

# Data Preparation and Exploratory Data Analysis

Citizen Analytics – An Initiative by Data Science Team

START >



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# **Learning Objectives**

By the end of this module, you will be able to:

01

Understand the concept of Exploratory Data Analysis (EDA) and Statistics.

02

Identify ways to treat outliers and clean the data.

03

Identify ways to treat class imbalance, normalize data and joining multiple data.



# **Content**

01.	<ul><li>Exploratory Data Analysis</li><li>a. Graphical Techniques used in EDA</li><li>b. Distribution in Box-Plot and</li></ul>	04	05.	<ul><li>Imbalance Data</li><li>a. Class imbalance</li></ul>
	Histogram		06.	<ul><li>Data Normalization</li><li>a. Data Normalization and Example</li></ul>
02.	Statistics	08		·
	<ul> <li>a. Statistics – Measures of Central Tendency</li> <li>b. Statistics – Measures of Dispersion</li> </ul>		07.	Joining Multiple Data  a. Joining Data and Example
	·		08.	Summary
03.	Outlier Treatment  a. Outlier Treatments and Example	12		
04.	Cleaning Missing Value  a. Converting data types and columns' name  b. Clean the dataset without NA values  c. Mathematical Operation	19		

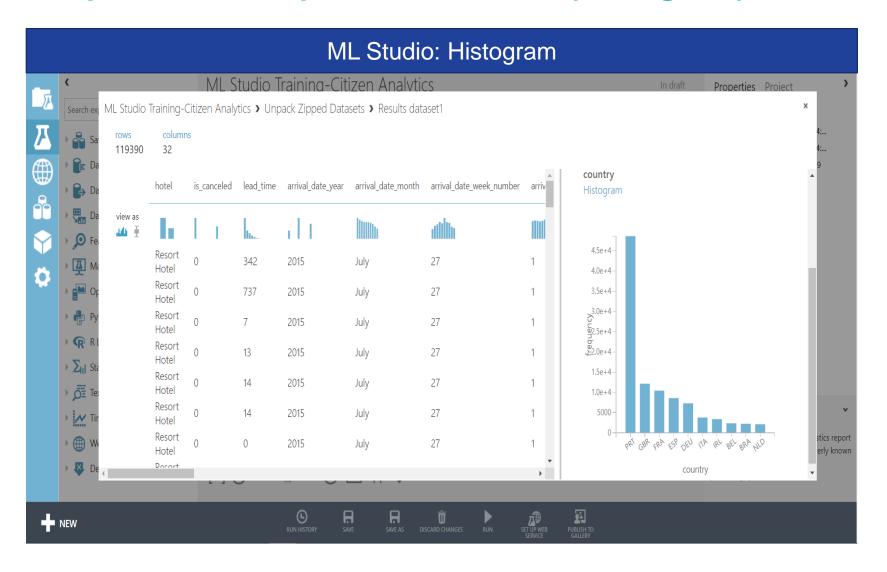


# **Exploratory Data Analysis**





# **Graphical Techniques used in EDA (Histogram)**

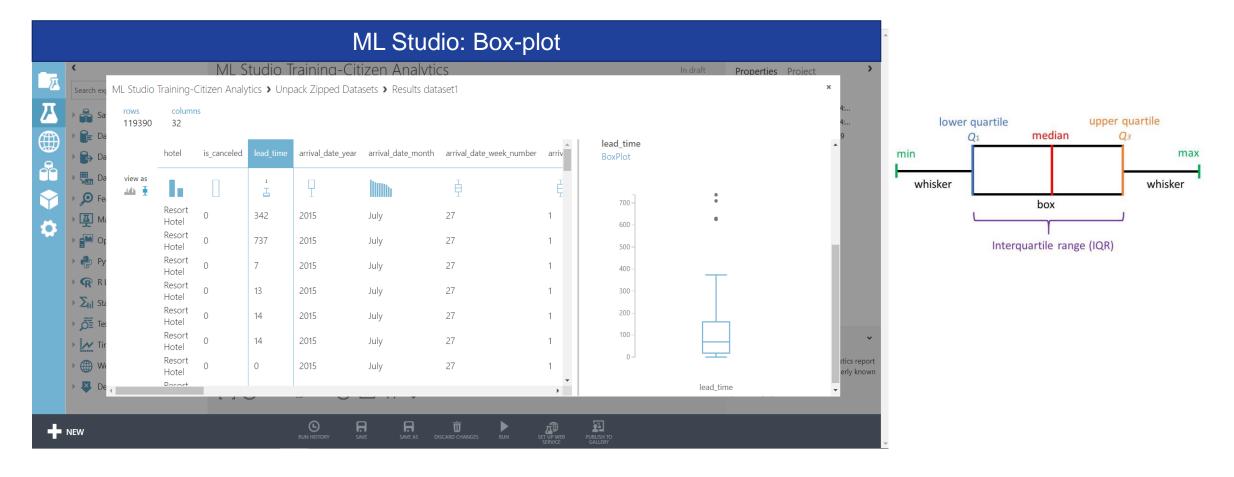


A histogram is a graphical display of data using **bars of different heights**, showing the shape and spread of continuous data. The taller the bar, the higher its occurrences.



# **Graphical Techniques used in EDA (Box-plot)**

• A box plot (also known as box and whisker plot) is a chart that used to show the distribution of numerical data and its skewness by showing its quartiles and mean as well as its range.





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# **Distribution in Box-Plot and Histogram**

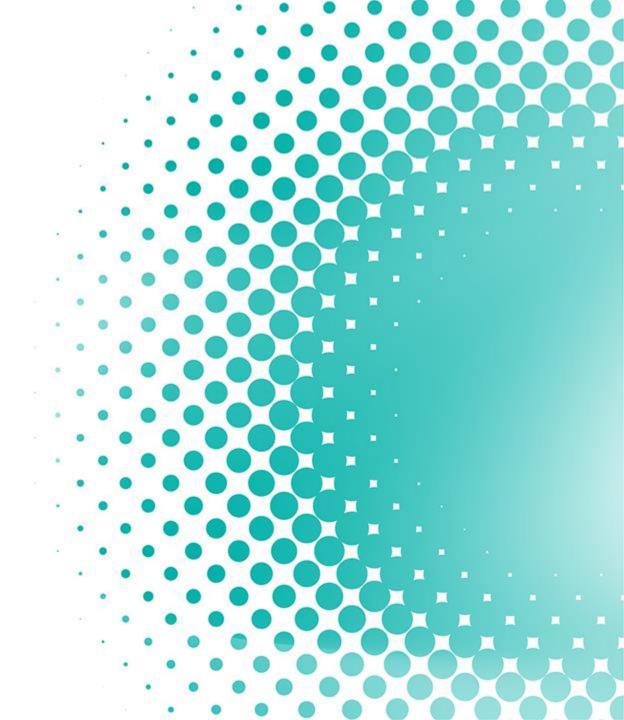
- In symmetric distributions (normal), the mean, median, and mode are the same.
- In skewed data, the mean and median lie further toward the skew than the mode, either skewed to the right or left.

Normally distributed Skewed



# **Statistics**

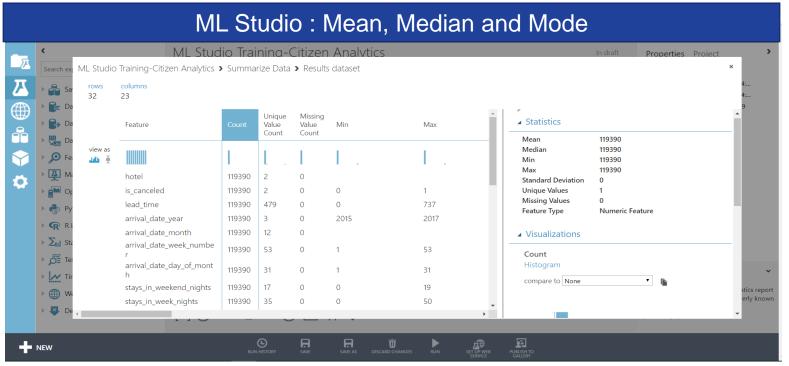




### **Statistics – Measures of Central Tendency**

A Measure of Central Tendency is a single value that attempts to describe the set of data by identifying the central position within that set of data.

- Mean
- Median
- Mode



Mean is the average value of a numeric dataset

Mode is a dataset that has the highest frequency

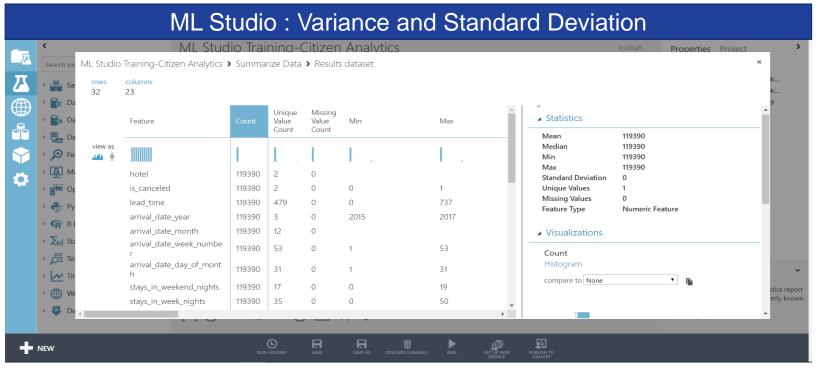
Median is the centric value of ascending ordered-dataset



# **Statistics – Measures of Dispersion**

Dispersion is a way of describing how spread out a set of data is.

- Range Difference of maximum and minimum value
- Interquartile Range Difference of first quartile and third quartile
- Variance
- Standard Deviation



Variance is the average squared distance between the mean and each data value.

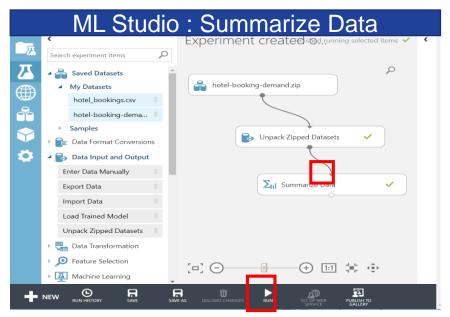
Standard deviation is just a square root of the variance, measuring how disperse is the data from the mean.



$$\sigma = \sqrt{\frac{\sum_{i=1}^{n} (x_i - \bar{x})^2}{n}}$$

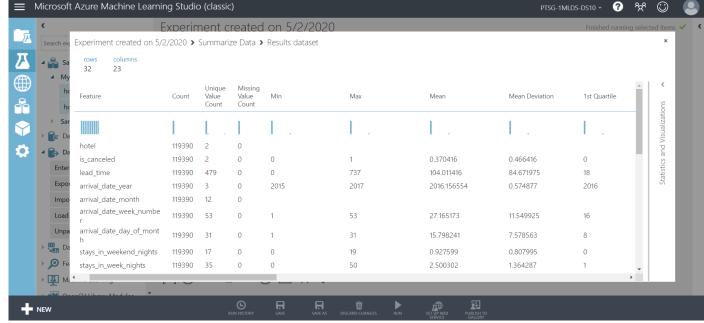
### **Summary of the data**

'Summarize Data' module allows us to see all related statistical measures such as mean, median, quartiles etc in just one pop out window.





Drag and drop "Summarize Data" module, click "Run" and later right-click on the circle to view Statistics Summary of your dataset of follows



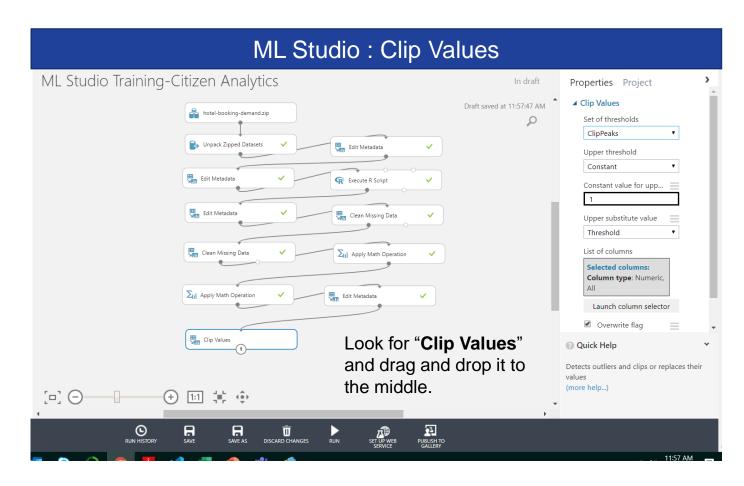


# **Outlier Treatment**



#### **Outliers treatment**

Outliers are observations that is far away from other observations and might create bias if it is not treated well



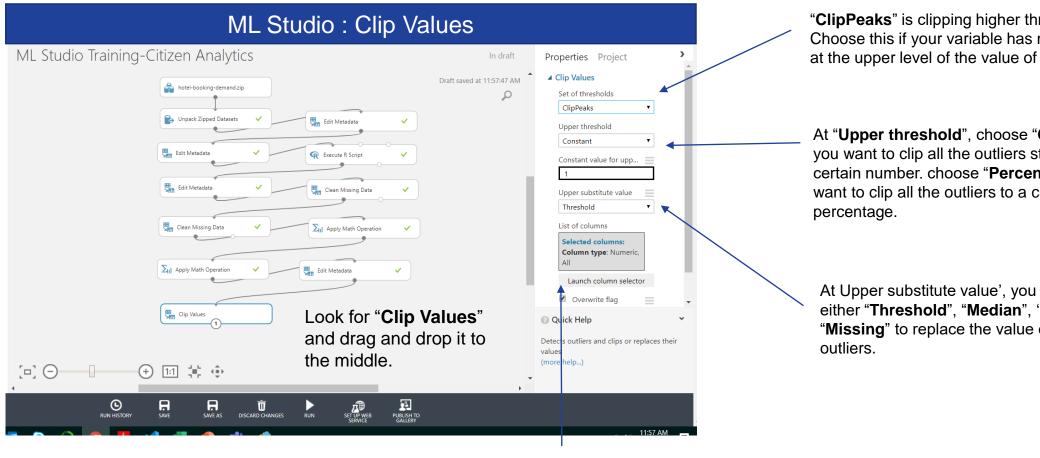
On "Set of thresholds", there are 3 choices which are "ClipPeaks", "ClipSubpeaks" and "ClipPeaksAndSubpeaks":

- "ClipPeaks" is clipping the higher threshold. Choose this if your variable has more outliers at the upper level of the value of the variable
- \*ClipSubpeaks" is clipping the lower threshold. Choose this if your variable has more outliers at the lower level of the value of the variable
- "ClipPeaksAndSubpeaks" is clipping threshold at the upper and lower level. Choose this if your variable has outliers both at the upper and lower level of the value of the variable



#### **Outliers treatment**

Outliers are observations that is far away from other observations and might create bias if it is not treated well



"ClipPeaks" is clipping higher threshold. Choose this if your variable has more outliers at the upper level of the value of the variable

At "Upper threshold", choose "Constant" if you want to clip all the outliers starting from a certain number. choose "Percentile" if you want to clip all the outliers to a certain

At Upper substitute value', you can choose either "Threshold", "Median", "Mean" and "Missing" to replace the value of the

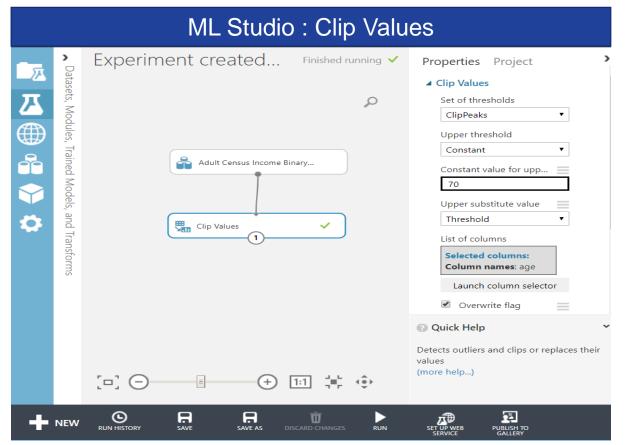


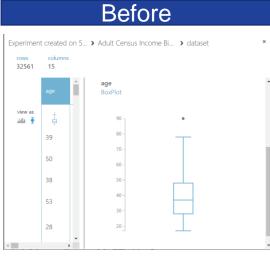
Lastly, select which column you want to treat for its outliers.

### **Outliers treatment - Example**

#### **Dataset = Adult Census Income Binary**

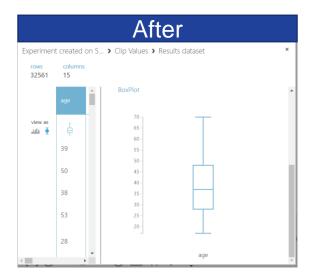
Set of thresholds (ClipPeaks) > Upper threshold (Constant) > Constant value for upper threshold (70) > Upper substitute value (Threshold) > Column (Age)





The dot represents the outliers in this data





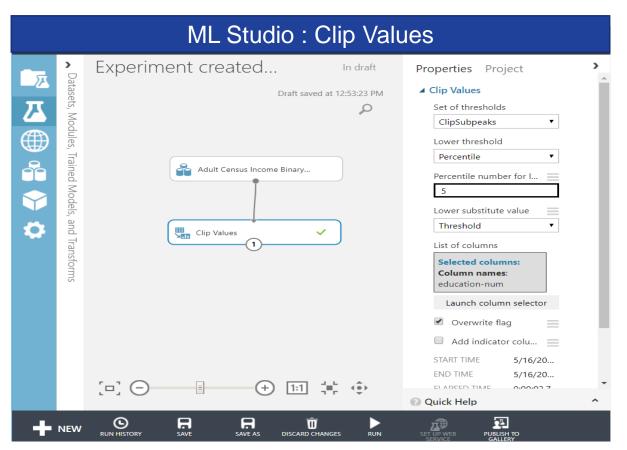
The data is already clipped and the outliers is no longer there as the value is replaced by using the new upper threshold.

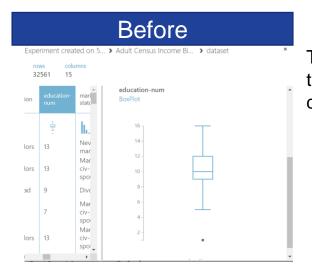


### **Outliers treatment - Example**

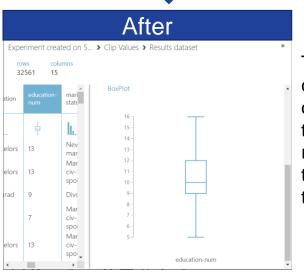
#### **Dataset = Adult Census Income Binary**

Set of thresholds (ClipSubpeaks) > Lower threshold (Percentile) > Percentile value for upper threshold (5) > Upper substitute value (Threshold) > Column (education-num)





The dot represents the outliers in this data



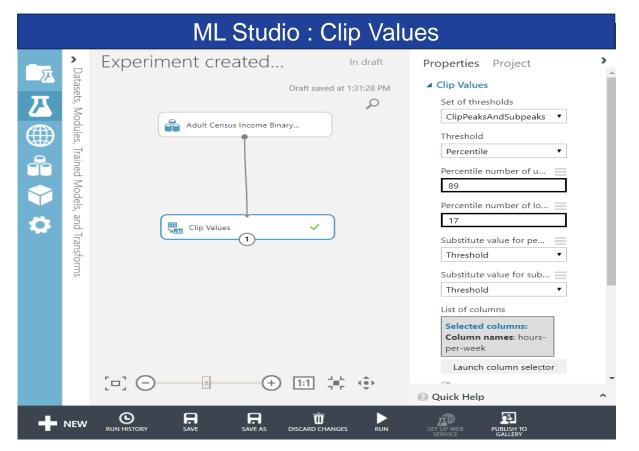
The data is already clipped and the outliers is no longer there as the value is replaced by using the new lower threshold.

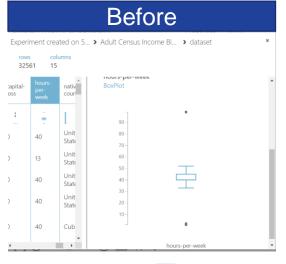


### **Outliers treatment - Example**

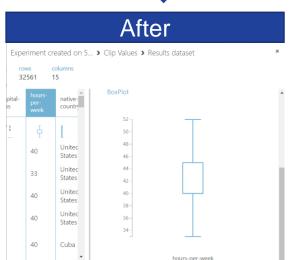
#### **Dataset = Adult Census Income Binary**

Set of thresholds (ClipPeaksAndSubpeaks) > Threshold (Percentile) > Percentile value for upper threshold (89) > Percentile value for lower threshold (17) > Upper substitute value (Threshold) > Column (hours-per-week)





The dot represents the outliers in this data

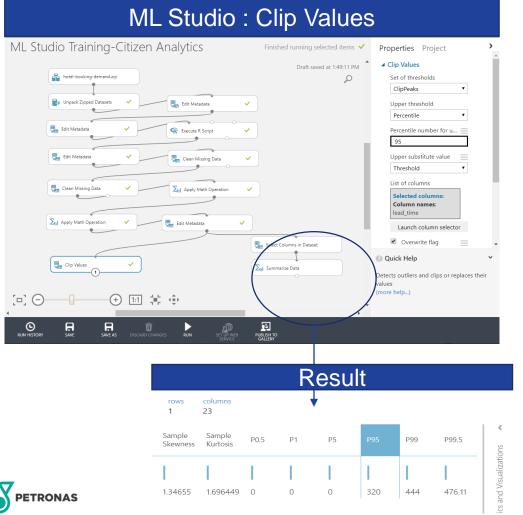


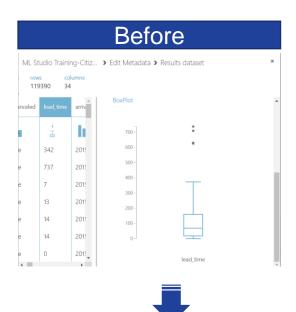
The data is already clipped and the outliers is no longer there as the value is replaced by using the new upper and lower percentile of threshold.



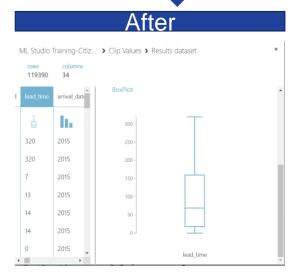
#### **Outliers treatment**

Set of threshols (ClipPeaks) > Upper threshold (Percentile) > Constant value for upper threshold (95) > Upper substitute value (Threshold) > Column (lead\_time)





The dot represents the outliers in this data



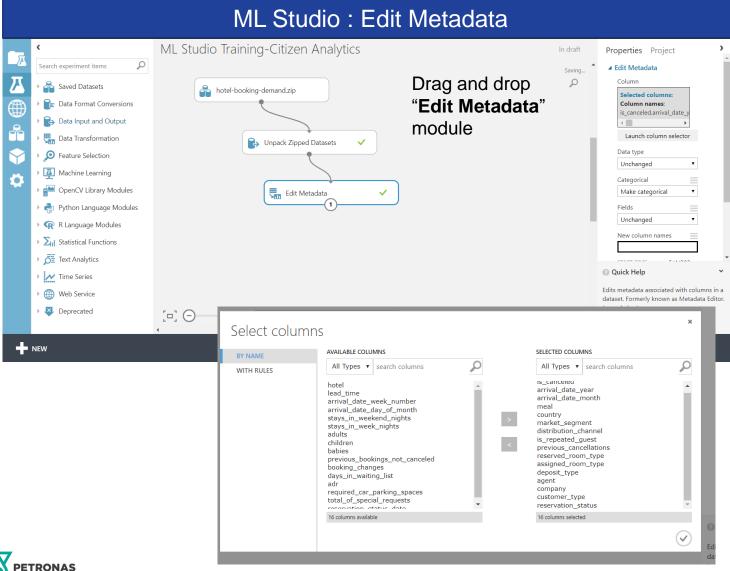
The data is already clipped and the outliers is no longer there as the value is replaced by using the new upper threshold.

# **Cleaning Missing Value**





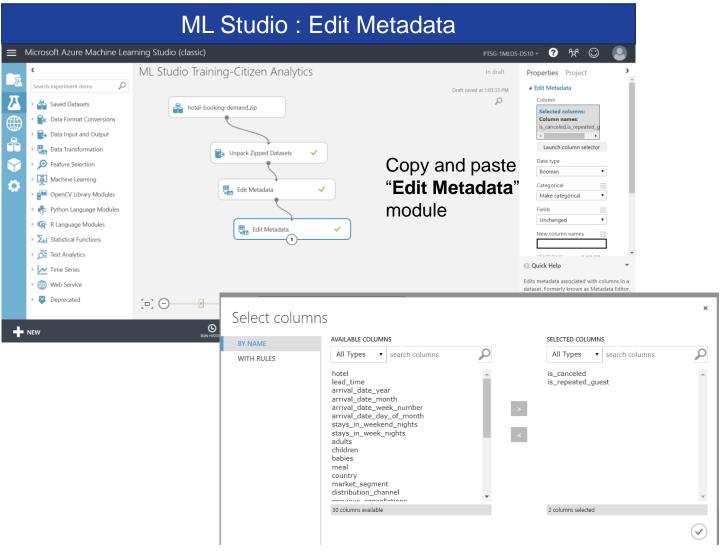
# Converting data types and columns' name



Click "Launch column selector" to select columns that you want to change either its datatype, data category or its name

Change these variables to Categorical data by selecting 'Make Categorical' in 'Categorical' section.

# Converting data types and columns' name

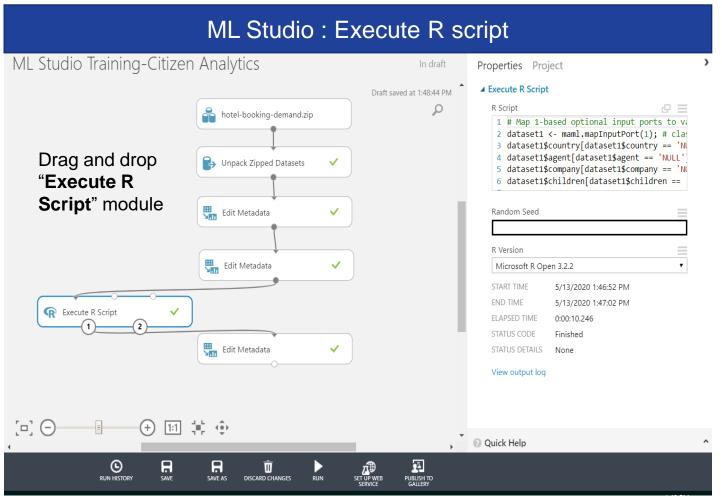


Click "Launch column selector" to select columns that you want to change either its datatype, data category or its name

Change these variables to Categorical data by selecting 'Make Categorical' in 'Categorical' section and 'Boolean' for datatype.



#### Clean the dataset without NA values



Change "**NULL**" and "**NA**" values in 'Children', 'Agent', 'Company' and 'Country' to NA

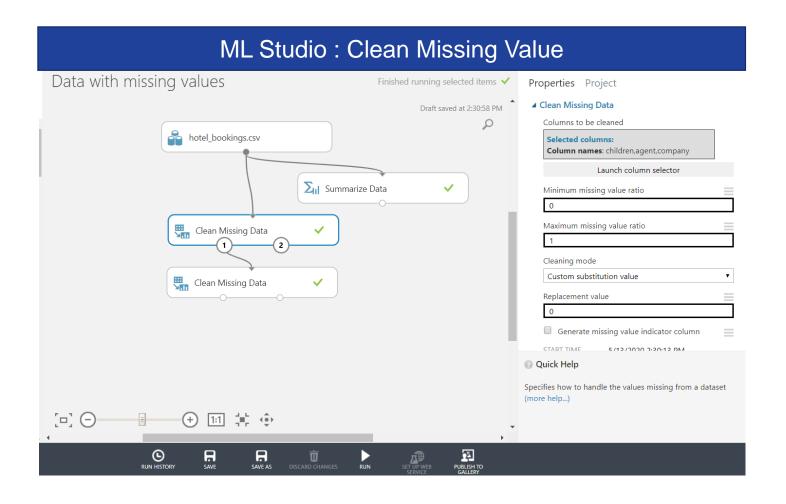
Copy and paste this R code in "R Script" space.

# Map 1-based optional input ports to variables dataset1 <- maml.mapInputPort(1); # class: data.frame dataset1\$country[dataset1\$country == 'NULL'] = NA; dataset1\$agent[dataset1\$agent == 'NULL'] = NA; dataset1\$company[dataset1\$company == 'NULL'] = NA; dataset1\$children[dataset1\$children == 'NA'] = NA; maml.mapOutputPort("dataset1");



Internal

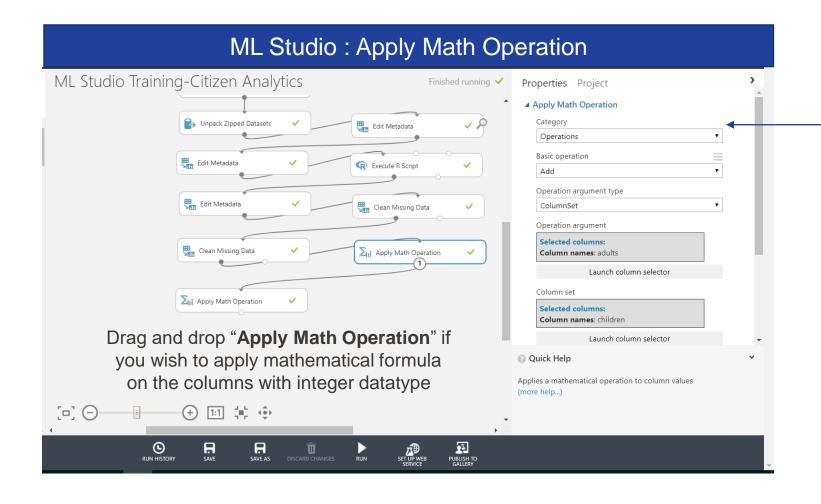
#### Clean the dataset with NA values



Replace NA values in 'Children', 'Agent', 'Company' to 0 while 'Country' to Unknown



# **Mathematical Operation**



On "Category", you can select what kind of mathematical function you want to apply to integer columns

Add values in columns 'Adult', 'Children' and then after getting the result, add to 'Babies' and rename the new columns

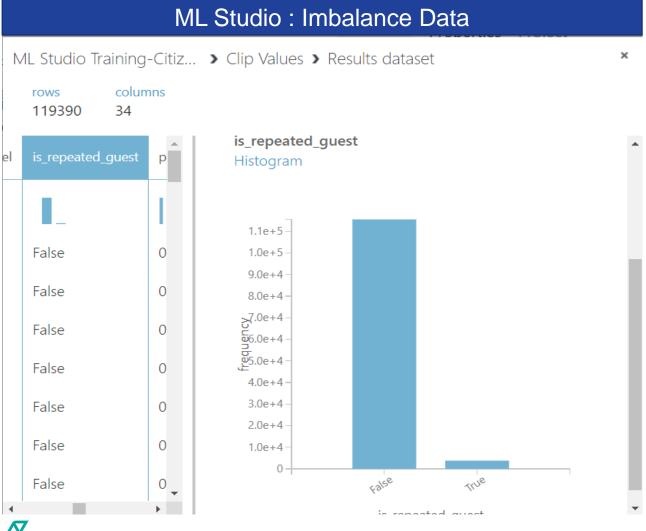
Internal

# **Imbalance Data**



#### Class Imbalance

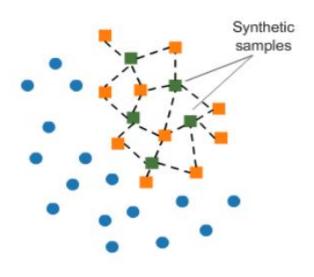
Imbalanced data happen when the classes are not represented equally



Dealing with imbalanced dataset using SMOTE

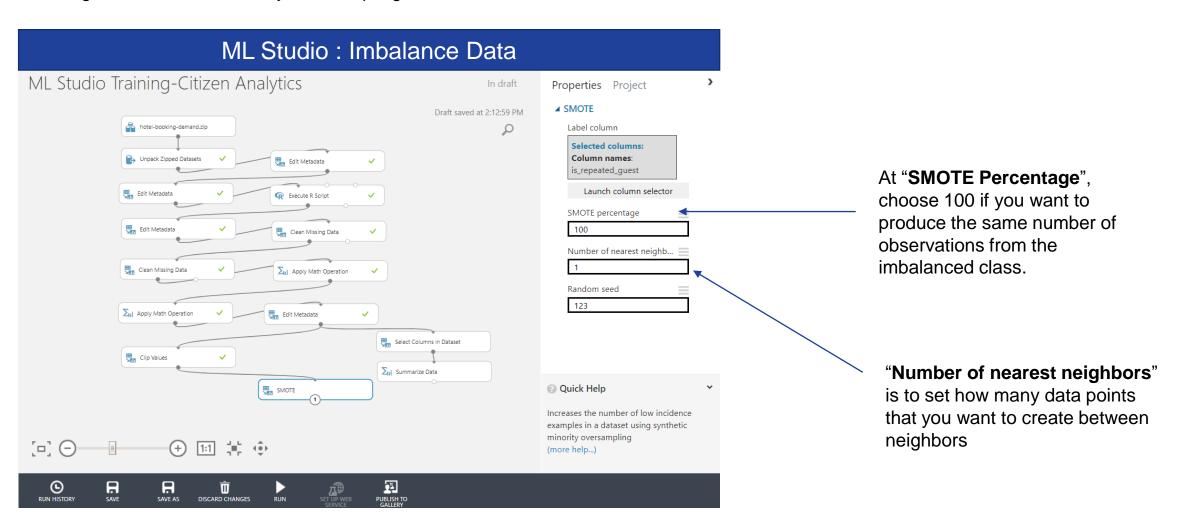
– Synthetic Minority Oversampling Technique

SMOTE is just creating new synthetics datapoints between two datapoints



#### Class Imbalance

Dealing with imbalance data by oversampling method





Internal

# **Data Normalization**





#### **Data Normalization**

#### Concept

A method to standardize the range of independent variables value, where mostly the resulting range will be fitted within 0 to 1.

#### Why need to normalize?

• To make data more stable and the coefficient that will be derived is more reliable as the data is on the same scale.

$$z = \frac{x - \bar{x}}{\sigma}$$

**Z-score Normalization** 

$$z = \frac{x - \min(x)}{\max(x) - \min(x)}$$

Min-Max Normalization

$$z = \frac{1}{1 + \exp(x)}$$

**Logistic Normalization** 

$$z = lognormal.CDF(x, \mu, \sigma)$$

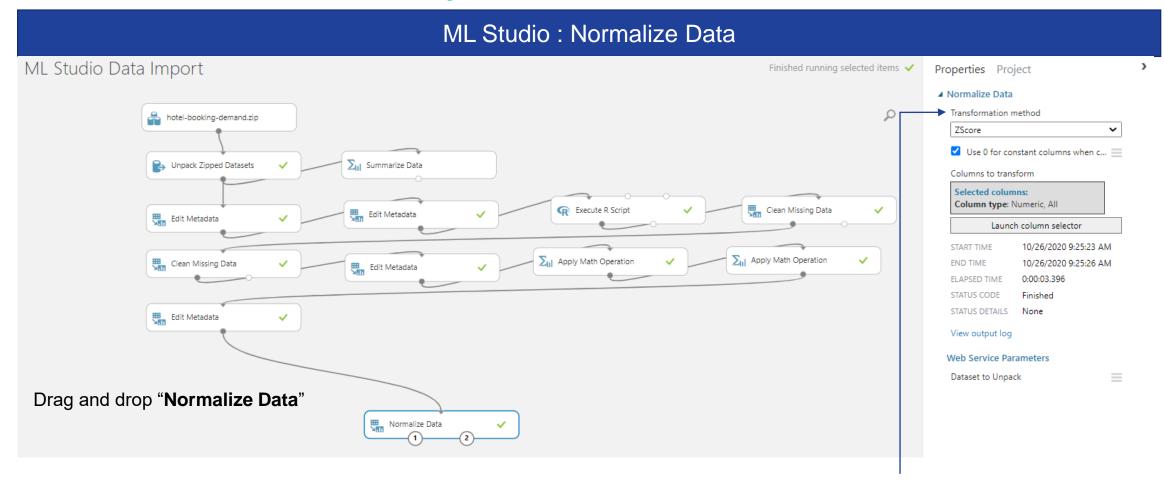
LogNormal Normalization

$$= \frac{p(k|x;\theta)}{[E(Y|x)]^k e^{E(Y|x)}}$$

**TanH Normalization** 



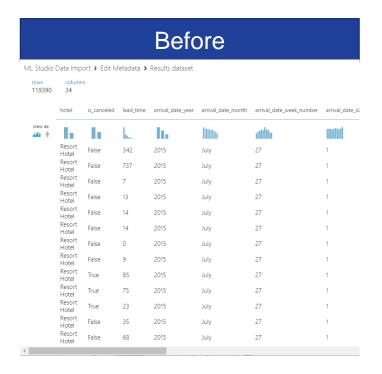
# **Data Normalization - Example**



At "**Transformation method**", select type of normalization and select columns



# **Data Normalization - Example**





The original data with values ranges in hundred.

Charles	After						
IL Studio Data Import > Normalize Data > Transformed dataset  rows columns  119390 34							
	hotel	is_canceled	lead_time	arrival_date_year	arrival_date_month	arrival_date_week_number	arrival_date_da
view as	Ъ	h	li.	li.		dille	
	Resort Hotel	False	2.227051	2015	July	-0.012141	-1.685297
	Resort Hotel	False	5.923385	2015	July	-0.012141	-1.685297
	Resort Hotel	False	-0.907814	2015	July	-0.012141	-1.685297
	Resort Hotel	False	-0.851667	2015	July	-0.012141	-1.685297
	Resort Hotel	False	-0.842309	2015	July	-0.012141	-1.685297
	Resort Hotel	False	-0.842309	2015	July	-0.012141	-1.685297
	Resort Hotel	False	-0.973319	2015	July	-0.012141	-1.685297
	Resort Hotel	False	-0.889098	2015	July	-0.012141	-1.685297
	Resort Hotel	True	-0.177905	2015	July	-0.012141	-1.685297
	Resort Hotel	True	-0.271483	2015	July	-0.012141	-1.685297
	Resort Hotel	True	-0.758089	2015	July	-0.012141	-1.685297
	Resort Hotel	False	-0.645795	2015	July	-0.012141	-1.685297
	Resort Hotel	False	-0.336988	2015	July	-0.012141	-1.685297

The normalized data with values ranges in tenth using Z score.



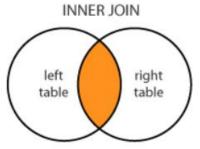
# **Joining Multiple Data**

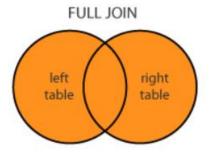




# Joining data

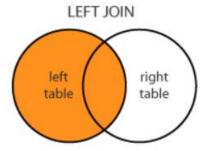
Data is merged on the common column on both tables

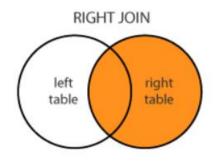




Data is combined regardless of the columns name.

Data is merged on the common column but returns only the values from the left table while the key columns match.





Data is merged on the common column but returns only the values from the right table while the key columns match.

#### Notes:

When a row in the left (right) table has no matching rows in the right (left) table, the returned row contains missing values for all columns that come from the right (left) table unless you specify a replacement value for missing values.



# Joining data - Example

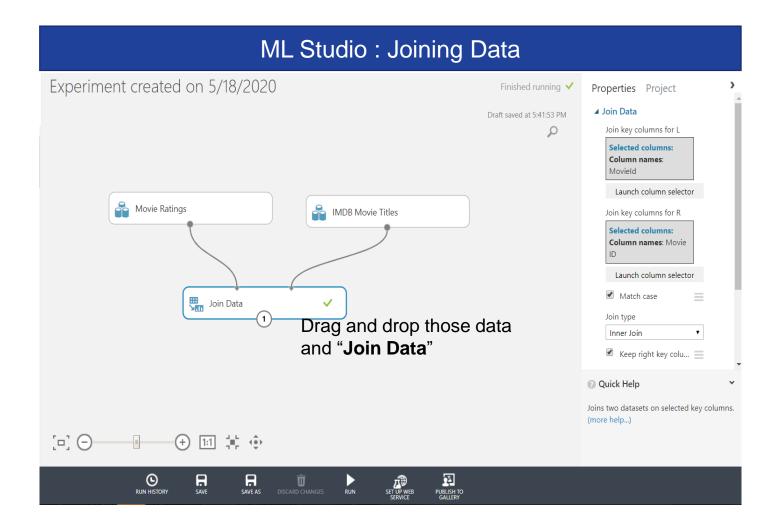


IMDB Movie Titles Dataset							
Experiment created on 5/18/2020 > IMDB Movie Titles > dataset							
rows 16614	columns 2						
	Movie ID	Movie Name					
view as	luttu						
	8	Edison Kinetoscopic Record of a Sneeze (1894)					
	91	Le manoir du diable (1896)					
	417	Le voyage dans la lune (1902)					
	628	The s of Dollie (1908)					
	833	The Country Doctor (1909)					
	1223	Frankenstein (1910)					
	1740	The Lonedale Operator (1911)					
	2101	Cleopatra (1912)					
	2130	Linferno (1911)					
1		Fantmas - lombre de la					

Both are original data from two different tables, aims to join this two tables on 'Movield' column.



# Joining data - Example



Select which column has the same data attribute

At "Join Type", select either "Inner Join", "Left Outer Join", "Left Semi-Join" or "Full Outer Join"

- •Left Outer Join = Left join
- •Full Outer Join = Outer join
- •For each of the rows in the left table that has no matching rows in the right table, the join results include a row containing missing values from the right table.
- •For each of the rows in the right table that has no matching rows in the left table, the join results include a row containing missing values for all columns from the left table.
- •Left Semi-Join = Left join but returns only the values when key columns match



# Joining data - Example

### ML Studio: Joining Data-Result

rows columns 227472 6

view	as
dist	Ŧ

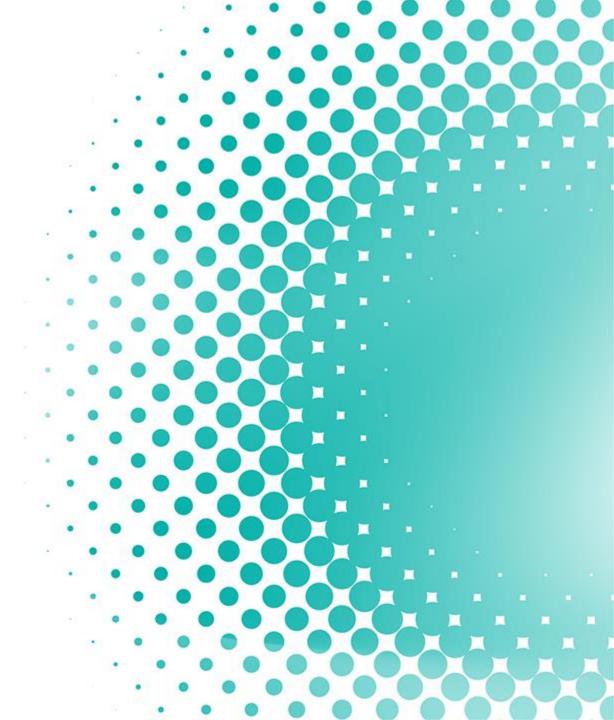
UserId	Movield	Rating	Timestamp	Movie ID	Movie Name
	lath.			lalla.	
1	68646	10	1381620027	68646	The Godfather (1972)
1	113277	10	1379466669	113277	Heat (1995)
2	454876	8	1394818630	454876	Life of Pi (2012)
2	790636	7	1389963947	790636	Dallas Buyers Club (2013)
2	816711	8	1379963769	816711	World Z (2013)
2	1091191	7	1391173869	1091191	Lone Survivor (2013)
2	1322269	7	1391529691	1322269	August Osage County (2013)
2	1433811	8	1380453043	1433811	Disconnect (2012)
2	1454468	8	1387016442	1454468	Gravity (2013)
2	1535109	8	1386350135	1535109	Captain Phillips (2013)Biography
2	1675434	8	1396688981	1675434	Intouchables (2011)

Both tables already joined by matching the 'Movield' values



Internal

# **Summary and References**



# **Summary**

1

#### **Exploratory Data Analysis**

- Graphical Techniques used in EDA
- Distribution in Box-Plot and Histogram

2

#### **Statistics**

- Statistics Measures of Central Tendency
- Statistics Measures of Dispersion

3

#### **Outlier Treatment**

 Outlier Treatments using "Clip Value" module

4

### **Cleaning Missing Value**

- Converting data types and columns' name
- Clean the dataset without NA values
- Mathematical Operation

https://docs.microsoft.com/en-us/azure/machine-learning/algorithm-module-reference/clip-values

https://docs.microsoft.com/en-us/azure/machine-learning/algorithm-module-reference/clean-missing-data

https://docs.microsoft.com/en-us/azure/machine-learning/algorithm-module-reference/apply-math-operation



### **Summary**

5

#### **Imbalance Data**

Treat Class imbalance using SMOTE

https://docs.microsoft.com/en-us/azure/machine-learning/algorithm-module-reference/smote

6

#### **Data Normalization**

 Data Normalization using "Transformation Method" module  https://docs.microsoft.com/en-us/azure/machinelearning/algorithm-module-reference/normalize-data

7

### **Joining Multiple Data**

 Joining Data using "Inner Join", "Left Outer Join", "Left Semi-Join" or "Full Outer Join"

 https://docs.microsoft.com/en-us/azure/machinelearning/algorithm-module-reference/join-data



# Thank you for your passion!

