Predict Patient diabetes using Microsoft Azure ML Work Bench

Prerequisites

- Valid Azure Subscription
- Azure Machine Learning Workbench

Steps Involved

- Download and Understanding DataSet
- Build Model

Objective

The objective of the dataset is to diagnostically predict whether a patient has diabetes, based on certain diagnostic measurements included in the dataset. Several constraints were placed on the selection of these instances from a larger database. All patients here are females at least 21 years old of Pima Indian heritage.

The Tool

Azure Machine Learning services (preview) aka Azure ML workbench are an integrated, end-to-end data science and advanced analytics solution. It helps **professional data scientists** prepare data, develop experiments, and deploy models at cloud scale.

Download and Understanding Dataset

The datasets consist of several medical predictor variables and one target variable, Outcome. Predictor variables includes the number of pregnancies the patient has had, their BMI, insulin level, age, and so on

// Please download dataset from https://aka.ms/aidevdaysds2

Column Name	Description	Data Type
Pregnancies	Number of times pregnant	Numeric
Glucose	Plasma glucose concentration a 2 hour in an oral glucose tolerance test	Numeric
BloodPressure	Diastolic blood pressure (mm Hg)	Numeric
SkinThickness	Triceps skin fold thickness (mm)	Numeric
Insulin	2-Hour serum insulin (mu U/ml)	Numeric
BMI	Body mass index (weight in kg/(height in m)^2)	Numeric
DiabetesPedigreeFunction	Diabetes pedigree function	Numeric
Age	Age (years)	Numeric
Outcome	Class variable (0 or 1)	Numeric

Note:- Datasource Reference: - https://www.kaggle.com/uciml/pima-indians-diabetes-database/data

Evaluating Data Fields

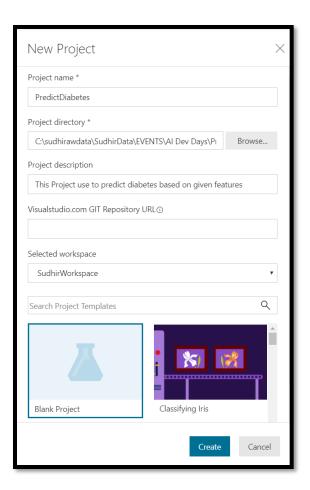
➤ Open "Azure Machine Learning Workbench"



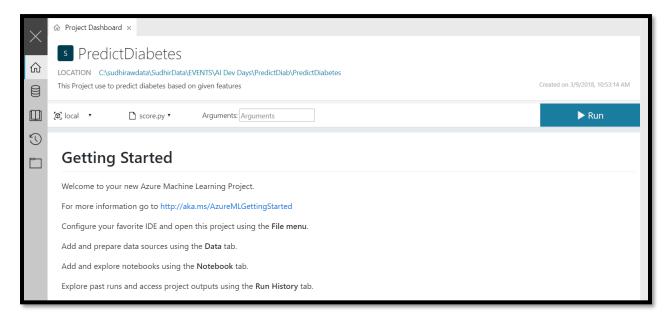
➤ Click + and **New Project** to add new project



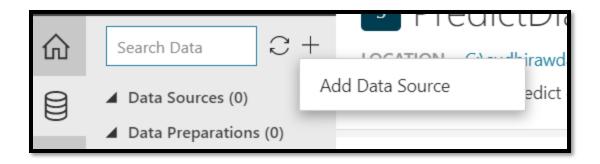
Provide Project Name, Project directory, Project Description, Selected Workspace and select Blank Project. Click Create



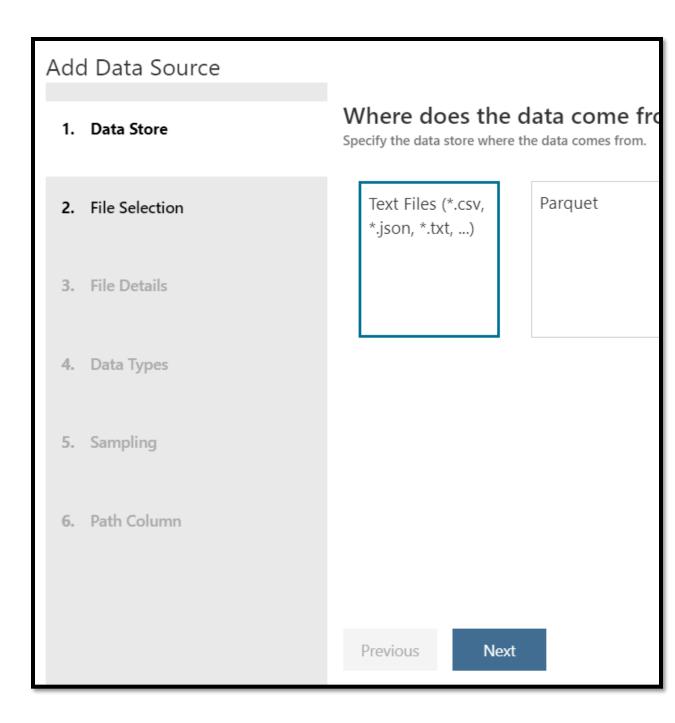
Once created it will look like below



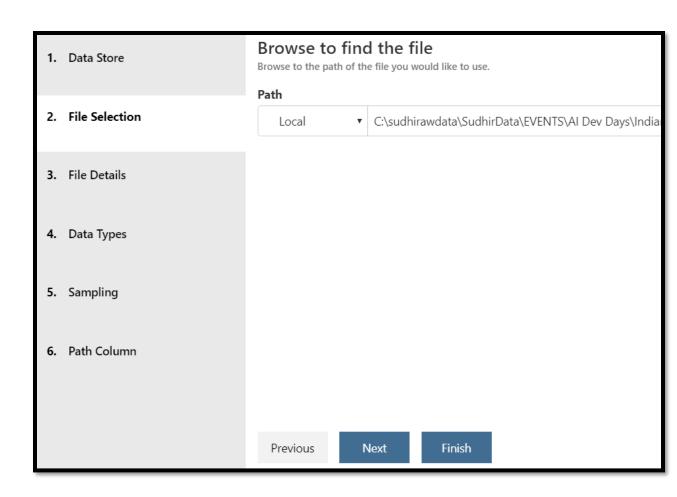
> Click Data (next icon after home) and Click + and Add Data Source



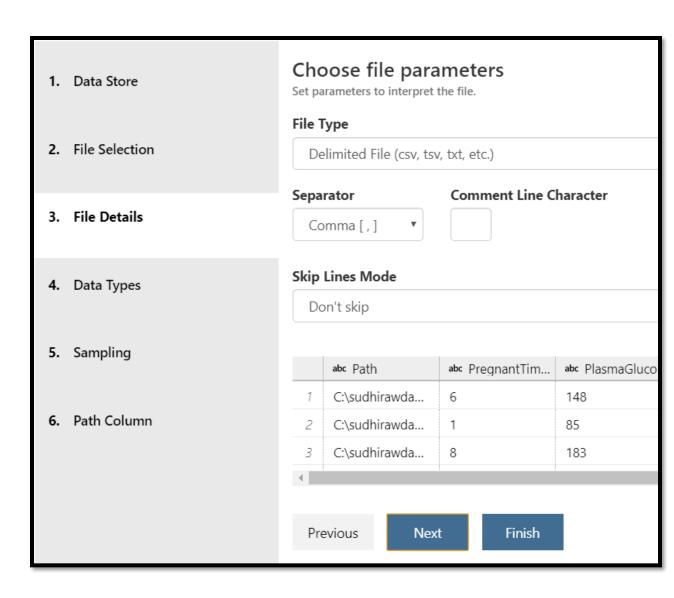
> Select **Text Files** and **Next**



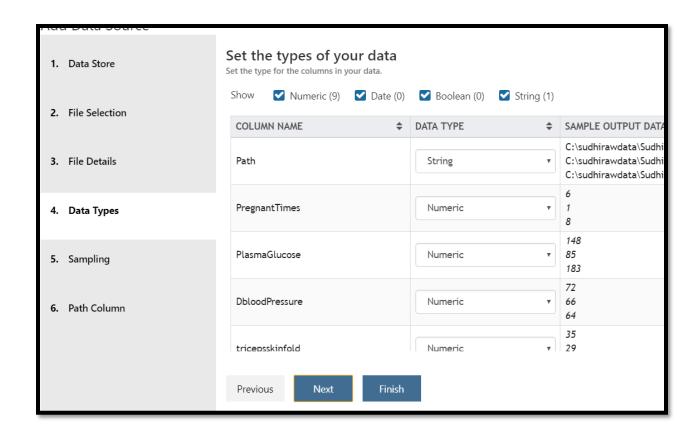
> Select the file which downloaded in earlier step, Click **Next**



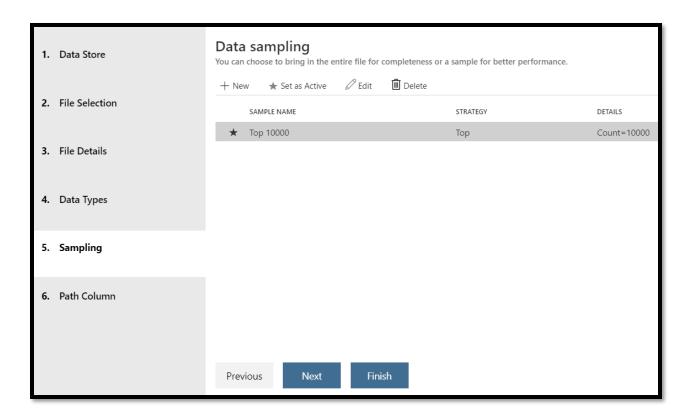
> On step number 3, leave it default and click Next



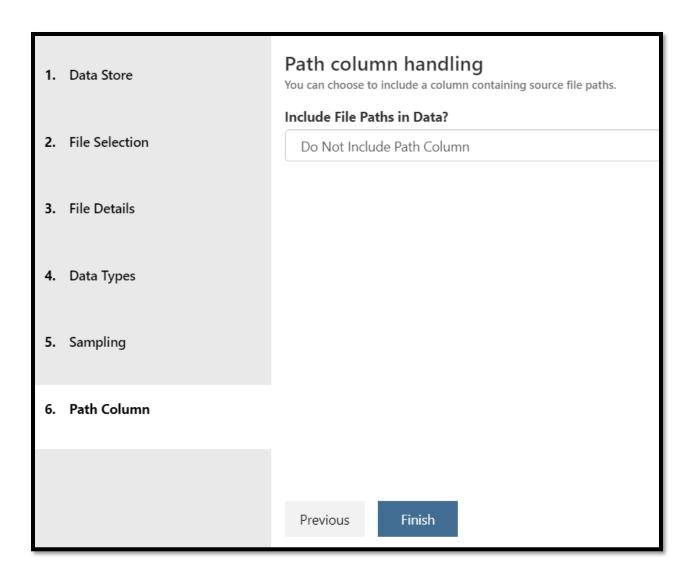
Check the datatype on step number 4. All should be numeric except first column (Path).
Click Next



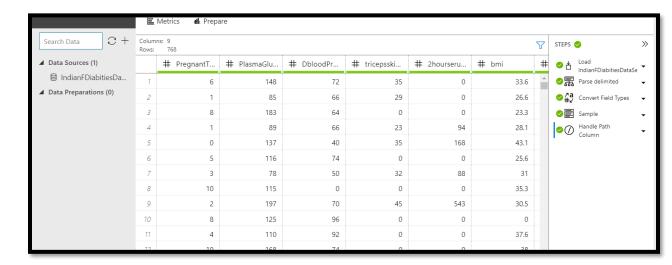
Click Next on step number 5



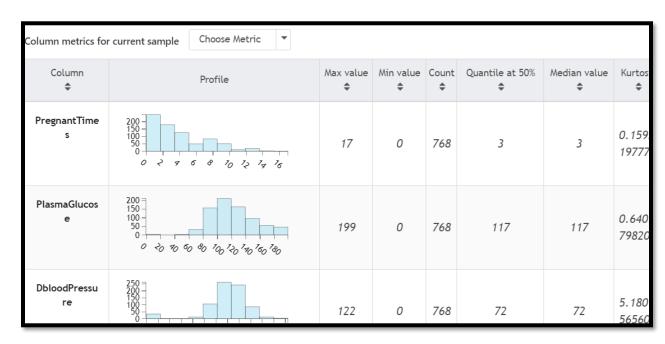
> On step number 6 (Path Column), Select "Do Not Include Path Column" and click Finish



> Once setup, screen will look like

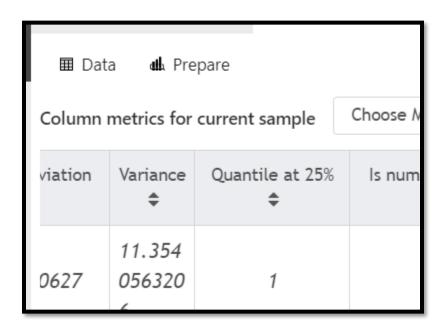


Click on Matrix. Data Profiling will help you identify problems in your data such as invalid dates, missing values etc.

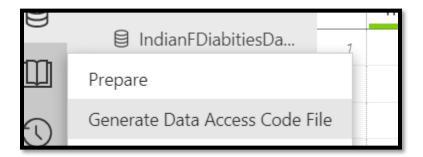


*Note: - Since data is clean we are not going to do any transformation

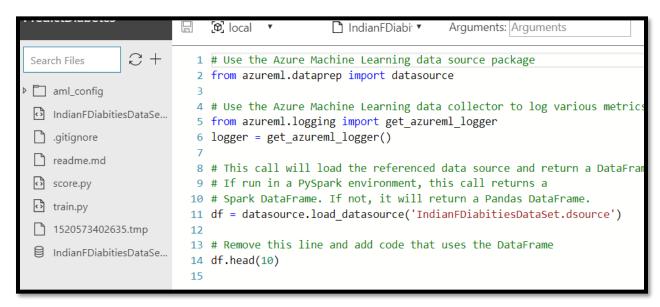
> Click Data



Right click on Data Source and click Generate Data Access Code File



> After generating data access code file, the screen will look like



Building Machine Learning Model using scikitlearn

Copy below code and paste it in editor

import references import pickle import pandas as pd from sklearn.metrics import accuracy_score from sklearn.model_selection import train_test_split from sklearn.tree import DecisionTreeClassifier import numpy as np

Use the Azure Machine Learning data source package from azureml.dataprep import datasource

The screen will look like below after adding above code

```
1 # import references
2 import pickle
3 import pandas as pd
4 from sklearn.metrics import accuracy_score
5 from sklearn.model_selection import train_test_split
6 from sklearn.tree import DecisionTreeClassifier
7 import numpy as np
9 # Use the Azure Machine Learning data source package
10 from azureml.dataprep import datasource
12 # Use the Azure Machine Learning data collector to log various metrics
13 from azureml.logging import get_azureml_logger
14 logger = get azureml logger()
16 # This call will load the referenced data source and return a DataFrame.
17 # If run in a PySpark environment, this call returns a
18 # Spark DataFrame. If not, it will return a Pandas DataFrame.
19 df = datasource.load_datasource('IndianFDiabitiesDataSet.dsource')
21 # Remove this line and add code that uses the DataFrame
22 df.head(10)
```

Comment code # df.head(10) and copy below code and paste it to the editor. In this piece of code we are defining columns which algorithm used to define patterns

> After changes

```
9 # Use the Azure Machine Learning data source package
10 from azureml.dataprep import datasource
12 # Use the Azure Machine Learning data collector to log various metrics
13 from azureml.logging import get_azureml_logger
14 logger = get_azureml_logger()
16 # This call will load the referenced data source and return a DataFrame.
17 # If run in a PySpark environment, this call returns a
18 # Spark DataFrame. If not, it will return a Pandas DataFrame.
19 df = datasource.load datasource('IndianFDiabitiesDataSet.dsource')
21 # Remove this line and add code that uses the DataFrame
22 # df.head(10)
23
24 # Capture features from the given dataset
25 features=['PlasmaGlucose', 'bmi', 'age', 'tricepsskinfold', 'PregnantTimes', 'diabetespedigreefunction',
          'DbloodPressure', '2hourseruminsulin']
27 X = df[features].copy()
28 print("Features")
29 print(X)
```

Copy below code and paste it in editor. In this piece of code, we are defining the target column

```
#Target column
label_features=['isdiabetic']
Y = df[label_features].copy()
```

Below is the screenshot after pasting the code

Copy below code and paste it in editor. In this example we are splitting the dataset to avoid any overfitting and underfitting scenario. Next we are defining the algorithm (DecisionTreeClassifier) to solve this problem and used fit method to build a decision tree classifier from the training set.

```
# Split Data into Training and Test
# test_size = .30 defines that training data is 70% and test is 30%

X_train, X_test, y_train, y_test = train_test_split(X, Y, test_size=0.30, random_state=42)

diabities_classifier = DecisionTreeClassifier(max_leaf_nodes=10, random_state=0)

diabities_classifier.fit(X_train, y_train)

print(diabities_classifier)
```

> Below is the screenshot after implementing above code

```
34 Y = df[label_features].copy()
35
36 # Split Data into Training and Test
37 # test_size = .30 defines that training data is 70% and test is 30%
38
39 X_train, X_test, y_train, y_test = train_test_split(X, Y, test_size=0.30, random_state=42)
40
41 diabities_classifier = DecisionTreeClassifier(max_leaf_nodes=10, random_state=0)
42
43 diabities_classifier.fit(X_train, y_train)
44
45 print(diabities_classifier)
46
```

Copy below code. Once Model is trained. Next step is to test it against the test dataset

```
# evaluate the test set
predictions = diabities_classifier.predict(X_test)
print("Prediction")
print(predictions)
accuracy_scored = accuracy_score(y_true = y_test, y_pred = predictions)
print ("Accuracy is {}".format(accuracy_scored))
```

> Below is the screenshot after implementing above code

```
diabities_classifier.fit(X_train, y_train)

print(diabities_classifier)

runder

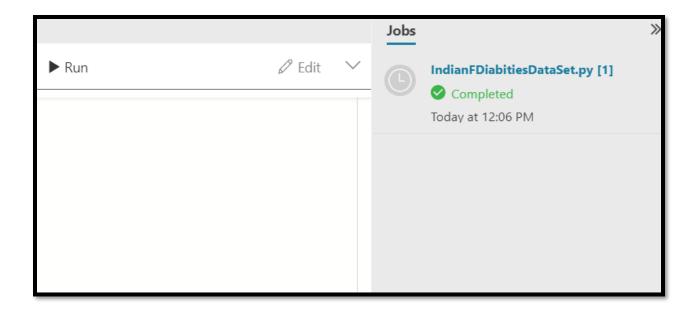
print(diabities_classifier)

runder

run
```

Let's save and run the code to test the accuracy of the model. Click "b" to save the file and click **Run**

On the right side of the screen it will show the progress of the job. Wait till job gets completed



Click on the completed Job



Scroll down and click on driver_log

```
☐ driver_log
 □ Control_log
driver_log
             splitter='best')
 Prediction
 [0.\ 0.\ 0.\ 0.\ 0.\ 0.\ 0.\ 1.\ 0.\ 0.\ 1.\ 0.\ 1.\ 0.\ 0.\ 0.\ 0.\ 0.\ 1.\ 0.\ 0.\ 0.\ 0.
  0. 1. 0. 0. 0. 0. 1. 0. 1. 1. 1. 1. 1. 0. 1. 0. 0. 1. 0. 0. 1. 1. 0.
  0. 1. 0. 1. 0. 0. 0. 0. 1. 0. 0. 1. 0. 0. 0. 0. 0. 0. 0. 0. 0. 1. 1. 0.
  0. 0. 0. 0. 0. 0. 0. 1. 0. 0. 0. 1. 0. 1. 0. 0. 0. 0. 0. 0. 0. 0.
  0.\ 0.\ 1.\ 0.\ 1.\ 0.\ 1.\ 0.\ 1.\ 0.\ 1.\ 0.\ 1.\ 0.\ 1.\ 0.\ 1.\ 0.\ 0.\ 0.\ 0.
  0.\ 0.\ 0.\ 0.\ 0.\ 0.\ 1.\ 1.\ 1.\ 1.\ 1.\ 1.\ 0.\ 0.\ 1.\ 0.\ 0.\ 1.\ 1.\ 0.\ 0.\ 0.\ 0.
  0. 0. 0. 0. 0. 1. 0. 0. 0. 0. 1. 1. 1. 0. 0. 1. 1. 0. 0. 0. 1. 0. 0.
  1. 1. 1. 0. 0. 1. 1. 0. 0. 0. 1. 1. 0. 0. 0. 0. 1. 0. 0. 0. 1. 1. 0. 0.
  1. 0. 0. 0. 0. 0. 1. 0. 0. 0. 0. 1. 0. 0. 1. 0. 0. 0. 0. 0. 0. 0. 0.
  1. 1. 0. 1. 1. 0. 0. 0. 0. 0. 0. 0. 0. 0. 0. 0.
 Accuracy is 0.7619047619047619
```

- It will show the accuracy around 76. You can try to add more random data and tweaking algorithm parameter to test for better accuracy. For now we are good with 76.
- Click on python file to build model file.



➤ Copy below code and paste it. Below code will serialize the model by inserting <u>pickle</u> file in output folder.

```
# serialize the model on disk in the special 'outputs' folder
print ("Export the model to model.pkl")
f = open('./outputs/model.pkl', 'wb')
pickle.dump(diabities_classifier, f)
f.close()
```

Below is the screenshot after implementing above code.

```
50 print(predictions)
51 accuracy_scored = accuracy_score(y_true = y_test, y_pred = predictions)
52 print ("Accuracy is {}".format(accuracy_scored))
53
54 # serialize the model on disk in the special 'outputs' folder
55 print ("Export the model to model.pkl")
56 f = open('./outputs/model.pkl', 'wb')
57 pickle.dump(diabities_classifier, f)
58 f.close()
59
```

Copy below code and past it in editor. In below code, we are loading pickle file in pass sample data to predict the outcome

```
# load the model back from the 'outputs' folder into memory
print("Import the model from model.pkl")

f2 = open('./outputs/model.pkl', 'rb')
diabities_classifier2 = pickle.load(f2)

# predict on a new sample

# ['PlasmaGlucose', 'bmi', 'age', 'tricepsskinfold', 'PregnantTimes', 'diabetespedigreefunction',

# 'DbloodPressure', '2hourseruminsulin']

X_new = [[390, 163.6, 102, 155,900,0.998,1270,10]]
print ('New sample: {}'.format(X_new))

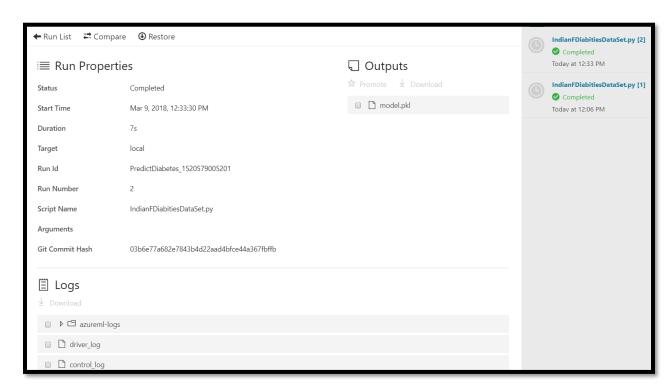
# score on the new sample
pred = diabities_classifier2.predict(X_new)
print('Predicted class is {}'.format(pred))
```

Below is the screenshot after pasting above code

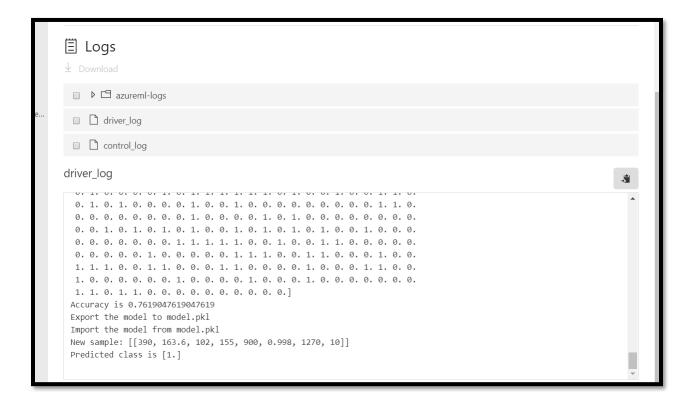
```
57 pickle.dump(diabities_classifier, f)
58 f.close()
59
60 # load the model back from the 'outputs' folder into memory
61 print("Import the model from model.pkl")
62 f2 = open('./outputs/model.pkl', 'rb')
63 diabities_classifier2 = pickle.load(f2)
64
65 # predict on a new sample
66 # ['PlasmaGlucose', 'bmi', 'age', 'tricepsskinfold', 'PregnantTimes', 'diabetespedigreefunction',
67 # 'DbloodPressure', '2hourseruminsulin']
68 X_new = [[390, 163.6, 102, 155,900,0.998,1270,10]]
69 print ('New sample: {}'.format(X_new))
70
71 # score on the new sample
72 pred = diabities_classifier2.predict(X_new)
73 print('Predicted class is {}'.format(pred))
74
```



➤ Click on latest job execution



Click on driver_log. In below screenshot you will notice exporting and importing of model file and test it against new sample data



Next Step is to operationalize it which you already did in previous HOL. If time permits you can try it again here. You can also look out <u>advance data preparation</u> to understand various data wrangling techniques.

Summary

Well done! In this hands-on lab we solved a problem to predict whether a person going to be a diabetic or not. This will help to take proactive measure and cure patient. You can deploy this model and make it available as a web service so that any application from anywhere can access it. We also learn the capabilities of Azure Machine Learning and how to work with it. You can download data from <u>UCI</u> and <u>Kaggle</u> to solve other problems. Let's move onto the next step.