



TASK 3 - P

Design and deploy model using
Azure machine learning

Survival prediction on
Titanic dataset

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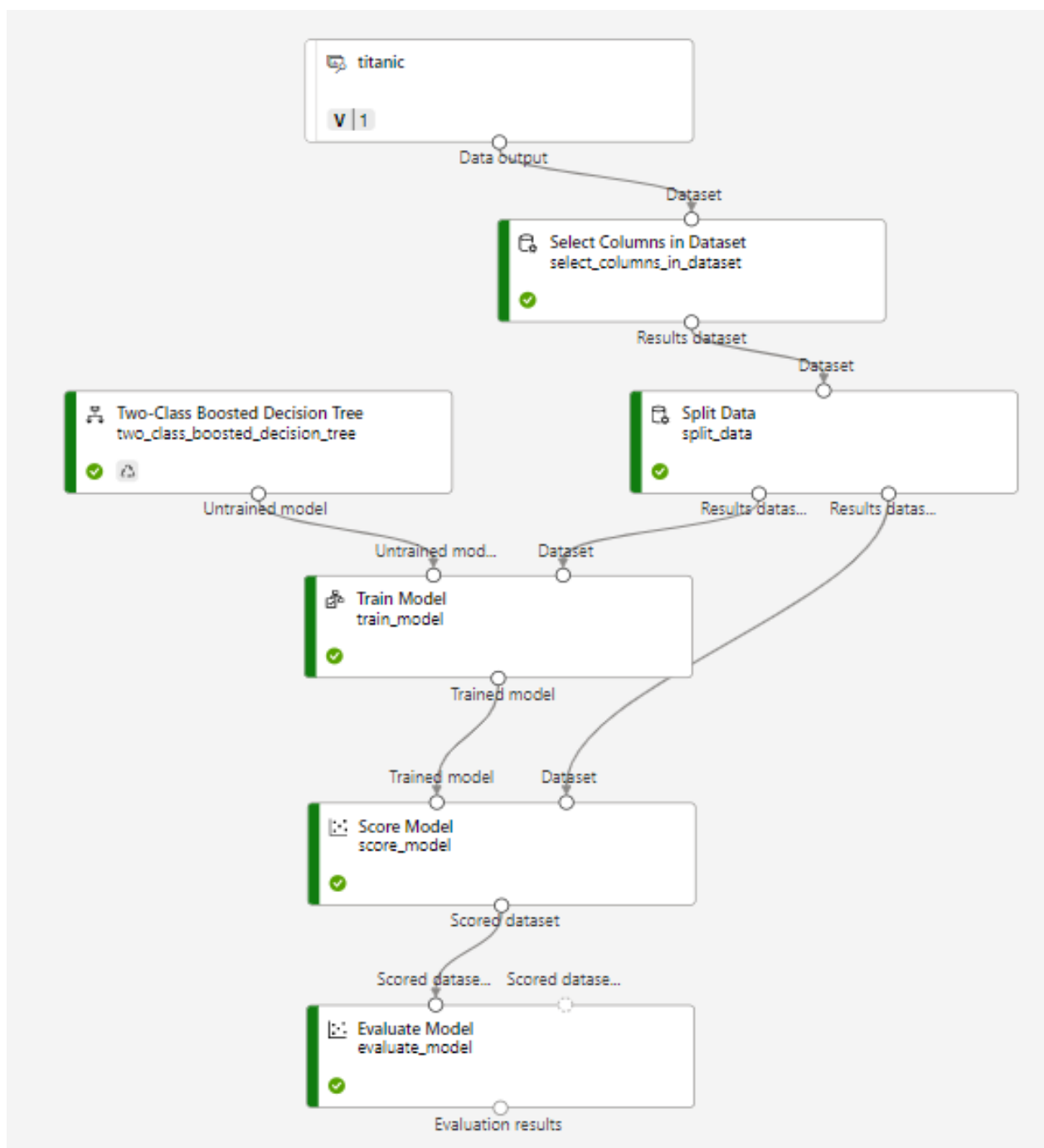
Question 1. For this task you need to design and deploy a machine learning model using Azure ML studio (designer). You need to use Microsoft Azure Machine Learning Studio to design and deploy your model. To complete this task, you need to select a dataset and design a decision tree model (classification tree or regression tree) and then deploy the built model and get the API key. To do this task you need to follow the workshop recording and slides and deploy your own model on Azure. You need to provide the screenshots of your designed model, training model, the performance of the built model (e.g., Accuracy, confusion matrix and etc) and deployed model with the API key and test the model. The screenshot of the model should include your Azure account name since the API key is unique to you.	2
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1. For this task you need to design and deploy a machine learning model using Azure ML studio (designer). You need to use Microsoft Azure Machine Learning Studio to design and deploy your model. To complete this task, you need to select a dataset and design a decision tree model (classification tree or regression tree) and then deploy the built model and get the API key. To do this task you need to follow the workshop recording and slides and deploy your own model on Azure. You need to provide the screenshots of your designed model, training model, the performance of the built model (e.g., Accuracy, confusion matrix and etc) and deployed model with the API key and test the model. The screenshot of the model should include your Azure account name since the API key is unique to you.

Introduction:

We have used titanic dataset to build a decision tree in Microsoft Azure Machine Learning Studio.

The below is the image showing the pipeline of building this model.



First of all we create a resource group called MDS-deakins with location as Central India.

Home > Resource groups

Deakin University

+ Create Manage view Refresh Export to CSV Open query Assign tags

Filter for any field... Subscription equals all Location equals all Add filter

Showing 1 to 4 of 4 records.

Name	Subscription	Location
DefaultResourceGroup-eastus	Azure for Students	East US
MC_MDS-deakins_aksttitanic2e469475a9e_eastus	Azure for Students	East US
MDS-deakins	Azure for Students	Central India
NetworkWatcherRG	Azure for Students	East US

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Give feedback

After creating the resource group, we create resource group we create a new Azure machine learning under the resource group we created.

Home > Azure Machine Learning

Deakin University

+ Create Recently deleted Manage view Refresh Export to CSV Open query Assign tags

Filter for any field... Subscription equals all Type equals all Resource group equals all Location equals all Add filter

Showing 1 to 1 of 1 records.

Name	Resource group	Location	Subscription	Type	Kind
Azuretitanic	MDS-deakins	East US	Azure for Students	Azure Machine Learning workspace	Default

< Previous Page 1 of 1 Next >

Give feedback

WE have created a work space in Azure machine learning.

We then launch studio and we create dataset in this environment using data option in designer.

Microsoft Azure Machine Learning Studio

Create data asset

Settings
These settings determine how the data is parsed. The initial settings are automatically detected; you can change them as needed to reparse the data.

File format: **Delimited** | Delimiter: **Comma** | Example: Field1,Field2,Field3 | Encoding: **UTF-8**

Column headers: **All files have same headers** | Skip rows: **None**

☐ Dataset contains multi-line data ⓘ

ⓘ Note: Processing tabular files with multi-line data is slower because multiple CPU cores cannot be used to ingest the data in parallel. Checking this option may result in slower processing times.

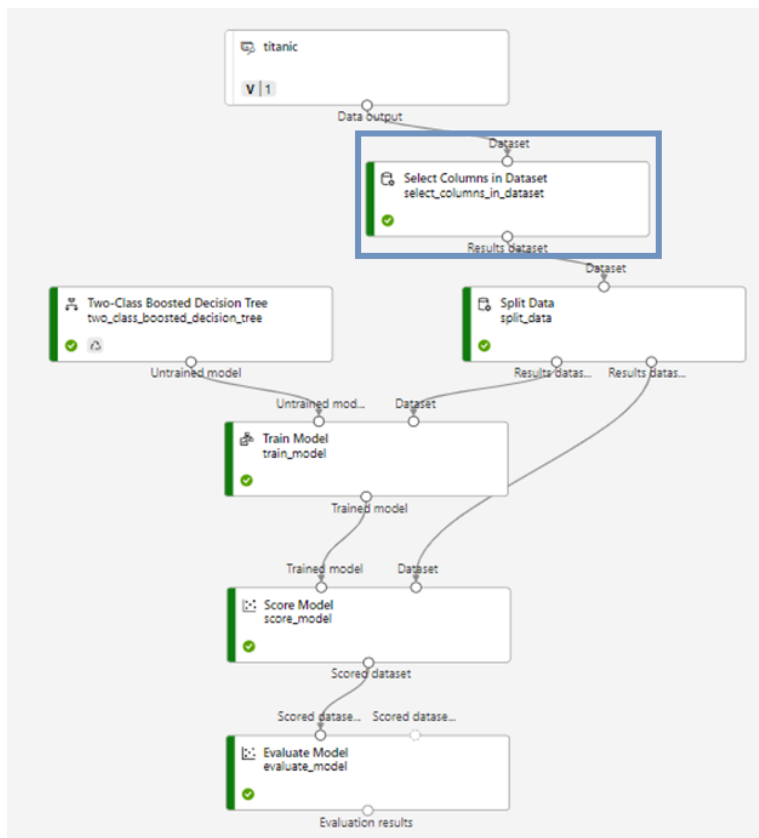
Data preview

pclass	survived	name	sex	age	sibsp	parch	ticket	fare	cabin	embarked	boat	body	home.d...
1	1	Allen, Mi...	female	29	0	0	24160	211.338	B5	S	2	null	St Louis, ...
1	1	Allison, ...	male	0.917	1	2	113781	151.55	C22 C26	S	11	null	Montreal...
1	0	Allison, ...	female	2	1	2	113781	151.55	C22 C26	S	null	null	Montreal...
1	0	Allison, ...	male	30	1	2	113781	151.55	C22 C26	S	null	135	Montreal...
1	0	Allison, ...	female	25	1	2	113781	151.55	C22 C26	S	null	null	Montreal...
1	1	Anderso...	male	48	0	0	19952	26.55	E12	S	3	null	New Yor...
1	1	Andrews...	female	63	1	0	13502	77.958	D7	S	10	null	Hudson, ...

Buttons: Back, Next, Review, Cancel

We have taken this data set from our local computer and in the above image we can see a preview of the dataset.

We can now drag and drop the dataset created in our work space to build a pipeline.



We then drag and drop Select Columns in Dataset from the component option in designer and connect it to dataset.

We select a total of 8 features in the Select Column in Dataset, namely:

1. **pclass**
2. **survived**
3. **sex**
4. **age**
5. **sibsp**
6. **parch**
7. **fare**
8. **embarked**

Including our target variable i.e., survived

We now drag and drop split data option from component and split the data into 80:20 ratio. Where 80 will be our train data and 20 will be our test data.

The screenshot displays the Microsoft Azure Machine Learning Studio interface. The central workspace shows a workflow for 'titanicProject-Job1'. The workflow consists of the following components: 'Select Columns in Dataset' (selecting columns from the dataset), 'Split Data' (splitting the data into training and testing sets), and 'Score Model' (evaluating the model). The 'Split Data' component is highlighted, and its parameters are shown on the right-hand side of the interface.

The 'Split Data' component parameters are as follows:

- Splitting mode:** Split Rows
- Fraction of rows in the first output dataset:** 0.8
- Randomized split:** True
- Random seed:** 0
- Stratified split:** False

The 'Output settings', 'Input settings', 'Run settings', and 'Component information' sections are also visible on the right-hand side of the interface.

After splitting the data, we drag and drop Two-Class Boosted Decision Tree here under parameters we have specified that:

1. Max number of leaves pre tree: 20
2. Min number of samples per leaf node: 10
3. Learning rate: 0.2
4. Number of trees constructed: 100

The screenshot shows the Microsoft Azure Machine Learning Studio interface. On the left, the 'Jobs' tab is selected, showing a list of jobs including 'two_class_boosted_decision_tree'. The main workspace displays a pipeline named 'titanicProject-Job1' with a 'Completed' status. The pipeline steps are: 'Select Columns in Dataset' (selecting columns from a dataset), 'Split Data' (splitting the data into training and testing sets), 'Train Model' (training a model), and 'Score Model' (scoring the model). A 'Two-Class Boosted Decision Tree' node is highlighted, and its parameters are shown in a side panel. The parameters include: 'Maximum number of leaves per tree' (20), 'Minimum number of samples per leaf node' (10), 'Learning rate' (0.2), 'Number of trees constructed' (100), and 'Random number seed'.

Hereafter we drag and drop train model and specify the label column as survived. We connect it to Two-Class Boost Decision Tree from where it takes the model for prediction and Split data from where it takes the train set.

The screenshot shows the 'Train Model' node configuration in the Microsoft Azure Machine Learning Studio. The 'Label column' is set to 'survived'. The pipeline diagram shows the 'Train Model' node connected to the 'Two-Class Boosted Decision Tree' node. The 'Train Model' node is highlighted, and its parameters are shown in a side panel. The parameters include: 'Label column' (survived), 'Model explanations' (False), 'Output settings', 'Input settings', 'Run settings', and 'Component information'.

Also, we drag and drop score model and connect it with train model and split data to feed it both the trained model and test model. Also, we drag and drop evaluate model and connect it to score model.

Now our pipeline is complete. We save this pipeline.

We now go to the compute option from the side bar and create a compute cluster named comclusterTitanic with location east US and 1 node.

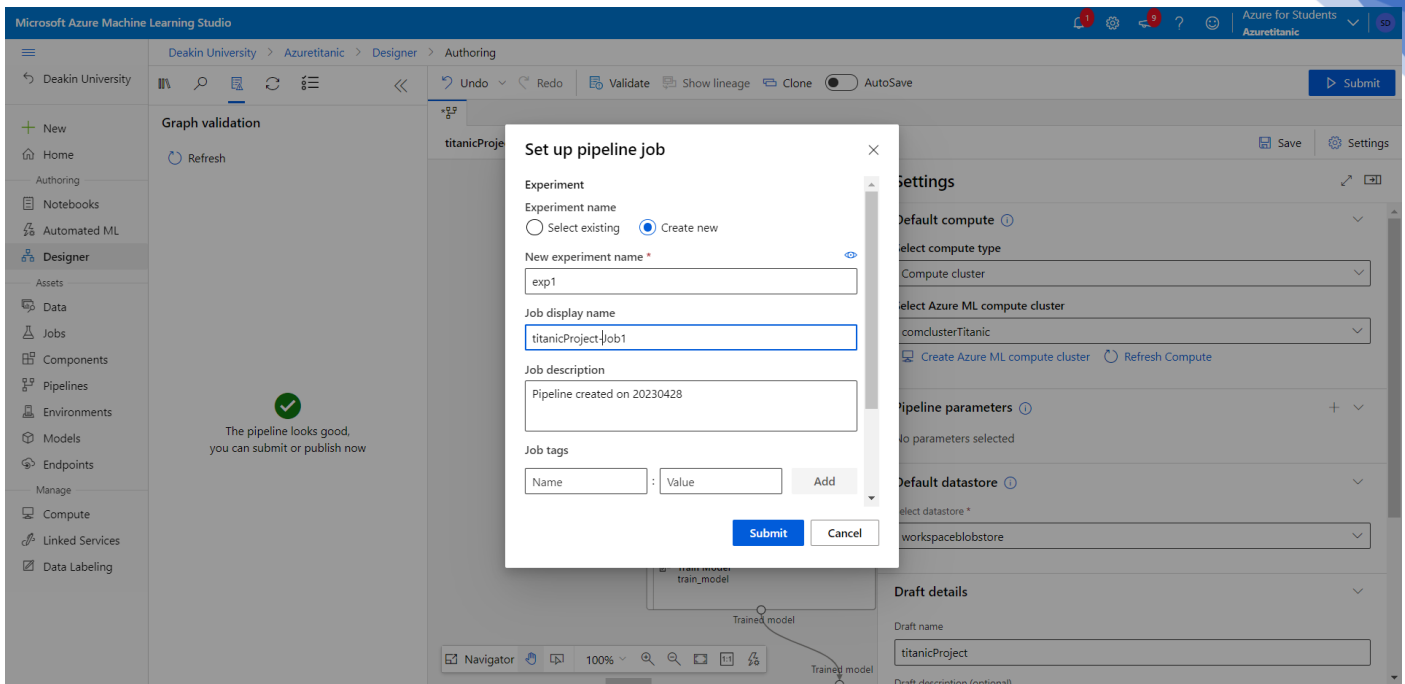
The screenshot shows the 'Compute' section of the Azure Machine Learning Studio interface. The left sidebar contains navigation options like 'New', 'Home', 'Notebooks', 'Automated ML', 'Designer', and 'Compute'. The main area displays a table of compute clusters. The cluster 'comclusterTitanic' is highlighted, showing it is in a 'Succeeded' state with 0 nodes. The table columns include Name, State, Size, Location, Created on, Active runs, Idle nodes, Busy nodes, and Unprovisioned nodes.

Name	State	Size	Location	Created on	Active runs	Idle nodes	Busy nodes	Unprovisioned nodes
comclusterTitanic	Succeeded (0 nodes)	STANDARD_DS3_V2	eastus	Apr 28, 2023 2:35 PM	0	0	0	1

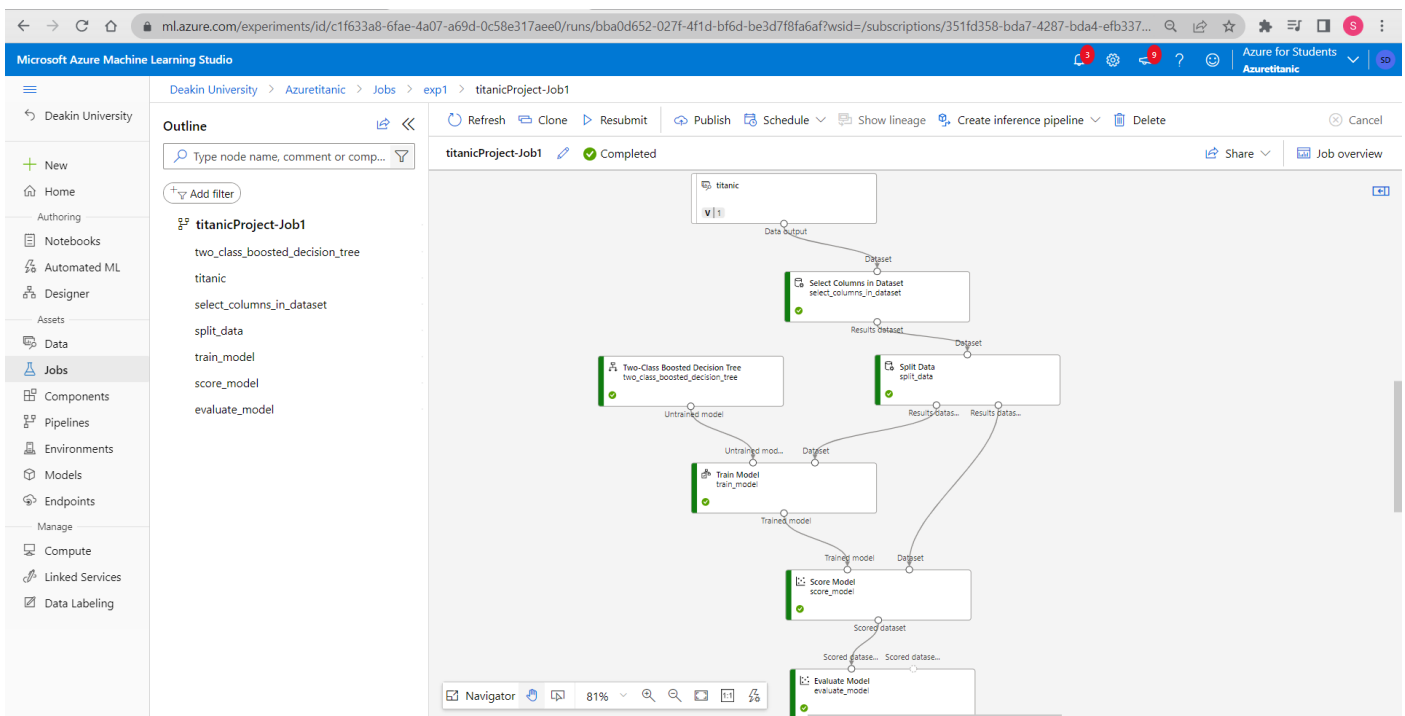
We go back to designer in our pipeline that we saved earlier and click settings on right side of the screen. Here we specify the cluster we created in the previous step and click submit.

The screenshot shows the 'Designer' section of the Azure Machine Learning Studio interface. The left sidebar contains navigation options like 'New', 'Home', 'Notebooks', 'Automated ML', 'Designer', and 'Compute'. The main area displays a pipeline diagram for 'titanicProject'. The 'Settings' panel on the right is open, showing the 'Default compute' section where 'comclusterTitanic' is selected as the Azure ML compute cluster. The 'Pipeline parameters' section is also visible.

We name the experiment as exp1 and also give it a display name and click on submit.



exp1 is created in jobs now so we go to jobs and click on exp1, we can see that the model is trained now.



We then double click on train model and download samplejson, conda dependencies and score.py.

The screenshot shows the Microsoft Azure Machine Learning Studio interface. The top navigation bar indicates the user is logged in as 'Azure for Students'. The left sidebar contains navigation options: Home, Notebooks, Automated ML, Designer, Assets, Data, Jobs, Components, Pipelines, Environments, Models, Endpoints, and Manage. The main workspace displays a workflow diagram for a 'titanicProject-Job1'. The workflow includes nodes for 'Select Columns', 'Split', 'Train Model', and 'Score Model'. The 'Train Model' node is highlighted, and the 'Outputs + logs' tab is active, showing a list of output files: '_meta.yaml', '_samples.json', '_schema.json', 'conda_env.yaml', 'data.learner', 'model_spec.yaml', and 'PY_score.py'. A terminal window on the right shows the execution logs, including file operations and a successful SystemExit.

Now we register the model as Azuretitanicml.

The screenshot shows the Microsoft Azure Machine Learning Studio interface, specifically the 'Model List' section. The top navigation bar indicates the user is logged in as 'Azure for Students'. The left sidebar contains navigation options: Home, Notebooks, Automated ML, Designer, Assets, Data, Jobs, Components, Pipelines, Environments, Models, Endpoints, and Manage. The main workspace displays a table with columns: Name, Version, Type, Source, Experiment, Job (Run ID), and Created on. A single model is listed: 'Azuretitanicml' with version 1, type CUSTOM, source 'This workspace', and experiment 'exp1'. The 'Job (Run ID)' is 'c7cbf3fc-0711-4ec4-b28b-b647...' and it was created on 'Apr 28, 2023 10:44 PM'.

Name	Version	Type	Source	Experiment	Job (Run ID)	Created on
Azuretitanicml	1	CUSTOM	This workspace	exp1	c7cbf3fc-0711-4ec4-b28b-b647...	Apr 28, 2023 10:44 PM

We can find the registered model in models option.

We will first create a Kubernetes cluster before deploying the model as shown below. Kubernetes cluster is named Akstitanic that we created for deployment.

The screenshot shows the Microsoft Azure Machine Learning Studio interface. The left sidebar contains navigation options like 'Deakin University', 'New', 'Home', 'Notebooks', 'Automated ML', 'Designer', 'Assets', 'Data', 'Jobs', 'Components', 'Pipelines', 'Environments', 'Models', 'Endpoints', 'Manage', 'Compute', 'Linked Services', and 'Data Labeling'. The main area is titled 'Compute' and shows a table of Kubernetes clusters. The cluster 'Akstitanic' is highlighted, showing it is in a 'Succeeded' state.

Name	State	Type	Attached/Created	Location	Created on
Akstitanic	Succeeded	AksCompute	Created	eastus	Apr 28, 2023 11:09 PM

We find the model created earlier in models after we enter the model, we can deploy it as webservice.

The screenshot shows the Microsoft Azure Machine Learning Studio interface. The left sidebar is the same as the previous screenshot. The main area is titled 'Models' and shows a model named 'AzureTitanicml:1'. The 'Deploy' button is highlighted, and a dropdown menu shows options for 'Real-time endpoint', 'Batch endpoint', and 'Web service'. The 'Attributes' section shows the model's name, version, creation date, creator, type, and job ID.

Attribute	Value
Name	AzureTitanicml
Version	1
Created on	Apr 28, 2023 10:44 PM
Created by	SAURABH DHARMADHIKARI
Type	CUSTOM
Created by job	c7cbf3fc-0711-4ec4-b28b-b647c22b1467
Asset ID	azureml://locations/eastus/workspaces/09ca9270-1373-46d5-8b39-dd99de140cdb/models/AzureTitanicml/versions/1

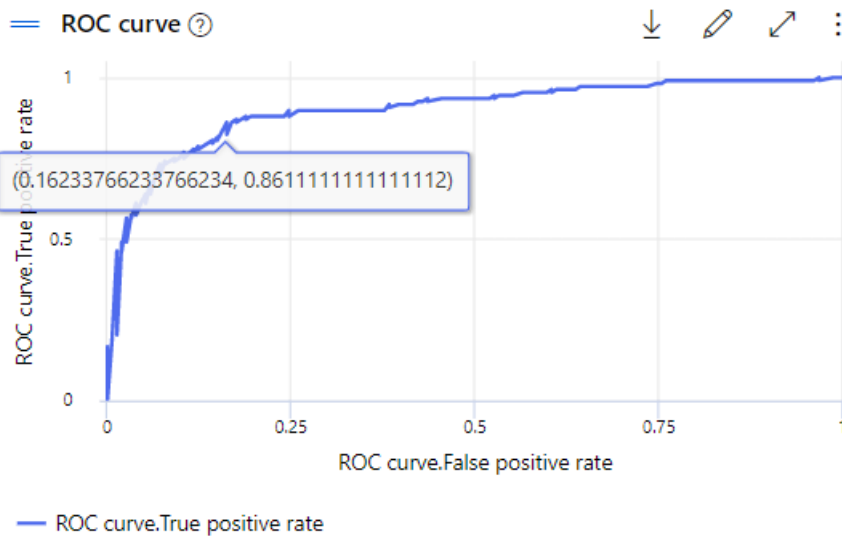
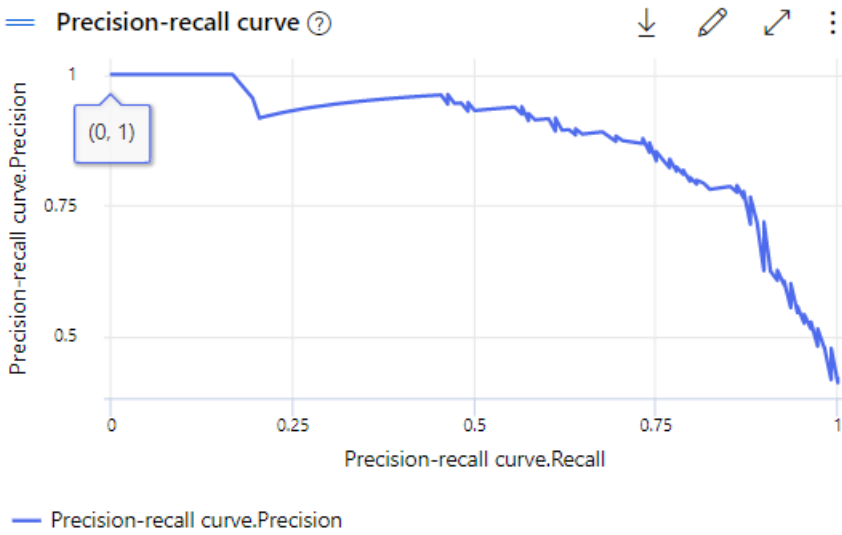
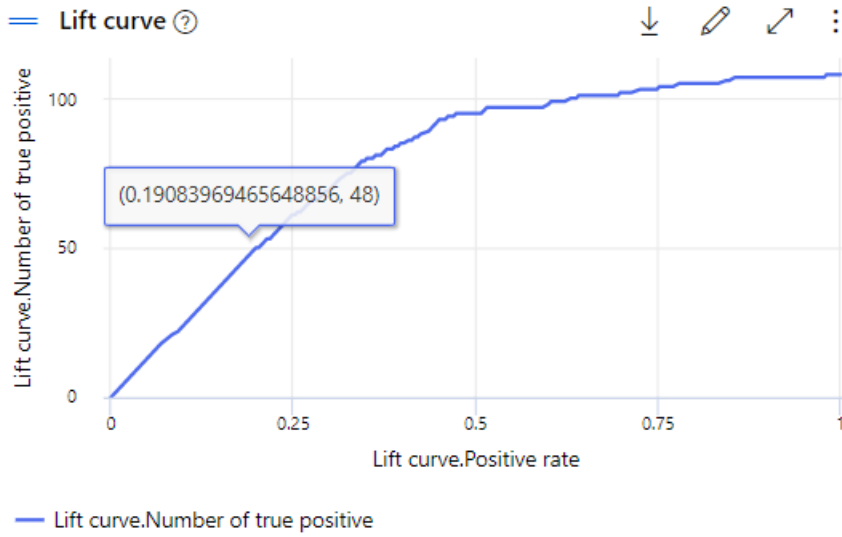
We put compute name as Akstitanic.

The screenshot shows the Microsoft Azure Machine Learning Studio interface. On the left, the 'Models' tab is selected in the sidebar. The main pane displays the details for 'Azuretitanicml:1', including its name, version (1), creation date (Apr 28, 2023 10:44 PM), and creator (SAURABH DHARMADHIKARI). On the right, the 'Deploy a model' dialog is open. The 'Compute name' is set to 'Akstitanic'. The 'Models' dropdown shows 'Azuretitanicml:1'. The 'Enable authentication' toggle is turned on. The 'Type' is set to 'Token-based authentication'. The 'Entry script file' is 'score (1).py' and the 'Conda dependencies file' is 'conda_env (1).yaml'. The 'Dependencies' section has an 'Add File' button. At the bottom right of the dialog are 'Deploy' and 'Cancel' buttons.

Once deployed we can now evaluate and check the performance of the model that we created.

The screenshot shows the Microsoft Azure Machine Learning Studio interface. On the left, the 'Jobs' tab is selected in the sidebar. The main pane displays the 'titanicProject-Job1' workflow, which includes steps like 'two_class_boosted_decision_tree', 'titanic', 'select_columns_in_dataset', 'split_data', 'train_model', 'score_model', and 'evaluate_model'. On the right, the 'Evaluate Model' dialog is open, showing the 'Metrics' tab. The metrics displayed are: Accuracy (0.8396947), AUC (0.9003427), F1 Score (0.7941176), Precision (0.84375), and Recall (0.75). A 'Confusion matrix' section is also visible at the bottom. The 'Chart visualization not available for' message is shown at the bottom right.

Double click evaluate model and after we click metrics we can understand the performance of the model that we created.



In the endpoint option we can find REST API endpoint required by the developer.

The screenshot shows the Microsoft Azure Machine Learning Studio interface. The browser address bar displays the URL: `ml.azure.com/endpoints/lists/realtimeendpoints/mltitanic/consume?wsid=/subscriptions/351fd358-bda7-4287-bda4-efb3375f0ab6/resourcegroups/MDS-deakins/providers/...`. The left sidebar contains navigation options: Deakin University, New, Home, Authoring, Notebooks, Automated ML, Designer, Assets, Data, Jobs, Components, Pipelines, Environments, Models, Endpoints (selected), Manage, Compute, Linked Services, and Data Labeling. The main content area shows the 'mltitanic' endpoint details. The 'Consume' tab is active, displaying the 'Basic consumption info' section with the 'REST endpoint' field containing the URL: `http://20.85.145.203:80/api/v1/service/mltitanic/score`. Below this, the 'Consumption option' section shows 'Consumption types' with tabs for Python, C#, and R. The Python tab is selected, showing a code editor with the following Python code:

```

1  import urllib.request
2  import json
3  import os
4  import ssl
5
6  def allowSelfSignedHttps(allowed):
7      # bypass the server certificate verification on client side
8      if allowed and not os.environ.get('PYTHONHTTPSVERIFY', '') and getattr(ssl, '_create_unverified_context'
9          ssl._create_default_https_context = ssl._create_unverified_context
10
11  allowSelfSignedHttps(True) # this line is needed if you use self-signed certificate in your scoring service.
12
13  # Request data goes here
14  # The example below assumes JSON formatting which may be updated
15  # depending on the format your endpoint expects.
16  # More information can be found here:
17  # https://docs.microsoft.com/azure/machine-learning/how-to-deploy-advanced-entry-script
18  data = {}
  
```

We can also find the codes in different languages here.

References

No references were taken.