- Set the MLflow tracking URI and the experiment name.



**Section:**

**Explanation:**

- Go to the workspace in the Azure portal.
- Retrieve the tracking URI of the workspace.
- Import MLflow and MLClient classes.
- Set the MLflow tracking URI and the experiment name.

**QUESTION 8**

**DRAG DROP**

You create an Azure Machine Learning workspace. You are training a classification model with no-code AutoML in Azure Machine Learning studio. The model must predict if a client of a financial institution will subscribe to a fixed-term deposit. You must preview the data profile in Azure Machine Learning studio once the dataset is created. You need to train the model.

Which four actions should you perform in sequence? To answer, move the appropriate actions from the list of actions to the answer area and arrange them in the correct order.

**Select and Place:**



**Actions**

- Create a tabular dataset.
- Create a file dataset.
- Create a compute cluster.
- Create an experiment.
- Create an automated ML job.



**Answer area**



**Correct Answer:**

**Actions**

- Create a tabular dataset.
- 
- 
- 
- 



**Answer area**

- Create a file dataset.
- Create a compute cluster.
- Create an experiment.
- Create an automated ML job.



**Section:**

**Explanation:**

- Create a file dataset.
- Create a compute cluster.
- Create an experiment.
- Create an automated ML job.

**QUESTION 9**

**HOTSPOT**

You are using the Azure Machine Learning designer to transform a dataset by using an Execute Python Script component and custom code.

You need to define the method signature for the Execute Python Script component and return value type.

What should you define? To answer, select the appropriate options in the answer area.

NOTE: Each correct selection is worth one point.

**Hot Area:**



**Answer Area**

| Script setting                                           | Value                                                                                                                                                                                                        |
|----------------------------------------------------------|--------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|
| Method signature for the Execute Python Script Component | <ul style="list-style-type: none"><li>azureml_main(dataframe1 = None, dataframe2 = None)</li><li>azureml_main(dataframe1 = None, dataframe2 = None)</li><li>main(dataframe1 = None)</li><li>main()</li></ul> |
| Return value type                                        | <ul style="list-style-type: none"><li>Pandas dataframe</li><li>Pandas dataframe</li><li>Pandas series</li><li>Named list</li></ul>                                                                           |

Answer Area:

**Answer Area**

| Script setting                                           | Value                                                                                                                                                                                                        |
|----------------------------------------------------------|--------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|
| Method signature for the Execute Python Script Component | <ul style="list-style-type: none"><li>azureml_main(dataframe1 = None, dataframe2 = None)</li><li>azureml_main(dataframe1 = None, dataframe2 = None)</li><li>main(dataframe1 = None)</li><li>main()</li></ul> |
| Return value type                                        | <ul style="list-style-type: none"><li>Pandas dataframe</li><li>Pandas dataframe</li><li>Pandas series</li><li>Named list</li></ul>                                                                           |

Section:

Explanation:

**QUESTION 10**

DRAG DROP

You are developing a machine learning solution by using the Azure Machine Learning designer.

You need to create a web service that applications can use to submit data feature values and retrieve a predicted label.

Which three actions should you perform in sequence? To answer, move the appropriate actions from the list of actions to the answer area and arrange them in the correct order.

Select and Place:

**Actions**

- Create and run a batch inference pipeline.
- Create and run a training pipeline.
- Deploy a service to an inference cluster.
- Create and run a real-time inference pipeline.



**Answer area**



**Correct Answer:**

**Actions**

- Create and run a batch inference pipeline.
- 
- 
- 



**Answer area**

- Create and run a training pipeline.
- Deploy a service to an inference cluster.
- Create and run a real-time inference pipeline.



**Section:**

**Explanation:**

Create and run a training pipeline.  
Deploy a service to an inference cluster.  
Create and run a real-time inference pipeline.

**QUESTION 11**

You manage an Azure Machine Learning workspace.  
You must provide explanations for the behavior of the models with feature importance measures.  
You need to configure a Responsible AI dashboard in Azure Machine Learning.  
Which dashboard component should you configure?

- A. Fairness assessment
- B. Counterfactual what-if
- C. Interpretability
- D. Casual inference

**Correct Answer: C**

**Section:**

**QUESTION 12**

**HOTSPOT**

You create an Azure Machine Learning workspace.

You plan to write an Azure Machine Learning SDK for Python v2 script that logs an image for an experiment. The logged image must be available from the images tab in Azure Machine Learning Studio.

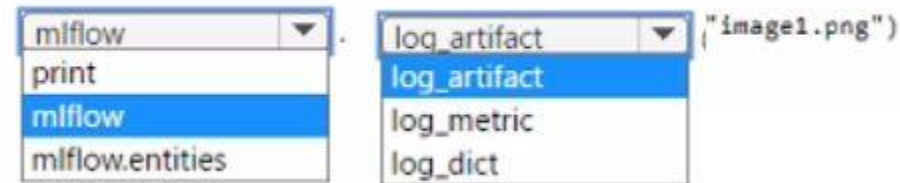
You need to complete the script.

Which code segments should you use? To answer, select the appropriate options in the answer area.

NOTE: Each correct selection is worth one point.

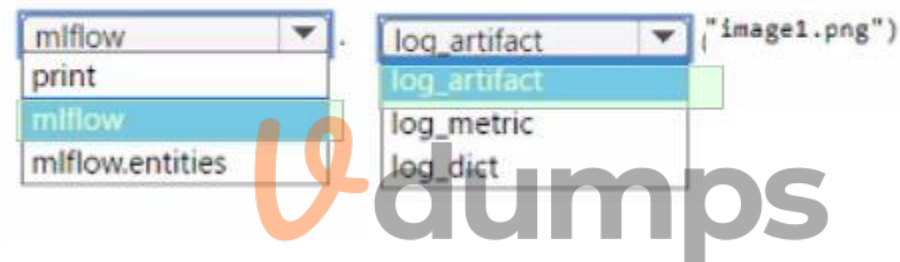
**Hot Area:**

**Answer Area**



**Answer Area:**

**Answer Area**



**Section:**

**Explanation:**

**QUESTION 13**

**HOTSPOT**

You use Azure Machine Learning to implement hyperparameter tuning with a Bandit early termination policy.

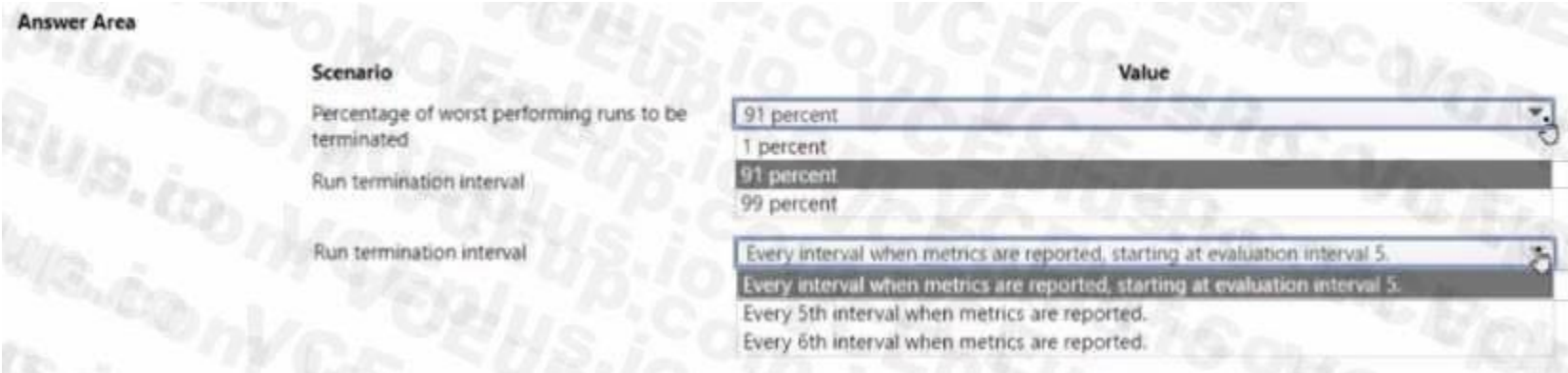
The policy uses a slack\_factor set to 0.1, an evaluation interval set to 1, and an evaluation delay set to b.

You need to evaluate the outcome of the early termination policy.

What should you evaluate? To answer, select the appropriate options in the answer area.

NOTE: Each correct selection is worth one point.

**Hot Area:**





**Answer Area:**

| Scenario                                             | Value                                                                        |
|------------------------------------------------------|------------------------------------------------------------------------------|
| Percentage of worst performing runs to be terminated | 91 percent                                                                   |
| Run termination interval                             | 1 percent                                                                    |
| Run termination interval                             | 91 percent                                                                   |
| Run termination interval                             | 99 percent                                                                   |
| Run termination interval                             | Every interval when metrics are reported, starting at evaluation interval 5. |
| Run termination interval                             | Every interval when metrics are reported, starting at evaluation interval 5. |
| Run termination interval                             | Every 5th interval when metrics are reported.                                |
| Run termination interval                             | Every 6th interval when metrics are reported.                                |

**Section:**

**Explanation:**

**QUESTION 14**

**HOTSPOT**

You create an Azure Machine learning workspace and load a Python training script named tram.py in the src subfolder. The dataset used to train your model is available locally. You run the following script to tram the model:

```
ws = Workspace.from_config()
experiment = Experiment(workspace=ws, name='nlp-experiment-train')

cpu_cluster_name = "cpu-cluster"
try:
 cpu_cluster = ComputeTarget(workspace=ws, name=cpu_cluster_name)
except ComputeTargetException:
 compute_config = AmlCompute.provisioning_configuration(vm_size='STANDARD_D2_V2', max_nodes=4)
 cpu_cluster = ComputeTarget.create(ws, cpu_cluster_name, compute_config)

cpu_cluster.wait_for_completion(show_output=True)

config = ScriptRunConfig(source_directory='./src',
 script='train.py',
 compute_target='cpu-cluster')

env = Environment.from_conda_specification(
 name='pytorch-env',
 file_path='./azureml/pytorch-env.yml'
)
config.run_config.environment = env

run = experiment.submit(config)
```

instructions: For each of the following statements, select Yes if the statement is true. Otherwise, select No.

NOTE: Each correct selection is worth one point

**Hot Area:**

**Answer Area**

| Statements                                                                                                         | Yes                   | No                    |
|--------------------------------------------------------------------------------------------------------------------|-----------------------|-----------------------|
| The script will use local compute resources and a new Azure Machine Learning compute will be created upon failure. | <input type="radio"/> | <input type="radio"/> |
| The dataset used during the training phase is automatically loaded in a new datastore.                             | <input type="radio"/> | <input type="radio"/> |
| A new environment object is created from a local Conda environment.                                                | <input type="radio"/> | <input type="radio"/> |

**Answer Area:**

**Answer Area**

| Statements                                                                                                         | Yes                   | No                               |
|--------------------------------------------------------------------------------------------------------------------|-----------------------|----------------------------------|
| The script will use local compute resources and a new Azure Machine Learning compute will be created upon failure. | <input type="radio"/> | <input checked="" type="radio"/> |
| The dataset used during the training phase is automatically loaded in a new datastore.                             | <input type="radio"/> | <input checked="" type="radio"/> |
| A new environment object is created from a local Conda environment.                                                | <input type="radio"/> | <input checked="" type="radio"/> |

**Section:**

**Explanation:**



**QUESTION 15**

**HOTSPOT**

You have a binary classifier that predicts positive cases of diabetes within two separate age groups.

The classifier exhibits a high degree of disparity between the age groups.

You need to modify the output of the classifier to maximize its degree of fairness across the age groups and meet the following requirements:

- Eliminate the need to retrain the model on which the classifier is based.
- Minimize the disparity between true positive rates and false positive rates across age groups.

Which algorithm and parity constraint should you use? To answer, select the appropriate options in the answer area. NOTE: Each correct selection is worth one point.

**Hot Area:**

**Answer Area**

| Setting           | Value                                                                                                                                                      |
|-------------------|------------------------------------------------------------------------------------------------------------------------------------------------------------|
| Algorithm         | <ul style="list-style-type: none"> <li>Exponentiated gradient</li> <li>Exponentiated gradient</li> <li>Grid search</li> <li>Threshold optimizer</li> </ul> |
| Parity constraint | <ul style="list-style-type: none"> <li>Bounded group loss</li> <li>Bounded group loss</li> <li>Equalized odds</li> <li>Error rate parity</li> </ul>        |

**Answer Area:**



Section:

Explanation:

### QUESTION 16

HOTSPOT

You have an Azure Machine learning workspace. The workspace contains a dataset with data in a tabular form.

You plan to use the Azure Machine Learning SDK for Python v1 to create a control script that will load the dataset into a pandas dataframe in preparation for model training. The script will accept a parameter designating the dataset. You need to complete the script.

How should you complete the script? To answer, select the appropriate options in the answer area.

NOTE: Each correct selection is worth one point.

Hot Area:



Answer Area:





Answer Area

```
import argparse
from azureml.core import Dataset, Run
parser = argparse.ArgumentParser()
parser.add_argument("--input-data", type=str)
args = parser.parse_args()
run = Run.get_context()
ws = run.experiment.workspace

get_by_id(ws, id=args.input_data)
to_pandas_dataframe()
from_pandas_dataframe()

ds = Dataset.from_pandas_dataframe()
df = ds.to_pandas_dataframe()
get_by_id(ws, id=args.input_data)
to_pandas_dataframe()
from_pandas_dataframe()
```



**Section:**

**Explanation:**

**QUESTION 17**

You train and register a machine learning model. You create a batch inference pipeline that uses the model to generate predictions from multiple data files. You must publish the batch inference pipeline as a service that can be scheduled to run every night. You need to select an appropriate compute target for the inference service. Which compute target should you use?

- A. Azure Machine Learning compute instance
- B. Azure Machine Learning compute cluster
- C. Azure Kubernetes Service (AKS)-based inference cluster
- D. Azure Container Instance (ACI) compute target

**Correct Answer: B**

**Section:**

**Explanation:**

Azure Machine Learning compute clusters is used for Batch inference. Run batch scoring on serverless compute. Supports normal and low-priority VMs. No support for real-time inference.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/concept-compute-target>

**QUESTION 18**

You use the Azure Machine Learning designer to create and run a training pipeline. The pipeline must be run every night to inference predictions from a large volume of files. The folder where the files will be stored is defined as a dataset. You need to publish the pipeline as a REST service that can be used for the nightly influencing run. What should you do?

- A. Create a batch inference pipeline
- B. Set the compute target for the pipeline to an inference cluster
- C. Create a real-time inference pipeline



D. Clone the pipeline

**Correct Answer: A**

**Section:**

**Explanation:**

Azure Machine Learning Batch Inference targets large inference jobs that are not time-sensitive. Batch Inference provides cost-effective inference compute scaling, with unparalleled throughput for asynchronous applications. It is optimized for high throughput, fire-and-forget inference over large collections of data.

You can submit a batch inference job by pipeline\_run, or through REST calls with a published pipeline.

Reference: <https://github.com/Azure/MachineLearningNotebooks/blob/master/how-to-use-azureml/machine-learning-pipelines/parallel-run/README.md>

#### QUESTION 19

You create a binary classification model. The model is registered in an Azure Machine Learning workspace. You use the Azure Machine Learning Fairness SDK to assess the model fairness.

You develop a training script for the model on a local machine.

You need to load the model fairness metrics into Azure Machine Learning studio.

What should you do?

- A. Implement the download\_dashboard\_by\_upload\_id function
- B. Implement the create\_group\_metric\_set function
- C. Implement the upload\_dashboard\_dictionary function
- D. Upload the training script

**Correct Answer: C**

**Section:**

**Explanation:**

import azureml.contrib.fairness package to perform the upload:

```
from azureml.contrib.fairness import upload_dashboard_dictionary, download_dashboard_by_upload_id
```

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/how-to-machine-learning-fairness-aml>

#### QUESTION 20

You have a dataset that includes confidential data. You use the dataset to train a model.

You must use a differential privacy parameter to keep the data of individuals safe and private.

You need to reduce the effect of user data on aggregated results.

What should you do?

- A. Decrease the value of the epsilon parameter to reduce the amount of noise added to the data
- B. Increase the value of the epsilon parameter to decrease privacy and increase accuracy
- C. Decrease the value of the epsilon parameter to increase privacy and reduce accuracy
- D. Set the value of the epsilon parameter to 1 to ensure maximum privacy

**Correct Answer: C**

**Section:**

**Explanation:**

Differential privacy tries to protect against the possibility that a user can produce an indefinite number of reports to eventually reveal sensitive data. A value known as epsilon measures how noisy, or private, a report is.

Epsilon has an inverse relationship to noise or privacy. The lower the epsilon, the more noisy (and private) the data is.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/concept-differential-privacy>

#### QUESTION 21

DRAG DROP

You are planning to host practical training to acquaint staff with Docker for Windows.

Staff devices must support the installation of Docker.

Which of the following are requirements for this installation? Answer by dragging the correct options from the list to the answer area.

Select and Place:

| Options                                                   | Answer |
|-----------------------------------------------------------|--------|
| 2 GB of system RAM                                        |        |
| 4 GB of system RAM                                        |        |
| BIOS-enabled virtualization                               |        |
| Microsoft Hardware-Assisted Virtualization Detection Tool |        |
| Windows 10 64-bit                                         |        |
| Windows 10 32-bit                                         |        |

Vdumps

Correct Answer:

| Options                                                   | Answer                      |
|-----------------------------------------------------------|-----------------------------|
| 2 GB of system RAM                                        | 4 GB of system RAM          |
|                                                           | BIOS-enabled virtualization |
|                                                           | Windows 10 64-bit           |
| Microsoft Hardware-Assisted Virtualization Detection Tool |                             |
|                                                           |                             |
| Windows 10 32-bit                                         |                             |



**Section:**

**Explanation:**

Reference: [https://docs.docker.com/toolbox/toolbox\\_install\\_windows/](https://docs.docker.com/toolbox/toolbox_install_windows/)  
<https://blogs.technet.microsoft.com/canitpro/2015/09/08/step-by-step-enabling-hyper-v-for-use-on-windows-10/>  
<https://docs.docker.com/docker-for-windows/install/>

**QUESTION 22**

**HOTSPOT**

You are using an Azure Machine Learning workspace. You set up an environment for model testing and an environment for production.

The compute target for testing must minimize cost and deployment efforts. The compute target for production must provide fast response time, autoscaling of the deployed service, and support real-time inferencing.

You need to configure compute targets for model testing and production.

Which compute targets should you use? To answer, select the appropriate options in the answer area.

NOTE: Each correct selection is worth one point.

**Hot Area:**

## Answer Area

### Environment

### Compute target

Testing

|                                         |   |
|-----------------------------------------|---|
|                                         | ▼ |
| Local web service                       |   |
| Azure Kubernetes Services (AKS)         |   |
| Azure Container Instances               |   |
| Azure Machine Learning compute clusters |   |

Production

|                                         |   |
|-----------------------------------------|---|
|                                         | ▼ |
| Local web service                       |   |
| Azure Kubernetes Services (AKS)         |   |
| Azure Container Instances               |   |
| Azure Machine Learning compute clusters |   |

Answer Area:



**Answer Area**

| Environment | Compute target                                                                                                                                                                                                                                                                                                                                                                                                                                         |
|-------------|--------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|
| Testing     | <div style="border: 1px solid black; padding: 5px;"> <div style="display: flex; justify-content: space-between; align-items: center;"> <span>▼</span> </div> <div style="background-color: #e0ffe0; padding: 2px;">Local web service</div> <div style="padding: 2px;">Azure Kubernetes Services (AKS)</div> <div style="padding: 2px;">Azure Container Instances</div> <div style="padding: 2px;">Azure Machine Learning compute clusters</div> </div> |
| Production  | <div style="border: 1px solid black; padding: 5px;"> <div style="display: flex; justify-content: space-between; align-items: center;"> <span>▼</span> </div> <div style="padding: 2px;">Local web service</div> <div style="background-color: #e0ffe0; padding: 2px;">Azure Kubernetes Services (AKS)</div> <div style="padding: 2px;">Azure Container Instances</div> <div style="padding: 2px;">Azure Machine Learning compute clusters</div> </div> |

**Section:**

**Explanation:**

Box 1: Local web service

The Local web service compute target is used for testing/debugging. Use it for limited testing and troubleshooting. Hardware acceleration depends on use of libraries in the local system.

Box 2: Azure Kubernetes Service (AKS)

Azure Kubernetes Service (AKS) is used for Real-time inference. Recommended for production workloads.

Use it for high-scale production deployments. Provides fast response time and autoscaling of the deployed service

Reference: <https://docs.microsoft.com/en-us/azure/machine-learning/concept-compute-target>

**QUESTION 23**

**DRAG DROP**

You are using a Git repository to track work in an Azure Machine Learning workspace.

You need to authenticate a Git account by using SSH.

Which three actions should you perform in sequence? To answer, move the appropriate actions from the list of actions to the answer area and arrange them in the correct order.

**Select and Place:**

| Actions                                                 | Answer Area |
|---------------------------------------------------------|-------------|
| Generate a public/private key pair                      |             |
| Add the private key to the Git account                  |             |
| Clone the Git repository by using an SSH repository URL |             |
| Add the public key to the Git account                   |             |
| Create a new Azure Key Vault resource                   |             |

Correct Answer:

| Actions                                | Answer Area                                             |
|----------------------------------------|---------------------------------------------------------|
|                                        | Generate a public/private key pair                      |
| Add the private key to the Git account | Add the public key to the Git account                   |
|                                        | Clone the Git repository by using an SSH repository URL |
|                                        |                                                         |
| Create a new Azure Key Vault resource  |                                                         |

Section:

Explanation:

Authenticate your Git Account with SSH:

Step 1: Generating a public/private key pair

Generate a new SSH key

1. Open the terminal window in the Azure Machine Learning Notebook Tab.

2. Paste the text below, substituting in your email address.

ssh-keygen -t rsa -b 4096 -C "your\_email@example.com" This creates a new ssh key, using the provided email as a label.

> Generating public/private rsa key pair.

Step 2: Add the public key to the Git Account

In your terminal window, copy the contents of your public key file.

Step 3: Clone the Git repository by using an SSH repository URL 1. Copy the SSH Git clone URL from the Git repo.

2. Paste the url into the git clone command below, to use your SSH Git repo URL. This will look something like:

git clone git@example.com:GitUser/azureml-example.git Cloning into 'azureml-example'.

Reference: <https://docs.microsoft.com/en-us/azure/machine-learning/concept-train-model-git-integration>

QUESTION 24

HOTSPOT

You are the owner of an Azure Machine Learning workspace.



You must prevent the creation or deletion of compute resources by using a custom role. You must allow all other operations inside the workspace.  
You need to configure the custom role.  
How should you complete the configuration? To answer, select the appropriate options in the answer area.  
NOTE: Each correct selection is worth one point.

Hot Area:

**Answer Area**

```
{
 "Name": "Data Scientist Custom",
 "IsCustom": true
 "Description": "Description"
 "Actions": [
 Microsoft.MachineLearningServices/workspaces/*/read
 Microsoft.MachineLearningServices/workspaces/computes/*/write
 Microsoft.MachineLearningServices/workspaces/delete
],
 "NotActions": [
 Microsoft.MachineLearningServices/workspaces/*/read
 Microsoft.MachineLearningServices/workspaces/*/write
 Microsoft.MachineLearningServices/workspaces/computes/*/delete
],
 "AssignableScopes": [
 "/subscriptions/<subscription_id>"
]
}
```

Answer Area:

## Answer Area

```
{
 "Name": "Data Scientist Custom",
 "IsCustom": true
 "Description": "Description"
 "Actions": [
 Microsoft.MachineLearningServices/workspaces/*/read
 Microsoft.MachineLearningServices/workspaces/computes/*/write
 Microsoft.MachineLearningServices/workspaces/delete
],
 "NotActions": [
 Microsoft.MachineLearningServices/workspaces/*/write
 Microsoft.MachineLearningServices/workspaces/computes/*/write
 Microsoft.MachineLearningServices/workspaces/delete
],
 "AssignableScopes": [
 "/subscriptions/<subscription_id>"
]
}
```

### Section:

### Explanation:

Box 1: Microsoft.MachineLearningServices/workspaces/\*/read

Reader role: Read-only actions in the workspace. Readers can list and view assets, including datastore credentials, in a workspace. Readers can't create or update these assets.

Box 2: Microsoft.MachineLearningServices/workspaces/\*/write

If the roles include Actions that have a wildcard (\*), the effective permissions are computed by subtracting the NotActions from the allowed Actions.

Box 3: Box 2: Microsoft.MachineLearningServices/workspaces/computes/\*/delete

Box 4: Microsoft.MachineLearningServices/workspaces/computes/\*/write

Reference: <https://docs.microsoft.com/en-us/azure/role-based-access-control/overview#how-azure-rbac-determines-if-a-user-has-access-to-a-resource>

### QUESTION 25

#### HOTSPOT

You create an Azure Machine Learning workspace named workspace1. You assign a custom role to a user of workspace1.

The custom role has the following JSON definition:



```
{
 "Name": "MyRole",
 "IsCustom": true,
 "Description": "New custom role description.",
 "Actions": ["*"],
 "NotActions": [
 "Microsoft.MachineLearningServices/workspaces/write",
 "Microsoft.MachineLearningServices/workspaces/computes/*/write",
 "Microsoft.MachineLearningServices/workspaces/computes/*/delete",
 "Microsoft.Authorization/*/write"
],
 "AssignableScopes": [
 "/subscriptions/<subscription_id>/resourceGroups/resourcegroup1/providers/
 Microsoft.MachineLearningServices/workspaces/workspace1"
]
}
```

Instructions: For each of the following statements, select Yes if the statement is true. Otherwise, select No.  
NOTE: Each correct selection is worth one point.

Hot Area:

| Statements                                              | Yes                   | No                    |
|---------------------------------------------------------|-----------------------|-----------------------|
| The user can perform all actions in the workspace       | <input type="radio"/> | <input type="radio"/> |
| The user can delete a compute resource in the workspace | <input type="radio"/> | <input type="radio"/> |
| The user can write metrics to the workspace             | <input type="radio"/> | <input type="radio"/> |

Answer Area:

| Statements                                              | Yes                              | No                               |
|---------------------------------------------------------|----------------------------------|----------------------------------|
| The user can perform all actions in the workspace       | <input type="radio"/>            | <input checked="" type="radio"/> |
| The user can delete a compute resource in the workspace | <input type="radio"/>            | <input checked="" type="radio"/> |
| The user can write metrics to the workspace             | <input checked="" type="radio"/> | <input type="radio"/>            |

**Section:**

**Explanation:**

Box 1: No

The actions listed in NotActions are prohibited.

If the roles include Actions that have a wildcard (\*), the effective permissions are computed by subtracting the NotActions from the allowed Actions.

Box 2: No

Deleting compute resources in the workspace is in the NotActions list.

Box 3: Yes

Writing metrics is not listed in NotActions.

Reference: <https://docs.microsoft.com/en-us/azure/role-based-access-control/overview#how-azure-rbac-determines-if-a-user-has-access-to-a-resource>

**QUESTION 26**

**HOTSPOT**

You create a new Azure Databricks workspace.

You configure a new cluster for long-running tasks with mixed loads on the compute cluster as shown in the image below.

Microsoft Azure

## Create Cluster

### New Cluster

Cancel **Create Cluster** 2-8 Workers: 28.0-112.0 GB Memory, 8-32 Cores, 1.5-6 DBU  
1 Driver: 14.0 GB Memory, 4 Cores, 0.75 DBU

Cluster Name  
mysparkcluster

Cluster Mode  
Standard

Pool  
None

Databricks Runtime Version [Learn more](#)  
Runtime: 6.4 (Scala 2.11, Spark 2.4.5)

**New** This Runtime version supports only Python 3.

Autopilot Options

- Enable autoscaling
- Terminate after 120 minutes of inactivity

Worker Type Min Workers Max Workers

Standard\_DS3\_v2 14.0 GB Memory, 4 Cores, 0.75 DBU 2 8

Driver Type  
Same as worker 14.0 GB Memory, 4 Cores, 0.75 DBU

Advanced Options

Use the drop-down menus to select the answer choice that completes each statement based on the information presented in the graphic.  
NOTE: Each correct selection is worth one point.

Hot Area:



### Answer Area

Code for each user runs as a separate process

|                          |   |
|--------------------------|---|
| <input type="checkbox"/> | ▼ |
| Yes                      |   |
| No                       |   |

The number of workers is fixed for the entire duration of the job

|                          |   |
|--------------------------|---|
| <input type="checkbox"/> | ▼ |
| Yes                      |   |
| No                       |   |

Answer Area:

### Answer Area

Code for each user runs as a separate process

|                          |   |
|--------------------------|---|
| <input type="checkbox"/> | ▼ |
| Yes                      |   |
| No                       |   |

The number of workers is fixed for the entire duration of the job

|                          |   |
|--------------------------|---|
| <input type="checkbox"/> | ▼ |
| Yes                      |   |
| No                       |   |

**Section:**

**Explanation:**

Box 1: No

Running user code in separate processes is not possible in Scala.

Box 2: No

Autoscaling is enabled. Minimum 2 workers, Maximum 8 workers.

Reference:

<https://docs.databricks.com/clusters/configure.html>

### QUESTION 27

You have machine learning models produce unfair predictions across sensitive features.

You must use a post-processing technique to apply a constraint to the models to mitigate their unfairness.

You need to select a post-processing technique and model type.

What should you use? To answer, select the appropriate options in the answer area.



Answer Area

| Setting    | Value                 |
|------------|-----------------------|
| Technique  | Grid Search           |
| Model type | Binary classification |

NOTE: Each correct selection is worth one point.

A. See below image

**Correct Answer: A**

**Section:**

**Explanation:**

Answer Area

| Setting    | Value                 |
|------------|-----------------------|
| Technique  | Grid Search           |
| Model type | Binary classification |

**QUESTION 28**

You have an Azure Machine Learning workspace

You plan to use the Azure Machine Learning SDK for Python v1 to submit a job to run a training script.

You need to complete the script to ensure that it will execute the training script.

How should you complete the script? To answer, select the appropriate options in the answer area.

NOTE: Each correct selection is worth one point

Answer Area

```

from azureml.core import Workspace, Environment, Experiment, ScriptRunConfig

ws = Workspace.from_config()
env = Environment.get(workspace=ws, name='AzureML-Minimal')
exp = Experiment(workspace=ws, name='experiment')

src = ScriptRunConfig(source_directory='./src',
 script='train.py',
 compute_target='compute-cluster',
 environment=env)

run = env . submit (config=src)

```

A. See below image

**Correct Answer: A**

**Section:**

**Explanation:**

### Answer Area

```
from azureml.core import Workspace, Environment, Experiment, ScriptRunConfig

ws = Workspace.from_config()
env = Environment.get(workspace=ws, name='AzureML-Minimal')
exp = Experiment(workspace=ws, name='experiment')

src = ScriptRunConfig(source_directory='./src',
 script='train.py',
 compute_target='compute-cluster',
 environment=env)

run = env.submit(src) (config=src)
```

### QUESTION 29

You create an Azure Machine Learning workspace. You train an MLflow-formatted regression model by using tabular structured data. You must use a Responsible AI dashboard to assess the model. You need to use the Azure Machine Learning studio UI to generate the Responsible AI dashboard. What should you do first?

- A. Deploy the model to a managed online endpoint.
- B. Register the model with the workspace.
- C. Create the model explanations.
- D. Convert the model from the MLflow format to a custom format.

**Correct Answer: B**

**Section:**

**Explanation:**

### QUESTION 30

You have an Azure Machine Learning workspace named workspaces. You must add a datastore that connects an Azure Blob storage container to workspaces. You must be able to configure a privilege level. You need to configure authentication. Which authentication method should you use?

- A. Account key
- B. SAS token
- C. Service principal
- D. Managed identity

**Correct Answer: D**

**Section:**

### QUESTION 31

You are developing a machine learning model.



You must inference the machine learning model for testing.

You need to use a minimal cost compute target

Which two compute targets should you use? Each correct answer presents a complete solution.

NOTE: Each correct selection is worth one point

- A. Local web service
- B. Remote VM
- C. Azure Databricks
- D. Azure Machine Learning Kubernetes
- E. Azure Container Instances

**Correct Answer: A, E**

**Section:**

### QUESTION 32

You are creating a compute target to train a machine learning experiment.

The compute target must support automated machine learning, machine learning pipelines, and Azure Machine Learning designer training.

You need to configure the compute target

Which option should you use?

- A. Azure HDInsight
- B. Azure Machine Learning compute cluster
- C. Azure Batch
- D. Remote VM

**Correct Answer: B**

**Section:**

### QUESTION 33

You create an Azure Machine Learning workspace. You use the Azure Machine Learning SDK for Python.

You must create a dataset from remote paths. The dataset must be reusable within the workspace.

You need to create the dataset.

How should you complete the following code segment? To answer, select the appropriate options in the answer area.

NOTE: Each correct selection is worth one point.

- A. See below image



### Answer Area

```
from azureml.core import Dataset
from azureml.data.dataset_factory import DataType
web_paths = ['https://domain.blob.core.windows.net/demo/dataset1.tsv',
 'https://domain.blob.core.windows.net/demo/dataset2.tsv']
ds = Dataset.Tabular.from_parquet_files(path=web_paths)
ds = ds.unregister_all_versions(workspace=workspace,
 name='ds',
 description='training data')
```

**Correct Answer: A**

**Section:**

#### QUESTION 34

You manage an Azure Machine Learning workspace by using the Azure CLI ml extension v2. You need to define a YAML schema to create a compute cluster. Which schema should you use?

- A. <https://azuremlschemas.azureedge.net/latest/computdnstarKeichema.json>
- B. <https://azuremlschemas.azureedge.net/latest/amlCompute.schema.json>
- C. <https://azuremlschemas.azureedge.net/latest/vmCompute.schema.json>
- D. <https://azuremlschemas.azureedge.net/latest/kubernetesCompute.schema.json>

**Correct Answer: B**

**Section:**

#### QUESTION 35

You are developing a machine learning model by using Azure Machine Learning. You are using multiple text files in tabular format for model data. You have the following requirements:

- You must use AutoML jobs to train the model.
- You must use data from specified columns.
- The data concept must support lazy evaluation.

You need to load data into a Pandas dataframe.

Which data concept should you use?

- A. Data asset
- B. URI
- C. Datastore
- D. MLTable

**Correct Answer: D**

**Section:**

#### QUESTION 36

You create an Azure Machine Learning dataset containing automobile price data. The dataset includes 10,000 rows and 10 columns. You use the Azure Machine Learning designer to transform the dataset by using an Execute Python Script component and custom code.

The code must combine three columns to create a new column.





You need to configure the code function.

Which configurations should you use? To answer, select the appropriate options in the answer area.

NOTE: Each correct selection is worth one point.

A. See below image

**Answer Area**

| Function setting          | Value        |
|---------------------------|--------------|
| Entry point function name | azureml_main |
| Function return type      | vector       |

**Correct Answer: A**

**Section:**

### QUESTION 37

You use differential privacy to ensure your reports are private. The calculated value of the epsilon for your data is 1.8. You need to modify your data to ensure your reports are private. Which epsilon value should you accept for your data?

- A. between 0 and 1
- B. between 2 and 3
- C. between 3 and 10
- D. more than 10



**Correct Answer: B**

**Section:**

### QUESTION 38

HOTSPOT

You must use in Azure Data Science Virtual Machine (DSVM) as a compute target.

You need to attach an existing DSVM to the workspace by using the Azure Machine Learning SDK for Python.

How should you complete the following code segment? To answer, select the appropriate options in the answer area.

NOTE: Each correct selection is worth one point.

**Hot Area:**

Answer Area

```
from azureml.core.compute import RemoteCompute, ComputeTarget
compute_target_name = "dsvm"
config = RemoteCompute.attach_configuration(resource_id='<resource_id>',
ssh_port=22, username='<username>', private_key_file='./ssh/id_rsa')
compute = ComputeTarget.attach(ws, compute_target_name, config)
compute.wait_for_completion(show_output=True)
```

Answer Area:

Answer Area

```
from azureml.core.compute import RemoteCompute, ComputeTarget
compute_target_name = "dsvm"
config = RemoteCompute.attach_configuration(resource_id='<resource_id>',
ssh_port=22, username='<username>', private_key_file='./ssh/id_rsa')
compute = ComputeTarget.attach(ws, compute_target_name, config)
compute.wait_for_completion(show_output=True)
```



Section:

Explanation:

Answer Area

```
from azureml.core.compute import RemoteCompute, ComputeTarget
compute_target_name = "dsvm"
config = RemoteCompute.attach_configuration(resource_id='<resource_id>',
ssh_port=22, username='<username>', private_key_file='./ssh/id_rsa')
compute = ComputeTarget.attach(ws, compute_target_name, config)
compute.wait_for_completion(show_output=True)
```

**QUESTION 39**

DRAG DROP

You need to implement source control for scripts in an Azure Machine Learning workspace. You use a terminal window in the Azure Machine Learning Notebook tab. You must authenticate your Git account with SSH. You need to generate a new SSH key.

Which four actions should you perform in sequence? To answer, move the appropriate actions from the list of actions to the answer area and arrange them in the correct order.

**Select and Place:**

| Actions                                                                                        | Answer area |
|------------------------------------------------------------------------------------------------|-------------|
| Run the ssh-keygen command.                                                                    |             |
| Press <b>Enter</b> when prompted to enter a file in which to save the key.                     |             |
| Verify that the default location is <code>/home/azureuser/.ssh</code> and press <b>Enter</b> . |             |
| Type a secure passphrase.                                                                      |             |

**Correct Answer:**

| Actions | Answer area                                                                                    |
|---------|------------------------------------------------------------------------------------------------|
|         | Run the ssh-keygen command.                                                                    |
|         | Press <b>Enter</b> when prompted to enter a file in which to save the key.                     |
|         | Verify that the default location is <code>/home/azureuser/.ssh</code> and press <b>Enter</b> . |
|         | Type a secure passphrase.                                                                      |

**Section:**

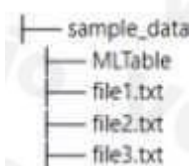
**Explanation:**

**QUESTION 40**

**HOTSPOT**

You manage an Azure Machine Learning workspace named workspace1 by using the Python SDK v2.

The default datastore of workspace1 contains a folder named sample\_data. The folder structure contains the following content:



You write Python SDK v2 code to materialize the data from the files in the sample.data folder into a Pandas data frame. You need to complete the Python SDK v2 code to use the MLTable folder as the materialization blueprint. How should you complete the code? To answer, select the appropriate options in the answer area.

NOTE: Each correct selection is worth one point.

**Hot Area:**



Answer Area

```
import mltable
tbl = mltable.load('/sample_data/MTable')
df = tbl.to_pandas_dataframe()
```

Answer Area:

Answer Area

```
import mltable
tbl = mltable.load('/sample_data/MTable')
df = tbl.to_pandas_dataframe()
```

Section:

Explanation:



Answer Area

```
import mltable
tbl = mltable.load('/sample_data/MTable')
df = tbl.to_pandas_dataframe()
```

**QUESTION 41**

You manage an Azure Machine Learning workspace named workspaces. You must develop Python SDK v2 code to attach an Azure Synapse Spark pool as a compute target in workspaces. The code must invoke the constructor of the SynapseSparkCompute class.

You need to invoke the constructor.

What should you use?

- A. Synapse workspace web URL and Spark pool name
- B. resource ID of the Synapse Spark pool and a user-defined name
- C. pool URL of the Synapse Spark pool and a system-assigned name
- D. Synapse workspace name and workspace web URL

**Correct Answer: B**

Section:

**QUESTION 42**

HOTSPOT

You plan to implement an Azure Machine Learning solution. You have the following requirements:



- Run a Jupyter notebook to interactively train a machine learning model.
- Deploy assets and workflows for machine learning proof of concept by using scripting rather than custom programming.

You need to select a development technique for each requirement

Which development technique should you use? To answer, select the appropriate options in the answer area.

NOTE: Each correct selection is worth one point.

Hot Area:

| Requirement                                                                                                          | Development tool                                                                                                                                                                                                              |
|----------------------------------------------------------------------------------------------------------------------|-------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|
| Run a Jupyter notebook to interactively train a machine learning model.                                              | <ul style="list-style-type: none"> <li>Azure Machine Learning Python SDK</li> <li>Azure CLI</li> <li>Azure Machine Learning studio</li> <li>Azure Machine Learning Python SDK</li> <li>Azure Machine Learning REST</li> </ul> |
| Deploy assets and workflows for machine learning proof of concept by using scripting rather than custom programming. | <ul style="list-style-type: none"> <li>Azure CLI</li> <li>Azure CLI</li> <li>Azure Machine Learning studio</li> <li>Azure Machine Learning Python SDK</li> <li>Azure Machine Learning REST</li> </ul>                         |

Answer Area:

| Requirement                                                                                                          | Development tool                                                                                                                                                                                                              |
|----------------------------------------------------------------------------------------------------------------------|-------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|
| Run a Jupyter notebook to interactively train a machine learning model.                                              | <ul style="list-style-type: none"> <li>Azure Machine Learning Python SDK</li> <li>Azure CLI</li> <li>Azure Machine Learning studio</li> <li>Azure Machine Learning Python SDK</li> <li>Azure Machine Learning REST</li> </ul> |
| Deploy assets and workflows for machine learning proof of concept by using scripting rather than custom programming. | <ul style="list-style-type: none"> <li>Azure CLI</li> <li>Azure CLI</li> <li>Azure Machine Learning studio</li> <li>Azure Machine Learning Python SDK</li> <li>Azure Machine Learning REST</li> </ul>                         |

Section:

Explanation:

#### QUESTION 43

You build a data pipeline in an Azure Machine Learning workspace by using the Azure Machine Learning SDK for Python.

You need to run a Python script as a pipeline step.

Which two classes could you use? Each correct answer presents a complete solution.

NOTE: Each correct selection is worth one point.

- A. PythonScriptStep
- B. AutoMLStep
- C. CommandStep

D. StepRun

Correct Answer: A, C

Section:

#### QUESTION 44

HOTSPOT

You manage an Azure Machine Learning workspace.

You must define the execution environments for your jobs and encapsulate the dependencies for your code.

You need to configure the environment from a Docker build context.

How should you complete the code segment? To answer, select the appropriate option in the answer area.

NOTE: Each correct selection is worth one point.

Answer:

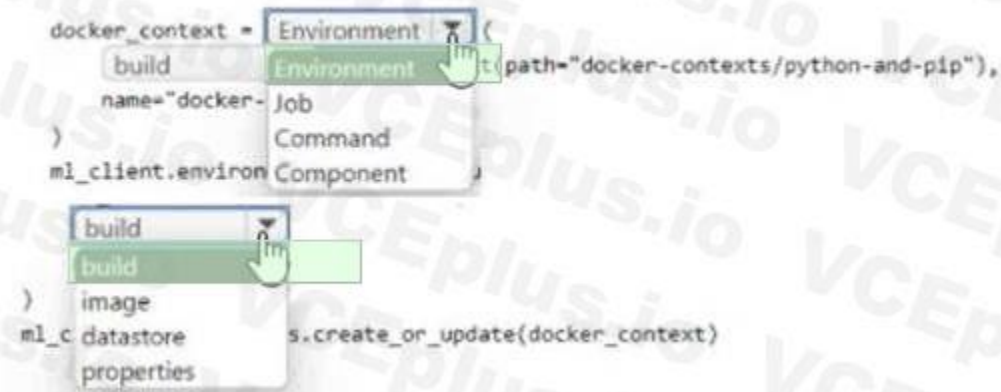
Hot Area:

Answer Area



Answer Area:

Answer Area



Section:

Explanation:

Answer Area

```
docker_context = Environment (
 build
 name="docker-context "
)
ml_client.environments.create_or_update(docker_context)
```

**QUESTION 45**

You have a dataset that contains records of patients tested for diabetes. The dataset includes the patient's age. You plan to create an analysis that will report the mean age value from the differentially private data derived from the dataset. You need to identify the epsilon value to use in the analysis that minimizes the risk of exposing the actual data. Which epsilon value should you use?

- A. -1.5
- B. -0.5
- C. 0.5
- D. 1.5

**Correct Answer: C**

**Section:**

**QUESTION 46**

You create a binary classification model. You use the Fairlearn package to assess model fairness. You must eliminate the need to retrain the model. You need to implement the Fairlearn package. Which algorithm should you use?

- A. fairlearn.reductions.ExponentiatedGradient
- B. fairlearn.reductions.GridSearch
- C. fairlearn.postprocessing.ThresholdOptimizer
- D. fairlearn.preprocessing.CorrelationRemover

**Correct Answer: D**

**Section:**



**QUESTION 47**

**HOTSPOT**

You manage an Azure Machine Learning workspace. You configure an automated machine learning regression training job by using the Azure Machine Learning Python SDK v2. You configure the regression job by using the following script:

```
regression_job.set_limits(
 timeout_minutes = 60,
 max_concurrent_trials = 5,
 enable_early_termination = True
)
```

For each of the following statements, select Yes if the statement is true. Otherwise, select No.

**Hot Area:**

**Answer Area**

| Statements                                                                              | Yes                   | No                    |
|-----------------------------------------------------------------------------------------|-----------------------|-----------------------|
| The job is terminated if the score is not improving in a specific number of iterations. | <input type="radio"/> | <input type="radio"/> |
| A maximum of five AutoML trials are run in parallel during the regression job.          | <input type="radio"/> | <input type="radio"/> |
| One AutoML trial can run for 60 minutes before it is terminated.                        | <input type="radio"/> | <input type="radio"/> |
| The AutoML trial run can take up to 1 month before it terminates.                       | <input type="radio"/> | <input type="radio"/> |

**Answer Area:**

**Answer Area**

**Statements**

- The job is terminated if the score is not improving in a specific number of iterations.
- A maximum of five AutoML trials are run in parallel during the regression job.
- One AutoML trial can run for 60 minutes before it is terminated.
- The AutoML trial run can take up to 1 month before it terminates.

| Yes                              | No                               |
|----------------------------------|----------------------------------|
| <input type="radio"/>            | <input checked="" type="radio"/> |
| <input checked="" type="radio"/> | <input type="radio"/>            |
| <input checked="" type="radio"/> | <input type="radio"/>            |
| <input type="radio"/>            | <input checked="" type="radio"/> |

**Section:**  
**Explanation:**

**QUESTION 48**

DRAG DROP

You manage an Azure Machine Learning workspace. You train a model named model1.

You must identify the features to modify for a differing model prediction result.

You need to configure the Responsible AI (RAI) dashboard for model1.

Which three actions should you perform in sequence? To answer move the appropriate actions from the list of actions to the answer area and arrange them in the correct order.

**Select and Place:**

**Actions**

- Add the explanation component to the Responsible AI Insights dashboard.
- Add the error analysis component to the Responsible AI Insights dashboard.
- Add the causal component to the Responsible AI Insights dashboard.
- Load and configure the Responsible AI Insights dashboard constructor component.
- Add the counterfactuals component to the Responsible AI Insights dashboard.
- Use the Gather Responsible AI Insights dashboard component to present the dashboard.



**Answer Area**



**Correct Answer:**

**Actions**

- Add the explanation component to the Responsible AI Insights dashboard.
- Add the error analysis component to the Responsible AI Insights dashboard.
- Add the causal component to the Responsible AI Insights dashboard.
- 
- 



**Answer Area**

- Load and configure the Responsible AI Insights dashboard constructor component.
- Add the counterfactuals component to the Responsible AI Insights dashboard.
- Use the Gather Responsible AI Insights dashboard component to present the dashboard.



**Section:**  
**Explanation:**

**QUESTION 49**

You have an Azure Machine Learning (ML) model deployed to an online endpoint.

You need to review container logs from the endpoint by using Azure ML Python SDK v2. The logs must include the console log from the inference server with print/log statements from the models scoring script.

What should you do first?

- A. Create an instance of the the MLClient class.
- B. Create an instance of the OnlineDeploymentOperations class.



- C. Connect by using SSH to the inference server.
- D. Connect by using Docker tools to the inference server.

**Correct Answer: A**

**Section:**

**QUESTION 50**

You train and publish a machine learning model.

You need to run a pipeline that retrains the model based on a trigger from an external system.

What should you configure?

- A. Azure Data Catalog
- B. Azure Batch
- C. Azure Logic App

**Correct Answer: C**

**Section:**

**QUESTION 51**

DRAG DROP

You manage an Azure Machine Learning workspace named workspace1 with a compute instance named compute1. You connect to compute1 by using a terminal window from workspace1. You create a file named "requirements.txt" containing Python dependencies to include Jupyter.

You need to add a new Jupyter kernel to compute1.

Which four commands should you use? To answer, move the appropriate actions from the list of actions to the answer area and arrange them in the correct order.

**Select and Place:**

**Commands**

- jupyter run
- conda create -n "python\_env"
- conda activate "python\_env"
- conda install -r "requirements.txt"
- ipython kernel install --user --name="python\_env"

**Answer Area**



**Correct Answer:**

### Commands

```
jupyter run
```

### Answer Area

```
conda create -n "python_env"
```

```
conda activate "python_env"
```

```
conda install -r "requirements.txt"
```

```
ipython kernel install --user --name="python_env"
```

### Section:

### Explanation:

### QUESTION 52

You create a workspace to include a compute instance by using Azure Machine Learning Studio. You are developing a Python SDK v2 notebook in the workspace. You need to use Intellisense in the notebook. What should you do?

- A. Start the compute instance.
- B. Run a %pip magic function on the compute instance.
- C. Run a !pip magic function on the compute instance.
- D. Stop the compute instance.

**Correct Answer: B**

### Section:

### QUESTION 53

### HOTSPOT

You use Azure Machine Learning to train a machine learning model.

You use the following training script in Python to perform logging:

```
import mlflow
mlflow.log_metric("accuracy", float(val_accuracy))
```

You must use a Python script to define a sweep job.

You need to provide the primary metric and goal you want hyperparameter tuning to optimize.

NOTE: Each correct selection is worth one point.

### Hot Area:




Answer Area

```
from azure.ai.ml.sweep import Uniform, Choice
command_job_for_sweep = command_job(
 learning_rate=Uniform(min_value=0.05, max_value=0.1),
 batch_size=Choice(values=[16, 32, 64, 128]),
)
sweep_job = command_job_for_sweep.sweep(
 compute="cpu-cluster",
 sampling_algorithm="bayesian",
 primary_metric="val_accuracy",
 goal="Maximize",
)
```



Answer Area:  
Answer Area

```
from azure.ai.ml.sweep import Uniform, Choice
command_job_for_sweep = command_job(
 learning_rate=Uniform(min_value=0.05, max_value=0.1),
 batch_size=Choice(values=[16, 32, 64, 128]),
)
sweep_job = command_job_for_sweep.sweep(
 compute="cpu-cluster",
 sampling_algorithm="bayesian",
 primary_metric="val_accuracy",
 goal="Maximize",
)
```



Section:  
Explanation:  
Answer Area

```
from azure.ai.ml.sweep import Uniform, Choice
command_job_for_sweep = command_job(
 learning_rate=Uniform(min_value=0.05, max_value=0.1),
 batch_size=Choice(values=[16, 32, 64, 128]),
)
sweep_job = command_job_for_sweep.sweep(
 compute="cpu-cluster",
 sampling_algorithm="bayesian",
 primary_metric="val_accuracy",
 goal="Maximize",
)
```

QUESTION 54  
HOTSPOT

You manage an Azure Machine Learning workspace named workspacel by using the Python SDK v2.

You must register datastores in workspace for Azure Blob and Azure Data Lake Gen2 storage to meet the following requirements:

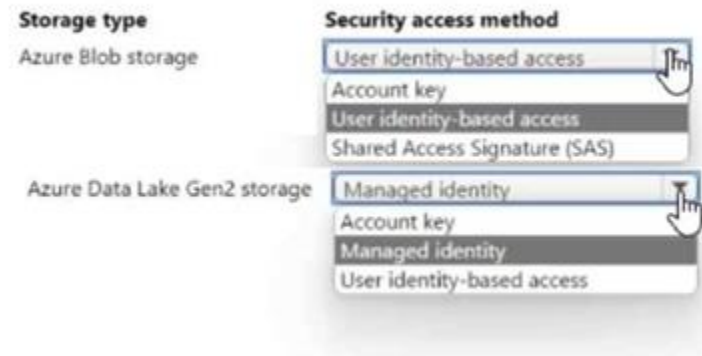
- Data scientists accessing the datastore must have the same level of access.
- Access must be restricted to specified containers or folders.

You need to configure a security access method used to register the Azure Blob and Azure Data lake Gen? storage in workspace. Which security access method should you configure? To answer, select the appropriate options in the answer area.

NOTE: Each correct selection is worth one point.

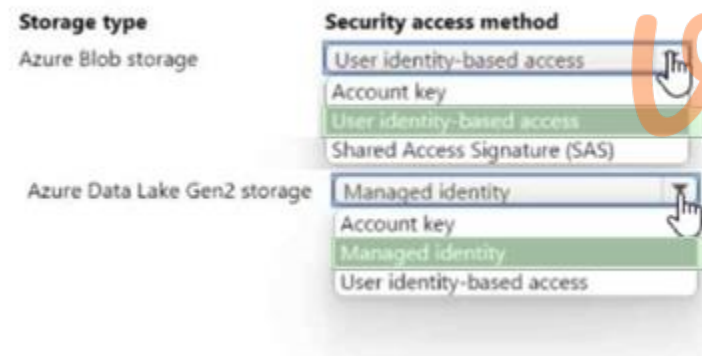
**Hot Area:**

Answer Area



**Answer Area:**

Answer Area



**vdumps**

**Section:**

**Explanation:**

**QUESTION 55**

**HOTSPOT**

You are creating data wrangling and model training solutions in an Azure Machine Learning workspace.

You must use the same Python notebook to perform both data wrangling and model training.

You need to use the Azure Machine Learning Python SDK v2 to define and configure the Synapse Spark pool asynchronously in the workspace as dedicated compute

How should you complete the code segment? To answer, select the appropriate options in the answer area.

NOTE: Each correct selection is worth one point.

**Hot Area:**



Answer Area

synapse\_compute =  (name=synapse\_name,  
resource\_id=synapse\_resource\_id)  (name=synapse\_name,  
resource\_id=synapse\_resource\_id)

(synapse\_compute)

Answer Area:

Answer Area

synapse\_compute =  (name=synapse\_name,  
resource\_id=synapse\_resource\_id)  (name=synapse\_name,  
resource\_id=synapse\_resource\_id)

(synapse\_compute)

Section:

Explanation:

Answer Area



synapse\_compute =  (name=synapse\_name,  
resource\_id=synapse\_resource\_id)

(synapse\_compute)

QUESTION 56

HOTSPOT

You create an Azure Machine Learning workspace

You are developing a Python SDK v2 notebook to perform custom model training in the workspace. The notebook code imports all required packages.

You need to complete the Python SDK v2 code to include a training script, environment, and compute information.

How should you complete the code? To answer, select the appropriate options in the answer area.

NOTE: Each correct selection is worth one point

Hot Area:

Answer Area

```
var1 = Environment.from_config(credential=DefaultAzureCredential())
var2 = command
inputs=dict(
 registered_model_name="sample_model",
)
code="/src/"
command="python main.py --registered_model_name
${(inputs.registered_model_name)}",
environment="aml-sclit-learn@latest",
compute="cpu-cluster",
experiment_name="sample_experiment",
display_name="sample_experiment",
)
var1.create_or_update(var2)
```

Answer Area:

Answer Area

```
var1 = Environment.from_config(credential=DefaultAzureCredential())
var2 = command
inputs=dict(
 registered_model_name="sample_model",
)
code="/src/"
command="python main.py --registered_model_name
${(inputs.registered_model_name)}",
environment="aml-sclit-learn@latest",
compute="cpu-cluster",
experiment_name="sample_experiment",
display_name="sample_experiment",
)
var1.create_or_update(var2)
```



Section:

Explanation:

QUESTION 57

DRAG DROP

You are creating a machine learning model that can predict the species of a penguin from its measurements. You have a file that contains measurements for free species of penguin in comma delimited format.

The model must be optimized for area under the received operating characteristic curve performance metric averaged for each class.

You need to use the Automated Machine Learning user interface in Azure Machine Learning studio to run an experiment and find the best performing model.

Which five actions should you perform in sequence? To answer, move the appropriate actions from the list of actions to the answer area and arrange them in the correct order.

Select and Place:

**Actions**

Select the **Regression** task type.

Set the Primary metric configuration setting to **AUC Weighted**.

Create and select a new dataset by uploading the comma-delimited file of penguin data.

Select the **Classification** task type.

Set the Primary metric configuration setting to **Accuracy**.

Configure the automated machine learning run by selecting the experiment name, target column, and compute target.

Run the automated machine learning experiment and review the results.



**Answer area**

Create and select a new dataset by uploading the comma-delimited file of penguin data.

Select the **Classification** task type.

Set the Primary metric configuration setting to **Accuracy**.

Configure the automated machine learning run by selecting the experiment name, target column, and compute target.

Run the automated machine learning experiment and review the results.



**Correct Answer:**

**Actions**

Select the **Regression** task type.

Set the Primary metric configuration setting to **AUC Weighted**.



**Answer area**

Create and select a new dataset by uploading the comma-delimited file of penguin data.

Select the **Classification** task type.

Set the Primary metric configuration setting to **Accuracy**.

Configure the automated machine learning run by selecting the experiment name, target column, and compute target.

Run the automated machine learning experiment and review the results.



**Section:**

**Explanation:**

Create and select an new dataset by uploading the comma-delimited file of penguin data.  
 Select the Classification task type.  
 Set the Primary metric configuration setting to Accuracy.  
 Configure the automated machine learning run by selecting the experiment name, target cloumn, and compute target.  
 Run the automated machine learning experiment and review the results.

**QUESTION 58**

**HOTSPOT**

You manage are Azure Machine Learning workspace by using the Python SDK v2.  
 You must create an automated machine learning job to generate a classification model by using data files stored in Parquet format. You must configure an auto scaling compute target and a data asset for the job.  
 You need to configure the resources for the job.  
 Which resource configuration should you use? to answer, select the appropriate options in the answer area.  
 NOTE: Each correct selection is worth one point.

**Hot Area:**

Answer Area

| Resource type  | Resource value                                                                                                                        |
|----------------|---------------------------------------------------------------------------------------------------------------------------------------|
| Compute target | <input type="text" value="Azure Databricks"/><br>Azure Databricks<br>Azure Data Lake Analytics<br>Azure HDInsight<br>Azure Databricks |
| Data asset     | <input type="text" value="uri_folder"/><br>uri_folder<br>mitable<br>uri_file<br>uri_folder                                            |

Answer Area:

Answer Area

| Resource type  | Resource value                                                                                                                        |
|----------------|---------------------------------------------------------------------------------------------------------------------------------------|
| Compute target | <input type="text" value="Azure Databricks"/><br>Azure Databricks<br>Azure Data Lake Analytics<br>Azure HDInsight<br>Azure Databricks |
| Data asset     | <input type="text" value="uri_folder"/><br>uri_folder<br>mitable<br>uri_file<br>uri_folder                                            |

Section:

Explanation:

QUESTION 59

You use the Azure Machine Learning SDK v2 for Python and notebooks to train a model. You use Python code to create a compute target, an environment, and a training script. You need to prepare information to submit a training job.

Which class should you use?

- A. MLClient
- B. command
- C. BuildContext
- D. EndpointConnection

Correct Answer: B

Section:

QUESTION 60

HOTSPOT

You load data from a notebook in an Azure Machine Learning workspace into a pandas DataFrame. The data contains 10,000 records. Each record consists of 10 columns.

You must identify the number of missing values in each of the columns.

You need to complete the Python code that will return the number of missing values in each of the columns.

Which code segments should you use? To answer, select the appropriate options in the answer area.

NOTE: Each correct selection is worth one point.

Hot Area:



Answer Area

|        | df.count() |
|--------|------------|
| values | 10         |
| index  | 0          |
| shape  | 10         |
| values | 100,000    |

Answer Area:

Answer Area

|        | df.count() |
|--------|------------|
| values | 10         |
| index  | 0          |
| shape  | 10         |
| values | 100,000    |

Section:

Explanation:

#### QUESTION 61

You create an Azure Machine learning workspace.

You use the Azure Machine Learning Python SDK v2 to define the search space for concrete hyperparameters. The hyper parameters must consist of a list of predetermined, comma-separated.

You need to import the class from the azure ai ml. sweep package used to create the list of values.

Which class should you import?

- A. Uniform
- B. Normal
- C. Randint
- D. Choice

Correct Answer: D

Section:

#### QUESTION 62

You create an Azure Machine learning workspace named workspace1. The workspace contains a Python SDK v2 notebook that uses Mallow to correct model coxiong men's anal arracks from your local computer.

You must reuse the notebook to run on Azure Machine Learning compute instance in workspace.

You need to commit to log training and artifacts from your data science code.

What should you do?

- A. Configure the tracking URL.
- B. Instantiate the MLClient class.
- C. Log in to workspace1.
- D. Instantiate the job class.

Correct Answer: A

Section:

#### QUESTION 63

HOTSPOT

You manage an Azure Machine Learning workspace named workspace 1 with a compute instance named computet.

You must remove a kernel named kernel 1 from computet1. You connect to compute 1 by using noa terminal window from workspace 1.



You need to enter a command in the terminal window to remove kernel 1.  
Which command should you use? To answer, select the appropriate options in the answer area.  
NOTE: Each correct selection it worth one point.

**Hot Area:**

Answer Area

jupyter kernelspec uninstall kernel1

**Answer Area:**

Answer Area

jupyter kernelspec uninstall kernel1

**Section:**

**Explanation:**

**QUESTION 64**

You use Azure Machine Learning to train a model.  
You must use Bayesian sampling to tune hyperparameters.  
You need to select a learning\_rate parameter distribution.  
Which two distributions can you use? Each correct answer presents a complete solution.  
NOTE Each correct selection is worth one point.

- A. Normal
- B. Uniform
- C. Choice
- D. LogUniform
- E. QNormal

**Correct Answer: B, C**

**Section:**

**QUESTION 65**

**DRAG DROP**

You manage an Azure Machine Learning workspace named workspace1 and a Data Science Virtual Machine (DSVM) named DSMV1.  
You must run an experiment in DSMV1 by using a Jupyter notebook and Python SDK v2 code. You must store metrics and artifacts in workspace 1. You start by creating Python SDK v2 code to import all required packages.  
You need to implement the Python SDK v2 code to store metrics and artifacts in workspace1.  
Which three actions should you perform in sequence? To answer, move the appropriate actions from the list of actions to the answer area and arrange them in the correct order.

**Select and Place:**



**Actions**

- Instantiate an object of the MLCClient class.
- Instantiate an object of the Output class.
- Retrieve the tracking URI of workspace1.
- Set the MLflow tracking URI.
- Set the URI parameter of the mlflow.projects.run method.

**Answer Area**

- 
- 
- 

**Correct Answer:**

**Actions**

- Instantiate an object of the MLCClient class.
- Instantiate an object of the Output class.

**Answer Area**

- Retrieve the tracking URI of workspace1.
- Set the MLflow tracking URI.
- Set the URI parameter of the mlflow.projects.run method.

**Section:**

**Explanation:**

Retrieve the tracking URI of workspace1.

Set the MLflow tracking URI.

Set the URI parameter of the mlflow.projects.run method.



**QUESTION 66**

You have the following Azure subscriptions and Azure Machine Learning service workspaces:

| Subscription                         | Workspace  | Comment                                                                                                                                 |
|--------------------------------------|------------|-----------------------------------------------------------------------------------------------------------------------------------------|
| 385bdf5-4cef-4ad4-b977-3f86d92727c9  | ml-default | This is the default subscription.                                                                                                       |
| 5a5891d1-557a-4234-9b83-2e90412b1068 | ml-project | The information required to uniquely identify this workspace is stored in the file config.json in the same folder as the Python script. |

You need to obtain a reference to the ml-project workspace

Solution: Run the following Python code.

```
from azure.ai.ml import MLCClient
ws = MLCClient.workspaces.get("ml-project")
```

Does the solution meet the goal?

A. Yes

B. No

**Correct Answer: A**

**Section:**

**QUESTION 67**

**HOTSPOT**

You are using hyperparameter tuning in Azure Machine Learning Python SDK v2 to train a model. You configure the hyperparameter tuning experiment by running the following code:

```

from azure.ai.ml.sweep import Normal, Uniform

command_job_for_sweep = command_job(
 learning_rate=Normal(10, 3),
 keep_probability=Uniform(0.05, 0.1),
 batch_size=Choice(values=[16, 32, 64, 128]),
 number_of_hidden_layers=Choice(range(3,5))
)

```

For each of the following statements select Yes if the statement is true. Otherwise, select No. NOTE: Each correct selection is worth one point.

Hot Area:  
Answer Area

| Statements                                                                                                                                          | Yes                   | No                    |
|-----------------------------------------------------------------------------------------------------------------------------------------------------|-----------------------|-----------------------|
| By defining sampling in this manner, every possible combination of the parameters will be tested.                                                   | <input type="radio"/> | <input type="radio"/> |
| Random values of the learning_rate parameter will be selected from a normal distribution with a mean of 10 and a standard deviation of 3.           | <input type="radio"/> | <input type="radio"/> |
| The keep_probability parameter value will always be either <b>0.05</b> or <b>0.1</b> .                                                              | <input type="radio"/> | <input type="radio"/> |
| Random values for the number_of_hidden_layers parameter will be selected from a normal distribution with a mean of 3 and a standard deviation of 5. | <input type="radio"/> | <input type="radio"/> |

Answer Area:  
Answer Area

| Statements                                                                                                                                          | Yes                              | No                               |
|-----------------------------------------------------------------------------------------------------------------------------------------------------|----------------------------------|----------------------------------|
| By defining sampling in this manner, every possible combination of the parameters will be tested.                                                   | <input type="radio"/>            | <input checked="" type="radio"/> |
| Random values of the learning_rate parameter will be selected from a normal distribution with a mean of 10 and a standard deviation of 3.           | <input checked="" type="radio"/> | <input type="radio"/>            |
| The keep_probability parameter value will always be either <b>0.05</b> or <b>0.1</b> .                                                              | <input type="radio"/>            | <input checked="" type="radio"/> |
| Random values for the number_of_hidden_layers parameter will be selected from a normal distribution with a mean of 3 and a standard deviation of 5. | <input type="radio"/>            | <input checked="" type="radio"/> |

Section:



**Explanation:**

**QUESTION 68**

You are authoring a notebook in Azure Machine Learning studio.

You must install packages from the notebook into the currently running kernel. The installation must be limited to the currently running kernel only.

You need to install the packages.

Which magic function should you use?

- A. !pip
- B. %load
- C. !conda
- D. %pip

**Correct Answer: D**

**Section:**

**QUESTION 69**

You plan to use automated machine learning by using Azure Machine Learning Python SDK v2 to train a regression model. You have data that has features with missing values, and categorical features with few distinct values.

You need to control whether automated machine learning automatically imputes missing values and encode categorical features as part of the training task. Which enemy of the autumn package should you use?

- A. ForecastHorizonMode
- B. RegressionPrimaryMetrics
- C. RegressionModels
- D. FeaturizationMode

**Correct Answer: D**

**Section:**

**QUESTION 70**

**HOTSPOT**

You download a .csv file from a notebook in an Azure Machine Learning workspace to a data/sample.csv folder on a compute instance. The file contains 10,000 records. You must generate the summary statistics for the data in the file. The statistics must include the following for each numerical column:

- \* number of non-empty values
- \* average value
- \* standard deviation
- \* minimum and maximum values
- \* 25th, 50th, and 75th percentiles

You need to complete the Python code that will generate the summary statistics.

Which code segments should you use? To answer, select the appropriate options in the answer area.

NOTE: Each correct selection is worth one point.

**Hot Area:**



Answer Area

```
import: CSV
CSV
numpy
pandas

datasample = lib.read_csv(data/sample.csv)
datasample. describe ()
describe
query
rank
```

Answer Area:

Answer Area

```
import: CSV
CSV
numpy
pandas

datasample = lib.read_csv(data/sample.csv)
datasample. describe ()
describe
query
rank
```

Section:

Explanation:

QUESTION 71

HOTSPOT

You perform hyper parameter tuning with Azure Machine Learning.

You create the following Python code:

```
from azure.ai.ml.sweep import Normal, Uniform

command_job_for_sweep = command_job(
 learning_rate=Normal(mu=10, sigma=3),
 keep_probability=Uniform(min_value=0.05, max_value=0.1),
)
```

For each of the following statements, select Yes if the statement is true. Otherwise, select No.

Hot Area:

**Answer Area**

| Statements                                                                                                                      | Yes                   | No                    |
|---------------------------------------------------------------------------------------------------------------------------------|-----------------------|-----------------------|
| The code defines a search space by using the <code>learning_rate</code> and <code>keep_probability</code> parameters.           | <input type="radio"/> | <input type="radio"/> |
| The logarithm of the <code>learning_rate</code> parameter has a normal distribution.                                            | <input type="radio"/> | <input type="radio"/> |
| The <code>keep_probability</code> parameter has a uniform distribution with a minimum value of 0.05 and a maximum value of 0.1. | <input type="radio"/> | <input type="radio"/> |

**Answer Area:**

**Answer Area**

| Statements                                                                                                                      | Yes                              | No                    |
|---------------------------------------------------------------------------------------------------------------------------------|----------------------------------|-----------------------|
| The code defines a search space by using the <code>learning_rate</code> and <code>keep_probability</code> parameters.           | <input checked="" type="radio"/> | <input type="radio"/> |
| The logarithm of the <code>learning_rate</code> parameter has a normal distribution.                                            | <input checked="" type="radio"/> | <input type="radio"/> |
| The <code>keep_probability</code> parameter has a uniform distribution with a minimum value of 0.05 and a maximum value of 0.1. | <input checked="" type="radio"/> | <input type="radio"/> |

**Section:**

**Explanation:**

**QUESTION 72**

**HOTSPOT**

You create an Azure Machine Learning model to include model files and a scoring script. You must deploy the model. The deployment solution must meet the following requirements:

- \* Provide near real-time inferencing.
- \* Enable endpoint and deployment level cost estimates.
- \* Support logging to Azure Log Analytics.

You need to configure the deployment solution.

What should you configure? To answer, select the appropriate options in the answer area.

NOTE: Each correct selection is worth one point.

**Hot Area:**



**Answer Area**

| Requirement          | Value                                                                                                                                  |
|----------------------|----------------------------------------------------------------------------------------------------------------------------------------|
| Endpoint type        | <input type="text" value="Managed online"/><br>Managed online<br>Kubernetes online<br>Batch                                            |
| Deployment component | <input type="text" value="Docker image"/><br>Docker image<br>Azure Container Instances (ACI)<br>Azure Kubernetes Service (AKS) cluster |

**Answer Area:**

**Answer Area**

| Requirement          | Value                                                                                                                                  |
|----------------------|----------------------------------------------------------------------------------------------------------------------------------------|
| Endpoint type        | <input type="text" value="Managed online"/><br>Managed online<br>Kubernetes online<br>Batch                                            |
| Deployment component | <input type="text" value="Docker image"/><br>Docker image<br>Azure Container Instances (ACI)<br>Azure Kubernetes Service (AKS) cluster |

**Section:**

**Explanation:**

**QUESTION 73**

You run Azure Machine Learning training experiments. The training scripts directory contains 100 files that includes a file named. amlignore. The directory also contains subdirectories named. /outputs and./logs. There are 20 files in the training scripts directory that must be excluded from the snapshot to the compute targets. You create a file named. gift ignore in the root of the directory. You add the names of the 20 files to the. gift ignore file. These 20 files continue to be copied to the compute targets. You need to exclude the 20 files. What should you do?

- A. Add the contents of the file named. amlignore to the file named. gift ignore.
- B. Move the file named. gift ignore to the. /logs directory.
- C. Copy the contents of the file named. gift ignore to the file named. amlignore.
- D. Move the file named. gift ignore to the. /outputs directory.

**Correct Answer: C**

**Section:**

**QUESTION 74**



## HOTSPOT

You manage an Azure Machine Learning workspace named workspace1.

You must register an Azure Blob storage datastore in workspace1 by using an access key. You develop Python SDK v2 code to import all modules required to register the datastore.

You need to complete the Python SDK v2 code to define the datastore.

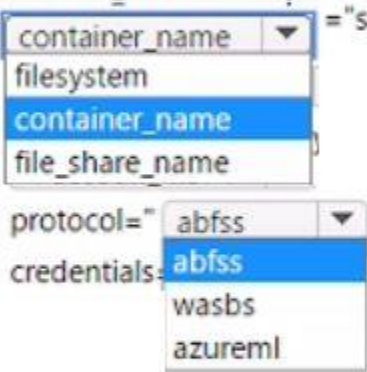
How should you complete the code? To answer, select the appropriate options in the answer area.

NOTE: Each correct selection is worth one point.

### Hot Area:

#### Answer Area

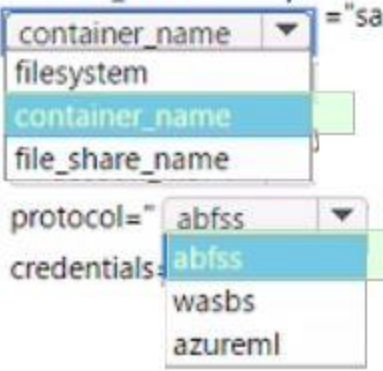
```
store = AzureBlobDatastore(
 name="blobdatastore",
 description="Sample blob datastore",
 account_name="sampleaccount1",
 container_name = "samplecontainer1",
 protocol="abfss",
 credentials=AccountKeyCredentials(
 "account_key":
 "6L4VGVXly30FM/RwmfLGppGj+xNEtUwDVBNPYwYDQDT1E+SlrZcjP1a+xCHI+D9BTwLnYKhrERo+AStap8yKw=="
),
)
```



### Answer Area:

## Answer Area

```
store = AzureBlobDatastore(
 name="blobdatastore",
 description="Sample blob datastore",
 account_name="sampleaccount1",
 container_name="samplecontainer1",
 protocol="abfss",
 credentials=AccountKeyCredentials(
 "account_key":
 "6L4VGVXly30FM/RwmfLGppGj+xNEtfUwDVBNPYwYDQDT1E+S1rZcjP1a+xCHI+D9BTwLnYKhrERo+AStap8yKw=="
)
)
```

A screenshot of a code editor showing a dropdown menu for the 'container\_name' property of an AzureBlobDatastore object. The dropdown menu is open, showing options: 'filesystem', 'container\_name', and 'file\_share\_name'. The 'container\_name' option is selected. Below it, another dropdown menu is open for the 'credentials' property, showing options: 'abfss', 'wasbs', and 'azureml'. The 'abfss' option is selected.

### Section:

### Explanation:

#### QUESTION 75

Note: This question is part of a series of questions that present the same scenario. Each question in the series contains a unique solution that might meet the stated goals. Some question sets might have more than one correct solution, while others might not have a correct solution.

After you answer a question in this section, you will NOT be able to return to it as a result, these questions will not appear in the review screen.

You train and register an Azure Machine Learning model.

You plan to deploy the model to an online end point.

You need to ensure that applications will be able to use the authentication method with a non-expiring artifact to access the model.

Solution:

Create a Kubernetes online endpoint and set the value of its auth-mode parameter to amyI Token. Deploy the model to the online endpoint.

Does the solution meet the goal?

A. Yes

B. No

**Correct Answer: B**

**Section:**

#### QUESTION 76

You manage an Azure Machine Learning workspace.

You must log multiple metrics by using MLflow.

You need to maximize logging performance.

What are two possible ways to achieve this goal? Each correct answer presents a complete solution.

NOTE: Each correct selection is worth one point.

The logo for 'Vdumps' features a stylized orange 'V' followed by the word 'dumps' in a grey, lowercase, sans-serif font.

- A. MLflowClient.log\_batch
- B. mlflowlog\_metrics
- C. mlflow.log\_param
- D. mlflow.log\_metric

**Correct Answer: A, B**

**Section:**

#### QUESTION 77

You manage an Azure Machine Learning workspace.

You need to define an environment from a Docker image by using the Azure Machine Learning Python SDK v2.

Which parameter should you use?

- A. conda\_file
- B. image
- C. build
- D. properties

**Correct Answer: B**

**Section:**

#### QUESTION 78

You use Azure Machine Learning studio to analyze an mltable data asset containing a decimal column named column1. You need to verify that the column1 values are normally distributed.

Which statistic should you use?

- A. Max
- B. Type
- C. Profile
- D. Mean

**Correct Answer: C**

**Section:**

#### QUESTION 79

You create an Azure Machine Learning workspace named woricspace1. The workspace contains a Python SDK v2 notebook that uses MLflow to collect model training metrics and artifacts from your local computer.

You must reuse the notebook to run on Azure Machine Learning compute instance in workspace1.

You need to continue to log metrics and artifacts from your data science code.

What should you do?

- A. Configure the tracking URI.
- B. Instantiate the job class.
- C. Log into workspace'!
- D. Instantiate the MLClient class.

**Correct Answer: A**

**Section:**



**QUESTION 80**

You use Azure Machine Learning Designer to load the following datasets into an experiment:

Data set 1

| Age | Length | Width |
|-----|--------|-------|
| 3   | 22     | 13    |
| 7   | 11     | 96    |
| 18  | 32     | 85    |

Dataset 2

| Age | Length | Width |
|-----|--------|-------|
| 11  | 101    | 65    |
| 6   | 98     | 23    |
| 33  | 22     | 54    |
| 17  | 52     | 12    |

You need to create a dataset that has the same columns and header row as the input datasets and contains all rows from both input datasets.

Solution: Use the Apply Transformation component.

Does the solution meet the goal?

- A. Yes
- B. No

**Correct Answer: B**

**Section:**

**QUESTION 81**

You manage an Azure Machine Learning workspace. You design a training job that is configured with a serverless compute. The serverless compute must have a specific instance type and count. You need to configure the serverless compute by using Azure Machine Learning Python SDK v2. What should you do?

- A. Specify the compute name by using the compute parameter of the command job
- B. Configure the tier parameter to Dedicated VM.
- C. Initialize and specify the ResourceConfiguration class
- D. Initialize AmlCompute class with size and type specification.

**Correct Answer: C**

**Section:**

**QUESTION 82**

HOTSPOT

You manage an Azure Machine Learning workspace. You create an experiment named experiment1 by using the Azure Machine Learning Python SDK v2 and MLflow. You are reviewing the results of experiment1 by using the following code segment:

```
runs = mlflow.search_runs(
 experiment_names=["experiment1"],
 max_results=5,
 order_by=["start_time ASC"])
```

For each of the following statements, Select Yes if the statement is true. Otherwise, select No.

**Hot Area:**



Answer Area

Statements

Aborted runs are returned.

Yes

No

The runs returned have been completed with errors.

The jobs that are returned have been canceled or killed by the user or system.

All metrics and their values are returned for the returned experiment runs.

Answer Area:

Answer Area

Statements

Aborted runs are returned.

Yes

No

The runs returned have been completed with errors.

The jobs that are returned have been canceled or killed by the user or system.

All metrics and their values are returned for the returned experiment runs.

Section:

Explanation:

