AIT-582

Application of Metadata in Complex Big Data Problems

Project Report

Fall-2017

K. Siva Naga Lakshmi G01099587 **Goal**: Hands-on experience to extract, and utilize metadata in the data mining process **Scenario**: You are a data scientist for an airline A, and you analyze a customer database. You want to identify the factors that are helpful to understand why some customers are flying your airline, and why others are canceling. Your data science team wants to recommend these factors to advertising team, such as demographic-specific packages to attract more customers.

Introduction

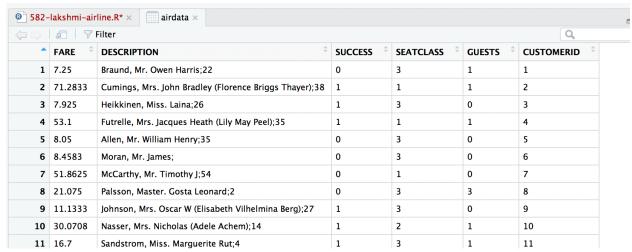
The Project has five milestones and they are described as follows:

1. Milestone - Data Acquisition and Conversion

In this step, the dataset is taken from the provided link. Using the R tool the dataset is converted from JSON to CSV File. Before converting the code, R packages need to be installed which are RCurl and RJSONIO.

Code (JSON to CSV) and deleting the extra header

Dataset



2. Milestone - Metadata Extraction and Imputation

The main objective of this milestone is to identify the metadata types. With "Description" data field, the data can be extracted and append them as additional data columns. The extracted metadata fields are as following Gender, age, first name, last name and the prefix.

The above metadata fields have been appended as separate columns using the R functions 'stringr', 'grepl' and 'sapply'. In the Gender column data has been derived using the prefixes "Mr. /Master" considered as male and female for all the remaining values.

| SUCCESS | SEATCLASS | GUESTS | CUSTOMERID * | LastName + | prefix + | FirstName | Gender | Age [‡] |
|---------|-----------|--------|--------------|-------------|----------|---------------------------------------|--------|------------------|
| 0 | 3 | 1 | 1 | Braund | Mr | Owen Harris | Male | 22.00 |
| 1 | 1 | 1 | 2 | Cumings | Mrs | John Bradley (Florence Briggs Thayer) | Female | 38.00 |
| 1 | 3 | 0 | 3 | Heikkinen | Miss | Laina | Female | 26.00 |
| 1 | 1 | 1 | 4 | Futrelle | Mrs | Jacques Heath (Lily May Peel) | Female | 35.00 |
| 0 | 3 | 0 | 5 | Allen | Mr | William Henry | Male | 35.00 |
| 0 | 3 | 0 | 6 | Moran | Mr | James | Male | NA |
| 0 | 1 | 0 | 7 | McCarthy | Mr | Timothy J | Male | 54.00 |
| 0 | 3 | 3 | 8 | Palsson | Master | Gosta Leonard | Male | 2.00 |
| 1 | 3 | 0 | 9 | Johnson | Mrs | Oscar W (Elisabeth Vilhelmina Berg) | Female | 27.00 |
| 1 | 2 | 1 | 10 | Nasser | Mrs | Nicholas (Adele Achem) | Female | 14.00 |
| 1 | 3 | 1 | 11 | Sandstrom | Miss | Marguerite Rut | Female | 4.00 |
| 1 | 1 | 0 | 12 | Bonnell | Miss | Elizabeth | Female | 58.00 |
| 0 | 3 | 0 | 13 | Saundercock | Mr | William Henry | Male | 20.00 |
| 0 | 3 | 1 | 14 | Andersson | Mr | Anders Johan | Male | 39.00 |
| 0 | 3 | 0 | 15 | Vestrom | Miss | Hulda Amanda Adolfina | Female | 14.00 |
| 1 | 2 | 0 | 16 | Hewlett | Mrs | (Mary D Kingcome) | Female | 55.00 |
| 0 | 3 | 4 | 17 | Rice | Master | Eugene | Male | 2.00 |
| 1 | า | ^ | 10 | Williams | Mr | Charles Eugene | Mala | A/A |

There were missing values in the age column, Using the mean imputation method NA values has been replaced with mean value of "29.56", using the floor function in R it has been rounded off to the nearest value.

After imputing the missing values in the age column, one more column is added here where the age category is defined. Now the data is ready to use for further analysis.

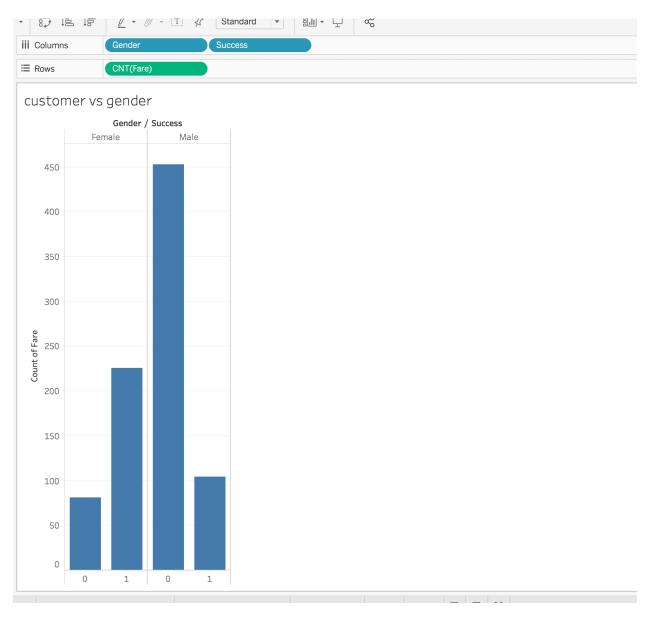
| Α | В | С | D | Е | F | G | Н | I | J | K | L |
|----|-----------|---------|-----------|--------|------------|--------------|--------|-----------------------------------|--------|-----|-------------|
| | FARE | SUCCESS | SEATCLASS | GUESTS | CUSTOMERII | LastName | prefix | FirstName | Gender | Age | agecategory |
| : | 7.25 | 0 | 3 | 1 | . 1 | Braund | Mr | Owen Harris | Male | 22 | Young |
| | 71.2833 | 1 | 1 | 1 | . 2 | Cumings | Mrs | John Bradley (Florence Briggs Tha | Female | 38 | Middle Aged |
| : | 7.925 | 1 | 3 | 0 | 3 | Heikkinen | Miss | Laina | Female | 26 | Young |
| | 4 53.1 | 1 | 1 | 1 | . 4 | Futrelle | Mrs | Jacques Heath (Lily May Peel) | Female | 35 | Middle Aged |
| | 8.05 | 0 | 3 | 0 | 5 | Allen | Mr | William Henry | Male | 35 | Middle Aged |
| (| 8.4583 | 0 | 3 | 0 | 6 | Moran | Mr | James | Male | 29 | Young |
| | 7 51.8625 | 0 | 1 | 0 | 7 | McCarthy | Mr | Timothy J | Male | 54 | Middle Aged |
| 1 | 3 21.075 | 0 | 3 | 3 | 8 | Palsson | Master | Gosta Leonard | Male | 2 | Infant |
| 9 | 11.1333 | 1 | 3 | 0 | 9 | Johnson | Mrs | Oscar W (Elisabeth Vilhelmina Ber | Female | 27 | Young |
| 10 | 30.0708 | 1 | 2 | 1 | . 10 | Nasser | Mrs | Nicholas (Adele Achem) | Female | 14 | Teenage |
| 1: | 1 16.7 | 1 | 3 | 1 | 11 | Sandstrom | Miss | Marguerite Rut | Female | 4 | minor |
| 12 | 26.55 | 1 | 1 | 0 | 12 | Bonnell | Miss | Elizabeth | Female | 58 | Middle Aged |
| 13 | 8.05 | 0 | 3 | 0 | 13 | Saundercock | Mr | William Henry | Male | 20 | Young |
| 14 | 4 31.275 | 0 | 3 | 1 | 14 | Andersson | Mr | Anders Johan | Male | 39 | Middle Aged |
| 1! | 7.8542 | 0 | 3 | 0 | 15 | Vestrom | Miss | Hulda Amanda Adolfina | Female | 14 | Teenage |
| 10 | 5 16 | 1 | 2 | 0 | 16 | Hewlett | Mrs | (Mary D Kingcome) | Female | 55 | Middle Aged |
| 17 | 7 29.125 | 0 | 3 | 4 | 17 | Rice | Master | Eugene | Male | 2 | Infant |
| 18 | 3 13 | 1 | 2 | 0 | 18 | Williams | Mr | Charles Eugene | Male | 29 | Young |
| 19 | 18 | 0 | 3 | 1 | 19 | Vander Plank | Mrs | Julius (Emelia Maria Vandemoorte | Female | 31 | Middle Aged |
| 20 | 7.225 | 1 | 3 | 0 | 20 | Masselmani | Mrs | Fatima | Female | 29 | Young |
| 2: | 1 26 | 0 | 2 | 0 | 21 | Fynney | Mr | Joseph J | Male | 35 | Middle Aged |

Above screenshot has shown that data is ready to be analyzed for further steps.

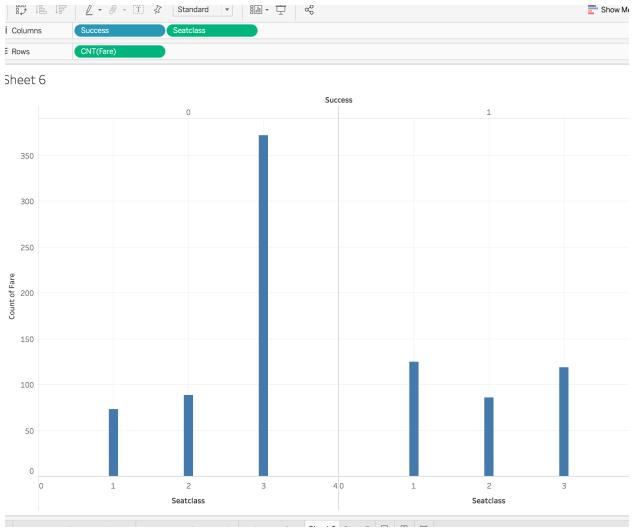
3. Metadata Exploration

In this milestone visualization of pattern between different metadata fields using the "Tableau" software and the results are interpreted.

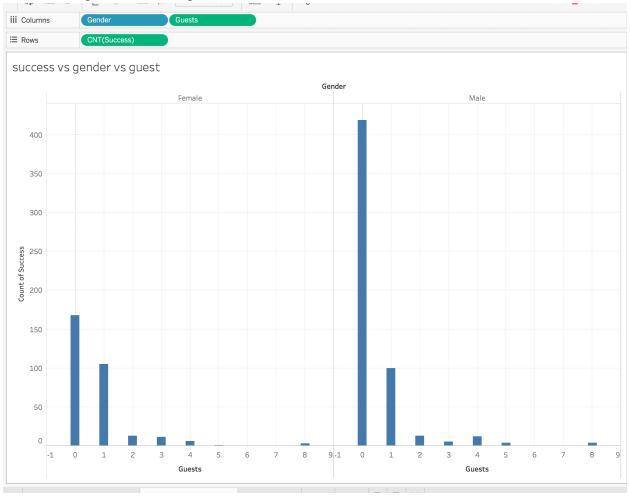
a) <u>Gender vs Fare with Success</u>- From the graph we can observe that males have the highest number of cancellations.



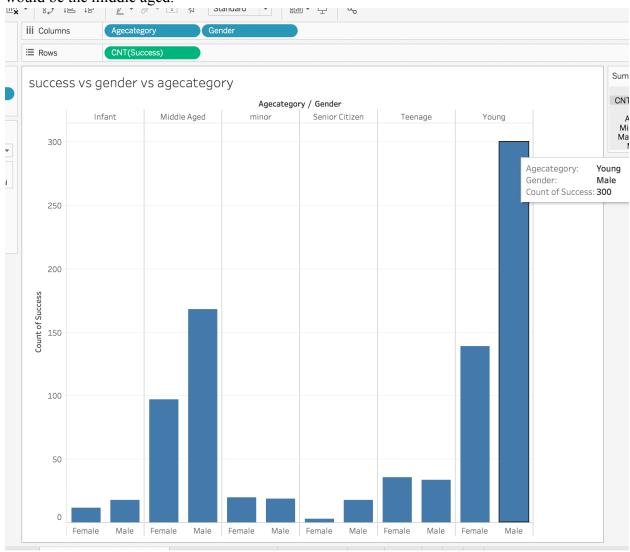
b) Success vs Fare with seat class- from the graph we can observe that there are lot of cancellations in the 3^{rd} seat class.



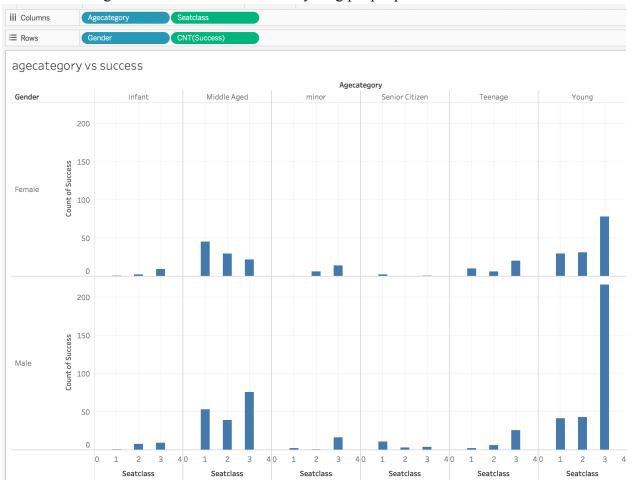
c) <u>Guest vs Success with gender</u>- From the graph we can observe that, males are travelling more with 0 guest when compare to female.



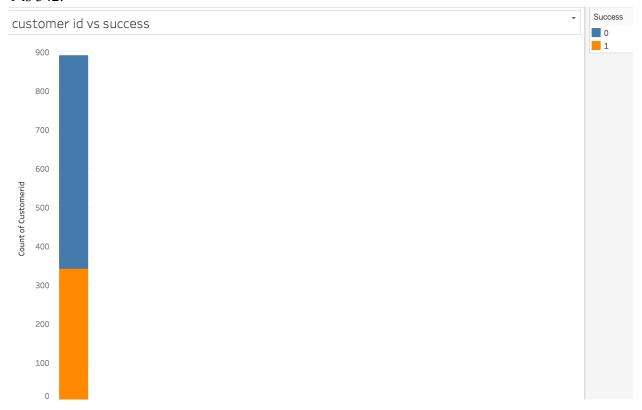
d) <u>Success vs Gender with Age category</u>- From the graph we can observe that, the count of success is high in male and in the young category compared to female. Second category would be the middle aged.



e) <u>Success vs Seat class with Age Category and Gender</u>- From the graph we can observe that male count is high in the seat-class 3where the young people preferred more.

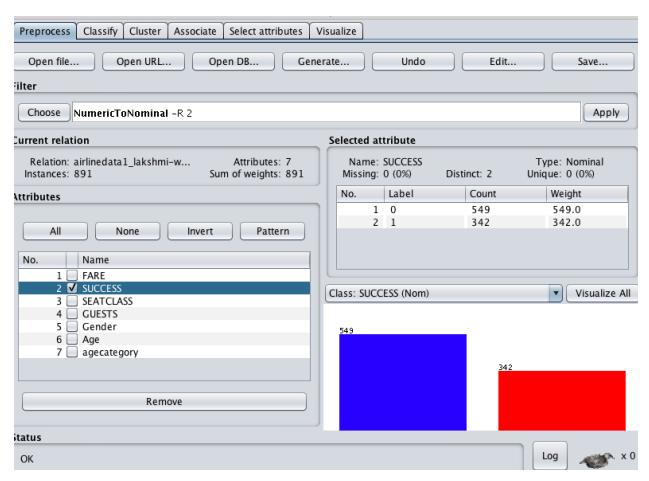


f) <u>Customer ID vs Success</u>- from the graph we can observe that where 0 indicates persons not willing to fly, and 1 indicates people willing to fly. Count of customers with 0 is 549 and 1 is 342.



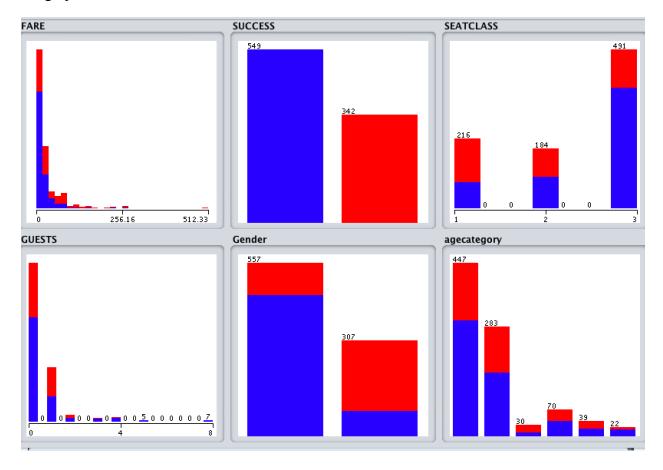
4. Attribute Preparation and Engineering

In this milestone, the extracted metadata file is converted to ARFF using the Weka tool. Loading the csv file in the ARFF viewer and saved it by using the ARFF extension.

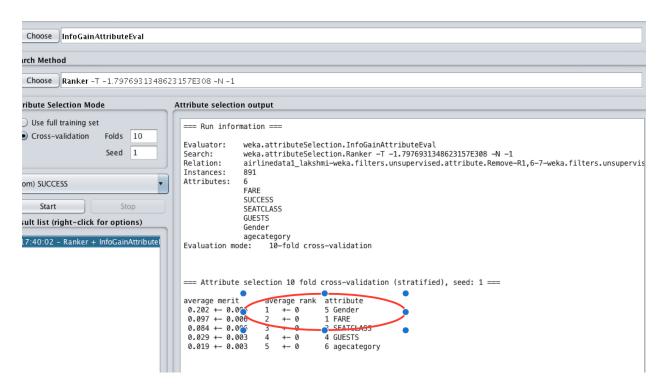


The above screenshot is the preprocessing step, in this data file is imported to Weka tool by clicking open file and giving the corresponding path where data file is located. After data is loaded, attribute type is changed to numeric to nominal using filter option. The result matches with the tableau one's in the Weka tool.

Similarly plotting other graphs can be done by clicking on the 'visualize all', where it displays all the graphs.



Attribute selection, here the objective of this step is to find the top attributes using 10-fold cross validation model. Where I opted the ranker method to find out the top attributes, which in-turn automatically takes the info gain attribute evaluation. Using this method, the top two attributes will be the Gender and the Fare. Using these two attributes, all other values or fields can be classified and they act as roots. Results are shown below in the screen shots

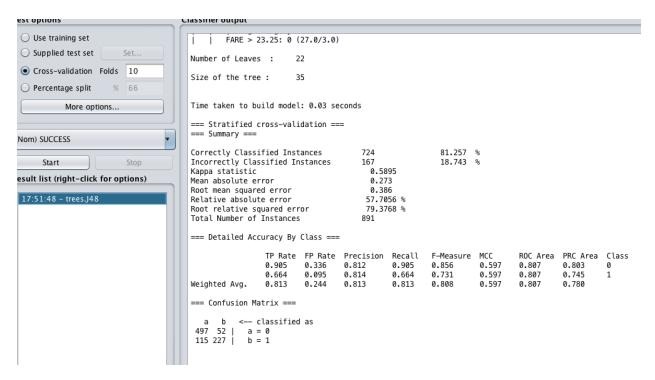


In the above screen shot, where gender, fare is represented in the eclipse and ranked as top two attributes.

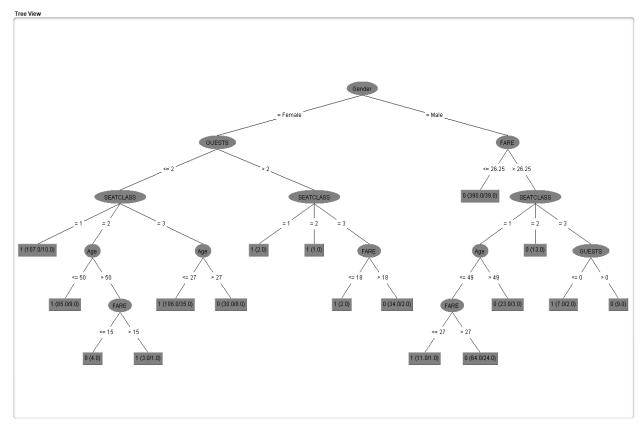
5. Prediction Modeling and Visualization

Objective of this milestone is to design a classification model using the J48 (decision tree) and Random forest algorithms. And also, to generate the ROC curve using 10-fold cross validation method.

For classification model, the data is divided into the training and the testing data by using the default classification feature of Weka. Data is divided into training as 80% and 20% test data. After running the **J48 model** on the training data the results are obtained as below in the screen shot. From the picture, we can say that accuracy of this model is **81.257%** with **724** correctly classified instances.



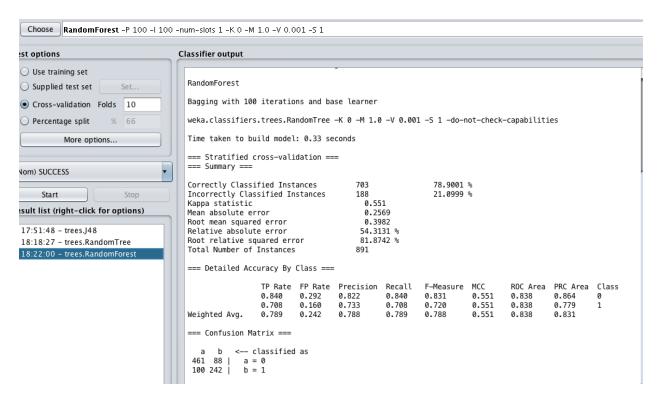
Above screen shot depicts classification has done using J48 algorithm



Decision Tree

After the Decision tree, I used the **Random Forest** method to classify the model and to check the significance of the method. The accuracy of this model gives as **78.90%** with **703** as correctly classified instances.

Random Forest Screenshot

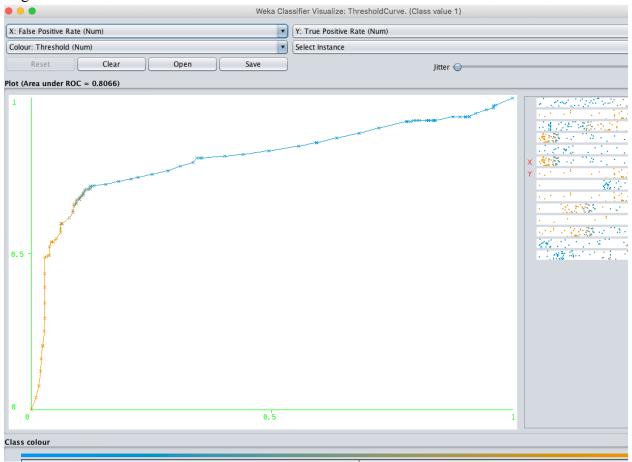


ROC Curves

It gives the graphical illustration of the performance of a binary classifier system as its discrimination threshold is varied. Plotting the ROC curve is done by decision tree and random forest algorithms using 10-fold cross validation.

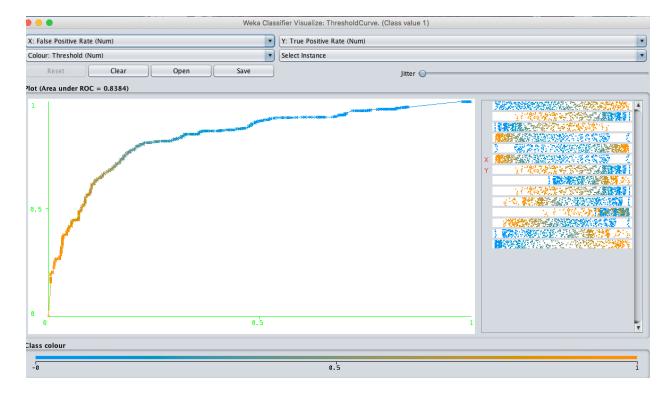
a) ROC-Decision Tree(J48)

For the threshold value 1 as shown below. The area under the curve is **0.8066**, which means it is a good model and if the curve is leaning towards more to the y-axis then it is considered as good model and more accurate.



b) ROC-Random Forest

For threshold value 1 is shown below, from the graph we can observe that area under curve is **0.834.**



Area under curve is more compared to decision tree. Accuracy is more for decision tree model and also the curve for random forest is smooth than decision tree model, but decision model is more better than random forest.

If we increase the true positive values, the curve will be more accurate and it will be towards the y-axis. When compared both the models, decision tree has truer positive value. The model will be robust.

Insights

Few insights can be drawn from the milestones

- Top Two attributes are the Gender and Fare where we can predict the values using these two factors.
- Males has higher success rate compared to females.
- The young age and middle age customers are more likely to travel.

If company works focuses on the above factors, it will be able to get more sales and the success rate will increase.