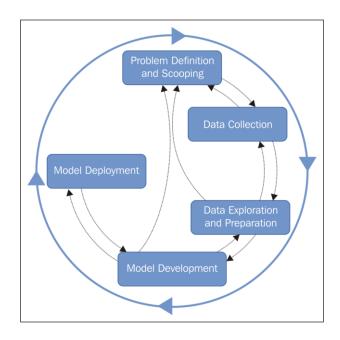
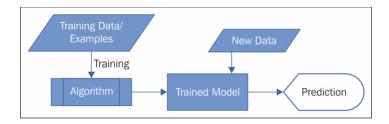
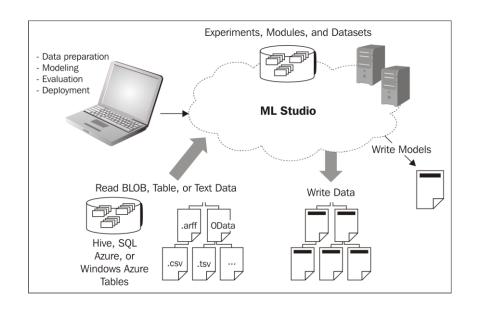
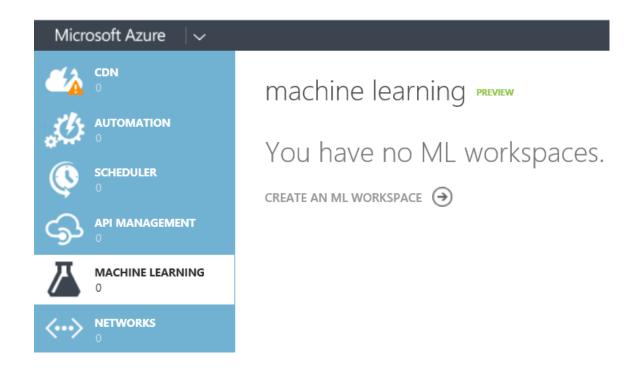
Chapter 1

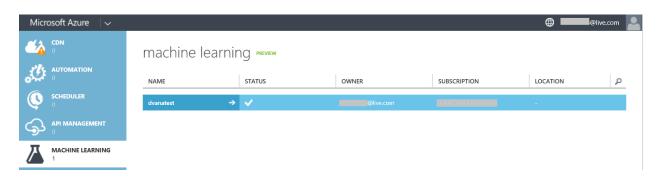


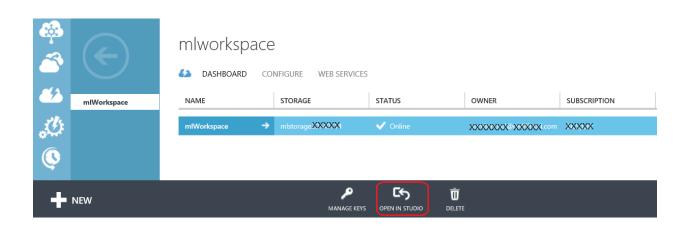


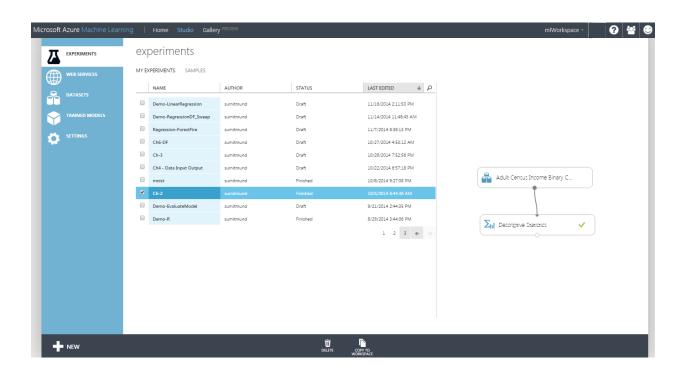
Chapter 2

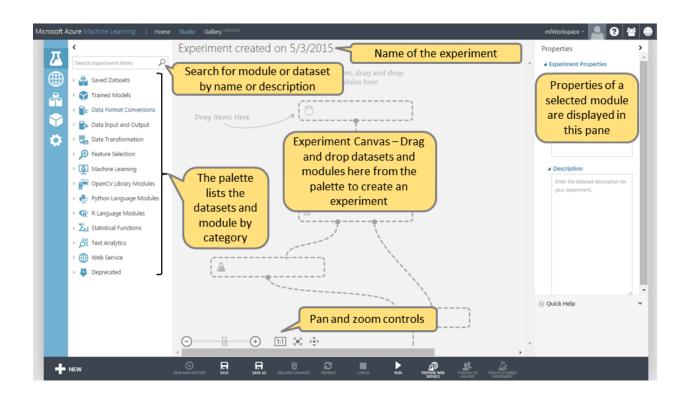


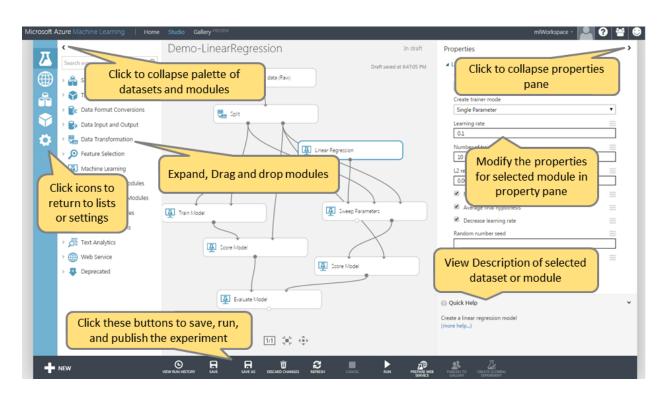


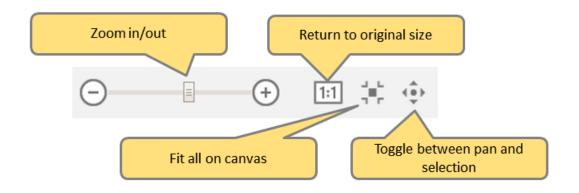


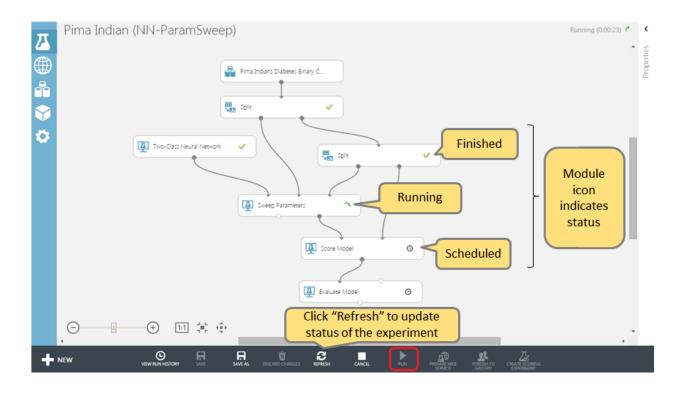


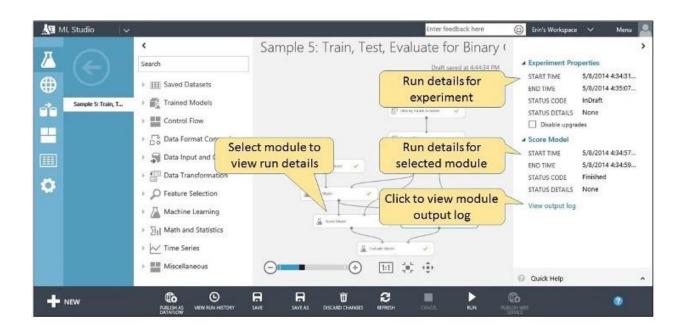


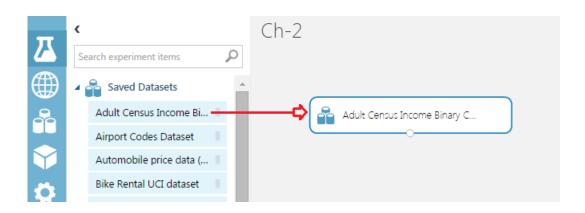


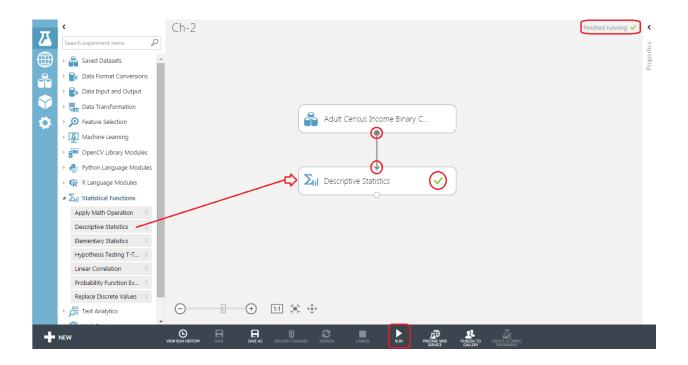


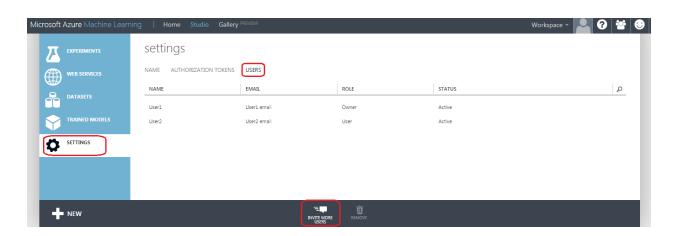






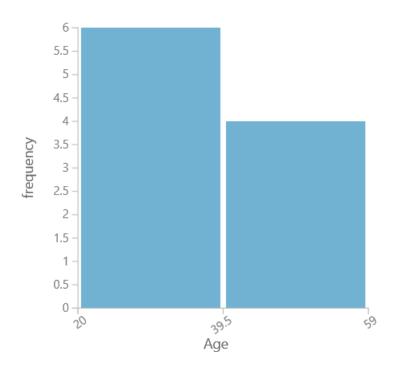


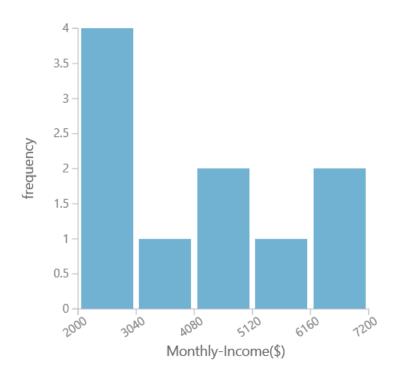


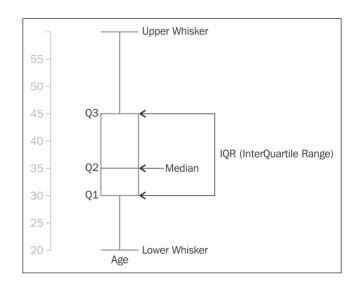


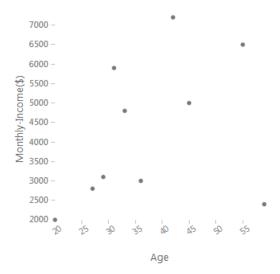
Chapter 3

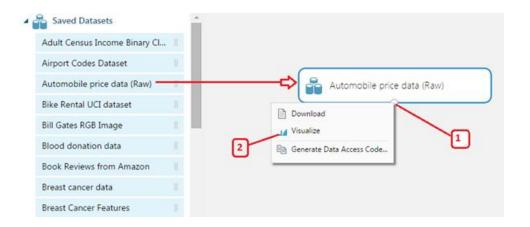
	Name	Age	Gender	Monthly-Income(\$)
Mean		37.7		4270
Median		34.5		3950
Min		20		2000
Max		59		7200
Standard Deviation		12.4637		1850.5555
Unique Values	10	10	2	10
Missing Values	0	0	0	0
Feature Type	String	Numeric	String	Numeric





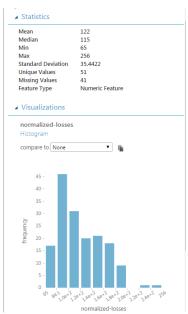








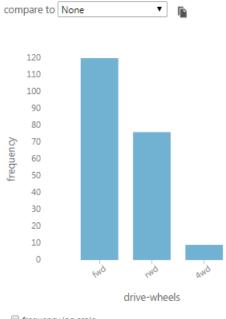
	symboling	normalized- losses	make	fuel- type	aspiration	num- of- doors	body-style	drive- wheels	engine- location	wheel- base	length	width	heic
view as	alli	du.	lum	L	I.	li.	li	h.	L				.lif
	3		alfa- romero	gas	std	two	convertible	rwd	front	88.6	168.8	64.1	48.8
	3		alfa- romero	gas	std	two	convertible	rwd	front	88.6	168.8	64.1	48.8
	1		alfa- romero	gas	std	two	hatchback	rwd	front	94.5	171.2	65.5	52.4
	2	164	audi	gas	std	four	sedan	fwd	front	99.8	176.6	66.2	54.1
	2	164	audi	gas	std	four	sedan	4wd	front	99.4	176.6	66.4	54.:
	2		audi	gas	std	two	sedan	fwd	front	99.8	177.3	66.3	53.1
	1	158	audi	gas	std	four	sedan	fwd	front	105.8	192.7	71.4	55.7
	1		audi	gas	std	four	wagon	fwd	front	105.8	192.7	71.4	55.7
	1	158	audi	gas	turbo	four	sedan	fwd	front	105.8	192.7	71.4	55.5
	0		audi	gas	turbo	two	hatchback	4wd	front	99.5	178.2	67.9	52
	2	192	bmw	gas	std	two	sedan	rwd	front	101.2	176.8	64.8	54.1
	0	192	bmw	gas	std	four	sedan	rwd	front	101.2	176.8	64.8	54.5
	0	188	bmw	gas	std	two	sedan	rwd	front	101.2	176.8	64.8	54.1
	0	188	bmw	gas	std	four	sedan	rwd	front	101.2	176.8	64.8	54.5
	1		bmw	gas	std	four	sedan	rwd	front	103.5	189	66.9	55.7
	0		bmw	gas	std	four	sedan	rwd	front	103.5	189	66.9	55.7
	0		bmw	gas	std	two	sedan	rwd	front	103.5	193.8	67.9	53.7
	0		bmw	gas	std	four	sedan	rwd	front	110	197	70.9	56.:
	2	121	chevrolet	gas	std	two	hatchback	fwd	front	88.4	141.1	60.3	53.1
	1	98	chevrolet	gas	std	two	hatchback	fwd	front	94.5	155.9	63.6	52 ▼



■ Visualizations

drive-wheels

Histogram

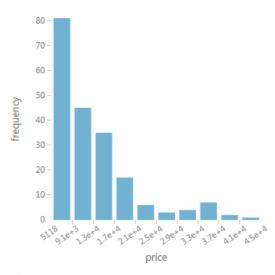


frequency log scale

bins 10

price Histogram





- price log scale
- frequency log scale

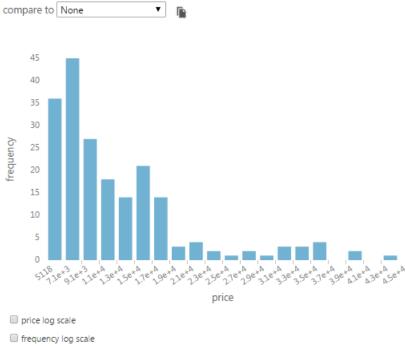
bins

10

- cumulative distribution
- probability density

price

Histogram



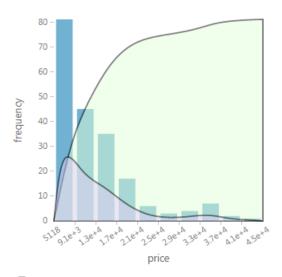


- cumulative distribution
- probability density

price

Histogram

compare to None ▼

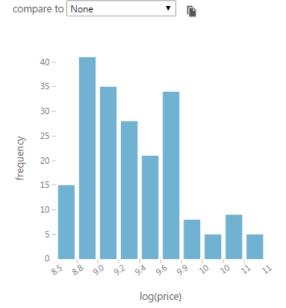


- price log scale
- frequency log scale

bins
10

cumulative distribution
probability density

price Histogram





bins 10

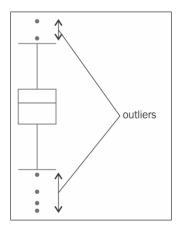
cumulative distribution

probability density

	symboling	normalized- losses	make	fuel- type	as
view as	自	İ Đ	liiiiiiiiiiiiiiiiiiiiiiiiiiiiiiiiiiiii	L	ш
	3		alfa- romero	gas	sto
	3		alfa- romero	gas	sto
	1		alfa- romero	gas	sto
	2	164	audi	gas	sto
	2	164	audi	gas	sto
	2		audi	gas	sto
	1	158	audi	gas	sto
	1		audi	gas	sto
	1	158	audi	gas	tu
	0		audi	gas	tu
	2	192	bmw	gas	sto
	0	192	bmw	gas	stc

✓ Visualizations symboling BoxPlot compare to None ✓ 1.5 2 1.5 1 0.5 -

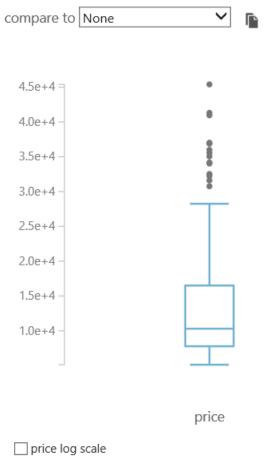
symboling



Visualizations

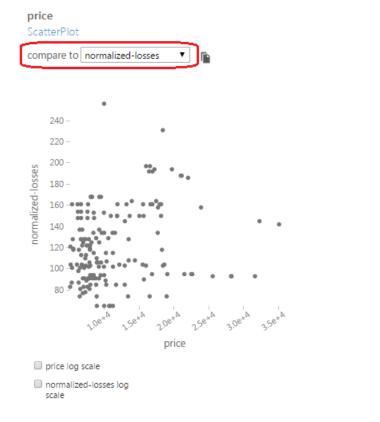
price

BoxPlot

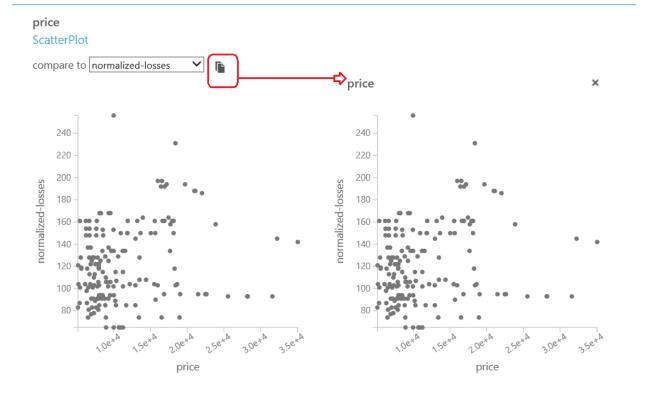


■ Visualizations

MultiboxPlot compare to make 4.5e+4 4.0e+4 3.5e+4 3.0e+4 8.2.5e+4 2.0e+4 1.5e+4 1.0e+4 1.oe+4 1.o

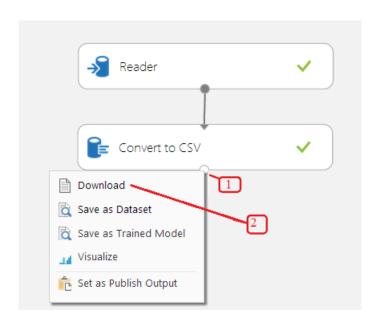


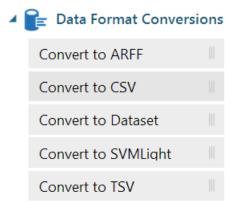
Visualizations

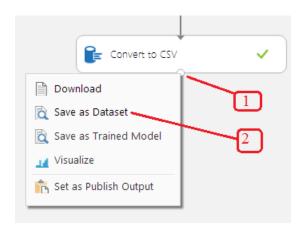


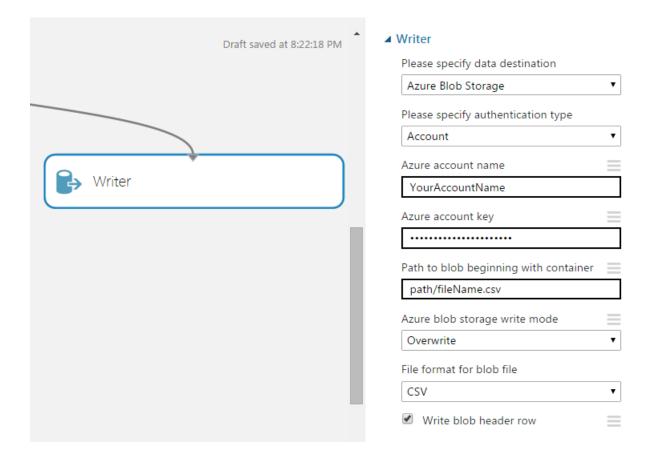
Chapter 4

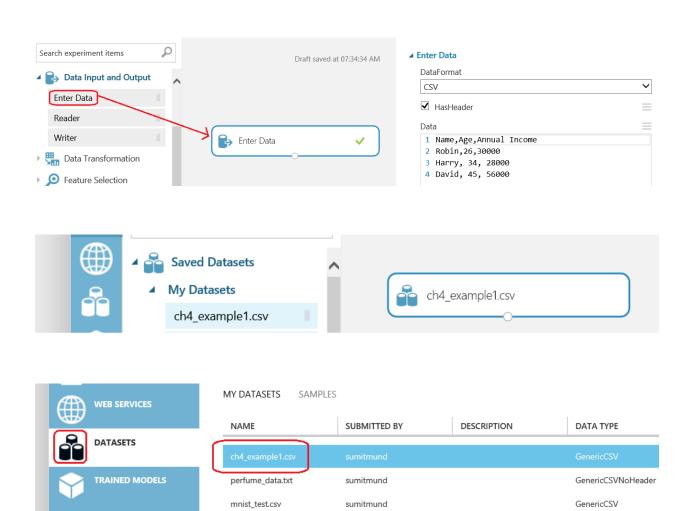








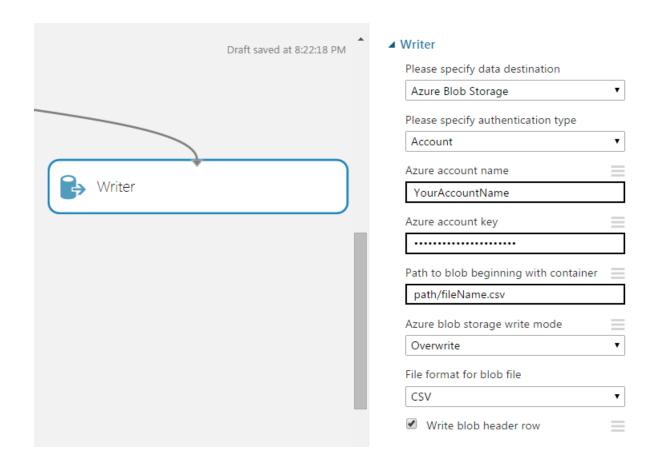




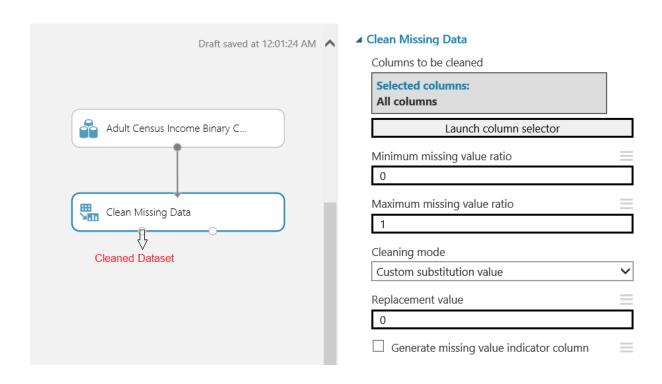
NEW

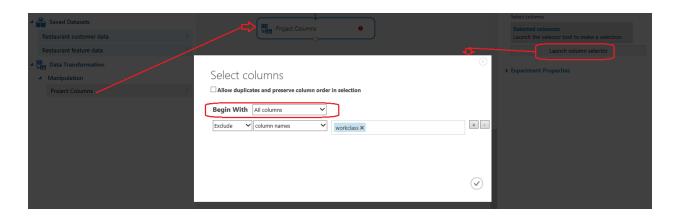
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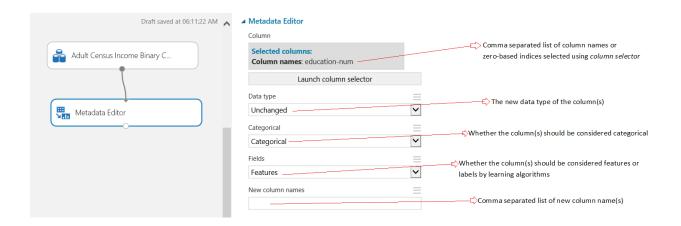
U DELETE

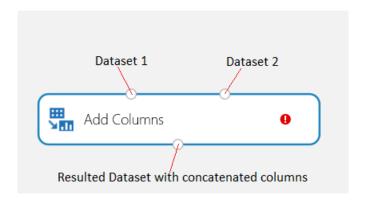


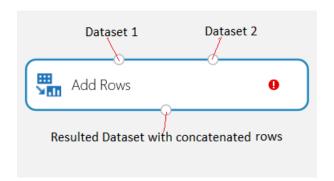
Chapter 5

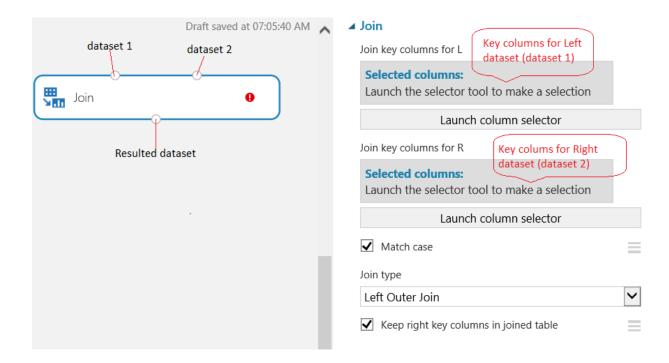


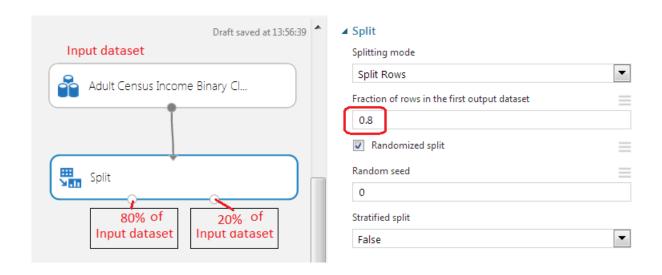


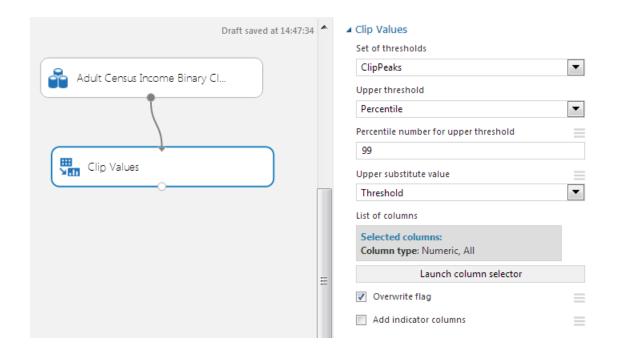


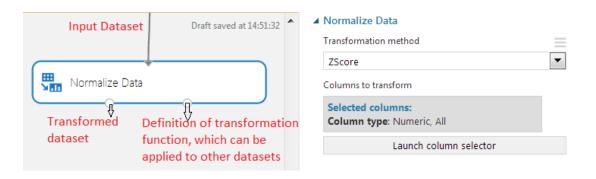


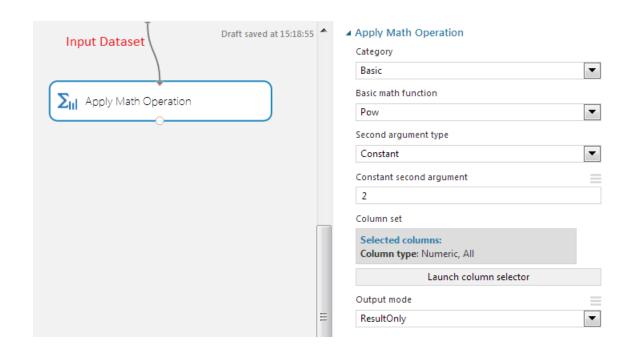


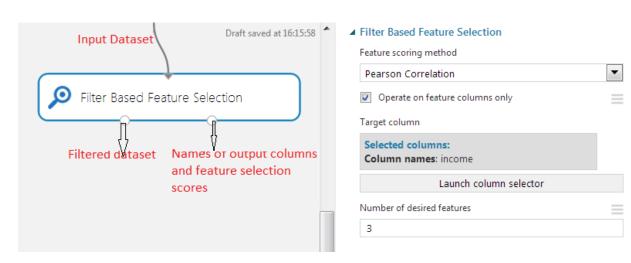


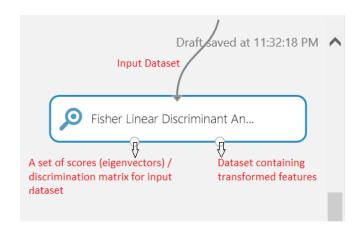


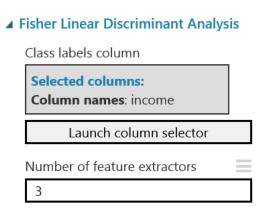


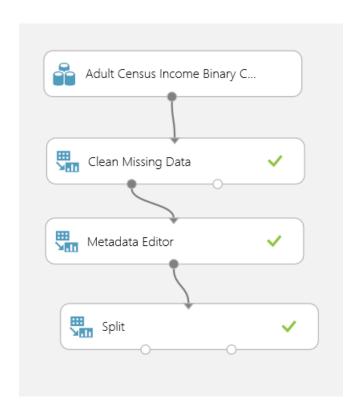


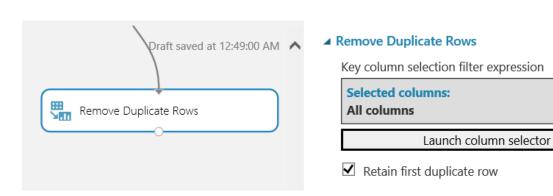












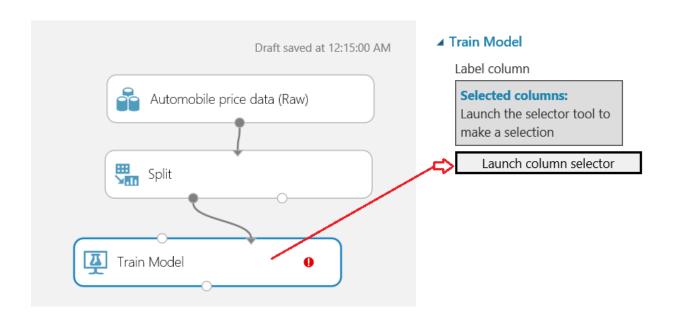


▲ Apply SQL Transformation

```
SQL Query Script

1 SELECT DISTINCT t1.placeid, t2.Name,
2 t2.City, AVG(rating) as 'AvgRating'
3 from t1 JOIN t2 ON t1.placeid = t2.placeid
4 group by t1.placeid;
```

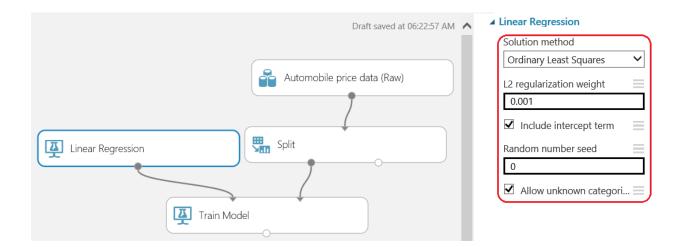
Chapter 6

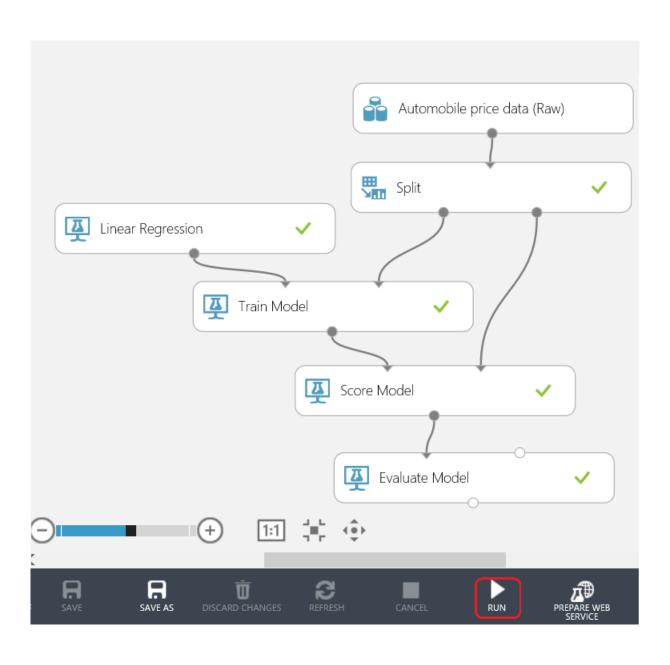


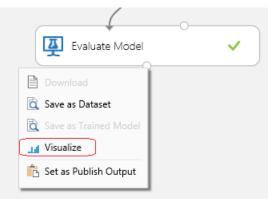
Select a single column

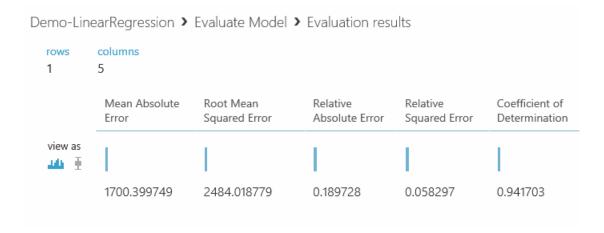


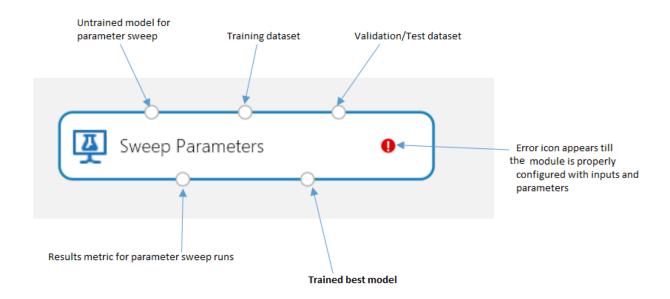


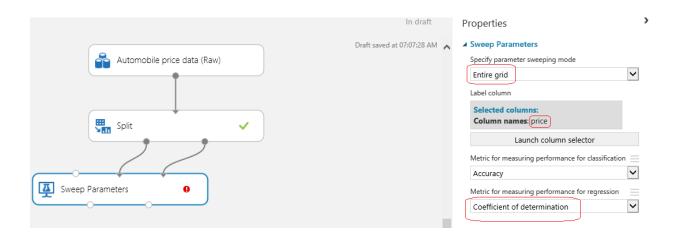


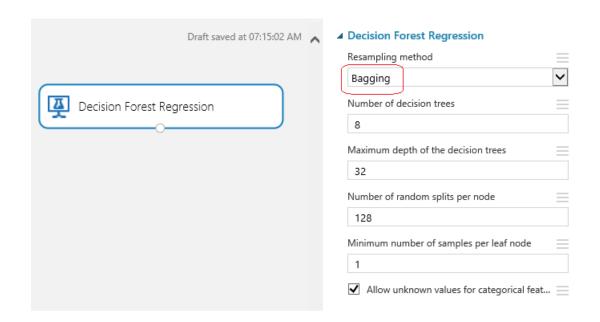


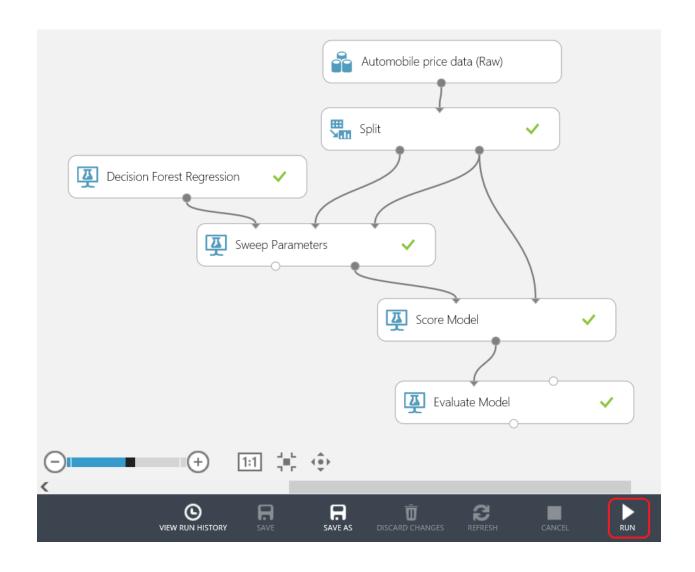


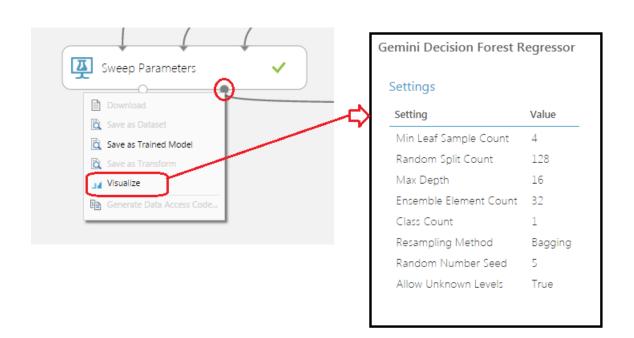


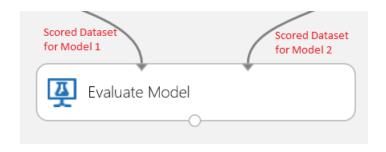


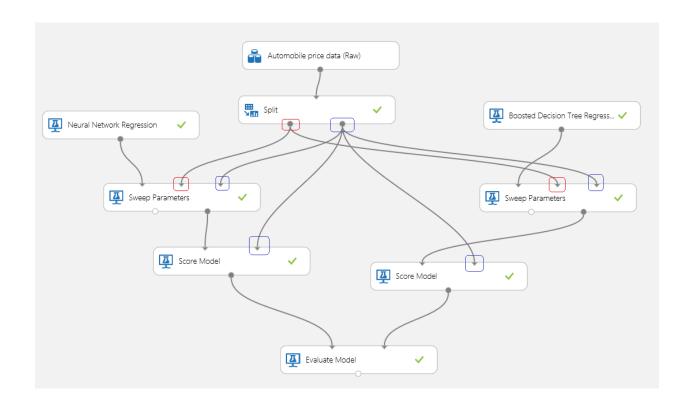








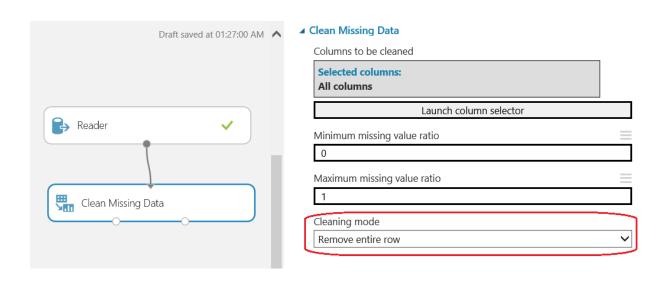


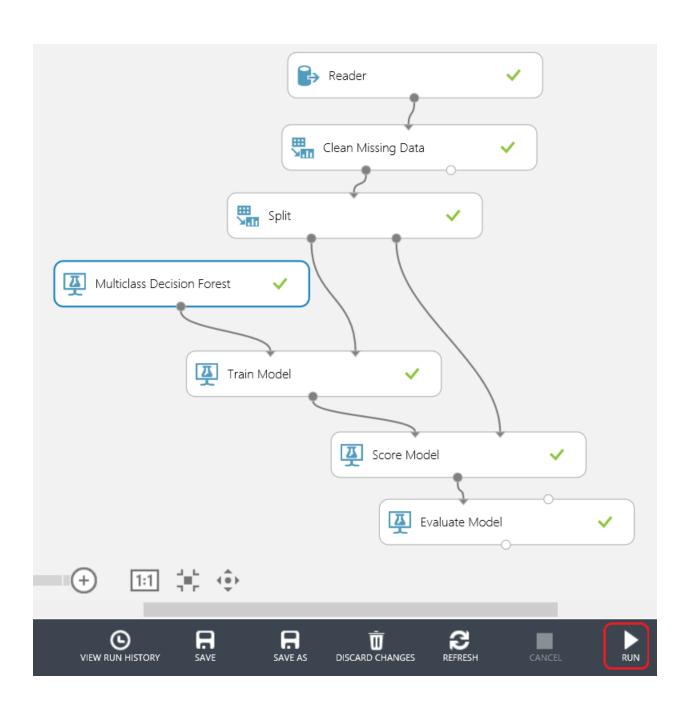


Mean Absolute Error	Root Mean Squared Error	Relative Absolute Error	Relative Squared Error	Coefficient of Determination	
			H		
5071.800876	7664.589306	0.565904	0.555025	0.444975	
1340.164661	1958.367584	0.149534	0.036235	0.963765	

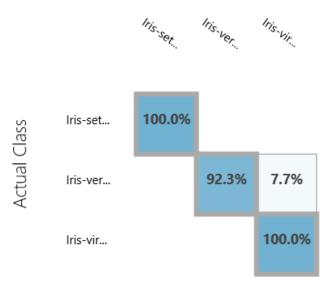
Chapter 7

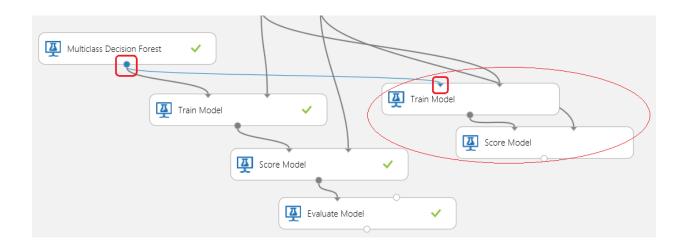


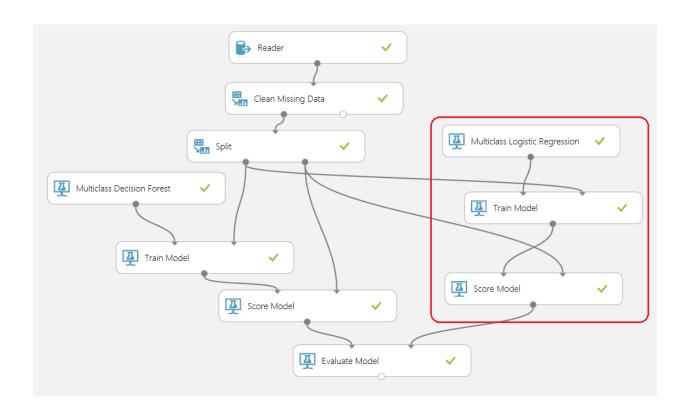




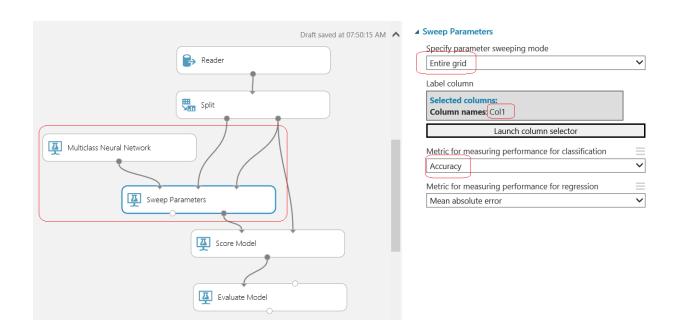
Predicted Class

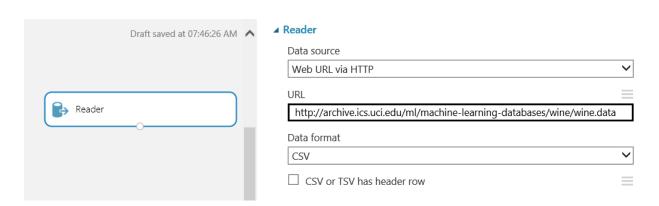


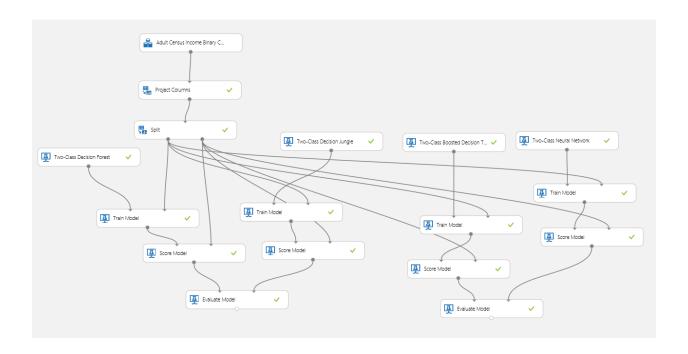


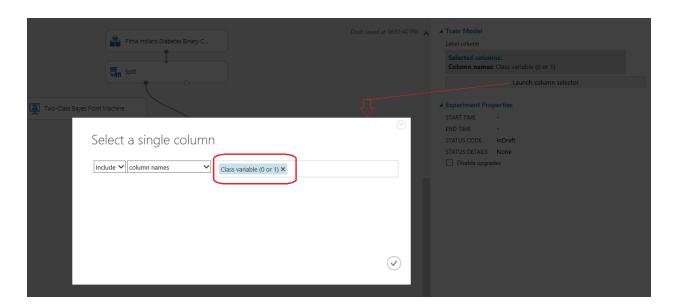


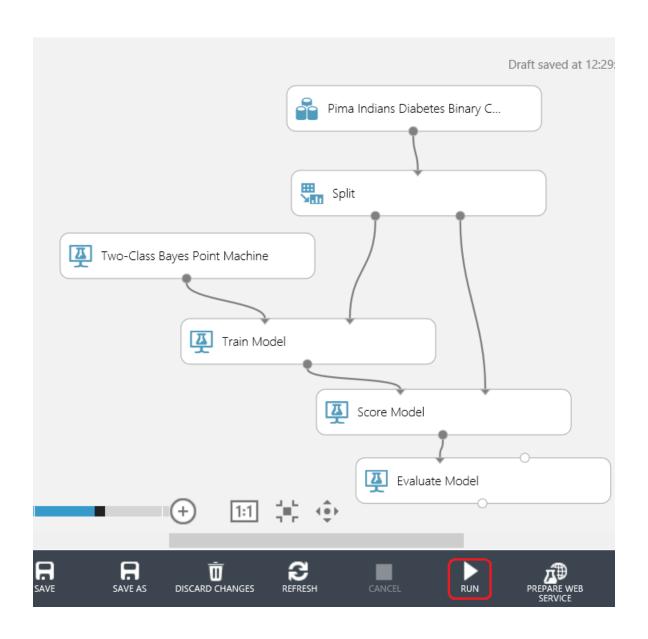






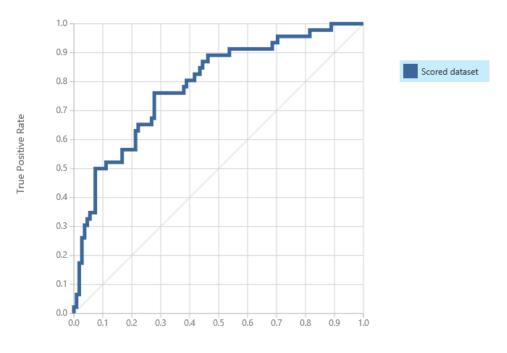






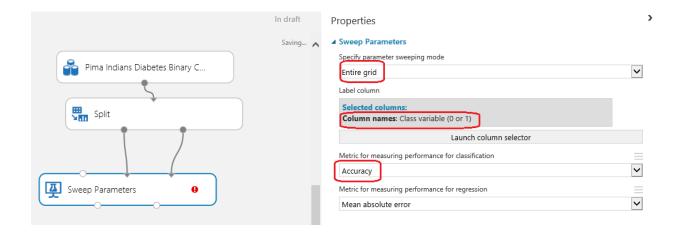
Classification - Pima Indians Diabetes Dataset > Evaluate Model > Evaluation results

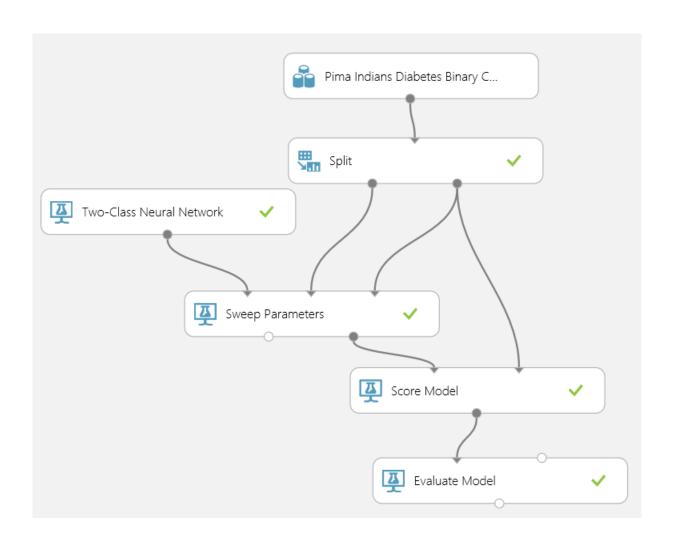
ROC PRECISION/RECALL LIFT

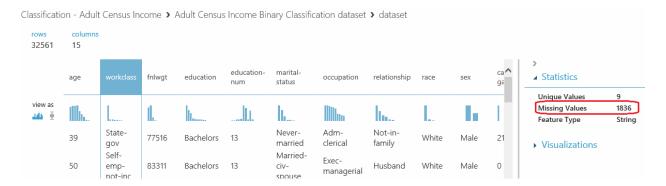


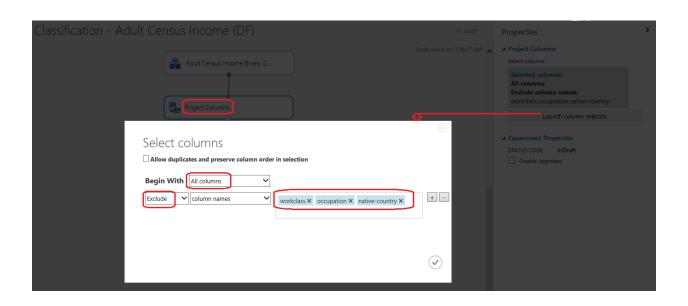
False Positive Rate

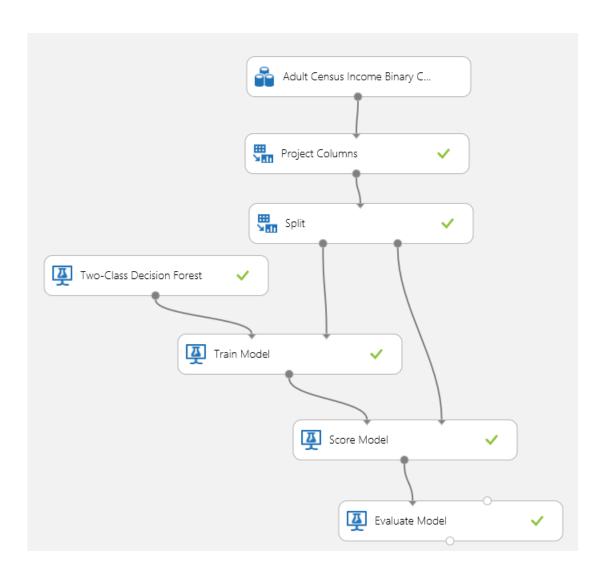
True Positive 23	False Negative 23	Accuracy 0.779	Precision 0.676	Threshold 0.51	AUC 0.788
False Positive	True Negative	Recall 0.500	F1 Score 0.575		

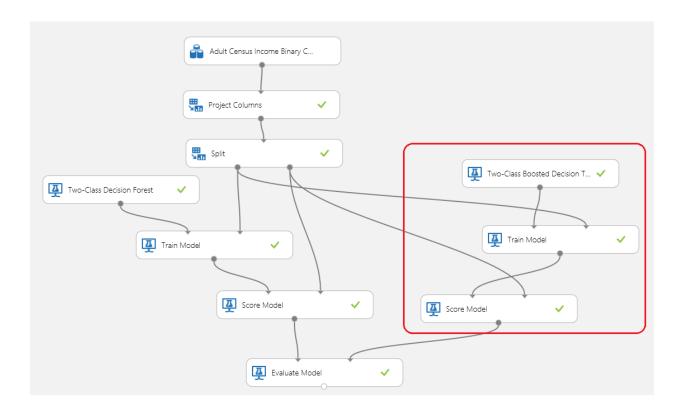


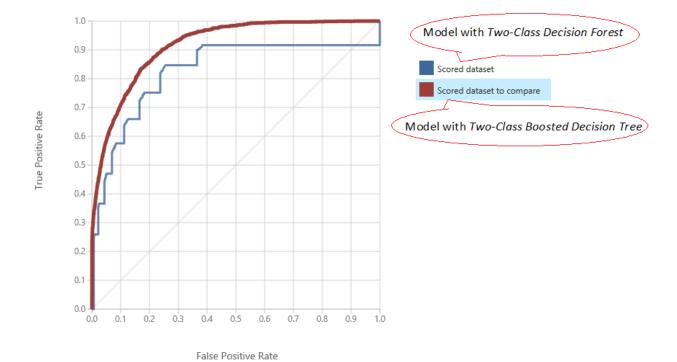




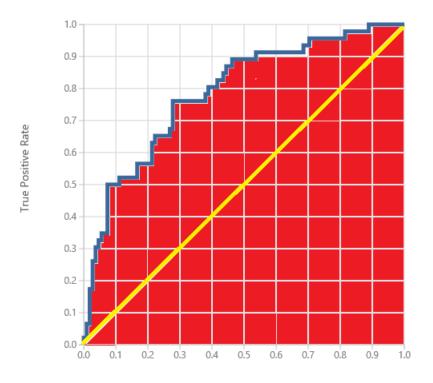




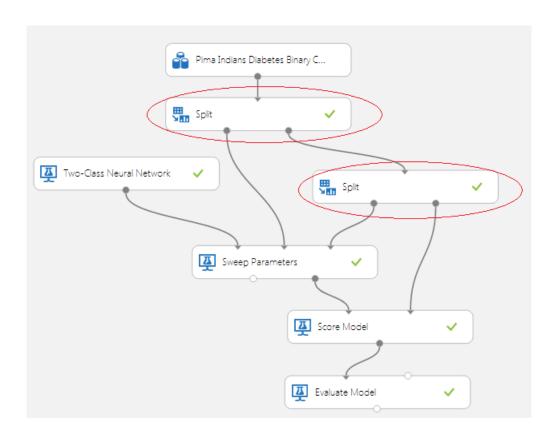


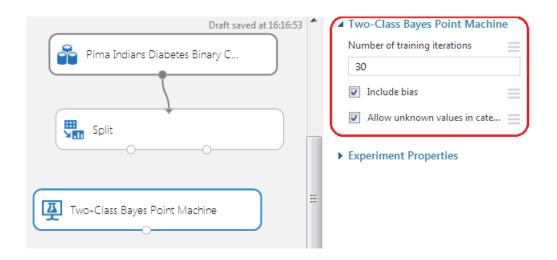


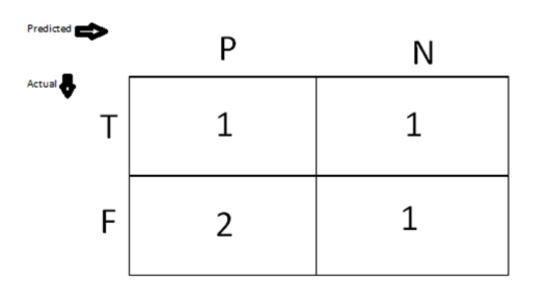
True Positive 1017	False Negative 561	Accuracy 0.859	Precision 0.741	Threshold 0.5	0.916
False Positive	True Negative	Recall 0.644	F1 Score 0.689		



False Positive Rate

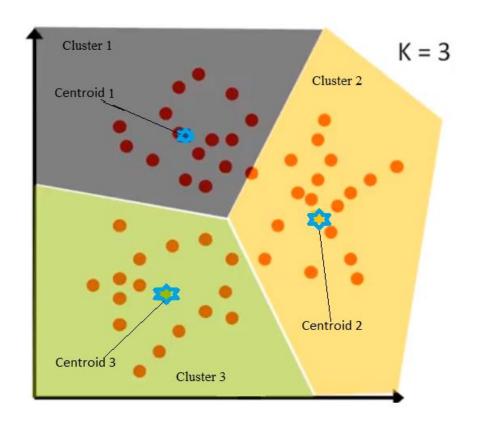






Predicted Low		Medium	High
Low	0 (0%)	1 (100%)	
Medium		2 (66.6%)	1(33.3%)
High			1 (100%)

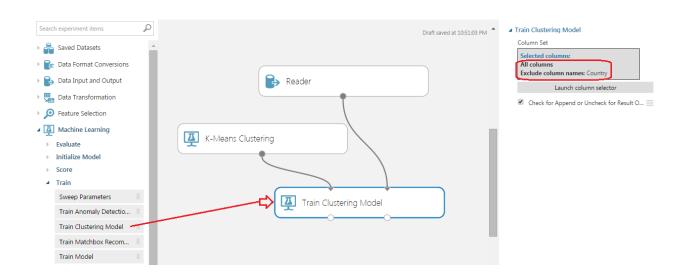
Chapter 8

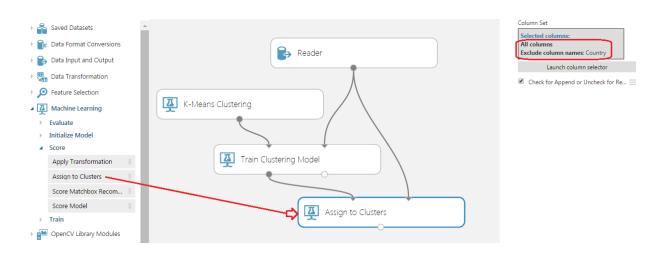


Country	RedMeat	WhiteMeat	Eggs	Milk	Fish	Cereals	Starch	Nuts	Fr&Veg
Albania	10.1	1.4	0.5	8.9	0.2	42.3	0.6	5.5	1.7
Austria	8.9	14	4.3	19.9	2.1	28	3.6	1.3	4.3
Belgium	13.5	9.3	4.1	17.5	4.5	26.6	5.7	2.1	4
Bulgaria	7.8	6	1.6	8.3	1.2	56.7	1.1	3.7	4.2
Czechoslovakia	9.7	11.4	2.8	12.5	2	34.3	5	1.1	4
Denmark	10.6	10.8	3.7	25	9.9	21.9	4.8	0.7	2.4
E Germany	8.4	11.6	3.7	11.1	5.4	24.6	6.5	0.8	3.6
Finland	9.5	4.9	2.7	33.7	5.8	26.3	5.1	1	1.4
France	18	9.9	3.3	19.5	5.7	28.1	4.8	2.4	6.5

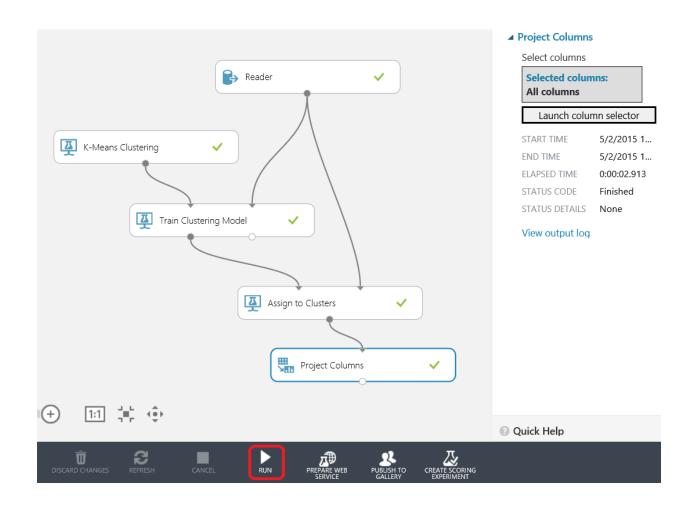




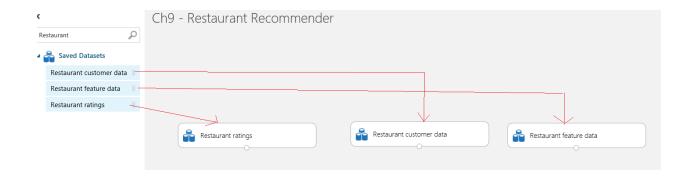


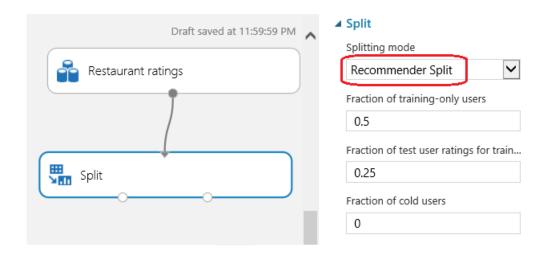


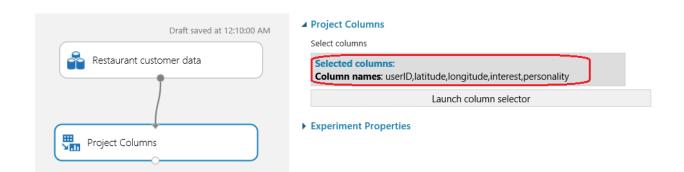
Country	RedMeat	WhiteMeat	Eggs	Milk	Fish	Cereals	Starch	Nuts	Fr&Veg	Assignments
	alla	ali da		Juli.	II II.	dua	nath	l	ulit. a.	lu.
Albania	10.1	1.4	0.5	8.9	0.2	42.3	0.6	5.5	1.7	0
Austria	8.9	14	4.3	19.9	2.1	28	3.6	1.3	4.3	1
Belgium	13.5	9.3	4.1	17.5	4.5	26.6	5.7	2.1	4	1
Bulgaria	7.8	6	1.6	8.3	1.2	56.7	1.1	3.7	4.2	0
Czechoslovakia	9.7	11.4	2.8	12.5	2	34.3	5	1.1	4	2
Denmark	10.6	10.8	3.7	25	9.9	21.9	4.8	0.7	2.4	1
E Germany	8.4	11.6	3.7	11.1	5.4	24.6	6.5	0.8	3.6	2
Finland	9.5	4.9	2.7	33.7	5.8	26.3	5.1	1	1.4	1
France	18	9.9	3.3	19.5	5.7	28.1	4.8	2.4	6.5	1
Greece	10.2	3	2.8	17.6	5.9	41.7	2.2	7.8	6.5	0
Hungary	5.3	12.4	2.9	9.7	0.3	40.1	4	5.4	4.2	2
	Albania Austria Belgium Bulgaria Czechoslovakia Denmark E Germany Finland France Greece	Albania 10.1 Austria 8.9 Belgium 13.5 Bulgaria 7.8 Czechoslovakia 9.7 Denmark 10.6 E Germany 8.4 Finland 9.5 France 18 Greece 10.2	Albania 10.1 1.4 Austria 8.9 14 Belgium 13.5 9.3 Bulgaria 7.8 6 Czechoslovakia 9.7 11.4 Denmark 10.6 10.8 E Germany 8.4 11.6 Finland 9.5 4.9 France 18 9.9 Greece 10.2 3	Albania 10.1 1.4 0.5 Austria 8.9 14 4.3 Belgium 13.5 9.3 4.1 Bulgaria 7.8 6 1.6 Czechoslovakia 9.7 11.4 2.8 Denmark 10.6 10.8 3.7 E Germany 8.4 11.6 3.7 Finland 9.5 4.9 2.7 France 18 9.9 3.3 Greece 10.2 3 2.8	Albania 10.1 1.4 0.5 8.9 Austria 8.9 14 4.3 19.9 Belgium 13.5 9.3 4.1 17.5 Bulgaria 7.8 6 1.6 8.3 Czechoslovakia 9.7 11.4 2.8 12.5 Denmark 10.6 10.8 3.7 25 E Germany 8.4 11.6 3.7 11.1 Finland 9.5 4.9 2.7 33.7 France 18 9.9 3.3 19.5 Greece 10.2 3 2.8 17.6	Albania 10.1 1.4 0.5 8.9 0.2 Austria 8.9 14 4.3 19.9 2.1 Belgium 13.5 9.3 4.1 17.5 4.5 Bulgaria 7.8 6 1.6 8.3 1.2 Czechoslovakia 9.7 11.4 2.8 12.5 2 Denmark 10.6 10.8 3.7 25 9.9 E Germany 8.4 11.6 3.7 11.1 5.4 Finland 9.5 4.9 2.7 33.7 5.8 France 18 9.9 3.3 19.5 5.7 Greece 10.2 3 2.8 17.6 5.9	Albania 10.1 1.4 0.5 8.9 0.2 42.3 Austria 8.9 14 4.3 19.9 2.1 28 Belgium 13.5 9.3 4.1 17.5 4.5 26.6 Bulgaria 7.8 6 1.6 8.3 1.2 56.7 Czechoslovakia 9.7 11.4 2.8 12.5 2 34.3 Denmark 10.6 10.8 3.7 25 9.9 21.9 E Germany 8.4 11.6 3.7 11.1 5.4 24.6 Finland 9.5 4.9 2.7 33.7 5.8 26.3 France 18 9.9 3.3 19.5 5.7 28.1 Greece 10.2 3 2.8 17.6 5.9 41.7	Albania 10.1 1.4 0.5 8.9 0.2 42.3 0.6 Austria 8.9 14 4.3 19.9 2.1 28 3.6 Belgium 13.5 9.3 4.1 17.5 4.5 26.6 5.7 Bulgaria 7.8 6 1.6 8.3 1.2 56.7 1.1 Czechoslovakia 9.7 11.4 2.8 12.5 2 34.3 5 Denmark 10.6 10.8 3.7 25 9.9 21.9 4.8 E Germany 8.4 11.6 3.7 11.1 5.4 24.6 6.5 Finland 9.5 4.9 2.7 33.7 5.8 26.3 5.1 France 18 9.9 3.3 19.5 5.7 28.1 4.8 Greece 10.2 3 2.8 17.6 5.9 41.7 2.2	Albania 10.1 1.4 0.5 8.9 0.2 42.3 0.6 5.5 Austria 8.9 14 4.3 19.9 2.1 28 3.6 1.3 Belgium 13.5 9.3 4.1 17.5 4.5 26.6 5.7 2.1 Bulgaria 7.8 6 1.6 8.3 1.2 56.7 1.1 3.7 Czechoslovakia 9.7 11.4 2.8 12.5 2 34.3 5 1.1 Denmark 10.6 10.8 3.7 25 9.9 21.9 4.8 0.7 E Germany 8.4 11.6 3.7 11.1 5.4 24.6 6.5 0.8 Finland 9.5 4.9 2.7 33.7 5.8 26.3 5.1 1 France 18 9.9 3.3 19.5 5.7 28.1 4.8 2.4 Greece 10.2 3 2.8 17.6 5.9 41.7 2.2 7.8	Albania 10.1 1.4 0.5 8.9 0.2 42.3 0.6 5.5 1.7 Austria 8.9 14 4.3 19.9 2.1 28 3.6 1.3 4.3 Belgium 13.5 9.3 4.1 17.5 4.5 26.6 5.7 2.1 4 Bulgaria 7.8 6 1.6 8.3 1.2 56.7 1.1 3.7 4.2 Czechoslovakia 9.7 11.4 2.8 12.5 2 34.3 5 1.1 4 Denmark 10.6 10.8 3.7 25 9.9 21.9 4.8 0.7 2.4 E Germany 8.4 11.6 3.7 11.1 5.4 24.6 6.5 0.8 3.6 Finland 9.5 4.9 2.7 33.7 5.8 26.3 5.1 1 1.4 France 18 9.9 3.3 19.5 5.7 28.1



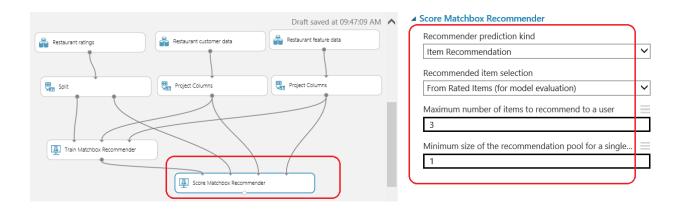
Chapter 9



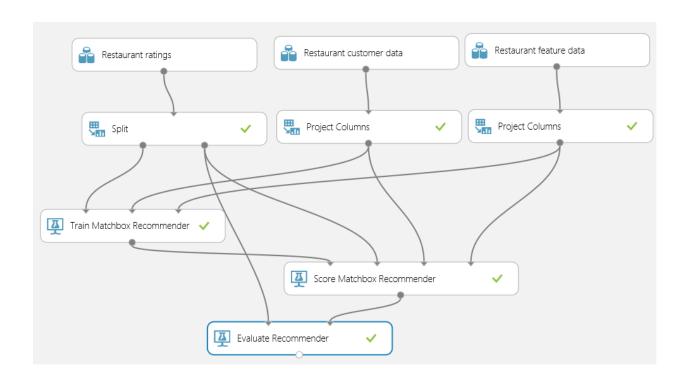


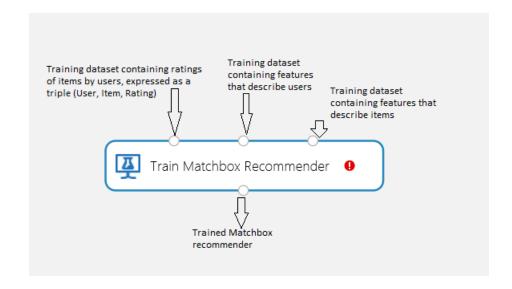


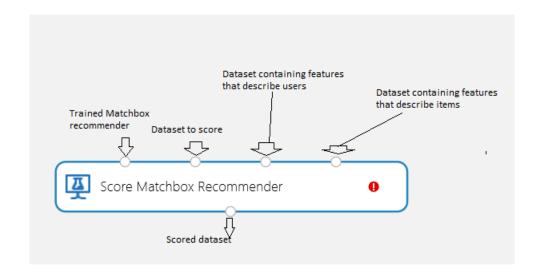


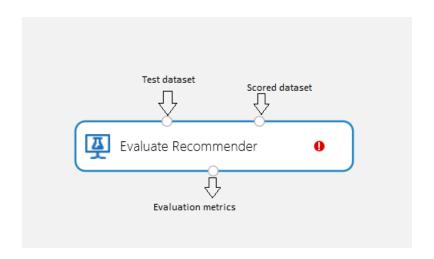


	User	Item 1	Item 2	Item 3
view as				
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	U1117	135018	132766	135088
	U1049	135052	132862	135051
	U1088	135057	135071	135032
	U1062	135052	135045	135062
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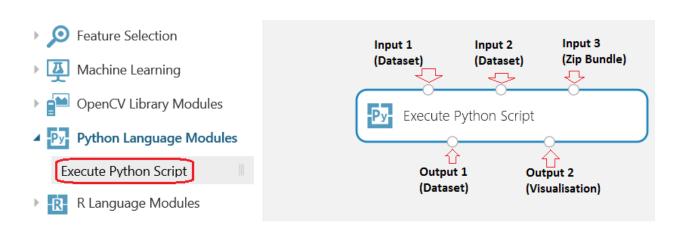




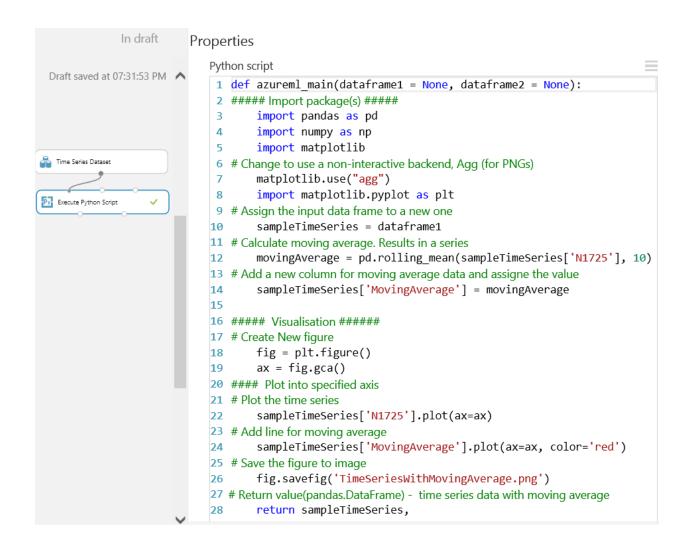


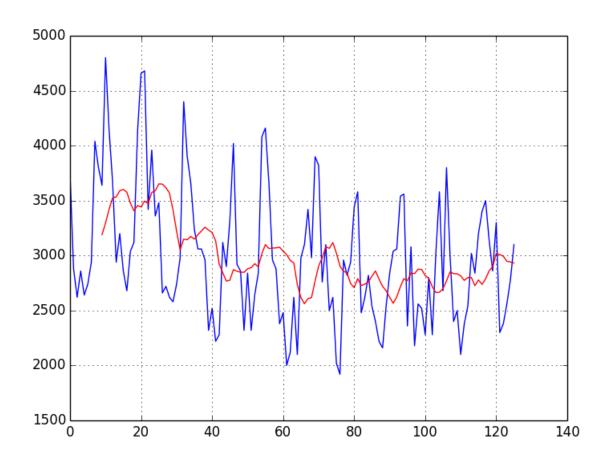


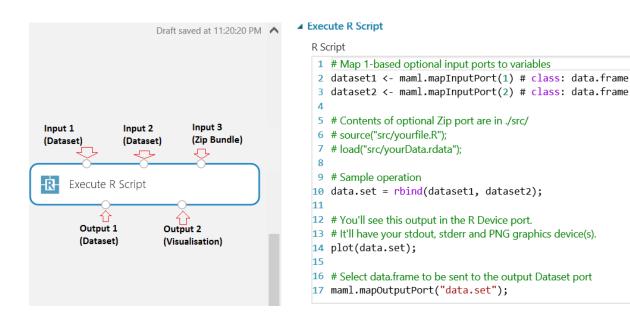
Chapter 10

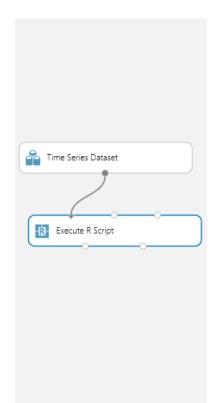


```
Python script
 1 # The script MUST contain a function named azureml main
 2 # which is the entry point for this module.
 4 # The entry point function can contain up to two input arguments:
 5 # Param < dataframe1>: a pandas.DataFrame
   # Param < dataframe 2 >: a pandas. DataFrame
 7
 8 def azureml_main(dataframe1 = None, dataframe2 = None):
 9
                                                 Input 2
10 ##### Execution logic goes here #########
# Code to get the result dataFrame
12
        resultDataFrame = ...
13
# Code to generate the visualisation (if any!)
15
# Return value must be of a sequence of pandas.DataFrame
17
        return resultDataFrame,
                          Output 1
```





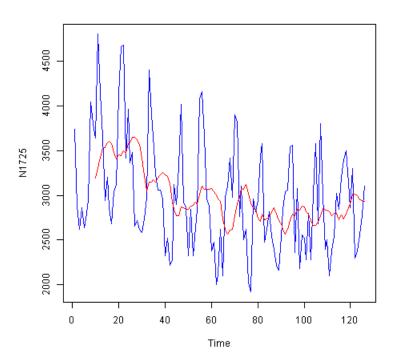




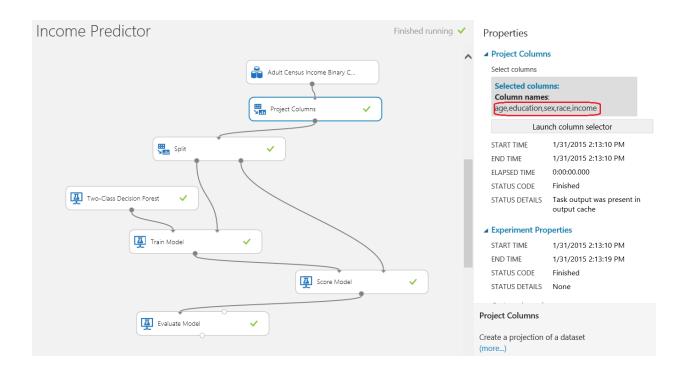
R Script

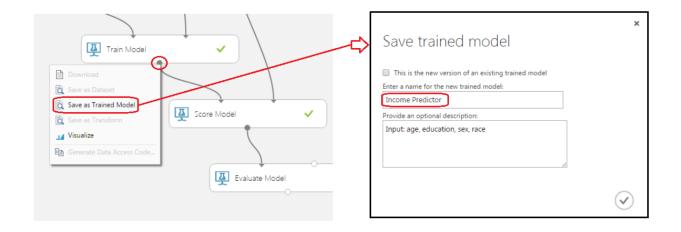
```
1 # Map 1-based optional input ports to variables
 2 dataset1 <- maml.mapInputPort(1) # class: data.frame</pre>
4 # Assign input dataset to new one
 5 timeSeries <- dataset1
7 # Load the zoo package to the session
8 library(zoo)
10 # Calculate moving average with moving window 10
11 ma <- rollmean(timeSeries[2],10)</pre>
12 # Add a new column and assign moving average values to it
13 timeSeries$MovingAverage[10:126] <- ma</pre>
14
15 #### Visualisation #####
16 # Plot time series
17 plot.ts(timeSeries['N1725'], col='blue')
18 # Add a line for moving average
19 lines(timeSeries['MovingAverage'], col='red')
20
21 # Select data.frame to be sent to the output Dataset port -
22 # time series data with moving average
23 maml.mapOutputPort("timeSeries");
```

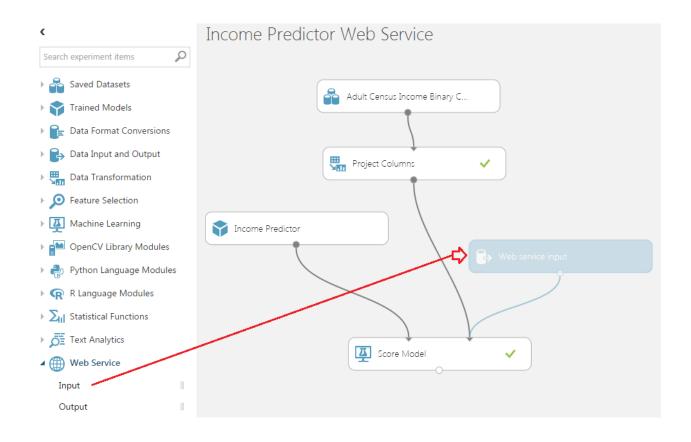
- ► Standard Output
- ► Standard Error
- ▲ Graphics Device

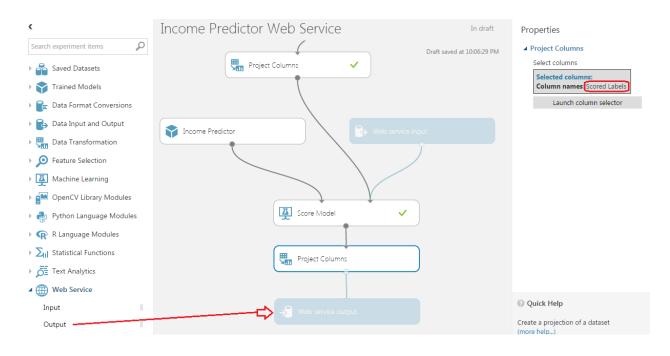


Chapter 11



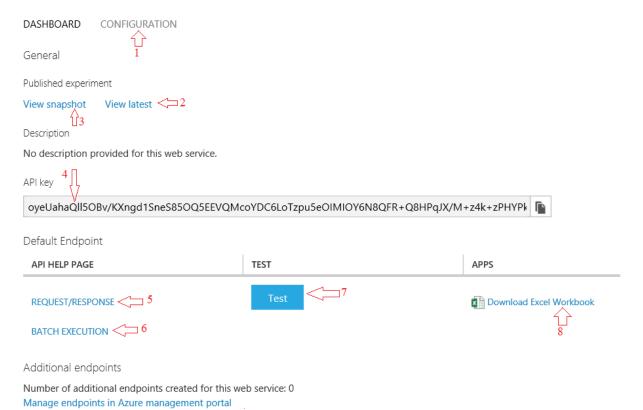








income predictor web service

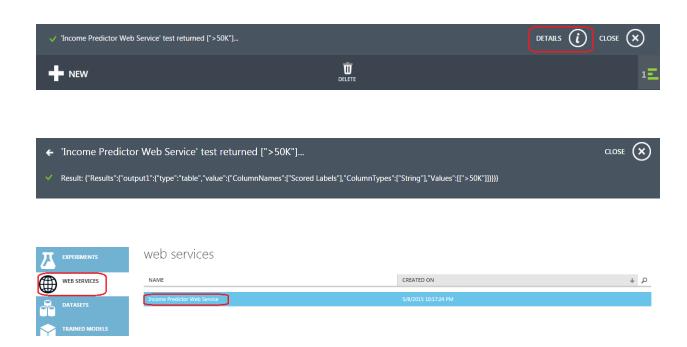


Test Income Predictor Web Service Service

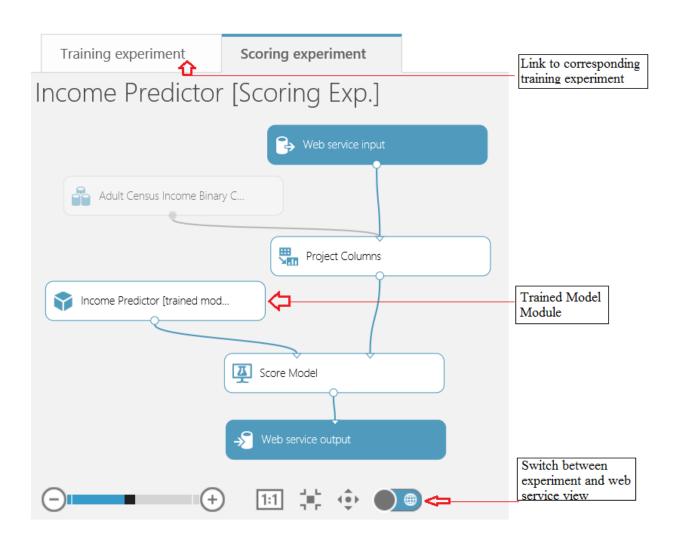
Enter data to predict

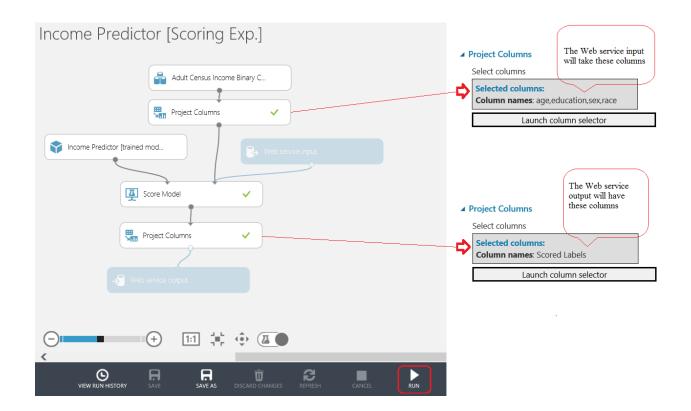
AGE	
49	
EDUCATION	
Masters	
RACE	
White	
SEX	
Male	



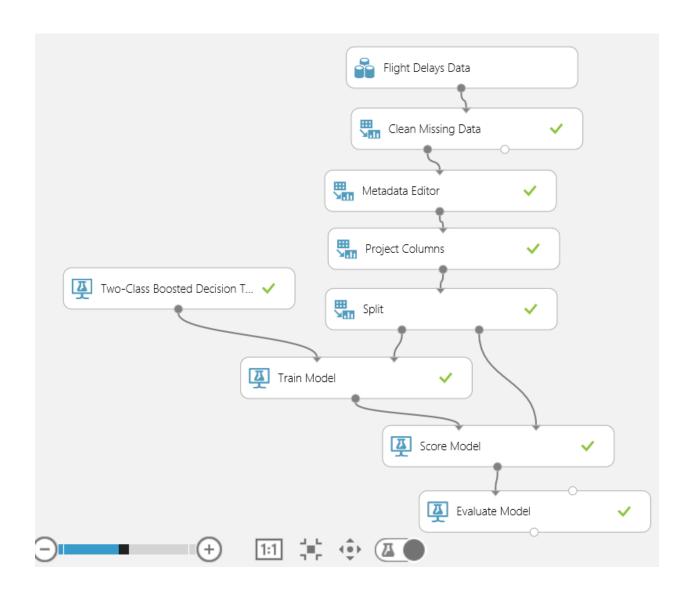


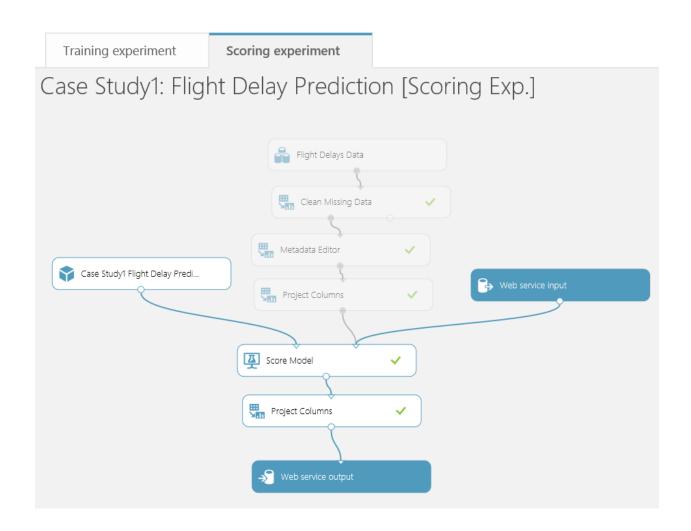






Chapter 12





Chapter 13

