



中原大學 雲端計算平台實務

12/17-作業報告

Microsoft Azure AI Fundamentals: Explore  
visual tools for machine learning

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中華民國一一〇年十二月

# 1. Learning Path Intro

Microsoft Azure AI Fundamentals: Explore visual tools for machine learning

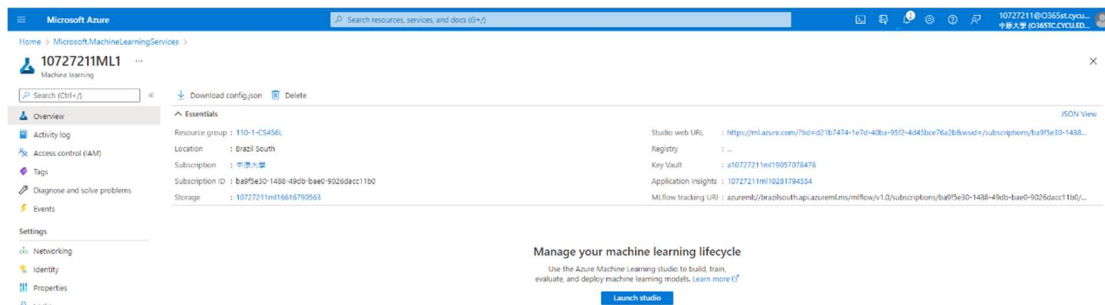
<https://docs.microsoft.com/en-us/learn/paths/create-no-code-predictive-models-azure-machine-learning/>

## Summary Homework Assignment

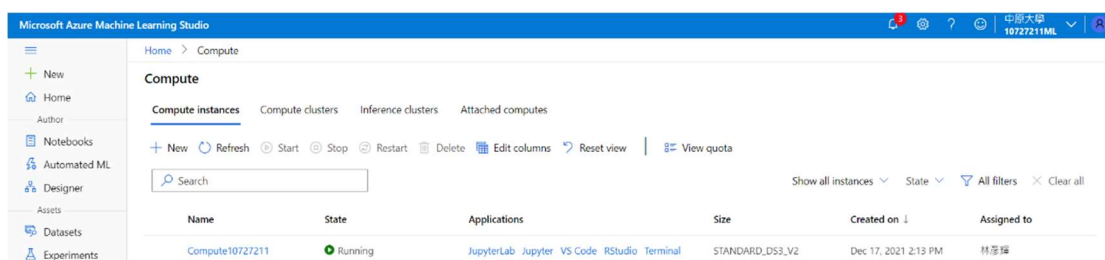
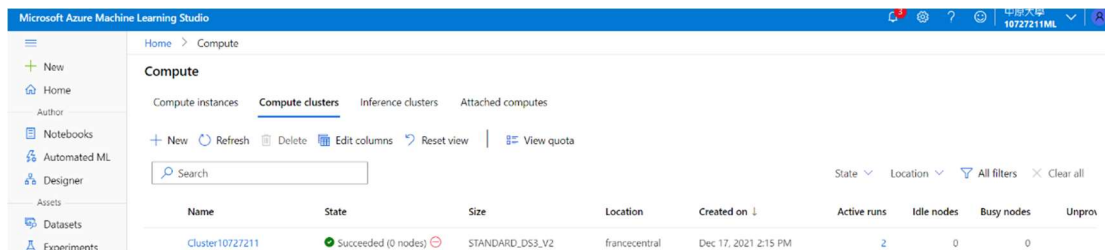
### Model 1: Use automated machine learning in Azure

## Machine Learning

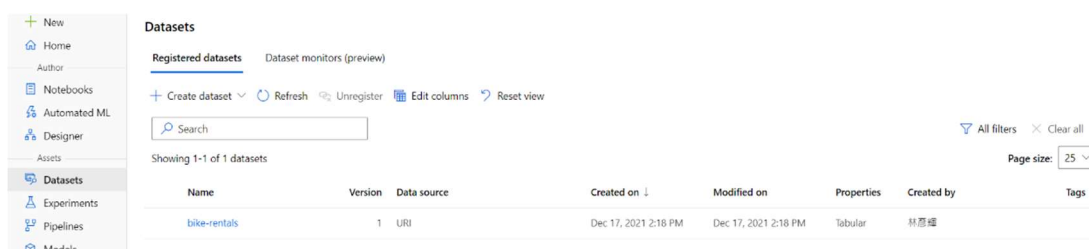
#### 1. Create a Azure Machine Learning workspace



#### 2. Create compute resource



#### 3. Create a dataset



#### 4. Run an automated machine learning experiment

**Learning Studio**

Home > Automated ML > mslearn-bike-rental > heroic\_cat\_81jb9wcl

heroic\_cat\_81jb9wcl

Refresh

Cancel

Delete

Details

Data guardrails

Models

Outputs + logs

Child runs

Snapshots

Properties

Status

Completed

Warning: User specified exit score reached, hence experiment is stopped. Current user specified exit\_score/Metric Score Threshold: 0.08

See more details

Created

Dec 18, 2021 12:08 AM

Started

Dec 18, 2021 12:08 AM

Duration

13m 47.44s

Compute duration

13m 47.44s

Compute target

cluster10727211

Run ID

AutoML\_a401ffd4-ee99-4730-a7b3-d62d2976a736

Script name

--

Created by

林登輝

Input datasets

Input name: training\_data, Dataset: bike-rentals: Version 1

Output datasets

None

Arguments

None

See all properties

Raw JSON

#### 5. Review the best model

Best model summary

Algorithm name

SparseNormalizer, LightGBM

Hyperparameters

View hyperparameters

Normalized root mean squared error

0.07840 View all other metrics

Sampling

100.00 % ⓘ

Registered models

No registration yet

Deploy status

No deployment yet

Run summary

Task type

Regression View configuration settings

Featurization

Auto

Primary metric

Normalized root mean squared error

Experiment name

mslearn-bike-rental

Run Metrics

Explained variance

0.83738

Mean absolute error

176.99

Mean absolute percentage error

39.485

Median absolute error

112.04

Normalized mean absolute error

0.051932

Normalized median absolute error

0.032875

Normalized root mean squared error

0.078396

Normalized root mean squared log error

0.056061

R2 score

0.83598

Root mean squared error

267.17

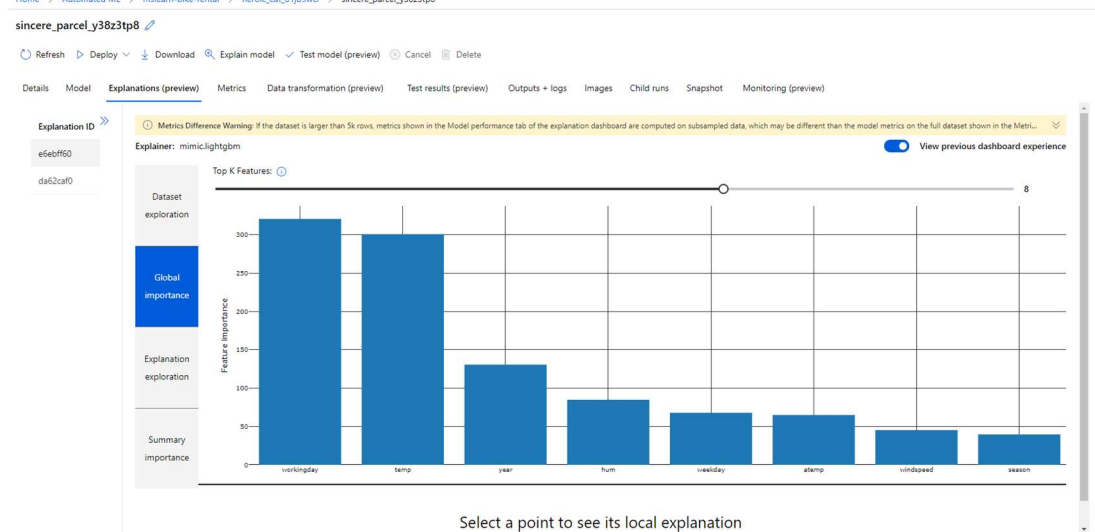
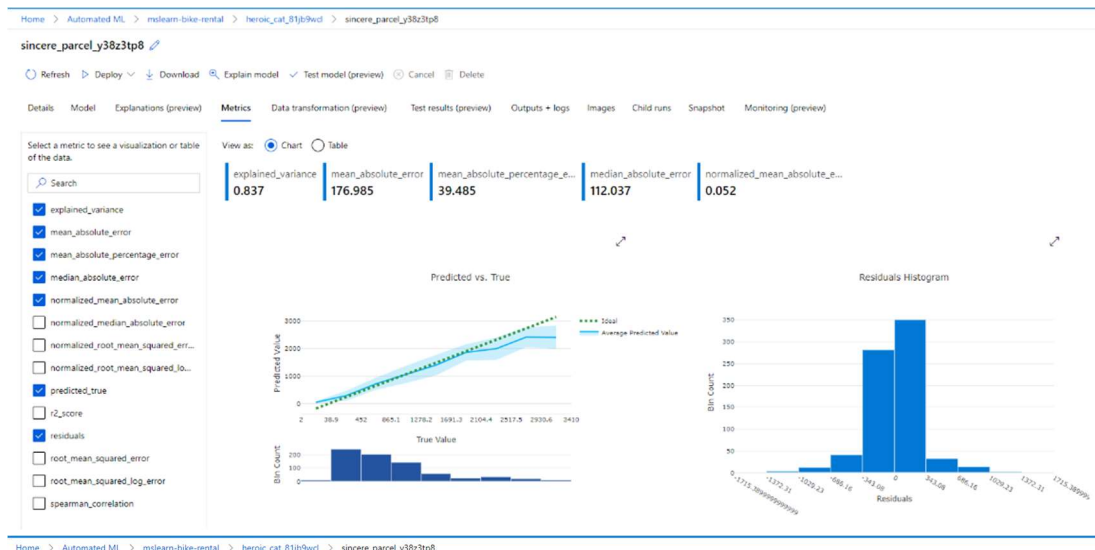
Root mean squared log error

0.39445

Spearman correlation

0.92799

Close



## 6. Deploy a model as a service

Learning Studio

Home > Experiments > mslearn-bike-rental > heroic\_cat\_81jb9wd > sincere\_parcel\_y38z3tp8

sincere\_parcel\_y38z3tp8

Refresh Deploy Download Explain model Test model (preview) Cancel Delete

Success: Model deployment is successfully triggered

Details **Model** Explanations (preview) Metrics Data transformation (preview) Test results (preview) Outputs + logs

**Model summary**

Algorithm name  
SparseNormalizer, LightGBM

Hyperparameters  
[View hyperparameters](#)

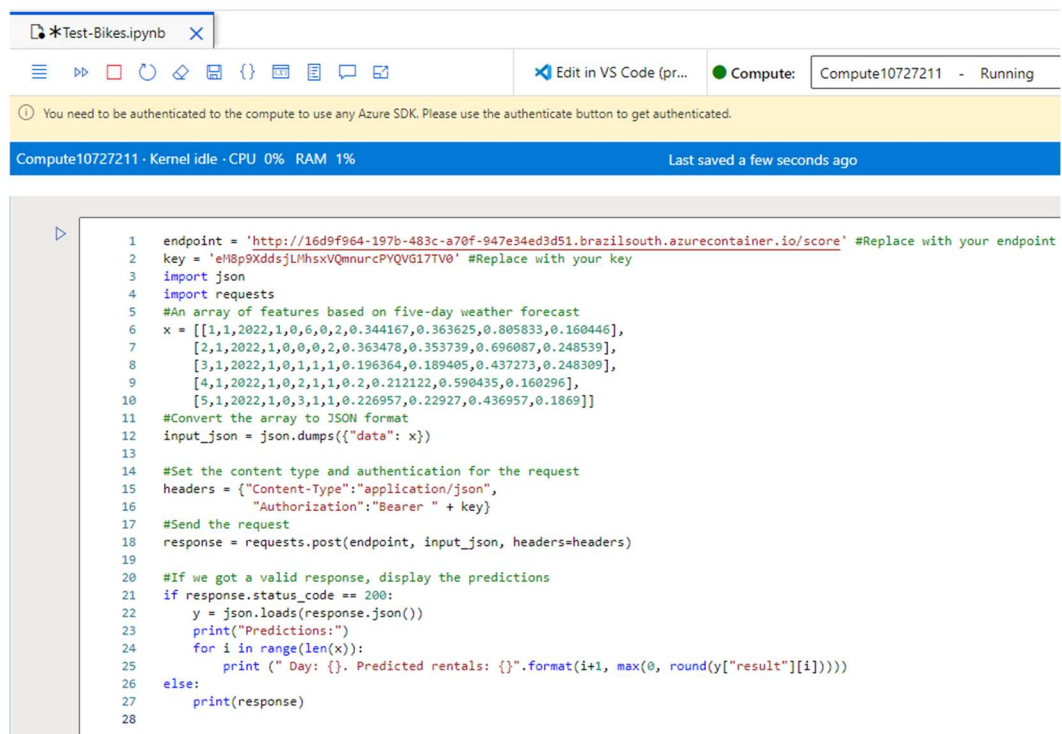
Normalized root mean squared error  
0.07840 [View all other metrics](#)

Sampling  
100.00 %

Registered models  
[AutoMLa401ffd4e9:1](#)

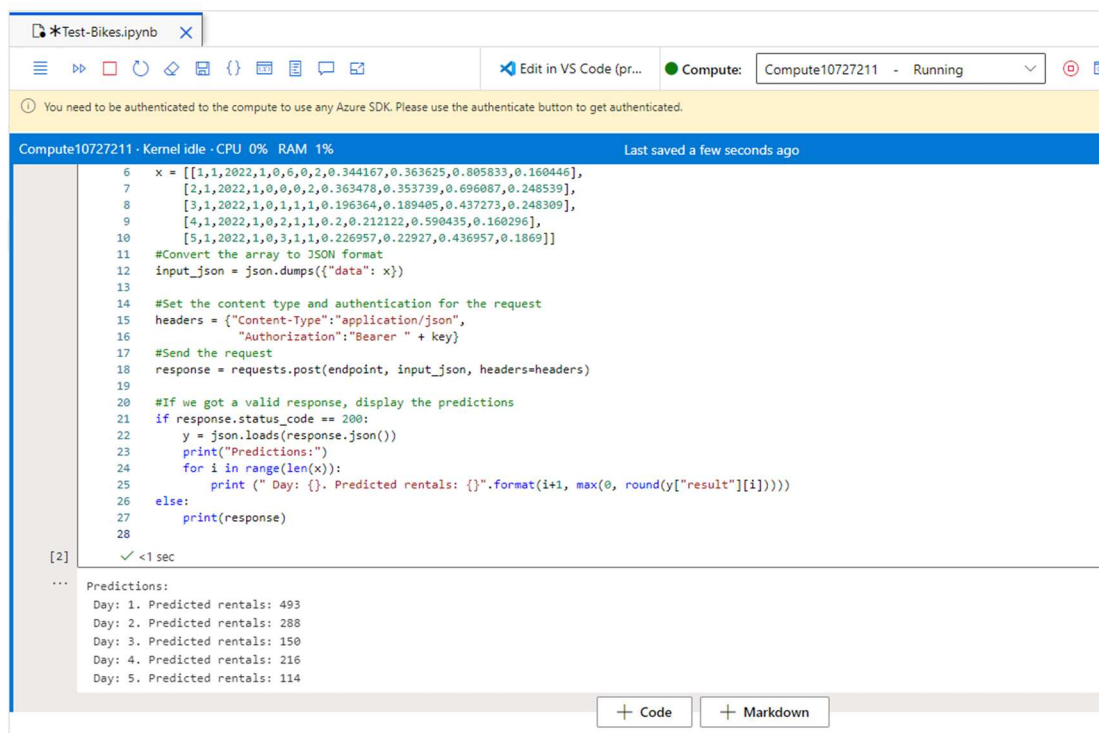
Deploy status  
[predict-rentals](#) Running

## 7. Test the deployed service



The image shows a Jupyter Notebook titled "Test-Bikes.ipynb" in a web-based IDE. The interface includes a top bar with navigation icons, a status bar showing "Compute: Compute10727211 - Running", and a message about authentication. The code in the notebook is as follows:

```
1 endpoint = 'http://16d9f964-197b-483c-a70f-947e34ed3d51.brazilsouth.azurecontainer.io/score' #Replace with your endpoint
2 key = 'eM8p9XddsJLMhsxVQmnrucPYQVG17TV0' #Replace with your key
3 import json
4 import requests
5 #An array of features based on five-day weather forecast
6 x = [[1,1,2022,1,0,6,0,2,0.344167,0.363625,0.805833,0.160446],
7      [2,1,2022,1,0,0,0,2,0.363478,0.353739,0.696087,0.248539],
8      [3,1,2022,1,0,1,1,1,0.196364,0.189405,0.437273,0.248309],
9      [4,1,2022,1,0,2,1,1,0.2,0.212122,0.590435,0.160296],
10     [5,1,2022,1,0,3,1,1,0.226957,0.22927,0.436957,0.1869]]
11 #Convert the array to JSON format
12 input_json = json.dumps({"data": x})
13
14 #Set the content type and authentication for the request
15 headers = {"Content-Type":"application/json",
16            "Authorization":"Bearer " + key}
17 #Send the request
18 response = requests.post(endpoint, input_json, headers=headers)
19
20 #If we got a valid response, display the predictions
21 if response.status_code == 200:
22     y = json.loads(response.json())
23     print("Predictions:")
24     for i in range(len(x)):
25         print (" Day: {}. Predicted rentals: {}".format(i+1, max(0, round(y["result"][i]))))
26 else:
27     print(response)
28
```



The image shows the same Jupyter Notebook after execution. The code is identical to the previous image, but the output is visible below the code cell. The status bar shows "Compute: Compute10727211 - Running" and "Last saved a few seconds ago". The output is as follows:

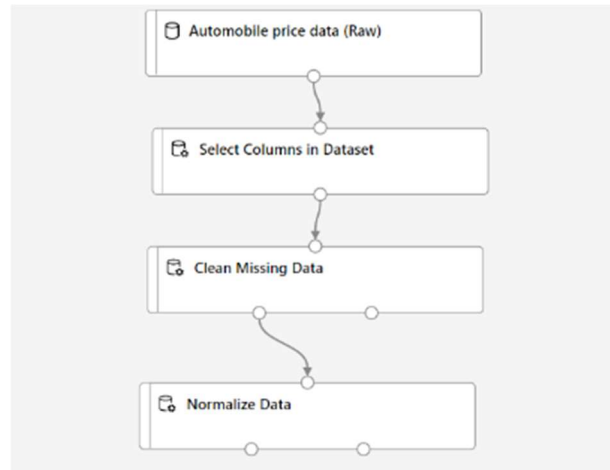
```
[2] ✓ <1 sec
... Predictions:
Day: 1. Predicted rentals: 493
Day: 2. Predicted rentals: 288
Day: 3. Predicted rentals: 150
Day: 4. Predicted rentals: 216
Day: 5. Predicted rentals: 114
```

At the bottom of the notebook, there are buttons for "+ Code" and "+ Markdown".

# Model 2: Create a Regression model with Azure Machine

## Learning designer

1. Create pipeline & Add and explore dataset & Run pipeline



**Set up pipeline run** ×

**Experiment**

☐ Select existing ☒ Create new

New experiment name \* <>

mslearn-auto-training

Job description

Pipeline created on 20211217

☒ Continue on failure step

**Compute target**

Default Cluster10727211

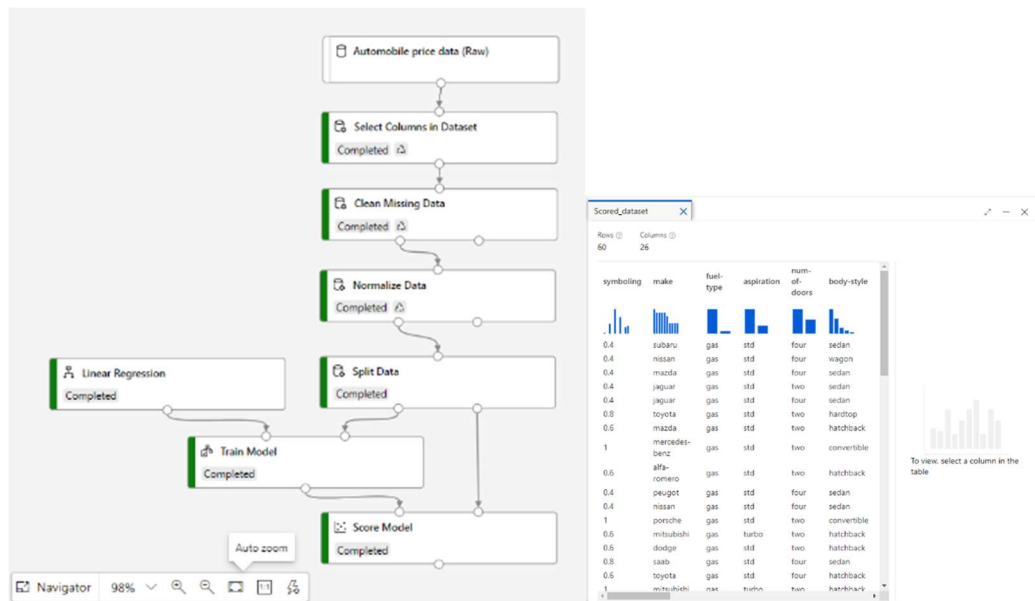
**Submit** **Cancel**

Transformed\_dataset ×

Rows 199 Columns 25

symboling	make	fuel-type	aspiration	num-of-doors	body-style	price
1	alfa-romero	gas	std	two	convertible	13500
1	alfa-romero	gas	std	two	convertible	13500
0.6	alfa-romero	gas	std	two	hatchback	12500
0.8	audi	gas	std	four	sedan	13500
0.8	audi	gas	std	four	sedan	13500
0.8	audi	gas	std	two	sedan	13500
0.6	audi	gas	std	four	sedan	13500
0.6	audi	gas	std	four	wagon	13500
0.6	audi	gas	turbo	four	sedan	13500
0.4	audi	gas	turbo	two	hatchback	13500
0.8	bmw	gas	std	two	sedan	13500
0.4	bmw	gas	std	four	sedan	13500
0.4	bmw	gas	std	two	sedan	13500
0.4	bmw	gas	std	four	sedan	13500
0.6	bmw	gas	std	four	sedan	13500
0.4	bmw	gas	std	four	sedan	13500

## 2. Create and run a training pipeline



## 3. Evaluate a regression model



#### 4. Create and run an inference model



#### 5. Deploy a predictive service & Test the service

```
1 endpoint = 'http://040b1948-5fb6-4ae6-a375-099bcdcf382e.brazilsouth.azurecontainer.io/score' #Replace with your endpoint
2 key = 'DSiUk0NrySpMQrxIeRZz7l0T0X0XVwIR' #Replace with your key
3
4 import urllib.request
5 import json
6 import os
7
8 # Prepare the input data
9 data = {
10     "Inputs": {
11         "WebServiceInput0":
12             [
13                 {
14                     'symboling': 3,
15                     'normalized-losses': None,
16                     'make': 'alfa-romero',
17                     'fuel-type': 'gas',
18                     'aspiration': 'std',
19                     'num-of-doors': 'two',
20                     'body-style': 'convertible',
21                     'drive-wheels': 'rwd',
22                     'engine-location': 'front',
23                     'wheel-base': 88.6,
24                     'length': 168.8,
25                     'width': 64.1,
26                     'height': 48.8,
27                     'curb-weight': 2548,
28                     'engine-type': 'dohc',
29                     'num-of-cylinders': 'four',
30                     'engine-size': 130,
31                     'fuel-system': 'mpfi',
32                     'bore': 3.47,
33                     'stroke': 2.68,
34                     'compression-ratio': 9,
35                     'horsepower': 111,
36                     'peak-rpm': 5000,
37                     'city-mpg': 21,
38                     'highway-mpg': 27,
39                 }
40             ]
41     }
42 }
43
44 body = str.encode(json.dumps(data))
45 headers = {'Content-Type': 'application/json', 'Authorization': ('Bearer ' + key)}
46 req = urllib.request.Request(endpoint, body, headers)
47
48 try:
49     response = urllib.request.urlopen(req)
50     result = response.read()
51     json_result = json.loads(result)
52     y = json_result["Results"]["WebServiceOutput0"][0]
53     print(y)
54 except urllib.error.HTTPError as error:
55     print("The request failed with status code: " + str(error.code))
56
57     # Print the headers to help debug the error
58     print(error.info())
59     print(json.loads(error.read().decode("utf8", 'ignore')))
```

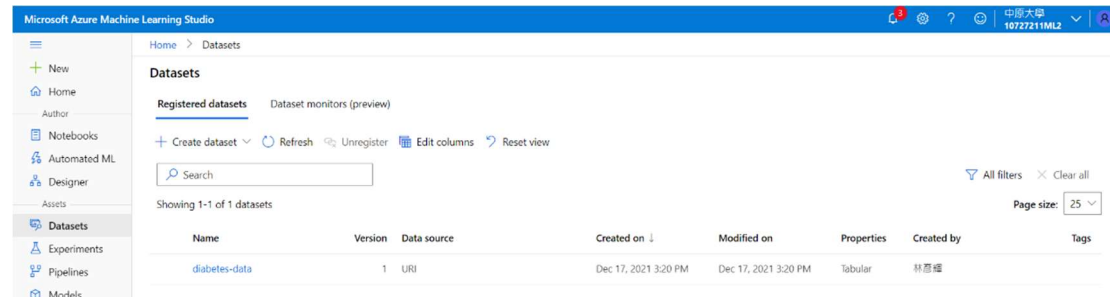
[1] ✓ 1 sec

... {'predicted\_price': 14997.00673125526}

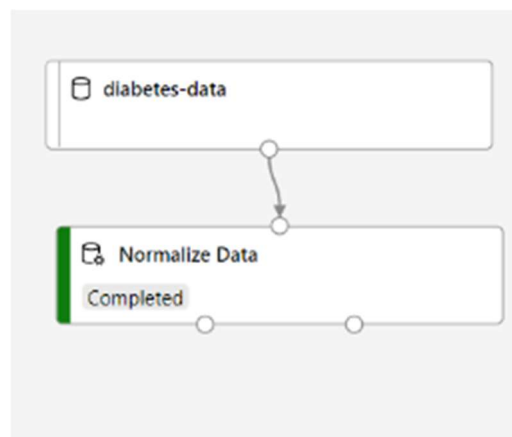


# Model 3: Create a classification model with Azure Machine Learning designer

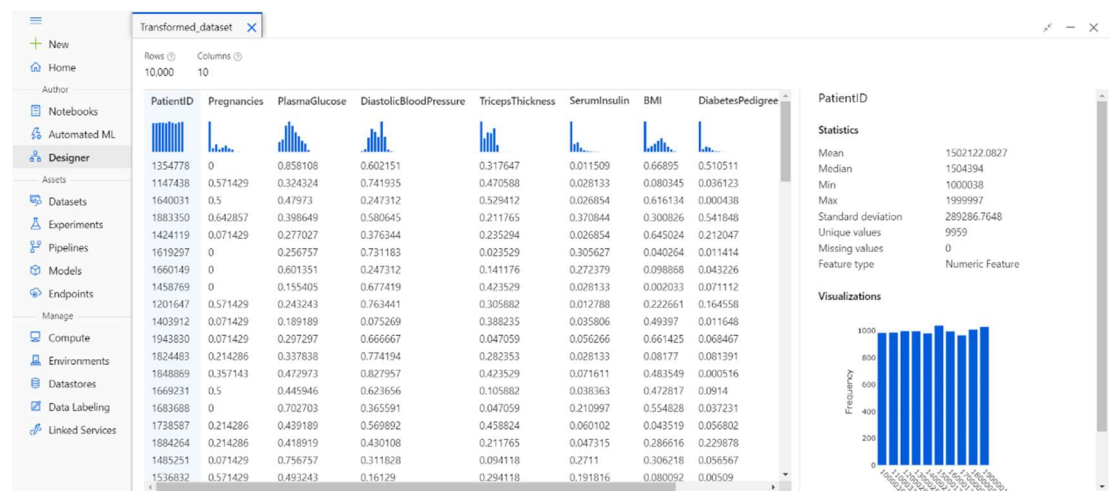
## 1. Create a dataset



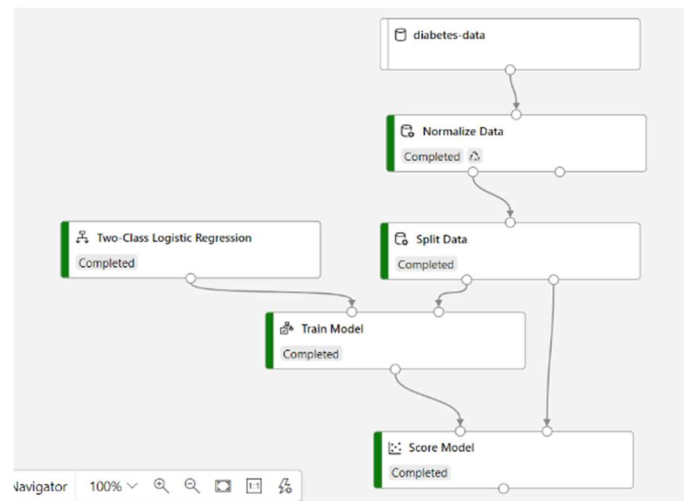
## 2. Create a pipeline & Add Transformations



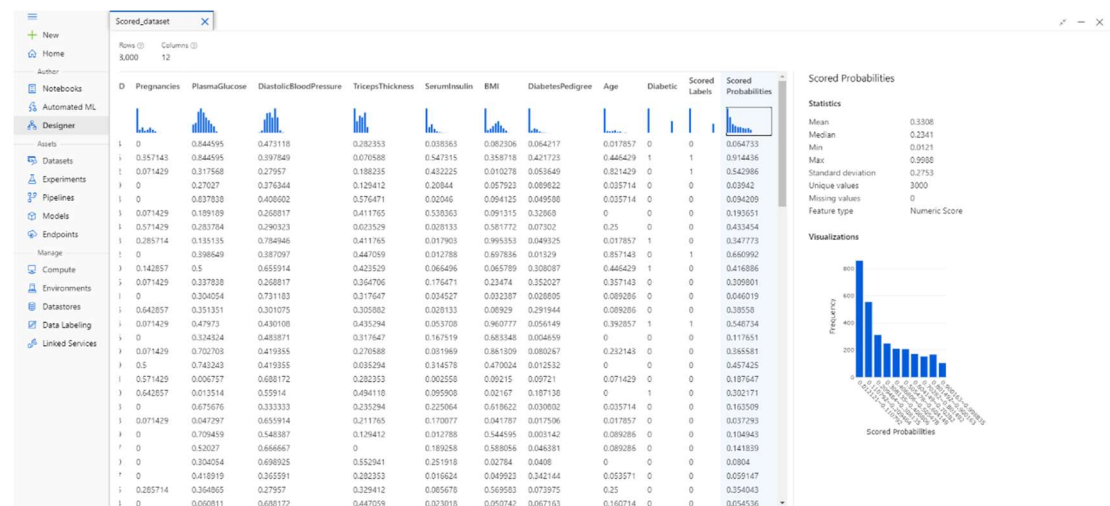
## 3. Run the pipeline & View the transform data



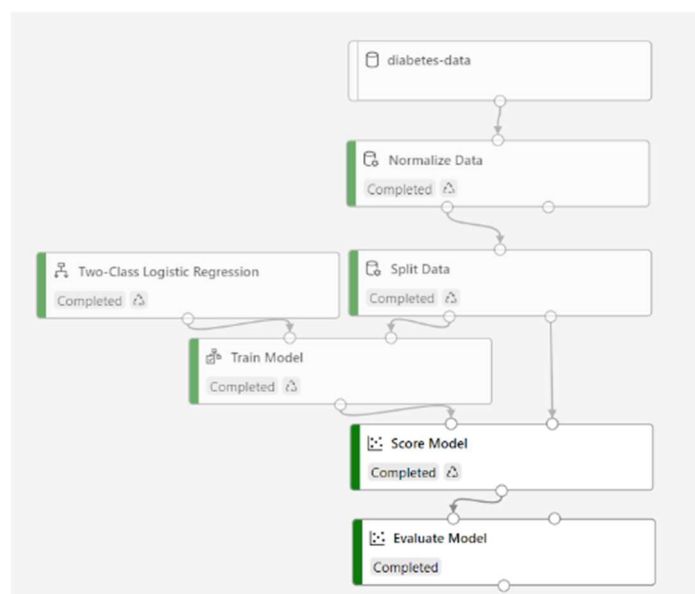
#### 4. Add training models



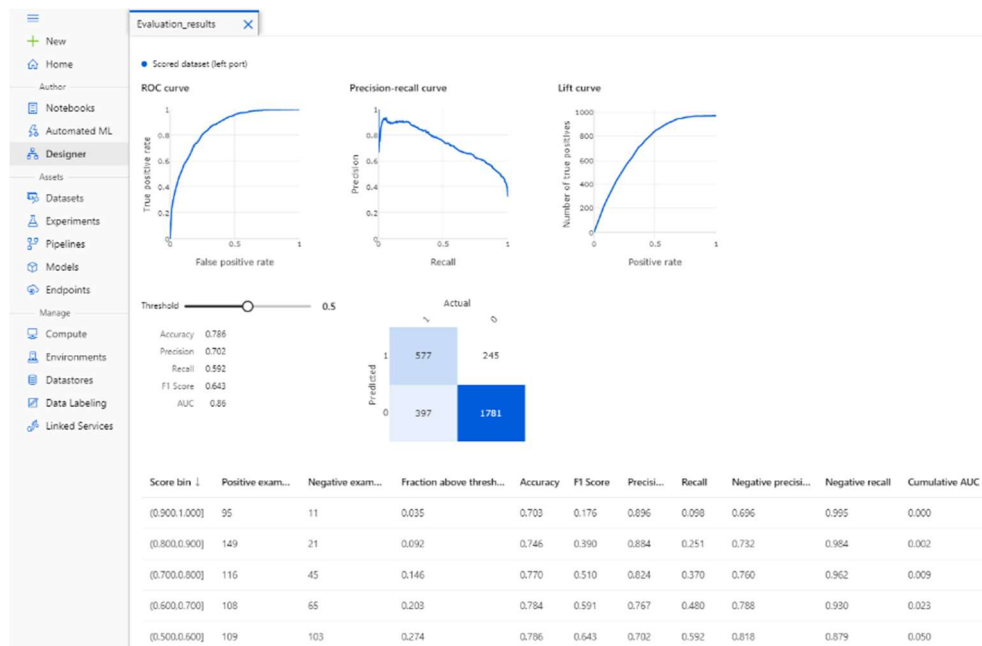
#### 5. Run the training pipeline & View result



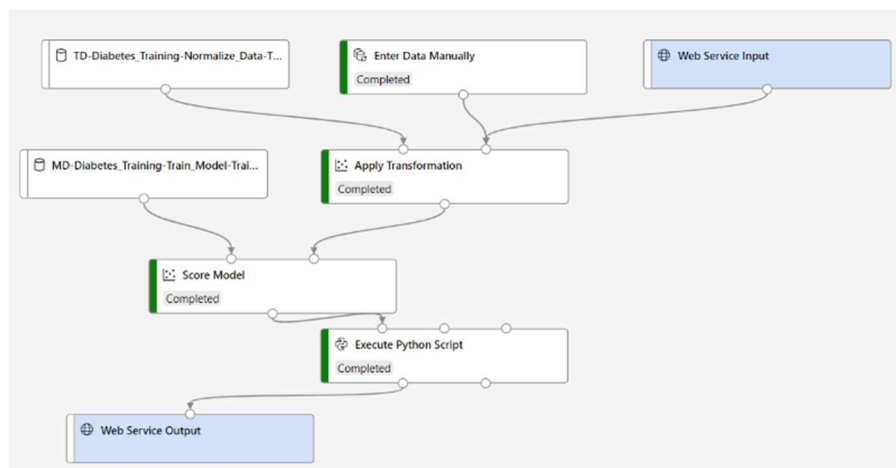
#### 6. Add an evaluate model



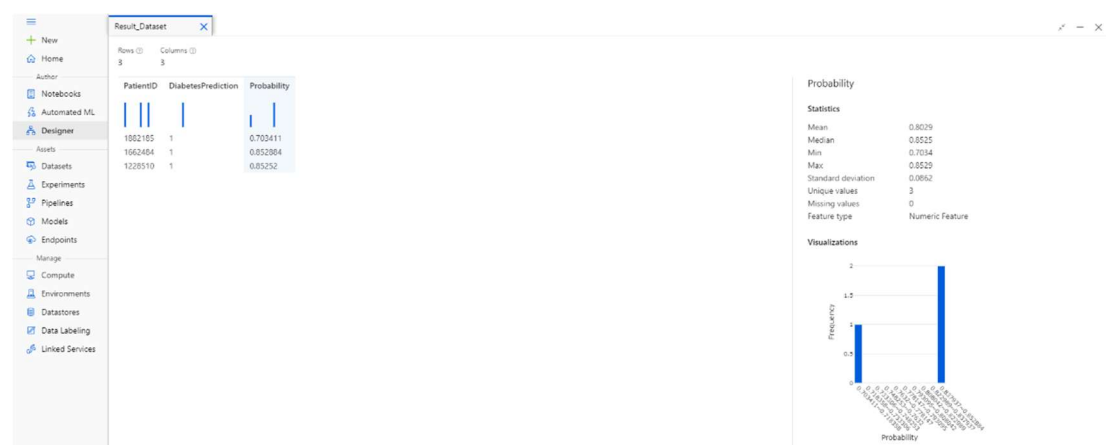
## 7. View Result



## 8. Create an inference pipeline



## 9. Run the pipeline & View result



## 10. Deploy a service & Test the service

Compute10727211 · Kernel idle · CPU 0% RAM 3%Last saved a few seconds ago

```
1 endpoint = "http://357c10ee-119b-436b-b5c9-263f255c17d1.brazilsouth.azurecontainer.io/score" #Replace with your endpoint
2 key = 'yM3TOwQjdhAJVbxQcZULV1Is0XPHICoE' #Replace with your key
3
4 import urllib.request
5 import json
6 import os
7
8 data = {
9     "Inputs": {
10         "WebServiceInput0":
11             [
12                 {
13                     'PatientID': 1882185,
14                     'Pregnancies': 9,
15                     'PlasmaGlucose': 104,
16                     'DiastolicBloodPressure': 51,
17                     'TricepsThickness': 7,
18                     'SerumInsulin': 24,
19                     'BMI': 27.36983156,
20                     'DiabetesPedigree': 1.3504720469999998,
21                     'Age': 43,
22                 },
23             ],
24     },
25     "GlobalParameters": {
26     }
27 }
28
29 body = str.encode(json.dumps(data))
30
31 headers = {'Content-Type': 'application/json', 'Authorization': ('Bearer ' + key)}
32
33 req = urllib.request.Request(endpoint, body, headers)
34
35
36 try:
37     response = urllib.request.urlopen(req)
38     result = response.read()
39     json_result = json.loads(result)
40     output = json_result["Results"]["WebServiceOutput0"][0]
41     print('Patient: {} \n Prediction: {} \n Probability: {:.2f}'.format(output["PatientID"],
42                                                                     output["DiabetesPrediction"],
43                                                                     output["Probability"]))
44 except urllib.error.HTTPError as error:
45     print("The request failed with status code: " + str(error.code))
46
47     # Print the headers to help debug
48     print(error.info())
49     print(json.loads(error.read().decode("utf8", 'ignore')))
```

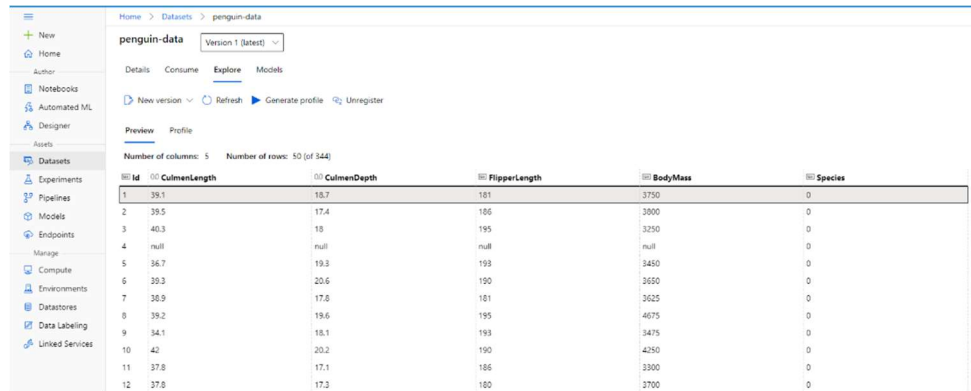
[1] ✓ <1 sec

... Patient: 1882185.0  
Prediction: 1.0  
Probability: 0.70

+ Code + Markdown

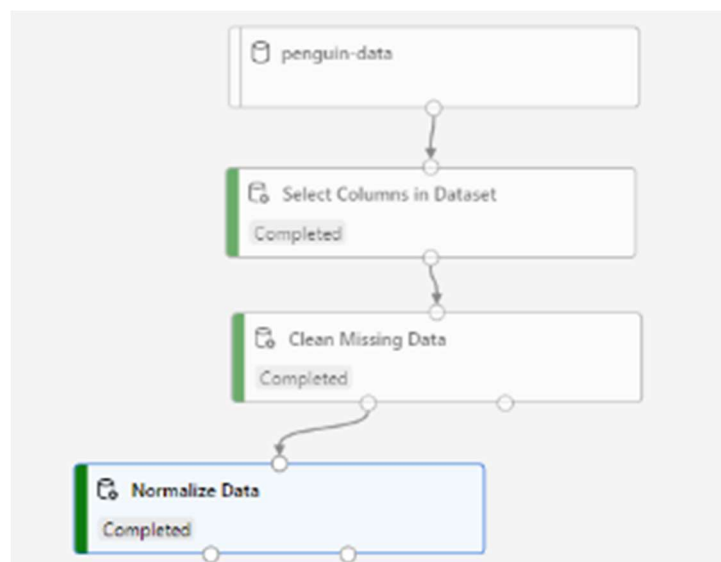
# Model 4: Create a Clustering model with Azure Machine Learning designer

## 1. Create a dataset

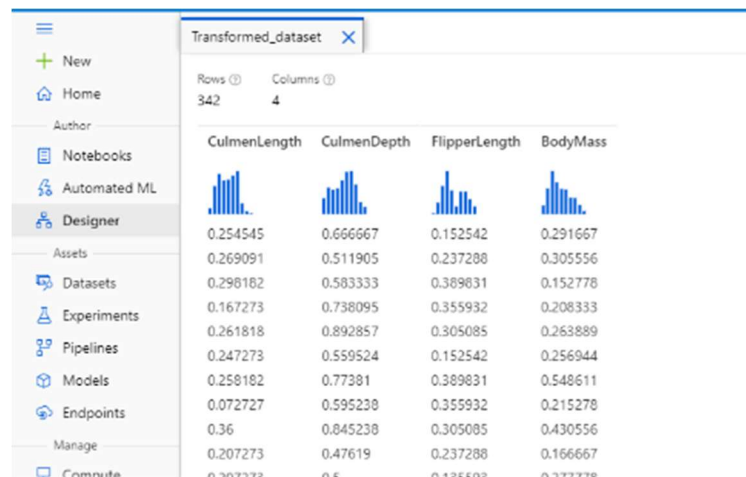


Id	CulmenLength	CulmenDepth	FlipperLength	BodyMass	Species
1	39.1	18.7	181	3750	0
2	39.5	17.4	186	3800	0
3	40.3	18	195	3250	0
4	null	null	null	null	0
5	36.7	19.3	193	3450	0
6	39.3	20.6	190	3650	0
7	38.9	17.8	181	3625	0
8	39.2	19.6	195	4675	0
9	34.1	18.1	193	3475	0
10	42	20.2	190	4250	0
11	37.8	17.1	186	3300	0
12	37.8	17.3	180	3700	0

## 2. Create a pipeline & Apply Transformations

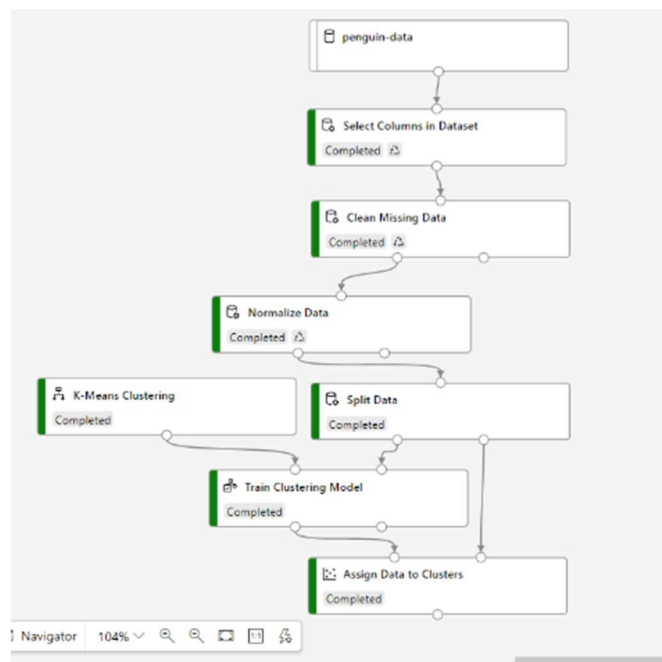


## 3. View the transformed data



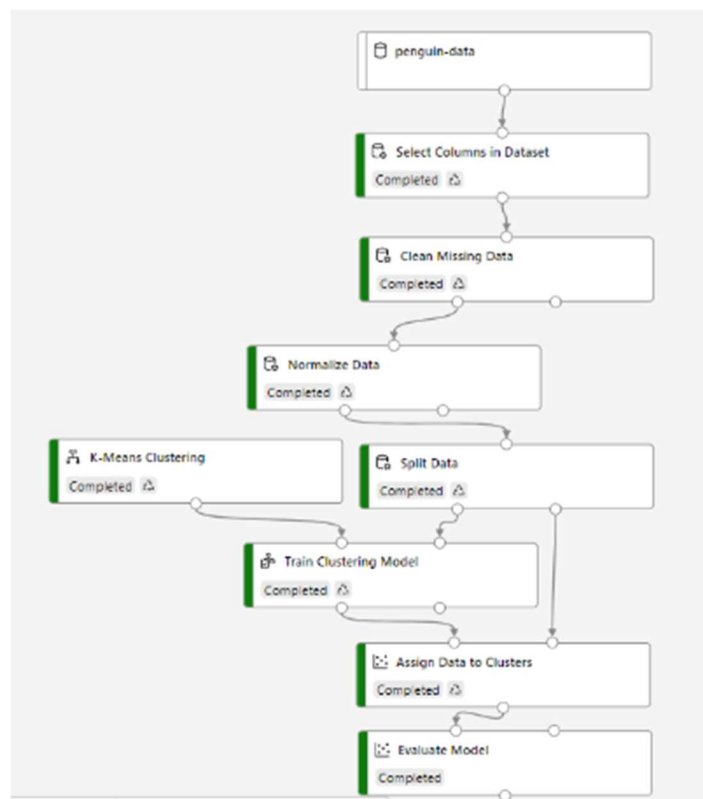
CulmenLength	CulmenDepth	FlipperLength	BodyMass
0.254545	0.666667	0.152542	0.291667
0.269091	0.511905	0.237288	0.305556
0.298182	0.583333	0.389831	0.152778
0.167273	0.738095	0.355932	0.208333
0.261818	0.892857	0.305085	0.263889
0.247273	0.559524	0.152542	0.256944
0.258182	0.77381	0.389831	0.548611
0.072727	0.595238	0.355932	0.215278
0.36	0.845238	0.305085	0.430556
0.207273	0.47619	0.237288	0.166667

#### 4. Add training models



#### 5. Run the training pipeline & View results

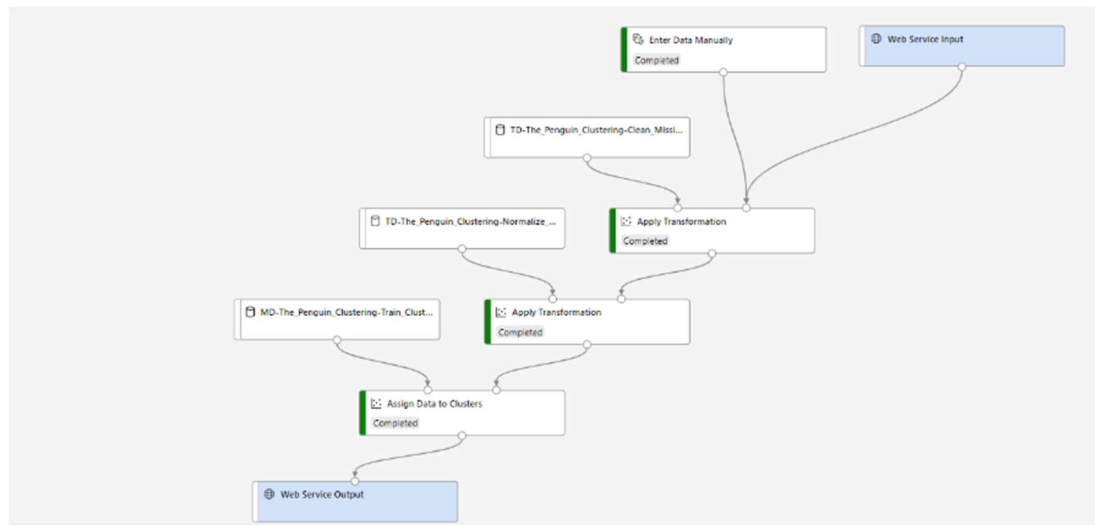
## 6. Add and Evaluate Model



## 7. View Result

Evaluation_results				
Result Description	Average Distance to Other Center	Average Distance to Cluster Center	Number of Points	Maximal Distance to Cluster Center
Evaluation For Cluster No.0	0.431223	0.193987	47	0.457334
Evaluation For Cluster No.1	0.725174	0.22048	33	0.399237
Evaluation For Cluster No.2	0.469312	0.256118	23	0.461864
Combined Evaluation	0.533907	0.216349	103	0.461864

## 8. Create an inference pipeline



## 9. View Result

Results_dataset								
Rows		Columns						
3		8						
CulmenLength	CulmenDepth	FlipperLength	BodyMass	Assignments	DistancesToClusterCenter no.0	DistancesToClusterCenter no.1	DistancesToClusterCenter no.2	
0.254545	0.666667	0.152542	0.291667	0	0.161106	0.945574	0.409129	
0.618182	0.202381	0.813559	0.680556	1	0.93688	0.081267	0.744505	
0.527273	0.559524	0.355932	0.305556	2	0.323781	0.683342	0.152137	



## 10. Deploy a service and test the service

```
1 endpoint = 'http://5ddb2211-1534-40ff-a7cd-b7e13779eddc.breilsouth.azurecontainer.io/score' #Replace with your endpoint
2 key = 'kqjensTXRdgdFPMx71qoalvnycyFsfco' #Replace with your key
3
4 import urllib.request
5 import json
6 import os
7
8 data = {
9     "Inputs": {
10         "WebServiceInput0":
11             [
12                 {
13                     "CulmenLength": 49.1,
14                     "CulmenDepth": 4.8,
15                     "FlipperLength": 1220,
16                     "BodyMass": 5150,
17                 },
18             ],
19         "GlobalParameters": {
20             }
21     }
22 }
23
24 body = str.encode(json.dumps(data))
25
26 headers = {'Content-Type': 'application/json', 'Authorization': ('Bearer ' + key)}
27
28 req = urllib.request.Request(endpoint, body, headers)
29
30
31 try:
32     response = urllib.request.urlopen(req)
33     result = response.read()
34     json_result = json.loads(result)
35     output = json_result["Results"]["WebServiceOutput0"][0]
36     print('Cluster: {}'.format(output["Assignments"]))
37
38 except urllib.error.HTTPError as error:
39     print("The request failed with status code: " + str(error.code))
40
41     # Print the headers to help debug
42     print(error.info())
43     print(json.loads(error.read().decode("utf8", 'ignore')))
```

[1] ✓ <1 sec

Cluster: 1

+ Code + Markdown

## Take screenshots of Badges and Trophies

The screenshot displays a user profile for Yan-Hui Lin (10727211@O365stg.cuedu.hk). The profile includes statistics: 39 Badges, 6 Trophies, 1 Reputation, 0 Answers accepted, 0 Following, and 0 Followers. A progress bar indicates Level 6 with a score of 55650/72099 XP. The 'Trophies' section lists six achievements:

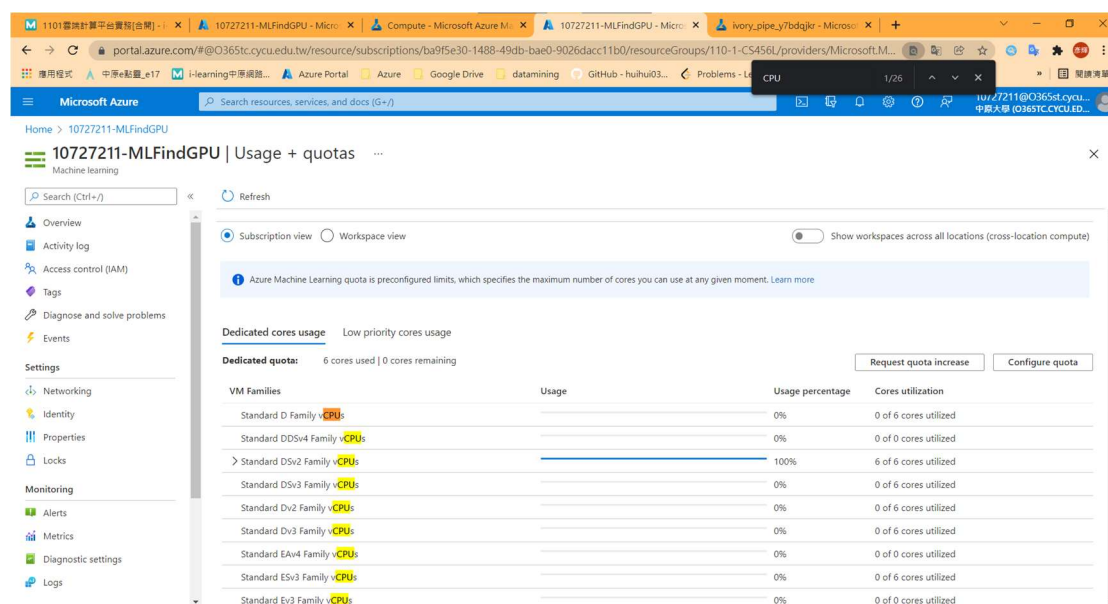
- Trophy:** Microsoft Azure AI Fundamentals: Explore visual tools for machine learning (Completed on 12/18/2021)
- Trophy:** Microsoft Azure AI Fundamentals: Explore natural language processing (Completed on 12/11/2021)
- Trophy:** Store data in Azure (Completed on 11/27/2021)
- Trophy:** Create serverless applications (Completed on 11/23/2021)
- Trophy:** Administer containers in Azure (Completed on 11/21/2021)
- Trophy:** Azure Fundamentals part 1: Describe core Azure concepts (Completed on 8/18/2021)

## Learned from the Learning Path

Azure Machine Learning 在機器學習中提供相當多的參數，從計算資源、演算法、GUI 介面、部署方式、計算模型、驗證模型到分析模型，Azure 的 ML 提供的計算相當多元，不管是資料前處理、模型訓練、驗證、甚至圖表分析都有 Model 支援，且分析的結果數據相當多元。Pipeline 的 GUI 概念讓開發人員可以更專注於他們的開發工作，Learning Path 過程中也從中習得不少 ML 的演算法知識與驗證觀念。

## Problems

1. 部分的 Region 可能資源不夠，Run Model 時可能因 Training 時間過長而導致系統主動判斷 Timeout Error。
2. 可以考慮開啟 GPU 模式（所有的地域都無法看見可用 GPU 之資源）。



## FeedBack

透過 Azure Machine Learning 的 Pipeline，從中體認 Azure 對於「No Code」支援的強大，使用者可以將實作需要的理論透過 GUI Block 的方式實作想要的應用，且這些 GUI 也支援防呆機制，在 Pipeline IO 加入審查機制，讓使用者可以減少 Debug 的過程，Azure 提供的演算法也相當多元，最後也可以透過 GUI 將 Model 做部署至 Container Instance 供及時 Inference、Predict，供對於機器學習的初心者而言，Azure Machine Learn 是一個相當適合的環境。