

Microsoft

Exam Questions DP-100

Designing and Implementing a Data Science Solution on Azure



NEW QUESTION 1

- (Exam Topic 3)

You have an Azure Machine Learning workspace that contains a training cluster and an inference cluster. You plan to create a classification model by using the Azure Machine Learning designer.

You need to ensure that client applications can submit data as HTTP requests and receive predictions as responses.

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.

Actions

Create a real-time inference pipeline and run the pipeline on the compute cluster.

Create a batch inference pipeline and run the pipeline on the compute cluster.

Deploy a service to the compute cluster.

Create a pipeline that trains a classification model and run the pipeline on the compute cluster.

Deploy a service to the inference cluster.

Answer area

- A. Mastered
- B. Not Mastered

Answer: A

Explanation:

Actions

Create a real-time inference pipeline and run the pipeline on the compute cluster.

Create a batch inference pipeline and run the pipeline on the compute cluster.

Deploy a service to the compute cluster.

Create a pipeline that trains a classification model and run the pipeline on the compute cluster.

Deploy a service to the inference cluster.

Answer area

Create a pipeline that trains a classification model and run the pipeline on the compute cluster.

Create a batch inference pipeline and run the pipeline on the compute cluster.

Deploy a service to the inference cluster.

NEW QUESTION 2

- (Exam Topic 3)

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.

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

- A. Mastered

B. Not Mastered

Answer: A

Explanation:

Graphical user interface, text, application Description automatically generated
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>

NEW QUESTION 3

- (Exam Topic 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 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

Answer: B

Explanation:

train_step is missing. Reference:
<https://docs.microsoft.com/en-us/python/api/azureml-pipeline-core/azureml.pipeline.core.pipelinedata?view=azu>

NEW QUESTION 4

- (Exam Topic 3)
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An IT department creates the following Azure resource groups and resources:

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

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

Answer: B

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>

NEW QUESTION 5

- (Exam Topic 3)

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 Kubernetes Service (AKS) inference cluster
- B. Azure Machine Learning compute cluster
- C. Azure Container Instance (ACI)
- D. attached Azure Databricks cluster

Answer: C

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>

NEW QUESTION 6

- (Exam Topic 3)

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. [spearman_correlation
- C. AUC.weighted
- D. accuracy
- E. normalized_root_mean_squared_error

Answer: C

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.

Reference:

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

NEW QUESTION 7

- (Exam Topic 3)

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. Register a new version of the model and update the entry script to load the new version of the model from its registered path.
- B. Modify the AKS service deployment configuration to enable application insights and re-deploy to AKS.
- C. Create an Azure Container Instances (ACI) web service deployment configuration and deploy the model on ACI.
- D. Add a breakpoint to the first line of the entry script and redeploy the service to AKS.
- E. Create a local web service deployment configuration and deploy the model to a local Docker container.

Answer: C

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).

NEW QUESTION 8

- (Exam Topic 3)

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.

Answer: A

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>

NEW QUESTION 9

- (Exam Topic 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.

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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

Answer: B

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>

NEW QUESTION 10

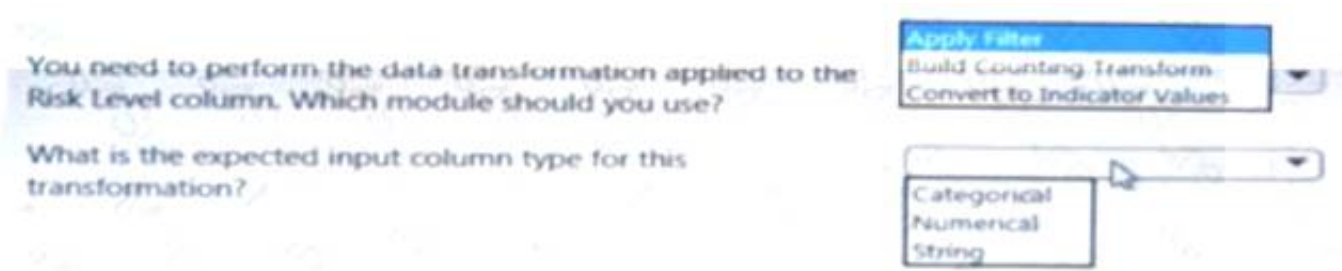
- (Exam Topic 3)

You are developing a machine learning experiment by using Azure. The following images show the input and output of a machine learning experiment:



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

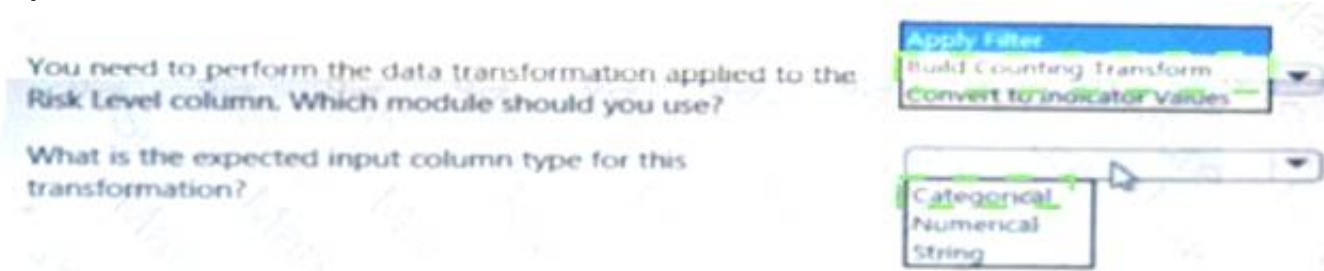
NOTE: Each correct selection is worth one point.



- A. Mastered
- B. Not Mastered

Answer: A

Explanation:



NEW QUESTION 10

- (Exam Topic 3)

You are developing deep learning models to analyze semi-structured, unstructured, and structured data types. You have the following data available for model building:

- > Video recordings of sporting events
 - > Transcripts of radio commentary about events
 - > Logs from related social media feeds captured during sporting events
- You need to select an environment for creating the model.

Which environment should you use?

- A. Azure Cognitive Services
- B. Azure Data Lake Analytics
- C. Azure HDInsight with Spark MLlib
- D. Azure Machine Learning Studio

Answer: A

Explanation:

Azure Cognitive Services expand on Microsoft's evolving portfolio of machine learning APIs and enable developers to easily add cognitive features – such as emotion and video detection; facial, speech, and vision recognition; and speech and language understanding – into their applications. The goal of Azure Cognitive Services is to help developers create applications that can see, hear, speak, understand, and even begin to reason. The catalog of services within Azure Cognitive Services can be categorized into five main pillars - Vision, Speech, Language, Search, and Knowledge.

References:

<https://docs.microsoft.com/en-us/azure/cognitive-services/welcome>

NEW QUESTION 14

- (Exam Topic 3)

A set of CSV files contains sales records. All the CSV files have the same data schema.

Each CSV file contains the sales record for a particular month and has the filename sales.csv. Each file is stored in a folder that indicates the month and year when the data was recorded. The folders are in an Azure blob container for which a datastore has been defined in an Azure Machine Learning workspace. The folders are organized in a parent folder named sales to create the following hierarchical structure:

```
/sales
  /01-2019
    /sales.csv
  /02-2019
    /sales.csv
  /03-2019
    /sales.csv
  ...
```

At the end of each month, a new folder with that month's sales file is added to the sales folder.

You plan to use the sales data to train a machine learning model based on the following requirements:

- > You must define a dataset that loads all of the sales data to date into a structure that can be easily converted to a dataframe.
- > You must be able to create experiments that use only data that was created before a specific previous month, ignoring any data that was added after that month.
- > You must register the minimum number of datasets possible.

You need to register the sales data as a dataset in Azure Machine Learning service workspace. What should you do?

- A. Create a tabular dataset that references the datastore and explicitly specifies each 'sales/mm-yyyy/ sales.csv' file every month
- B. Register the dataset with the name sales_dataset each month, replacing the existing dataset and specifying a tag named month indicating the month and year it

was registere

C. Use this dataset for all experiments.

D. Create a tabular dataset that references the datastore and specifies the path 'sales/*/sales.csv', register the dataset with the name sales_dataset and a tag named month indicating the month and year it was registered, and use this dataset for all experiments.

E. Create a new tabular dataset that references the datastore and explicitly specifies each 'sales/mm-yyyy/ sales.csv' file every mont

F. Register the dataset with the name sales_dataset_MM-YYYY each month with appropriate MM and YYYY values for the month and yea

G. Use the appropriate month-specific dataset for experiments.

H. Create a tabular dataset that references the datastore and explicitly specifies each 'sales/mm-yyyy/ sales.csv' fil

I. Register the dataset with the name sales_dataset each month as a new version and with a tag named month indicating the month and year it was registere

J. Use this dataset for all experiments, identifying the version to be used based on the month tag as necessary.

Answer: B

Explanation:

Specify the path. Example:

The following code gets the workspace existing workspace and the desired datastore by name. And then passes the datastore and file locations to the path parameter to create a new TabularDataset, weather_ds.

```
from azureml.core import Workspace, Datastore, Dataset
datastore_name = 'your datastore name'
```

```
# get existing workspace
```

```
workspace = Workspace.from_config()
```

```
# retrieve an existing datastore in the workspace by name
datastore = Datastore.get(workspace, datastore_name)
```

```
# create a TabularDataset from 3 file paths in datastore
datastore_paths = [(datastore, 'weather/2018/11.csv'), (datastore, 'weather/2018/12.csv'), (datastore, 'weather/2019/*.csv')]
```

```
weather_ds = Dataset.Tabular.from_delimited_files(path=datastore_paths)
```

NEW QUESTION 19

- (Exam Topic 3)

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 versionwith 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 mode

D. Do not register the new model.

E. Delete the existing model and register the new one with the same name.

Answer: B

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>

NEW QUESTION 21

- (Exam Topic 3)

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 inferencing 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

Answer: A

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-pipeli>

NEW QUESTION 22

- (Exam Topic 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.

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You are analyzing a numerical dataset which contains missing values in several columns.

You must clean the missing values using an appropriate operation without affecting the dimensionality of the feature set.

You need to analyze a full dataset to include all values.

Solution: Remove the entire column that contains the missing data point. Does the solution meet the goal?

A. Yes

B. No

Answer: B

Explanation:

Use the Multiple Imputation by Chained Equations (MICE) method. References: <https://www.ncbi.nlm.nih.gov/pmc/articles/PMC3074241/>
<https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/clean-missing-data>

NEW QUESTION 24

- (Exam Topic 3)

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.

```
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)
```

- A. Mastered
- B. Not Mastered

Answer: A

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>

NEW QUESTION 27

- (Exam Topic 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 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

Answer: B

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>

NEW QUESTION 32

- (Exam Topic 3)

You are conducting feature engineering to prepuce data for further analysis. The data includes seasonal patterns on inventory requirements.

You need to select the appropriate method to conduct feature engineering on the data. Which method should you use?

- A. Exponential Smoothing (ETS) function.
- B. One Class Support Vector Machine module
- C. Time Series Anomaly Detection module
- D. Finite Impulse Response (FIR) Filter module.

Answer: D

NEW QUESTION 34

- (Exam Topic 3)

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.

	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 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"/>

- A. Mastered
- B. Not Mastered

Answer: A

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>

NEW QUESTION 37

- (Exam Topic 3)

You are preparing to use the Azure ML SDK to run an experiment and need to create compute. You run the following code:

```
from azureml.core.compute import ComputeTarget, AmlCompute
from azureml.core.compute_target import ComputeTargetException
ws = Workspace.from_config()
cluster_name = 'aml-cluster'
try:
    training_compute = ComputeTarget(workspace=ws, name=cluster_name)
except ComputeTargetException:
    compute_config = AmlCompute.provisioning_configuration(vm_size='STANDARD_D2_V2', vm_priority='lowpriority',
max_nodes=4)
    training_compute = ComputeTarget.create(ws, cluster_name, compute_config)
    training_compute.wait_for_completion(show_output=True)
```

For each of the following statements, select Yes if the statement is true. Otherwise, select No.
NOTE: Each correct selection is worth one point.

	Yes	No
If a training cluster named aml-cluster already exists in the workspace, it will be deleted and replaced.	<input type="radio"/>	<input type="radio"/>
The wait_for_completion() method will not return until the aml-cluster compute has four active nodes.	<input type="radio"/>	<input type="radio"/>
If the code creates a new aml-cluster compute target, it may be preempted due to capacity constraints.	<input type="radio"/>	<input type="radio"/>
The aml-cluster compute target is deleted from the workspace after the training experiment completes.	<input type="radio"/>	<input type="radio"/>

- A. Mastered
- B. Not Mastered

Answer: A

Explanation:

Box 1: No
If a training cluster already exists it will be used. Box 2: Yes
The wait_for_completion method waits for the current provisioning operation to finish on the cluster. Box 3: Yes
Low Priority VMs use Azure's excess capacity and are thus cheaper but risk your run being pre-empted.
Box 4: No
Need to use training_compute.delete() to deprovision and delete the AmlCompute target. Reference:
<https://notebooks.azure.com/azureml/projects/azureml-getting-started/html/how-to-use-azureml/training/train-on> <https://docs.microsoft.com/en-us/python/api/azureml-core/azureml.core.compute.computetarget>

NEW QUESTION 39

- (Exam Topic 3)

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

Answer: B

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>

NEW QUESTION 43

- (Exam Topic 3)

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)
```

- A. Option A  
B. Option B  
C. Option C  
D. Option D  
E. Option E

**Answer:** DE

**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>

**NEW QUESTION 46**

- (Exam Topic 3)

You plan to build a team data science environment. Data for training models in machine learning pipelines will be over 20 GB in size.

You have the following requirements:

- Models must be built using Caffe2 or Chainer frameworks.
- Data scientists must be able to use a data science environment to build the machine learning pipelines and train models on their personal devices in both connected and disconnected network environments.
- Personal devices must support updating machine learning pipelines when connected to a network. You need to select a data science environment.

Which environment should you use?

- A. Azure Machine Learning Service  
B. Azure Machine Learning Studio  
C. Azure Databricks  
D. Azure Kubernetes Service (AKS)

**Answer:** A

**Explanation:**

The Data Science Virtual Machine (DSVM) is a customized VM image on Microsoft's Azure cloud built specifically for doing data science. Caffe2 and Chainer are supported by DSVM.

DSVM integrates with Azure Machine Learning.

**NEW QUESTION 50**

- (Exam Topic 3)

You create an Azure Machine Learning workspace and a new Azure DevOps organization. You register a model in the workspace and deploy the model to the target environment.

All new versions of the model registered in the workspace must automatically be deployed to the target environment.

You need to configure Azure Pipelines to deploy 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.



Actions

Create a service connection

Create a release pipeline

Create a build pipeline

Create an Azure DevOps project

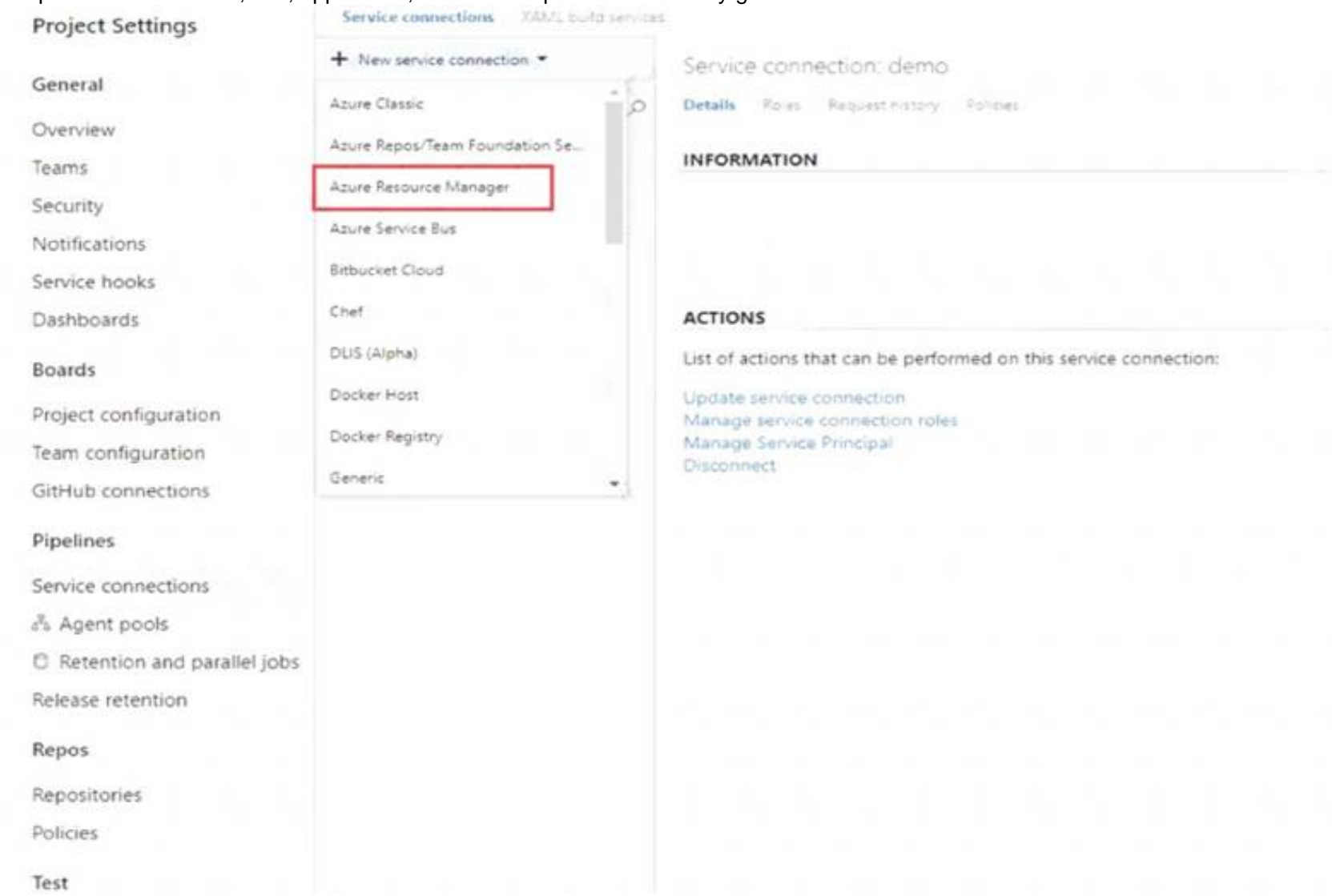
Install the Machine Learning extension for Azure Pipelines

Answer Area

- A. Mastered
- B. Not Mastered

Answer: A

**Explanation:**  
Graphical user interface, text, application, email Description automatically generated  
Step 1: Create an Azure DevOps project  
Step 2: Create a release pipeline  
> Sign in to your Azure DevOps organization and navigate to your project.  
> Go to Pipelines, and then select New pipeline.  
Step 3: Install the Machine Learning extension for Azure Pipelines  
You must install and configure the Azure CLI and ML extension.  
Step 4: Create a service connection  
How to set up your service connection  
Graphical user interface, text, application, email Description automatically generated



Select AzureMLWorkspace for the scope level, then fill in the following subsequent parameters. Graphical user interface, text, application Description automatically generated  
Note: How to enable model triggering in a release pipeline  
> Go to your release pipeline and add a new artifact. Click on AzureML Model artifact then select the appropriate AzureML service connection and select from the available models in your workspace.  
> Enable the deployment trigger on your model artifact as shown here. Every time a new version of that model is registered, a release pipeline will be triggered.  
Reference:  
<https://marketplace.visualstudio.com/items?itemName=ms-air-aiagility.vss-services-azureml> <https://docs.microsoft.com/en-us/azure/devops/pipelines/targets/azure-machine-learning>

**NEW QUESTION 55**  
- (Exam Topic 3)

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

**Answer:** A

**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. 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))

**NEW QUESTION 57**

- (Exam Topic 3)

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')
```

- A. Option A
- B. Option B
- C. Option C
- D. Option D
- E. Option E

**Answer: B**

**Explanation:**

Besides passing the dataset through the inputs parameter 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)
Reference:
https://docs.microsoft.com/es-es/azure/machine-learning/how-to-train-with-datasets
```

**NEW QUESTION 58**

- (Exam Topic 3)

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.

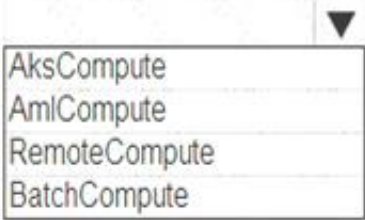
Client applications consuming the deployed web service must be authenticated based on their Azure Active Directory service principal.


You need to write a script that uses the Azure Machine Learning SDK to deploy the model. The necessary modules have been imported.

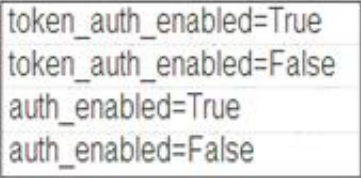
How should you complete the code? To answer, select the appropriate options in the answer area.

NOTE: Each correct selection is worth one point.



```
Assume the necessary modules have been imported
deploy_target =  (ws, "service-compute")

deployment_config =  .deploy_configuration(cpu_cores=1, memory_gb=1,

)

service = Model.deploy(ws, "ml-service",
 [model], inference_config, deployment_config, deploy_target)
service.wait_for_deployment(show_output = True)
```

- A. Mastered
- B. Not Mastered

**Answer: A**

**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.

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>

**NEW QUESTION 59**

- (Exam Topic 3)

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.

## Answer Area

```
{
 "Name": "Data Scientist Custom",
 "IsCustom": true
 "Description": "Description"
 "Actions": [
 Microsoft.MachineLearningServices/workspaces/*/read
 Microsoft.MachineLearningServices/workspaces/computes/*/write
 Microsoft.MachineLearningServices/workspaces/delete
 Microsoft.MachineLearningServices/workspaces/*/write
 Microsoft.MachineLearningServices/workspaces/computes/*/write
 Microsoft.MachineLearningServices/workspaces/delete
],
 "NotActions": [
 Microsoft.MachineLearningServices/workspaces/*/read
 Microsoft.MachineLearningServices/workspaces/*/write
 Microsoft.MachineLearningServices/workspaces/computes/*/delete
 Microsoft.MachineLearningServices/workspaces/*/read
 Microsoft.MachineLearningServices/workspaces/*/write
 Microsoft.MachineLearningServices/workspaces/computes/*/write
],
 "AssignableScopes": [
 "/subscriptions/<subscription_id>"
]
}
```

- A. Mastered
- B. Not Mastered

**Answer:** A

### Explanation:

Graphical user interface, application Description automatically generated

Graphical user interface, application Description automatically generated

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-use>

### NEW QUESTION 61

- (Exam Topic 3)

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.

#### 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

- A. Mastered
- B. Not Mastered

**Answer:** A

#### 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>

#### NEW QUESTION 64

- (Exam Topic 3)

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

**Answer:** ADE

#### Explanation:

References:

<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>

#### NEW QUESTION 68

- (Exam Topic 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 are analyzing a numerical dataset which contains missing values in several columns.

You must clean the missing values using an appropriate operation without affecting the dimensionality of the feature set.

You need to analyze a full dataset to include all values.

Solution: Replace each missing value using the Multiple Imputation by Chained Equations (MICE) method. Does the solution meet the goal?

- A. Yes
- B. NO

**Answer:** A

#### Explanation:

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.

Note: Multivariate imputation by chained equations (MICE), sometimes called "fully conditional specification" or "sequential regression multiple imputation" has emerged in the statistical literature as one principled method of addressing missing data. Creating multiple imputations, as opposed to single imputations, accounts for the statistical uncertainty in the imputations. In addition, the chained equations approach is very flexible and can handle variables of varying types (e.g., continuous or binary) as well as complexities such as bounds or survey skip patterns.

References: <https://www.ncbi.nlm.nih.gov/pmc/articles/PMC3074241/>

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



**NEW QUESTION 69**

- (Exam Topic 3)

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 ()

**Answer:** B

**Explanation:**

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

**NEW QUESTION 73**

- (Exam Topic 3)

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.

Target datasets

training-dataset

predictions-dataset

features-dataset

Answer Area

Baseline dataset

Target dataset

Target dataset

Target dataset

- A. Mastered
- B. Not Mastered

**Answer:** A

**Explanation:**

Text Description automatically generated with medium confidence  
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>

**NEW QUESTION 78**

- (Exam Topic 3)

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.

**Answer Area**

| Tree Depth | Bias                                                                    | Variance                                                                |
|------------|-------------------------------------------------------------------------|-------------------------------------------------------------------------|
| 5          | <div><div></div><div>High</div><div>Low</div><div>Identical</div></div> | <div><div></div><div>High</div><div>Low</div><div>Identical</div></div> |
| 15         | <div><div></div><div>High</div><div>Low</div><div>Identical</div></div> | <div><div></div><div>High</div><div>Low</div><div>Identical</div></div> |

- A. Mastered
- B. Not Mastered

Answer: A

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/>

NEW QUESTION 81

- (Exam Topic 3)  
You are creating an experiment by using Azure Machine Learning Studio.  
You must divide the data into four subsets for evaluation. There is a high degree of missing values in the data. You must prepare the data for analysis.  
You need to select appropriate methods for producing the experiment.  
Which three modules 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.  
NOTE: More than one order of answer choices is correct. You will receive credit for any of the correct orders you select.

Actions

Build Counting Transform

Missing Values Scrubber

Feature Hashing

Clean Missing Data

Replace Discrete Values

Import Data

Latent Dirichlet Transformation

Partition and Sample

Answer Area

- A. Mastered
- B. Not Mastered

Answer: A

Explanation:

The Clean Missing Data module in Azure Machine Learning Studio, to remove, replace, or infer missing values.

NEW QUESTION 86

- (Exam Topic 3)  
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 changes 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.

Answer: CE

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>

**NEW QUESTION 91**

- (Exam Topic 3)

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.

**Answer:** BD

**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. 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>

**NEW QUESTION 94**

- (Exam Topic 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 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

**Answer:** B

**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=azu>

**NEW QUESTION 98**

- (Exam Topic 3)

You train and register a model by using the Azure Machine Learning SDK on a local workstation. Python 3.6 and Visual Studio Code are installed on the workstation.

When you try to deploy the model into production as an Azure Kubernetes Service (AKS)-based web service, you experience an error in the scoring script that causes deployment to fail.

You need to debug the service on the local workstation before deploying the service to production.

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.



## Actions

## Answer Area

Create an AksWebservice deployment configuration for the service and deploy the model to it

Install Docker on the workstation

Create a LocalWebservice deployment configuration for the service and deploy the model to it

Debug and modify the scoring script as necessary. Use the reload() method of the service after each modification

Create an AciWebservice deployment configuration for the service and deploy the model to it



- A. Mastered
- B. Not Mastered

**Answer: A**

### Explanation:

Graphical user interface, text, application, email Description automatically generated

Step 1: Install Docker on the workstation

Prerequisites include having a working Docker installation on your local system. Build or download the dockerfile to the compute node.

Step 2: Create an AksWebservice deployment configuration and deploy the model to it

To deploy a model to Azure Kubernetes Service, create a deployment configuration that describes the compute resources needed.

# 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) service.wait_for_deployment(show_output = True)
```

```
print(service.state) print(service.get_logs())
```

Step 3: Create a LocalWebservice deployment configuration for the service and deploy the model to it

To deploy locally, modify your code to use LocalWebservice.deploy\_configuration() to create a deployment configuration. Then use Model.deploy() to deploy the service.

Step 4: Debug and modify the scoring script as necessary. Use the reload() method of the service after each modification.

During local testing, you may need to update the score.py file to add logging or attempt to resolve any problems that you've discovered. To reload changes to the score.py file, use reload(). For example, the following code reloads the script for the service, and then sends data to it.

Reference:

<https://docs.microsoft.com/en-us/azure/machine-learning/how-to-deploy-azure-kubernetes-service> <https://docs.microsoft.com/en-us/azure/machine-learning/how-to-troubleshoot-deployment-local>

### NEW QUESTION 103

- (Exam Topic 3)

You write code to retrieve an experiment that is run from your Azure Machine Learning workspace.

The run used the model interpretation support in Azure Machine Learning to generate and upload a model 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.

```
Assume required modules are imported
```

```
ws = Workspace.from_config()
feature_importances = explanation.
```

|                            |   |                                      |
|----------------------------|---|--------------------------------------|
|                            | ▼ | ( workspace = ws,                    |
| from_run                   |   | experiment_name='train_and_explain', |
| list_model_explanations    |   | run_id='train_and_explain_12345')    |
| from_run_id                |   |                                      |
| download_model_explanation |   |                                      |

```
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)
```

- A. Mastered
- B. Not Mastered

**Answer:** A

**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 explanations Client.

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>.

**NEW QUESTION 104**

- (Exam Topic 3)

You plan to provision an Azure Machine Learning Basic edition workspace for a data science project. You need to identify the tasks you will be able to perform in the workspace.

Which three tasks will you be able to perform? Each correct answer presents a complete solution.

NOTE: Each correct selection is worth one point. D

- A. Create a Compute Instance and use it to run code in Jupyter notebooks.
- B. Create an Azure Kubernetes Service (AKS) inference cluster.
- C. Use the designer to train a model by dragging and dropping pre-defined modules.
- D. Create a tabular dataset that supports versioning.
- E. Use the Automated Machine Learning user interface to train a model.

**Answer:** ABD

**Explanation:**

Reference:

<https://azure.microsoft.com/en-us/pricing/details/machine-learning/>

**NEW QUESTION 105**

- (Exam Topic 3)

You are developing a hands-on workshop to introduce Docker for Windows to attendees. You need to ensure that workshop attendees can install Docker on their devices.

Which two prerequisite components should attendees install on the devices? Each correct answer presents part of the solution.

NOTE: Each correct selection is worth one point.

- A. Microsoft Hardware-Assisted Virtualization Detection Tool
- B. Kitematic
- C. BIOS-enabled virtualization
- D. VirtualBox

E. Windows 10 64-bit Professional

**Answer:** CE

**Explanation:**

C: Make sure your Windows system supports Hardware Virtualization Technology and that virtualization is enabled. Ensure that hardware virtualization support is turned on in the BIOS settings. For example:



E: To run Docker, your machine must have a 64-bit operating system running Windows 7 or higher. References:  
[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/>

**NEW QUESTION 106**

- (Exam Topic 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 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

**Answer:** B

**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>

**NEW QUESTION 108**

- (Exam Topic 3)

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()

**Answer:** B

**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.

**NEW QUESTION 113**

- (Exam Topic 3)

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.

▲ 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

- A. Mastered
- B. Not Mastered

**Answer:** A

**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>

**NEW QUESTION 115**

- (Exam Topic 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 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)
```

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

**Answer: A**

**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>

**NEW QUESTION 119**

- (Exam Topic 3)

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.

- A. Mastered
- B. Not Mastered

**Answer: A**

**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>

**NEW QUESTION 123**

- (Exam Topic 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.

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 | A virtual machine named mlvm with the following configuration: <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.

- A. Yes
- B. No

**Answer:** A

**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>

**NEW QUESTION 125**

- (Exam Topic 3)

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

**Answer:** D

**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>

**NEW QUESTION 126**

- (Exam Topic 3)

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

The training loss, validation loss, training accuracy, and validation accuracy of each training epoch has been provided. You need to identify whether the classification model is over fitted.

Which of the following is correct?

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

**Answer:** B

**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.

References:

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

**NEW QUESTION 131**

- (Exam Topic 3)

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

**Answer:** BD

**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>

### NEW QUESTION 136

- (Exam Topic 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 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

**Answer: B**

#### 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>

### NEW QUESTION 140

- (Exam Topic 3)

You configure a Deep Learning Virtual Machine for Windows.

You need to recommend tools and frameworks to perform the following:

- Build deep neural network (DNN) models
- Perform interactive data exploration and visualization

Which tools and frameworks should you recommend? To answer, drag the appropriate tools to the correct tasks. Each tool 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.

| Tools                                                 | Answer Area                                                                                                                                                                                                                                          |      |      |                  |                 |                                                       |                 |
|-------------------------------------------------------|------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|------|------|------------------|-----------------|-------------------------------------------------------|-----------------|
|                                                       | <table border="1"><thead><tr><th>Task</th><th>Tool</th></tr></thead><tbody><tr><td>Build DNN models</td><td><div>Tool</div></td></tr><tr><td>Enable interactive data exploration and visualization</td><td><div>Tool</div></td></tr></tbody></table> | Task | Tool | Build DNN models | <div>Tool</div> | Enable interactive data exploration and visualization | <div>Tool</div> |
| Task                                                  | Tool                                                                                                                                                                                                                                                 |      |      |                  |                 |                                                       |                 |
| Build DNN models                                      | <div>Tool</div>                                                                                                                                                                                                                                      |      |      |                  |                 |                                                       |                 |
| Enable interactive data exploration and visualization | <div>Tool</div>                                                                                                                                                                                                                                      |      |      |                  |                 |                                                       |                 |
| <div>Vowpal Wabbit</div>                              |                                                                                                                                                                                                                                                      |      |      |                  |                 |                                                       |                 |
| <div>PowerBI Desktop</div>                            |                                                                                                                                                                                                                                                      |      |      |                  |                 |                                                       |                 |
| <div>Azure Data Factory</div>                         |                                                                                                                                                                                                                                                      |      |      |                  |                 |                                                       |                 |
| <div>Microsoft Cognitive Toolkit</div>                |                                                                                                                                                                                                                                                      |      |      |                  |                 |                                                       |                 |

- A. Mastered
- B. Not Mastered

**Answer: A**

#### Explanation:

Box 1: Vowpal Wabbit

Use the Train Vowpal Wabbit Version 8 module in Azure Machine Learning Studio (classic), to create a machine learning model by using Vowpal Wabbit.

Box 2: PowerBI Desktop

Power BI Desktop is a powerful visual data exploration and interactive reporting tool

BI is a name given to a modern approach to business decision making in which users are empowered to find, explore, and share insights from data across the enterprise.

References:

<https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/train-vowpal-wabbit-version-8> <https://docs.microsoft.com/en-us/azure/architecture/data-guide/scenarios/interactive-data-exploration>

### NEW QUESTION 143

- (Exam Topic 3)

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

**Answer: B**

**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.parall>

**NEW QUESTION 147**

- (Exam Topic 3)

You plan to use a Data Science Virtual Machine (DSVM) with the open source deep learning frameworks Caffe2 and Theano. You need to select a pre configured DSVM to support the framework.

What should you create?

- A. Data Science Virtual Machine for Linux (CentOS)
- B. Data Science Virtual Machine for Windows 2012
- C. Data Science Virtual Machine for Windows 2016
- D. Geo AI Data Science Virtual Machine with ArcGIS
- E. Data Science Virtual Machine for Linux (Ubuntu)

**Answer: E**

**NEW QUESTION 149**

- (Exam Topic 3)

You create a classification model with a dataset that contains 100 samples with Class A and 10,000 samples with Class B

The variation of Class B is very high. You need to resolve imbalances. Which method should you use?

- A. Partition and Sample
- B. Cluster Centroids
- C. Tomek links
- D. Synthetic Minority Oversampling Technique (SMOTE)

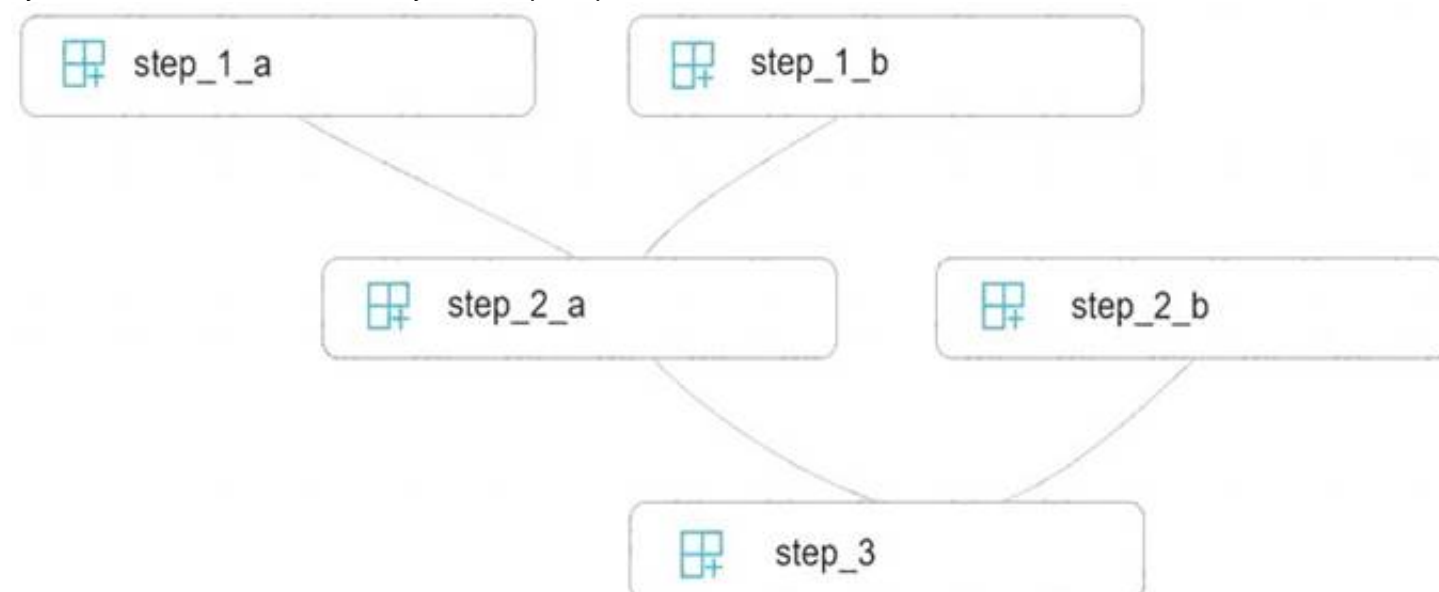
**Answer: D**

**NEW QUESTION 151**

- (Exam Topic 3)

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])`

- A. Option A  
 B. Option B  
 C. Option C  
 D. Option D

**Answer:** A

**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>

**NEW QUESTION 153**

- (Exam Topic 3)

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.

|                                                                                                                               | Yes                   | No                    |
|-------------------------------------------------------------------------------------------------------------------------------|-----------------------|-----------------------|
| The SHAP TreeExplainer will be used to interpret the model.                                                                   | <input type="radio"/> | <input type="radio"/> |
| If you omit the features and classes parameters in the TabularExplainer instantiation, the explainer still works as expected. | <input type="radio"/> | <input type="radio"/> |
| You could interpret the model by using a MimicExplainer instead of a TabularExplainer.                                        | <input type="radio"/> | <input type="radio"/> |

- A. Mastered  
 B. Not Mastered

**Answer:** A

**Explanation:**

s 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>



**NEW QUESTION 158**

- (Exam Topic 3)

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.

**Compute types**

Attached compute

Inference cluster

Training cluster

**Answer Area**

| Requirement                                                                  | Compute type            |
|------------------------------------------------------------------------------|-------------------------|
| Train models by using the Azure Machine Learning designer.                   | <div>Compute type</div> |
| Score new data through a trained model published as a real-time web service. | <div>Compute type</div> |
| Train models by using an Azure Databricks cluster.                           | <div>Compute type</div> |
| Deploy models by using the Azure Machine Learning designer.                  | <div>Compute type</div> |

- A. Mastered  
B. Not Mastered

**Answer:** A

**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                             |
| Azure Machine Learning compute instance | yes & hyperparameter tuning | yes          | yes                             |

Box 2: Inference cluster Box 3: Training cluster Box 4: Attached compute

**NEW QUESTION 159**

- (Exam Topic 3)

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.

**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**

⏪

⏩

⏴

⏵

- A. Mastered  
B. Not Mastered

**Answer:** A

**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/>

**NEW QUESTION 161**

- (Exam Topic 3)

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.

**Answer:** D

**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.

Reference:

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

**NEW QUESTION 162**

- (Exam Topic 3)

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. Preprocess Text
- C. Execute Python Script
- D. Latent Dirichlet Allocation

**Answer:** A

**Explanation:**

Typical metadata changes might include marking columns as features. References:

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

**NEW QUESTION 167**

- (Exam Topic 3)

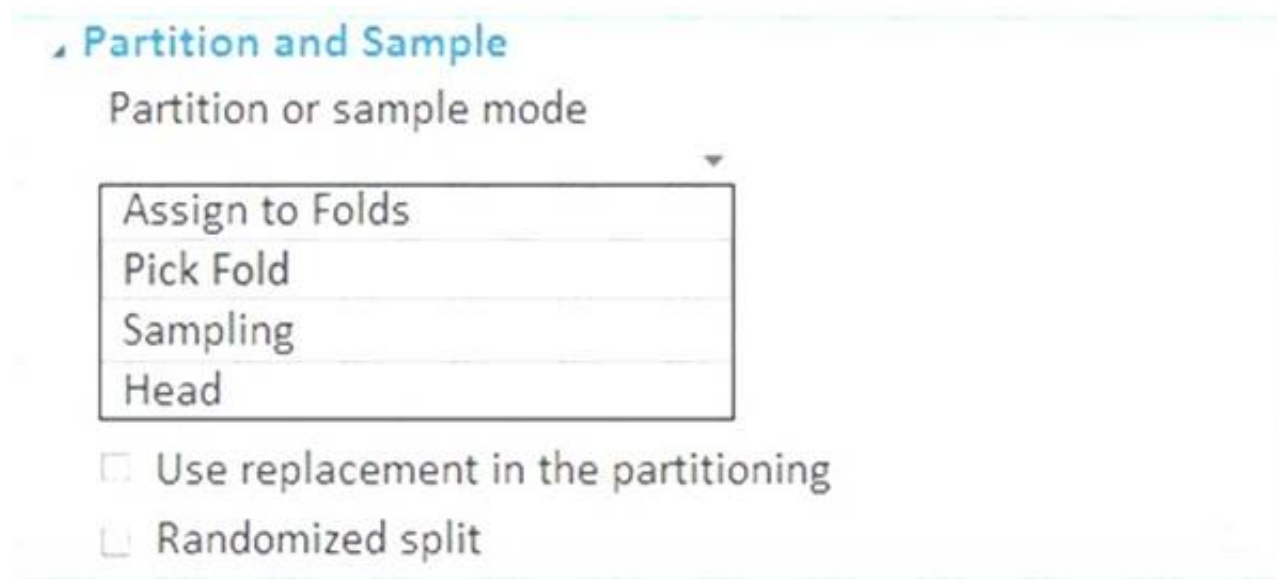
You have a dataset that contains 2,000 rows. You are building a machine learning classification model by using Azure Learning Studio. You add a Partition and Sample module to the experiment.

You need to configure the module. You must meet the following requirements:

- Divide the data into subsets
- Assign the rows into folds using a round-robin method
- Allow rows in the dataset to be reused

How should you configure the module? To answer, select the appropriate options in the dialog box in the answer area.

NOTE: Each correct selection is worth one point.



- A. Mastered
- B. Not Mastered

**Answer:** A

**Explanation:**

Use the Split data into partitions 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.

- For Partition or sample mode, select Assign to Folds.
- Use replacement in the partitioning: Select this option if you want the sampled row to be put back into the pool of rows for potential reuse. As a result, the same row might be assigned to several folds.
- If you do not use replacement (the default option), the sampled row is not put back into the pool of rows for potential reuse. As a result, each row can be assigned to only one fold.
- Randomized split: Select this option if you want rows to be randomly assigned to folds. If you do not select this option, rows are assigned to folds using the round-robin method. References:  
<https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/partition-and-sample>

**NEW QUESTION 171**

- (Exam Topic 3)

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. HDInsight
- B. Azure Databricks
- C. Azure Kubernetes Services
- D. the existing Machine Learning Compute resource
- E. a new Machine Learning Compute resource

**Answer:** C

**Explanation:**

Azure Kubernetes Service (AKS) can be used real-time inference. Reference:  
<https://docs.microsoft.com/en-us/azure/machine-learning/concept-compute-target>

**NEW QUESTION 173**

- (Exam Topic 3)

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

**Answer:** BC

**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.

References:

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

**NEW QUESTION 177**

- (Exam Topic 3)

You are building a regression model tot 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?

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

**Answer:** CE

**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.

References:

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

**NEW QUESTION 179**

- (Exam Topic 3)

You are performing a filter based feature selection for a dataset 10 build a multi class classifies 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. Chi-squared
- B. Spearman correlation
- C. Kendall correlation
- D. Person correlation

**Answer:** D

**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.

Reference:

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

**NEW QUESTION 180**

- (Exam Topic 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 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

**Answer:** B

**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>

**NEW QUESTION 182**

- (Exam Topic 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 are analyzing a numerical dataset which contain missing values in several columns.

You must clean the missing values using an appropriate operation without affecting the dimensionality of the feature set.

You need to analyze a full dataset to include all values.

Solution: Use the last Observation Carried Forward (IOCF) method to impute the missing data points. Does the solution meet the goal?

- A. Yes

B. No

**Answer:** B

**Explanation:**

Instead use the Multiple Imputation by Chained Equations (MICE) method.

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.

Note: Last observation carried forward (LOCF) is a method of imputing missing data in longitudinal studies. If a person drops out of a study before it ends, then his or her last observed score on the dependent variable is used for all subsequent (i.e., missing) observation points. LOCF is used to maintain the sample size and to reduce the bias caused by the attrition of participants in a study.

References:

<https://methods.sagepub.com/reference/encyc-of-research-design/n211.xml> <https://www.ncbi.nlm.nih.gov/pmc/articles/PMC3074241/>

**NEW QUESTION 186**

- (Exam Topic 3)

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_classification | 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')
automl_run = AutoMLRun(automl_ex, 'AutoML_1234567890-123')
best_iter = automl_run.get_output()[0]
```

B)

```
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]
```

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
```

A. Option A

B. Option B

C. Option C

D. Option D

**Answer:** A

**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-mach>

**NEW QUESTION 191**

- (Exam Topic 3)

You create an Azure Machine Learning compute resource to train models. The compute resource is configured as follows:

➤ Minimum nodes: 2

➤ Maximum nodes: 4

You must decrease the minimum number of nodes and increase the maximum number of nodes to the following values:

➤ Minimum nodes: 0

➤ Maximum nodes: 8

You need to reconfigure the compute resource.

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. Use the Azure Machine Learning studio.
- B. Run the update method of the AmlCompute class in the Python SDK.
- C. Use the Azure portal.
- D. Use the Azure Machine Learning designer.
- E. Run the refresh\_state() method of the BatchCompute class in the Python SDK

**Answer:** ABC

**Explanation:**

Reference:

[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))

**NEW QUESTION 194**

- (Exam Topic 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 are a data scientist using Azure Machine Learning Studio.

You need to normalize values to produce an output column into bins to predict a target column. Solution: Apply a Quantiles binning mode with a PQuantile normalization.

Does the solution meet the goal?

- A. Yes
- B. No

**Answer:** B

**Explanation:**

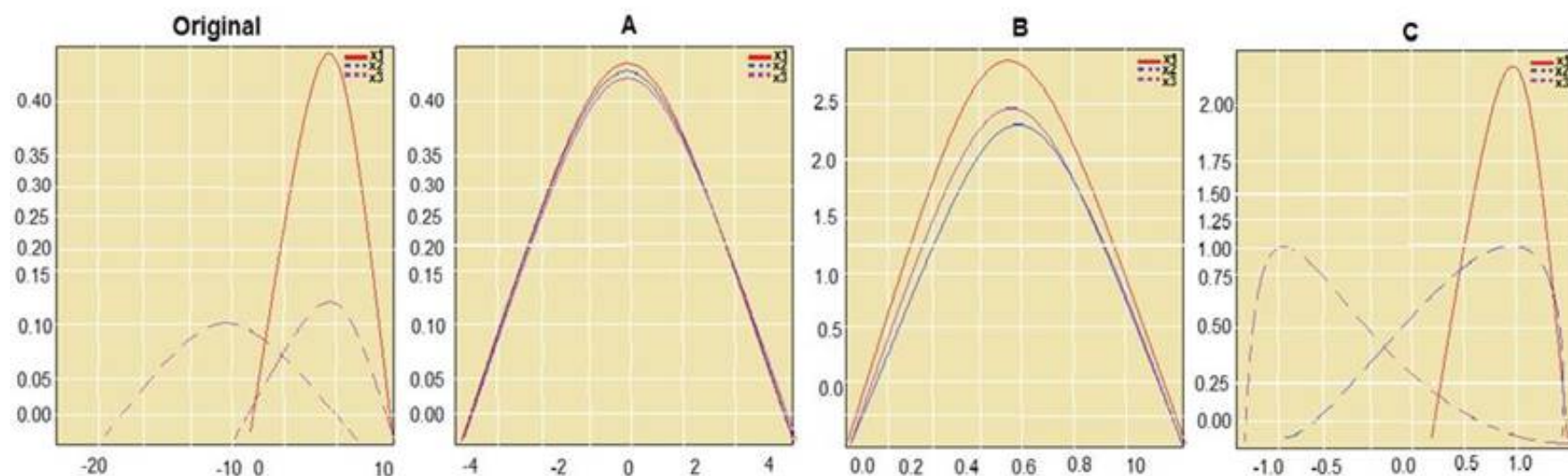
Use the Entropy MDL binning mode which has a target column. References:

<https://docs.microsoft.com/en-us/azure/machine-learning/studio-module-reference/group-data-into-bins>

**NEW QUESTION 195**

- (Exam Topic 3)

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.

**Question**

**Answer choice**

Which scaler is used in graph A?

▼

Standard Scaler

Min Max Scale

Normalizer

Which scaler is used in graph B?

▼

Standard Scaler

Min Max Scale

Normalizer

Which scaler is used in graph C?

▼

Standard Scaler

Min Max Scale

Normalizer

- A. Mastered
- B. Not Mastered

**Answer:** A

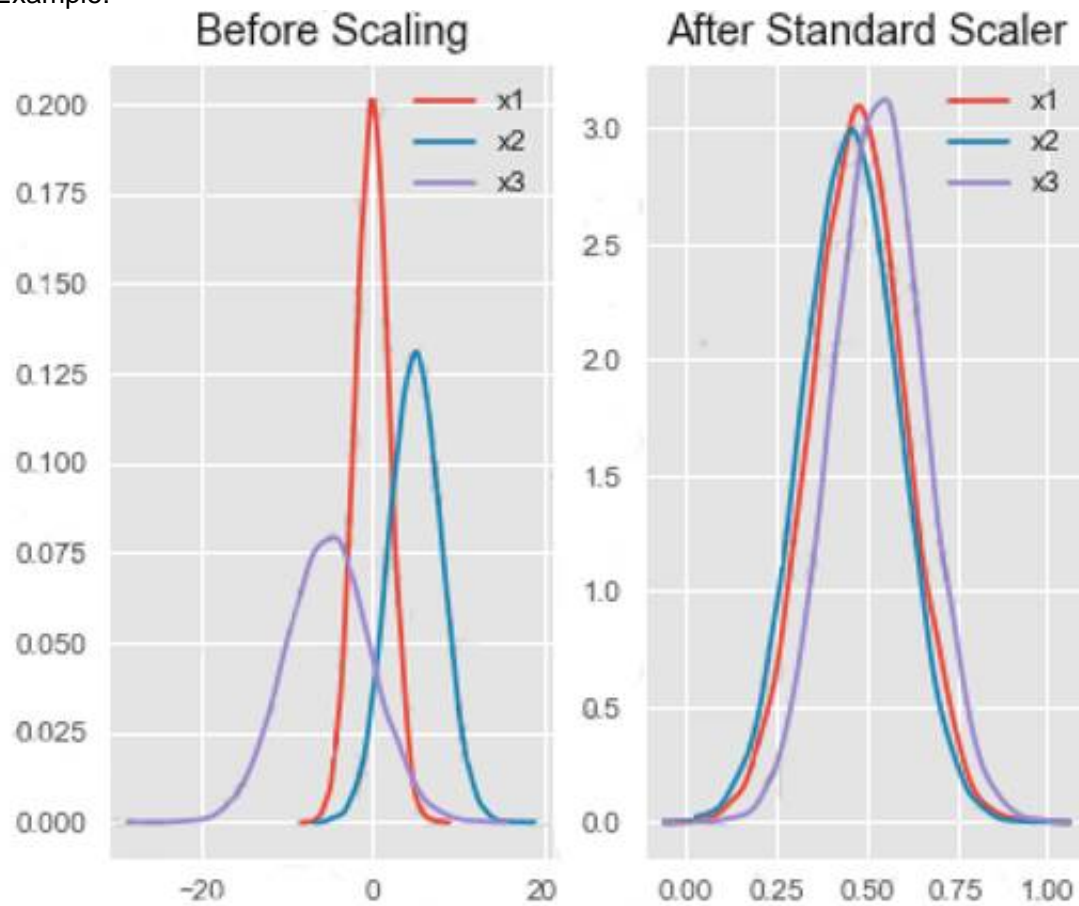


**Explanation:**

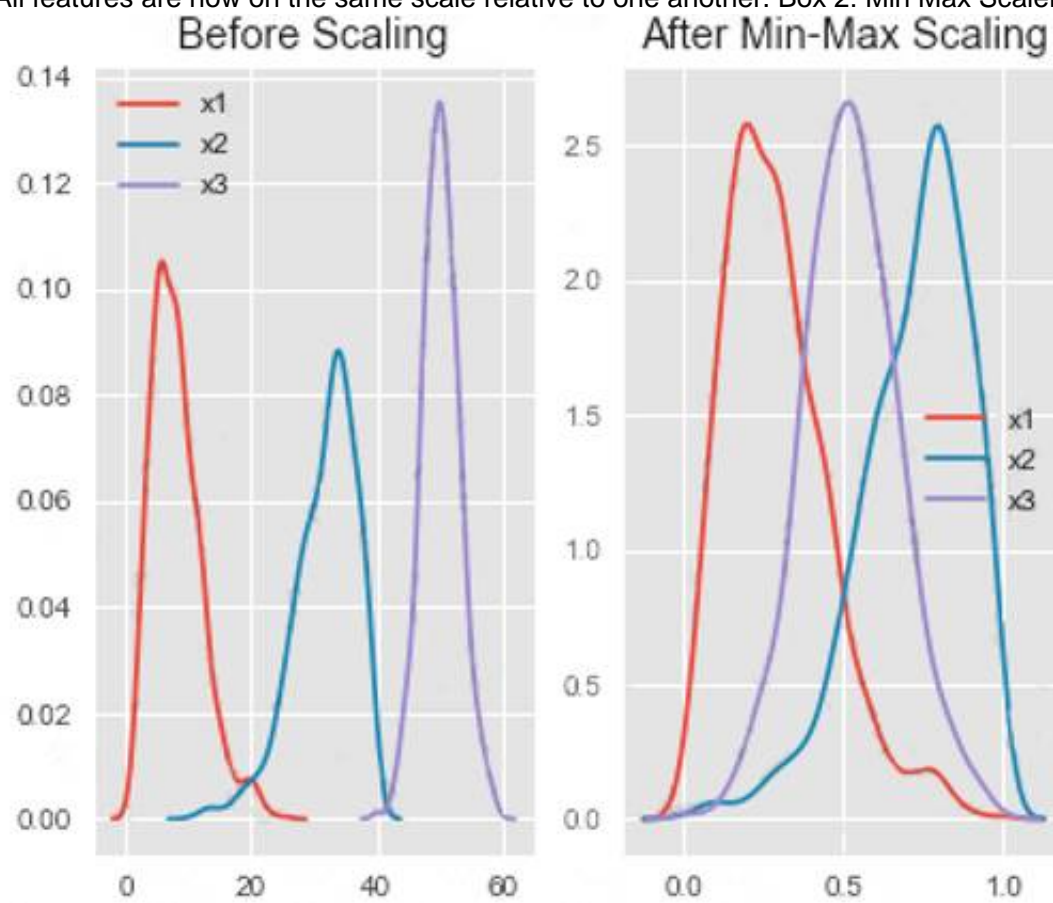
Box 1: StandardScaler

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/>

**NEW QUESTION 200**

- (Exam Topic 3)

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.

**Answer: D**

**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>

### NEW QUESTION 205

- (Exam Topic 3)

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\_sec function
- C. Implement the upload\_dashboard\_dictionary function
- D. Upload the training script

**Answer:** C

#### 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>

### NEW QUESTION 210

- (Exam Topic 2)

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.

#### Modules

Score Matchbox Recommender

Apply Transformation

Evaluate Recommender

Evaluate Model

Train Model

Sweep Clustering

Score Model

Load Trained Model

#### Answer Area



- A. Mastered
- B. Not Mastered

**Answer:** A

#### 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>

### NEW QUESTION 214

- (Exam Topic 1)

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.

**Answer:** D

#### 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.

### NEW QUESTION 215

- (Exam Topic 1)

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.

**Answer:** A

#### 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 References:

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

### NEW QUESTION 217

- (Exam Topic 1)

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.

| 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.                               |   |             |

- A. Mastered
- B. Not Mastered

**Answer:** A

#### 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>

### NEW QUESTION 218

- (Exam Topic 1)

You need to resolve the local machine learning pipeline performance issue. What should you do?

- A. Increase Graphic Processing Units (GPUs).
- B. Increase the learning rate.
- C. Increase the training iterations,
- D. Increase Central Processing Units (CPUs).

**Answer:** A



**NEW QUESTION 220**

- (Exam Topic 1)

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.

```
import pytorch as deeplearninglib
import tensorflow as deeplearninglib
import cntk as deeplearninglib
```

```
train_smampler = 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_smampler, 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)
```

- A. Mastered  
B. Not Mastered

**Answer:** A

**Explanation:**

Box 1: import pytorch 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)

**NEW QUESTION 225**

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