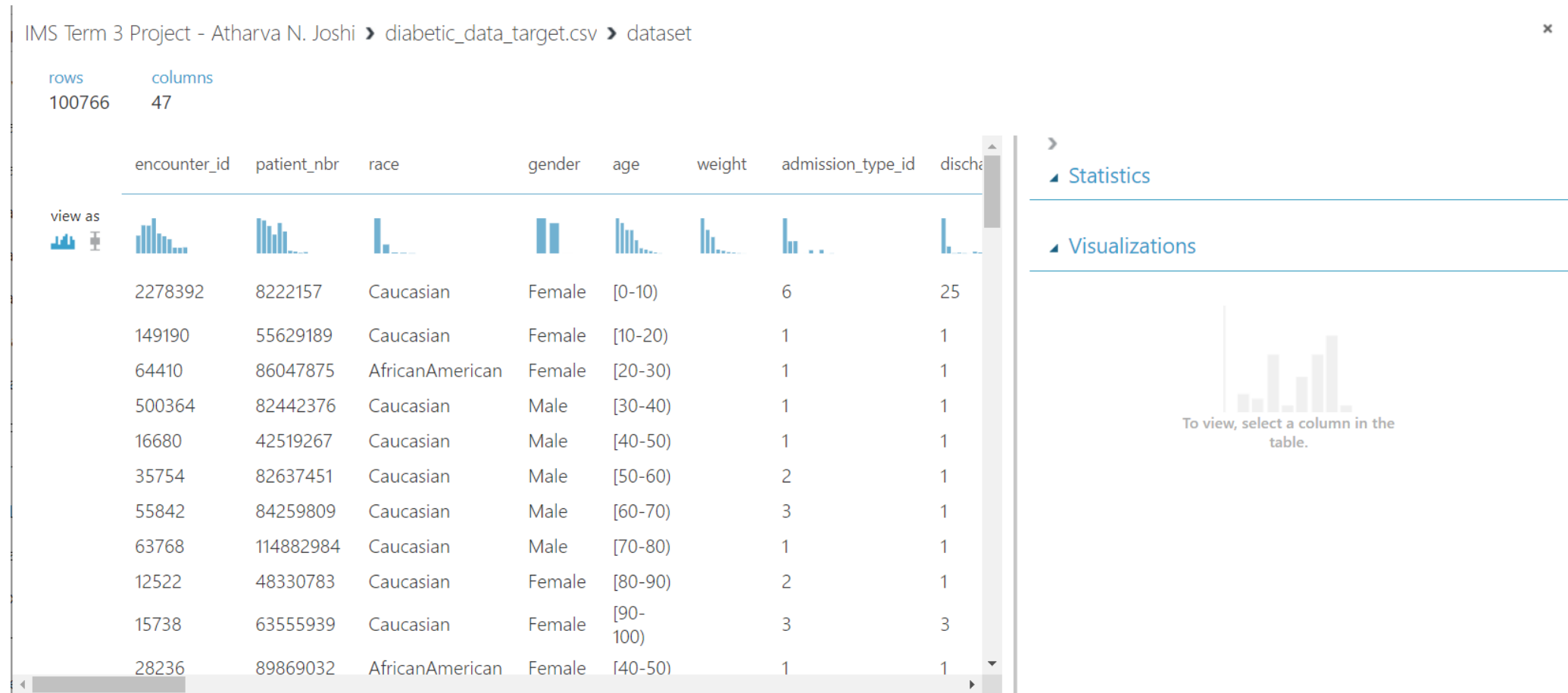
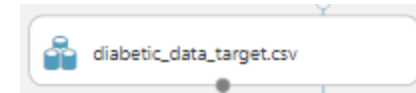


# IMS Term 3 Project

- Atharva N. Joshi

# First we load the dataset and try to do EDA. Azure gives us all necessary information without any extra commands



We see that some columns have many null values. We remove those columns except race which has comparatively less null values and hence, will be imputed

#### Statistics

Unique Values	5
Missing Values	2234
Feature Type	String Feature

#### Visualizations

race

#### Statistics

Unique Values	17
Missing Values	40115
Feature Type	String Feature

#### Visualizations

payer\_code

...

#### Statistics

Unique Values	9
Missing Values	97606
Feature Type	String Feature

#### Visualizations

weight

#### Statistics

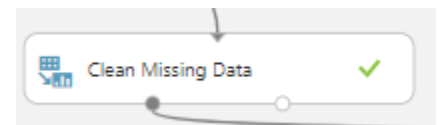
Unique Values	72
Missing Values	49269
Feature Type	String Feature

#### Visualizations

medical\_specialty



To  
include/exclude  
certain columns



To impute null  
values

We can also see that the below two columns have only one value. So we can exclude those columns as well.

#### Statistics

---

Unique Values	1
Missing Values	0
Feature Type	String Feature

#### Visualizations

---

examide

#### Statistics

---

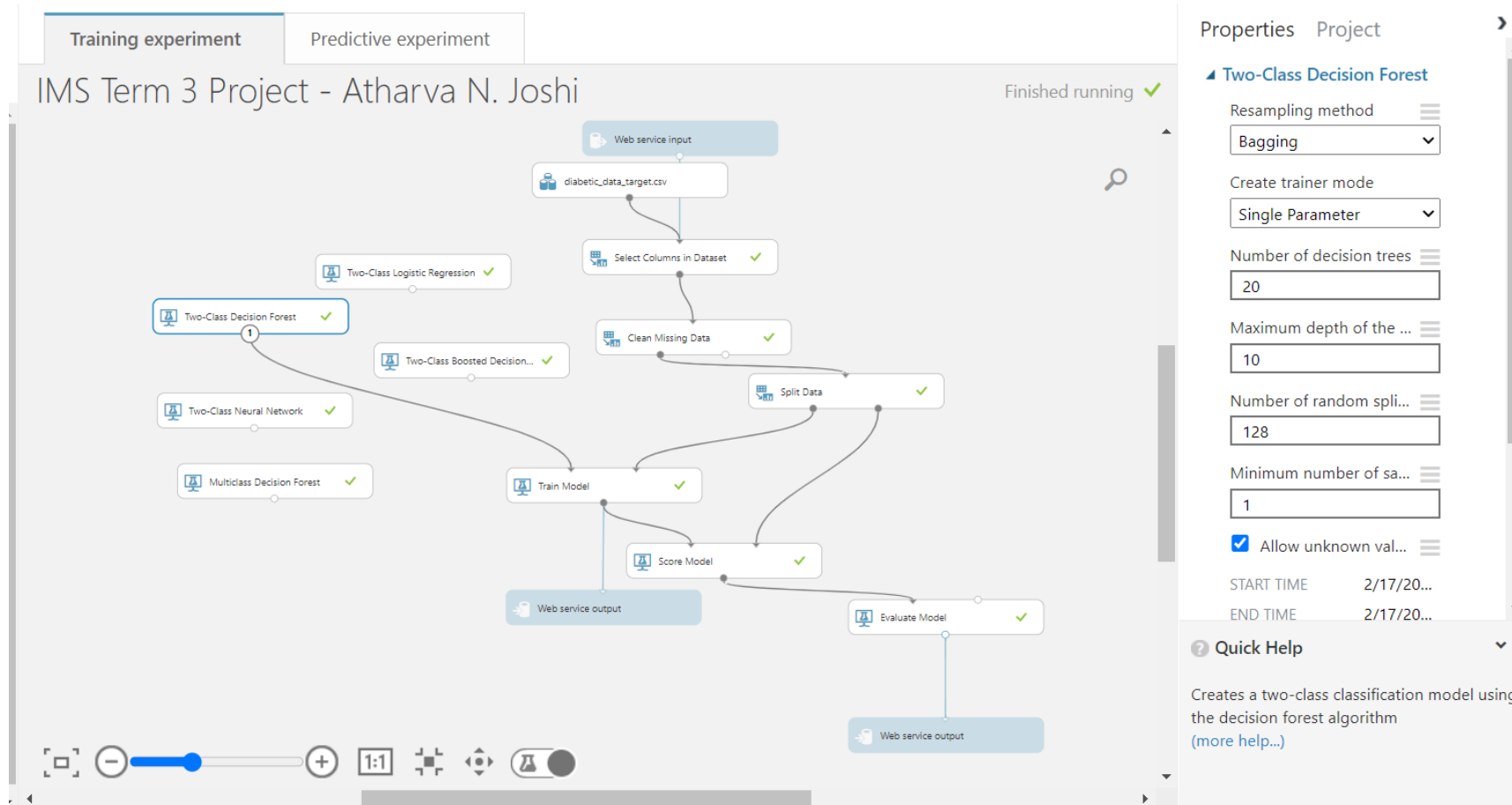
Unique Values	1
Missing Values	0
Feature Type	String Feature

#### Visualizations

---

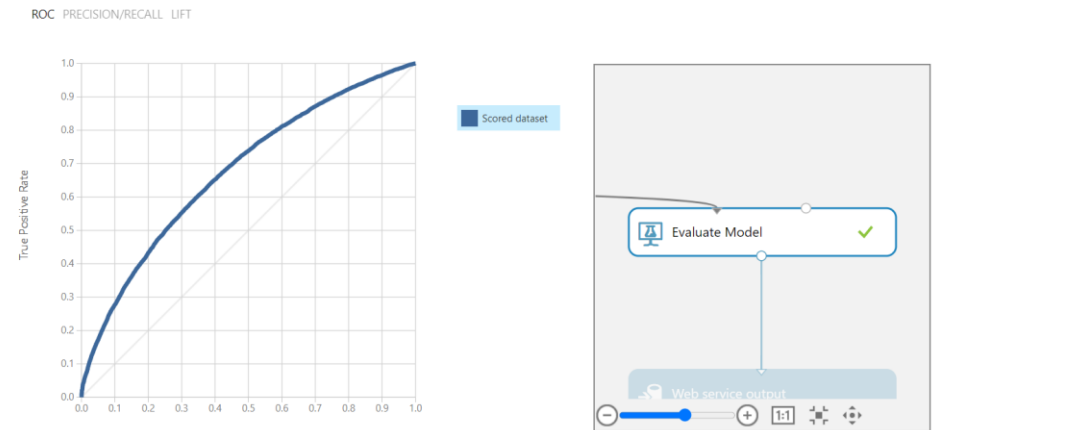
citoglipton

# Two Class Decision Forest

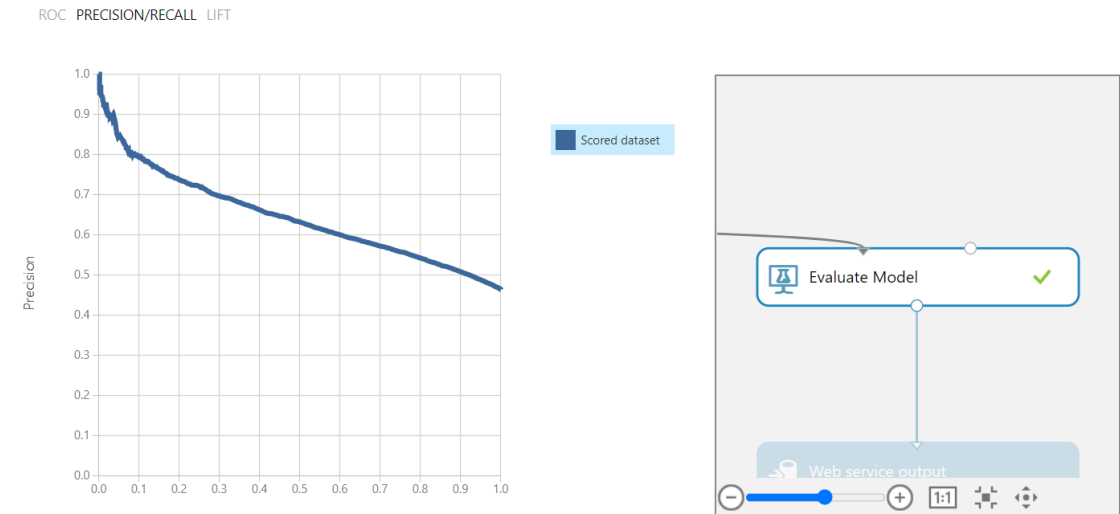


# Result – Best among all Algorithms

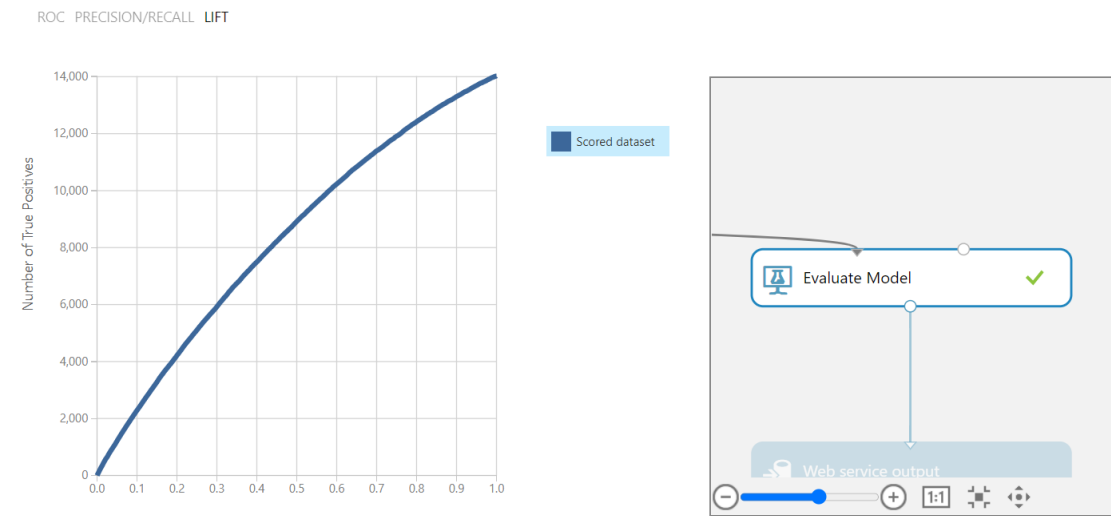
IMS Term 3 Project - Atharva N. Joshi > Evaluate Model > Evaluation results



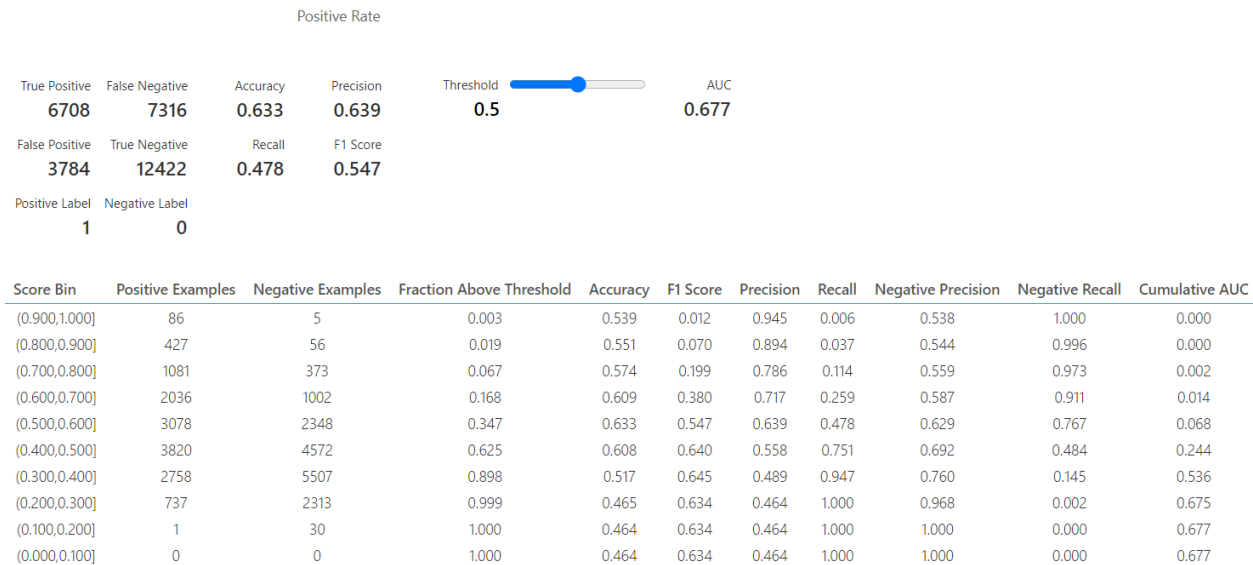
IMS Term 3 Project - Atharva N. Joshi > Evaluate Model > Evaluation results



IMS Term 3 Project - Atharva N. Joshi > Evaluate Model > Evaluation results



IMS Term 3 Project - Atharva N. Joshi > Evaluate Model > Evaluation results



# Two Class Logistic Regression

Training experiment

Predictive experiment

IMS Term 3 Project - Atharva N. Joshi

In draft

Draft saved at 1:37:02 PM

The workflow diagram illustrates the process of training and evaluating a two-class logistic regression model. It begins with a 'Web service input' node, which feeds into a 'diabetic\_data\_target.csv' file. This file is then processed by a 'Select Columns in Dataset' node, followed by a 'Clean Missing Data' node. The data is then split into training and testing sets using a 'Split Data' node. The training set is used to train a 'Two-Class Logistic Regression' model, which is then evaluated using an 'Evaluate Model' node. The testing set is used to score the model using a 'Score Model' node, which produces a 'Web service output'.

```
graph TD; Input[Web service input] --> CSV[diabetic_data_target.csv]; CSV --> Select[Select Columns in Dataset]; Select --> Clean[Clean Missing Data]; Clean --> Split[Split Data]; Split --> Train[Train Model]; Split --> Score[Score Model]; Train --> Score; Score --> Output[Web service output];
```

Properties

Project

Two-Class Logistic Regression

Create trainer mode

Single Parameter

Optimization tolerance

1E-07

L1 regularization weight

1

L2 regularization weight

1

Memory size for L-BFGS

20

Random number seed

☒ Allow unknown cat...

START TIME

2/17/20...

END TIME

2/17/20...

Quick Help

Creates a two-class logistic regression model

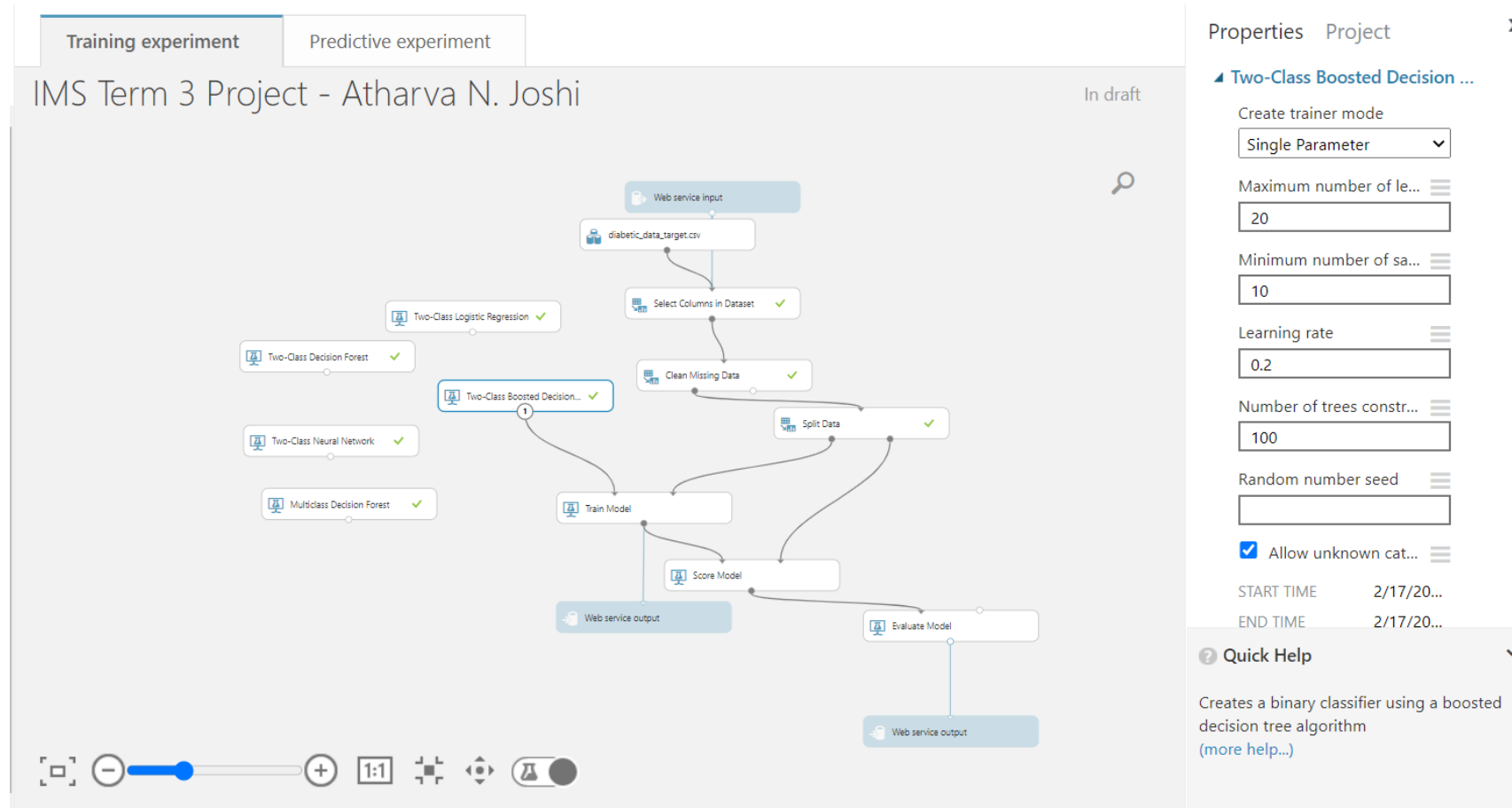
(more help...)

# Result

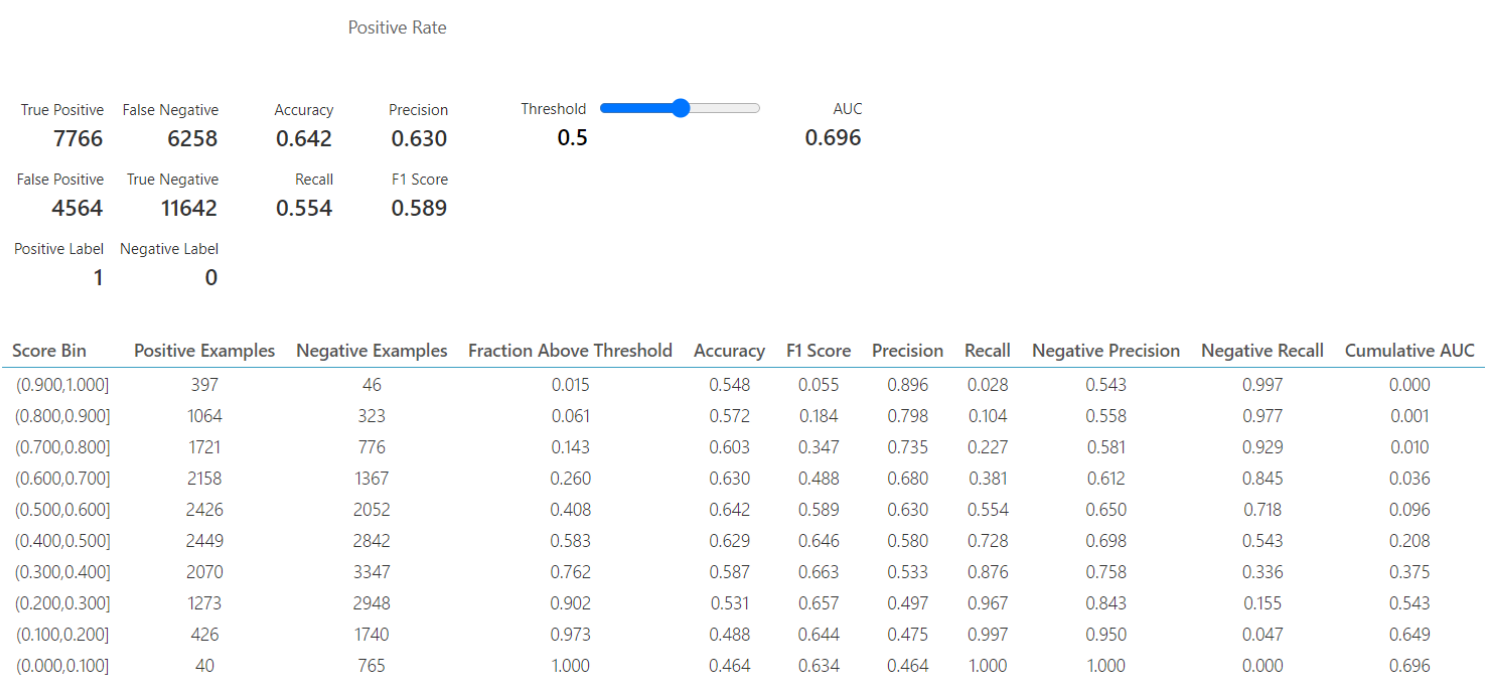
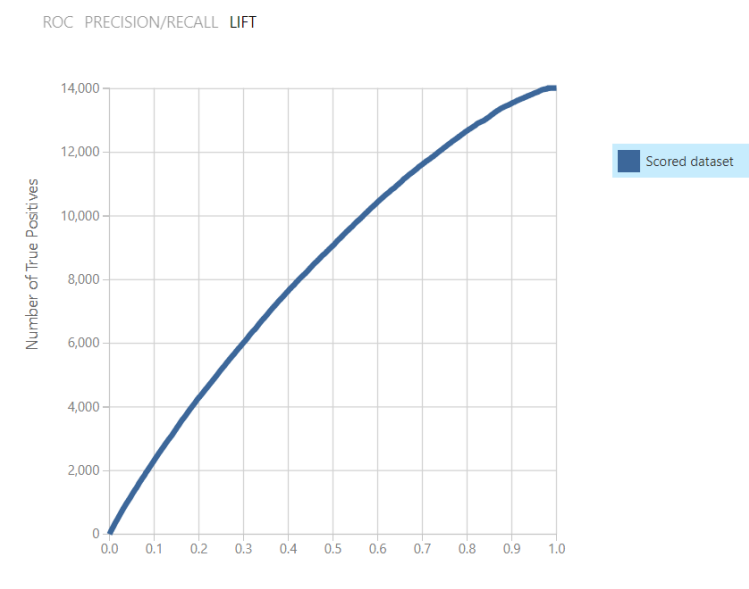
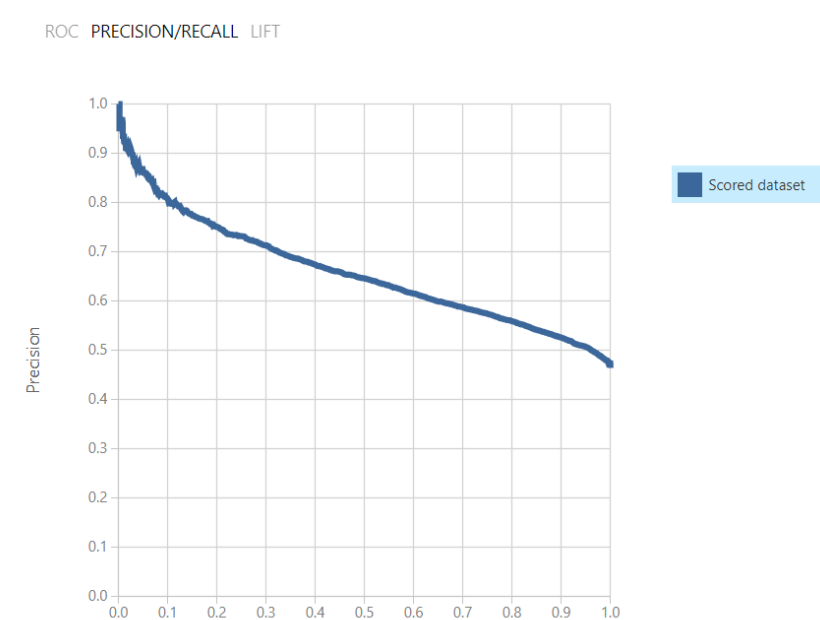
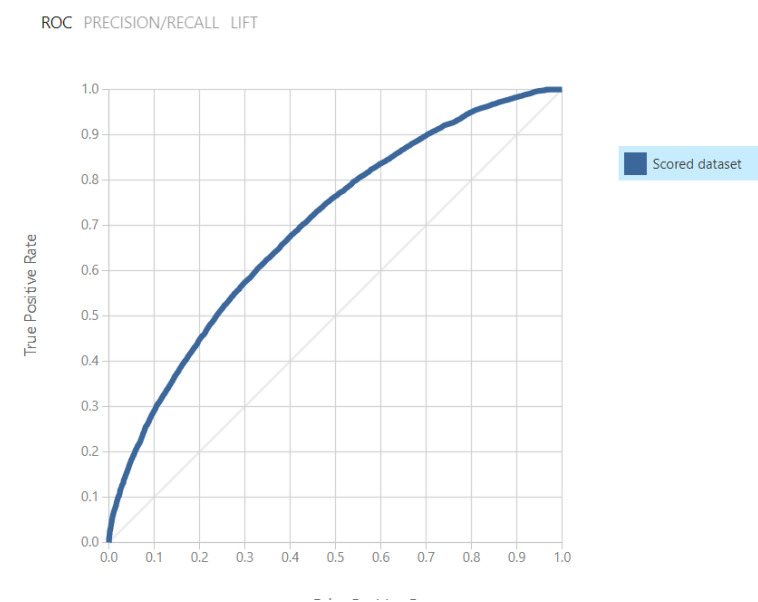




# Two Class Boosted Decision Forest



# Result



# Two Class Neural Network

Training experiment

Predictive experiment

IMS Term 3 Project - Atharva N. Joshi

In draft

Draft saved at 1:53:36 PM

Properties

Project

Two-Class Neural Network

Create trainer mode

Single Parameter

Hidden layer specification

Fully-connected case

Number of hidden nodes

128

Learning rate

0.1

Number of learning iter...

100

The initial learning wei...

0.1

The momentum

0

The type of normalizer

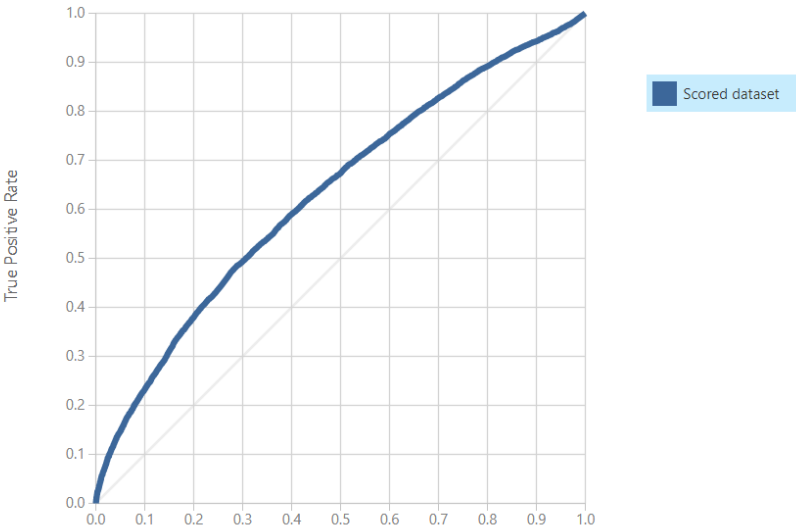
Quick Help

Creates a binary classifier using a neural network algorithm

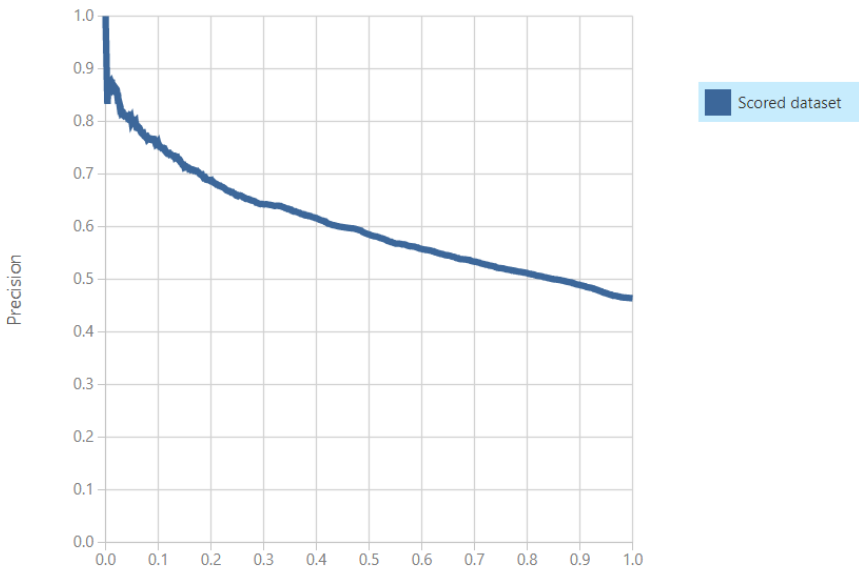
(more help...)

# Result

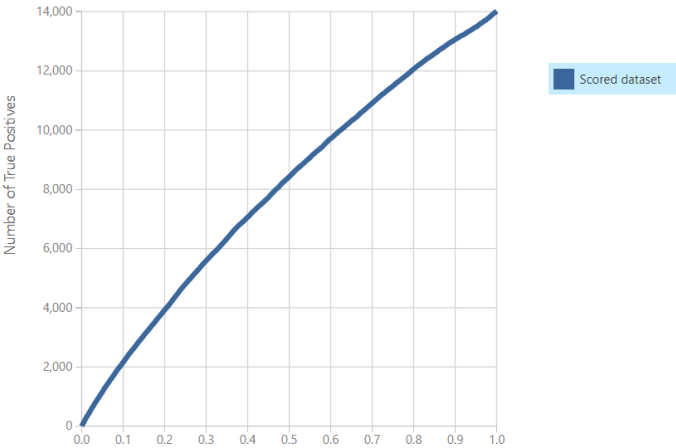
ROC PRECISION/RECALL LIFT



ROC PRECISION/RECALL LIFT



ROC PRECISION/RECALL LIFT



Positive Rate

True Positive	False Negative	Accuracy	Precision	Threshold	AUC
6978	7046	0.604	0.586	0.5	0.632
False Positive	True Negative	Recall	F1 Score		
4938	11268	0.498	0.538		
Positive Label	Negative Label				
1	0				

Score Bin	Positive Examples	Negative Examples	Fraction Above Threshold	Accuracy	F1 Score	Precision	Recall	Negative Precision	Negative Recall	Cumulative AUC
(0.900,1.000]	921	253	0.039	0.558	0.121	0.784	0.066	0.549	0.984	0.001
(0.800,0.900]	997	461	0.087	0.576	0.230	0.729	0.137	0.561	0.956	0.004
(0.700,0.800]	1058	688	0.145	0.588	0.323	0.680	0.212	0.573	0.913	0.011
(0.600,0.700]	1665	1219	0.240	0.603	0.436	0.639	0.331	0.591	0.838	0.031
(0.500,0.600]	2337	2317	0.394	0.604	0.538	0.586	0.498	0.615	0.695	0.091
(0.400,0.500]	1955	2425	0.539	0.588	0.589	0.548	0.637	0.635	0.546	0.176
(0.300,0.400]	1492	2180	0.661	0.565	0.613	0.522	0.743	0.649	0.411	0.269
(0.200,0.300]	1286	1997	0.769	0.542	0.628	0.504	0.835	0.669	0.288	0.367
(0.100,0.200]	1059	1957	0.869	0.512	0.634	0.486	0.911	0.684	0.167	0.472
(0.000,0.100]	1254	2709	1.000	0.464	0.634	0.464	1.000	1.000	0.000	0.632

# Multi Class Decision Forest

Training experiment

Predictive experiment

IMS Term 3 Project - Atharva N. Joshi

In draft

Saving...

```
graph TD; Input[Web service input] --> CSV[diabetic_data_target.csv]; CSV --> Select[Select Columns in Dataset]; Select --> Clean[Clean Missing Data]; Clean --> Split[Split Data]; Split --> Train[Train Model]; Split --> Eval[Evaluate Model]; Train --> Score[Score Model]; Score --> Eval; Eval --> Output[Web service output];
```

Properties

Project

▲ Multiclass Decision Forest

Resampling method  
Bagging

Create trainer mode  
Single Parameter

Number of decision trees  
8

Maximum depth of the ...  
32

Number of random spli...  
128

Minimum number of sa...  
1

☒ Allow unknown val...

START TIME 2/17/20...

END TIME 2/17/20...

Quick Help

Creates a multiclass classification model using the decision forest algorithm  
(more help...)

# Result

IMS Term 3 Project - Atharva N. Joshi > Evaluate Model > Evaluation results

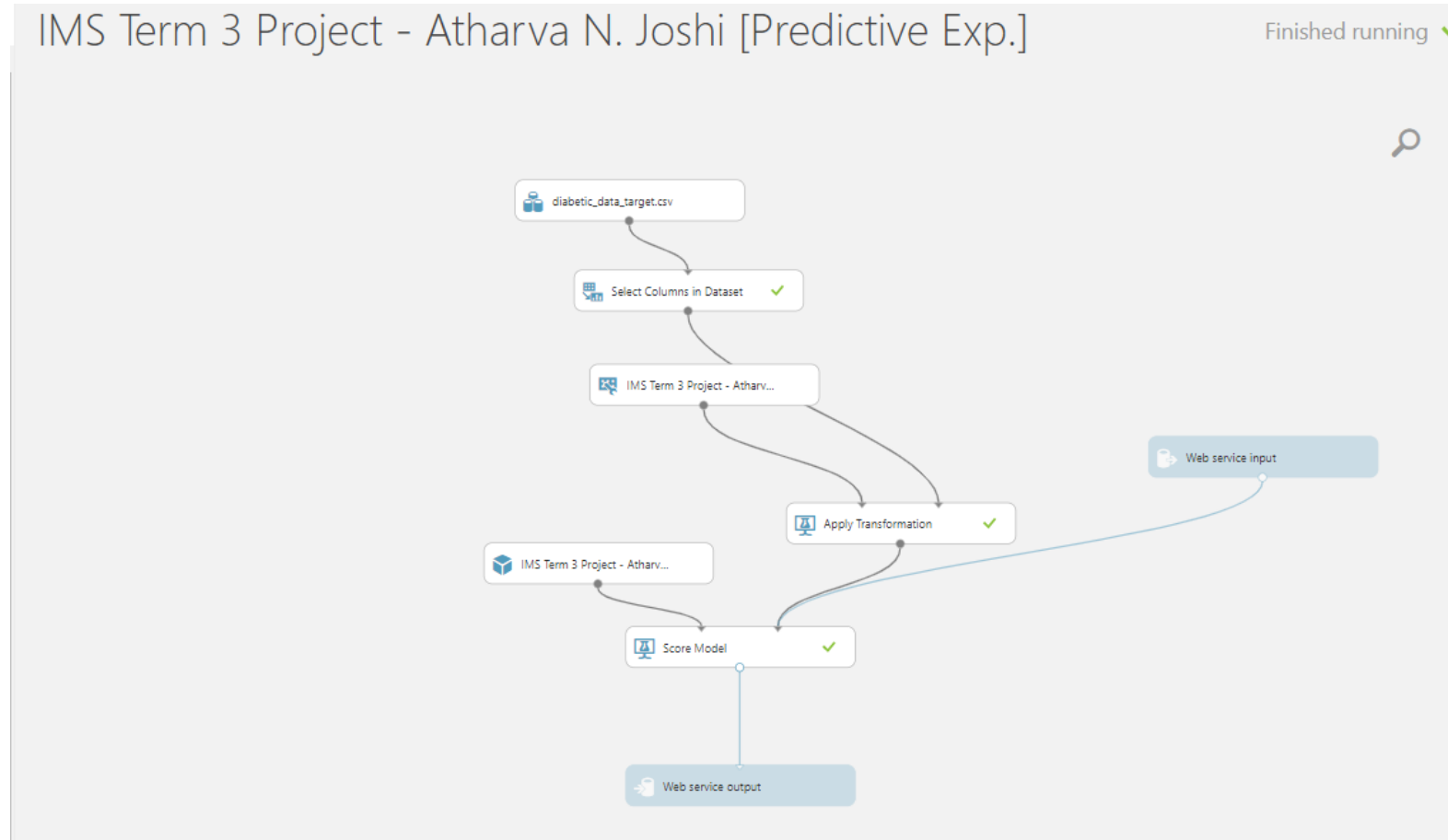
## Metrics

Overall accuracy	0.592987
Average accuracy	0.592987
Micro-averaged precision	0.592987
Macro-averaged precision	0.589345
Micro-averaged recall	0.592987
Macro-averaged recall	0.585973

## Confusion Matrix

		Predicted Class	
		0	1
Actual Class	0	68.3%	31.7%
	1	51.1%	48.9%

# Deployment over Azure



# Deployment Link

<https://gallery.azure.ai/Experiment/IMS-Term-3-Project-Atharva-N-Joshi-Predictive-Exp>