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TEXAS
The University of Texas at Austin

Trips and Travel.com customer analysis

MIS 382N: Group Project

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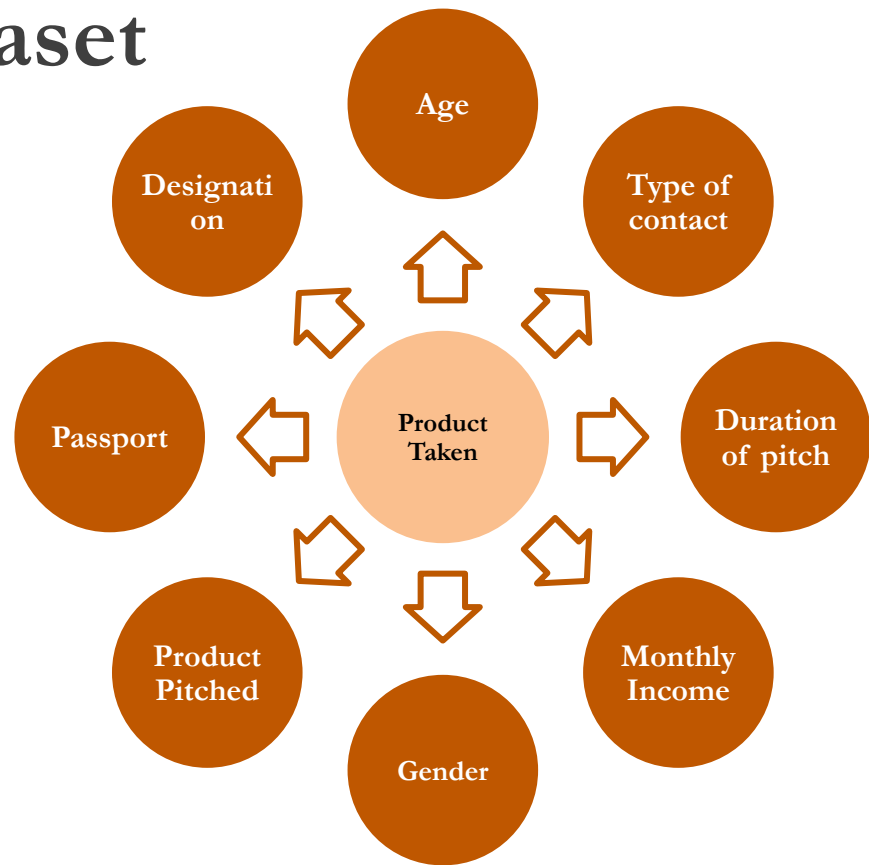
Problem Statement

Trips & Travel.Com wants to expand its customer base by adding a new product to its product offerings.

We aim to identify the potential customer base to market this product to, in order to increase the ROI on the marketing expenditure.

Understanding the Dataset

- What are the variables that affecting the target - “Product Taken”
- The dataset consists of 4888 rows with 18 features
 - Categorical
 - Quantitative



Approach

Exploratory Data Analysis

- Scatter Plots

Feature Selection

- 5 variables out of 21 variables selected

Data Preparation

- Filtering the dataset

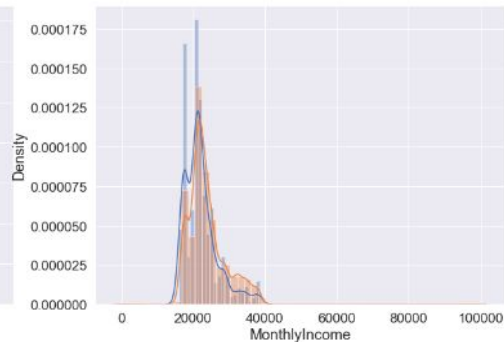
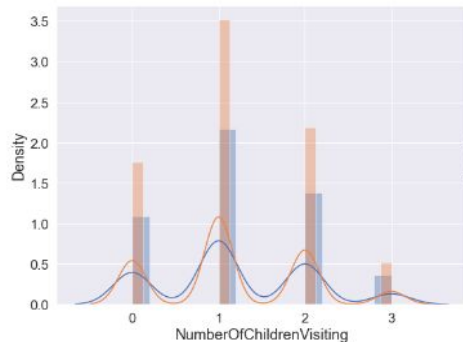
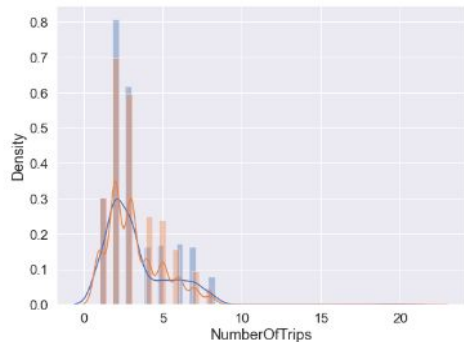
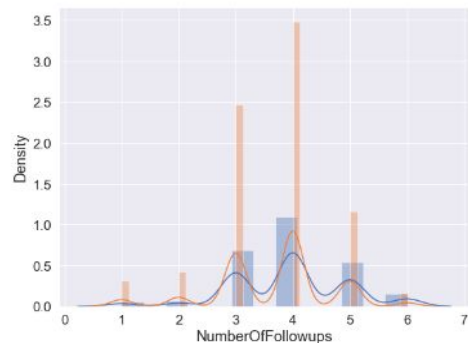
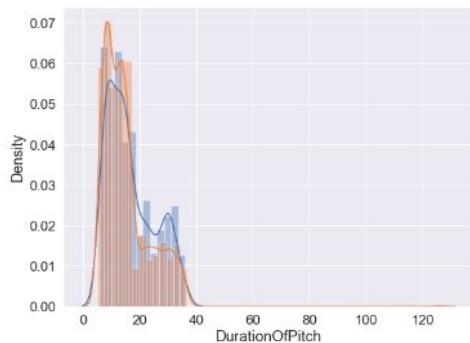
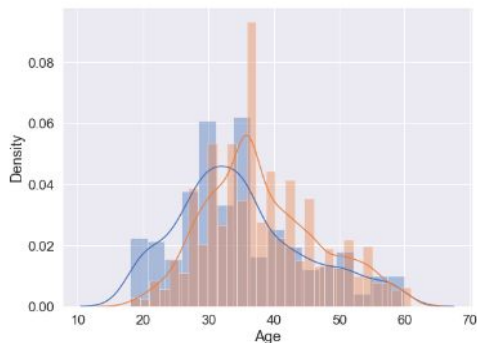
Modeling

- Logistic Regression
- Customer Scoring
- Random Forests

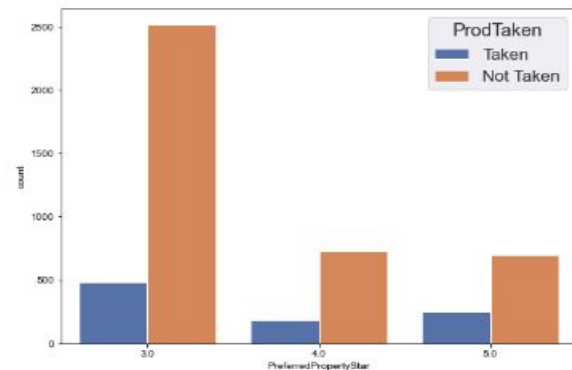
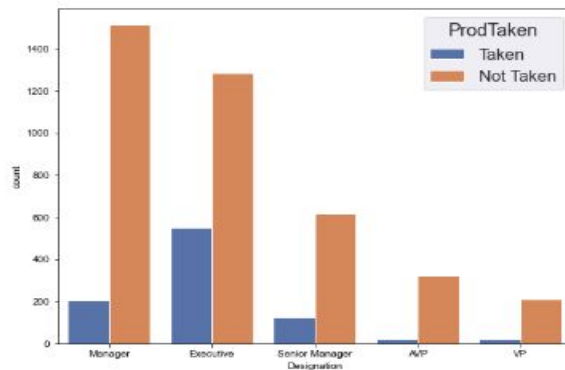
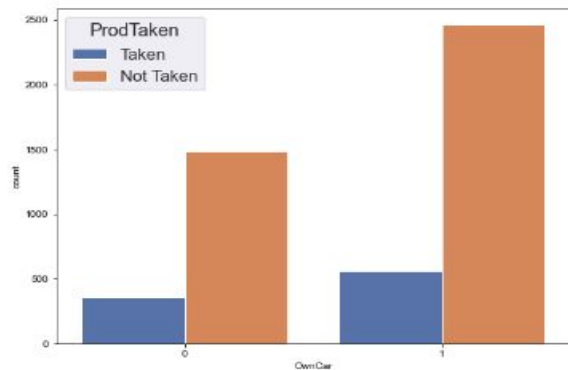
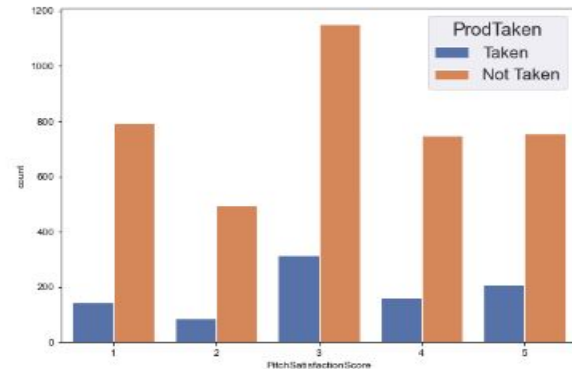
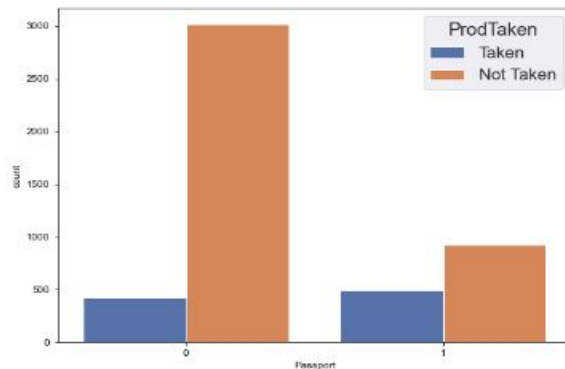
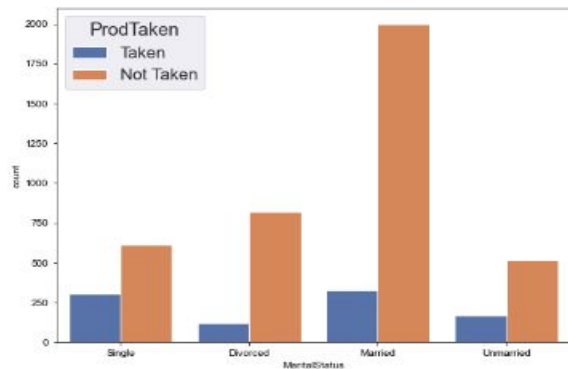
Results

Exploratory Data Analysis - Continuous

■ Taken
■ Not Taken



Exploratory Data Analysis - Categorical



Exploratory Data
Analysis

Feature Selection

Data Preparation

Modeling

Model
Improvement

Feature Importance

- Feature ranking based on **Chi2 test**
- **MonthlyIncome** has the most impact on the target

MonthlyIncome	91649.5
Passport	235.8253
Age	222.9664
DurationOfPitch	128.6842
Marital Status	127.5178
Designation	29.22755
CityTier	18.72158
NumberOfFollowups	16.33239
Occupation	11.98029
TypeOfContact	8.497068
PreferredPropertyStar	8.452716
PitchSatisfcationScore	7.822561

Exploratory Data
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- Replaced missing values with median
- Categorical Columns and Target encoded as 0 & 1
- Dropped irrelevant columns

Exploratory Data
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Improvement

Models

- Logistic Regression
- Customer Scoring
- Random Forests

Variable Selection

- Tried 4 models - top 2 features, top 5 features, top 10 features, and all features.
- Model with top 5 features produced best results

MonthlyIncome	91649.5
Passport	235.8253
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PitchSatisfactionScore	7.822561

XLStat Logistic Regression Coefficients

Source	Value
Intercept	0.490
Duration of Pitch	0.026
Marital Status	-0.335
Monthly Income	-0.402
Passport	1.448
Age	-0.033

Average Response Rates and Lifts

Avg Response Rate	Avg Lift
18.8216%	87.2344%

- Average Response Rate of 18.8216% tells us that there is a 18.8216% chance that a customer responded positively
- Average lift of 87.23% tells us that there is an increase of 87.23% in sales

Average Marginal Effects

Avg ME Dur	Avg ME MS	Avg ME I	Avg ME P	Avg ME A
0.3438%	-4.4053%	-5.2913%	19.0438%	-0.4376%

- Average Marginal Effect for each variable tells us the unit change in that variable for a one unit change in the response

Profit Calculation

Profits
2800
-200
2800
-200
-200
-200

ECLV	3000
Solicitation	200
Cutoff	0.066667

- Calculated profit for each customer, based on Expected Customer Lifetime Value and Solicitation Cost to make a cutoff.
- ECLV and Solicitation Cost were taken as the average for the travel industry according to:

<https://blog.datumize.com/what-is-the-value-of-a-new-customer-in-the-travel-industry>

Profits with Targeting	\$1,653,000.00
Profits without Targeting	\$1,182,400.00

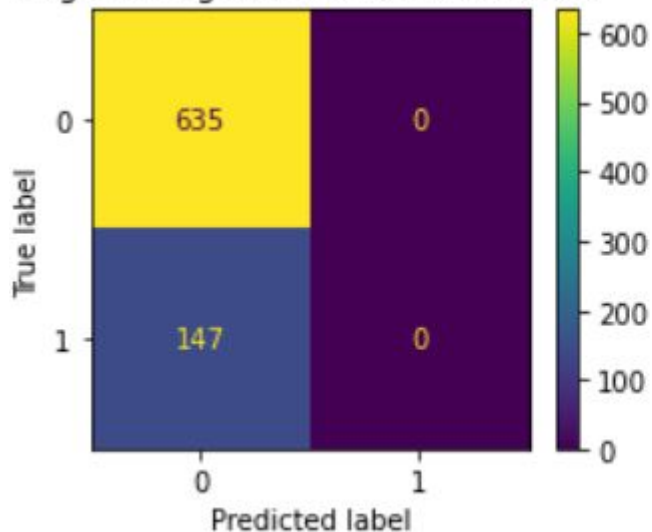
%Targeted and Profits with and without targeting

%Targeted	77.8435%
Profits with Targeting	\$1,653,000.00
Profits without Targeting	\$1,182,400.00

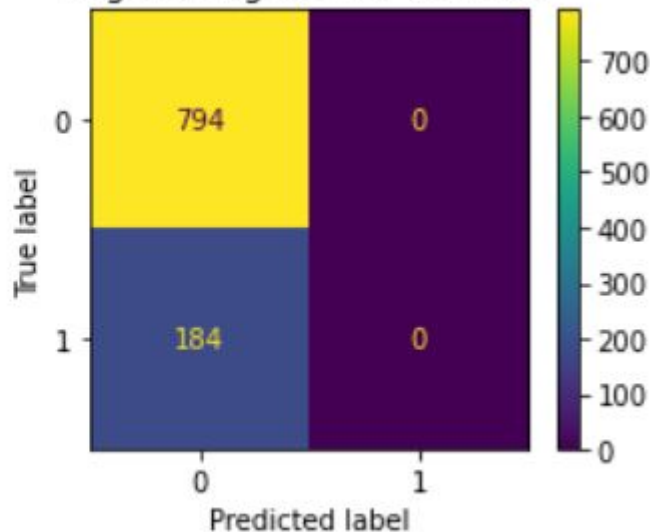
- Percent of customers targeted - 77.84%
- Difference in profit of approximately \$471,000.

Logistic Regression

Logistic Regression: Validation data



Logistic Regression: Test data



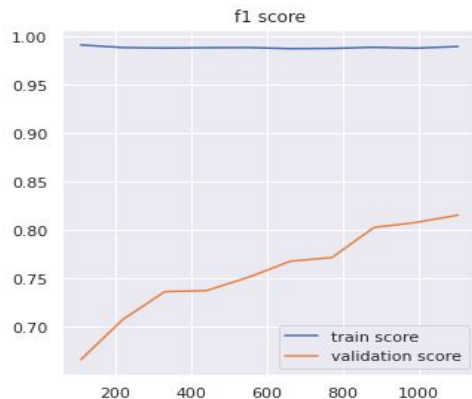
Accuracy Score:
0.8289641943734015

Random Forest Classifier

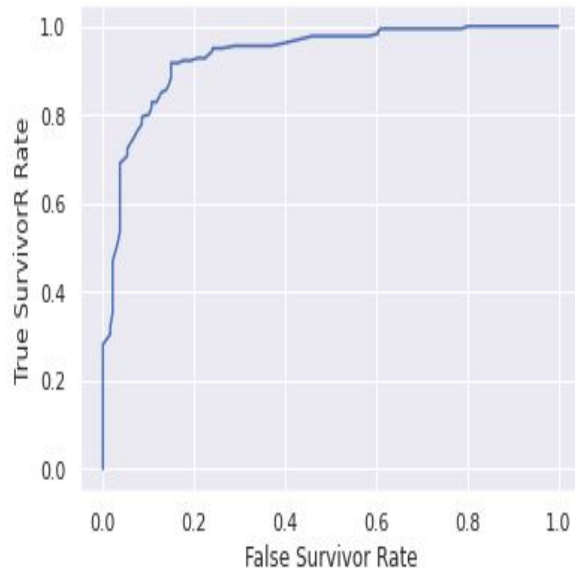
Accuracy = 0.8559782608695652

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	precision	recall	f1-score	support
0	0.85	0.87	0.86	187
1	0.86	0.85	0.85	181
accuracy			0.86	368
macro avg	0.86	0.86	0.86	368
weighted avg	0.86	0.86	0.86	368



Random Forests ROC Curve



Exploratory Data
Analysis

Feature Selection

Data Preparation

Modelling

Model Selection

Results

- By Customer Scoring we are able predict that **18%** of the response rate is positive, which we think is pretty good, considering most people's negative response to a sales pitch.
- From the Logistic Regression model we can predict future customers from the featured selection with **82%** accuracy.
- We were further able to improve this accuracy to **86%** with a Random Forest Classifier.
- By targeting specific customers, we can improve profit by almost half a million dollars.