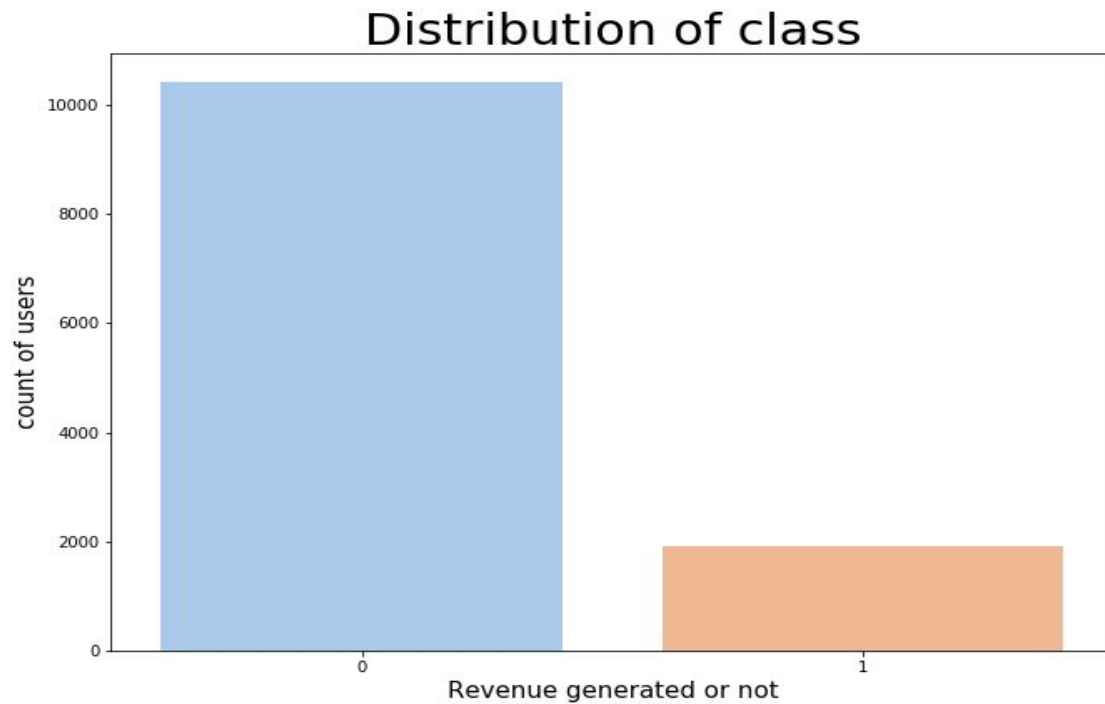


# **CS6220 - Data Mining Techniques**

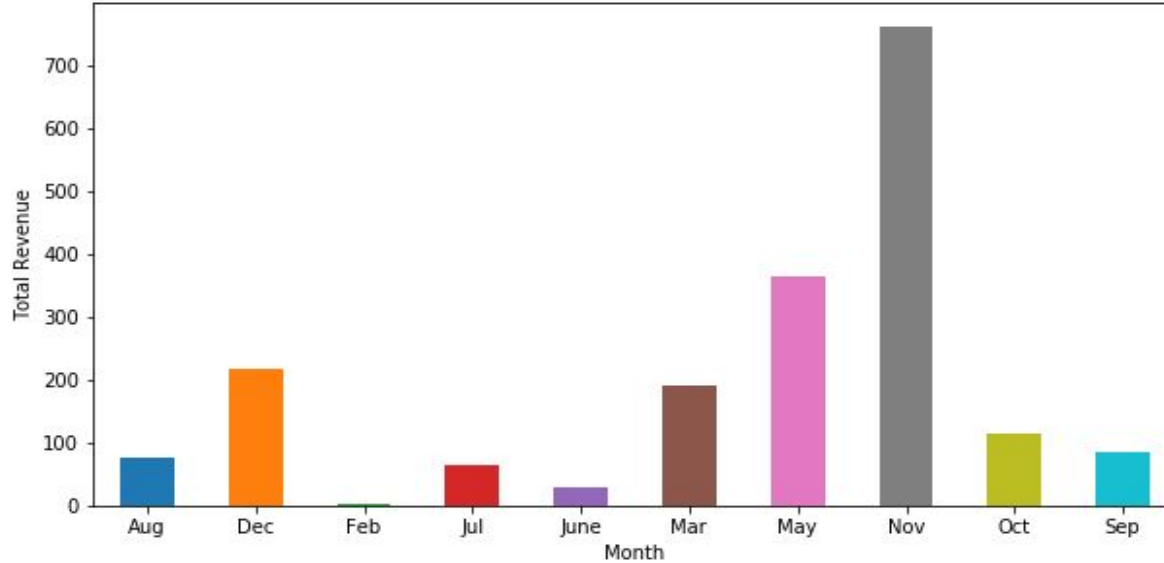
## **Online Shoppers Purchasing Intention - RESULTS**

*Rajath Kashyap  
Mukund Wagh  
Bishwarup Neogy*



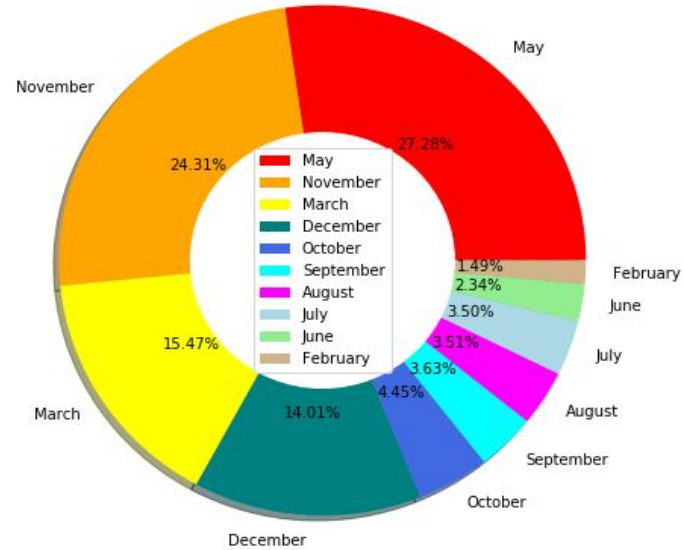
- Unbalanced dataset
- Get valuable insight from the available data.

## Revenue Per Month



- More the number of visitors more is the sale.
- We have products that fulfill the needs of the customers.

## Users per month

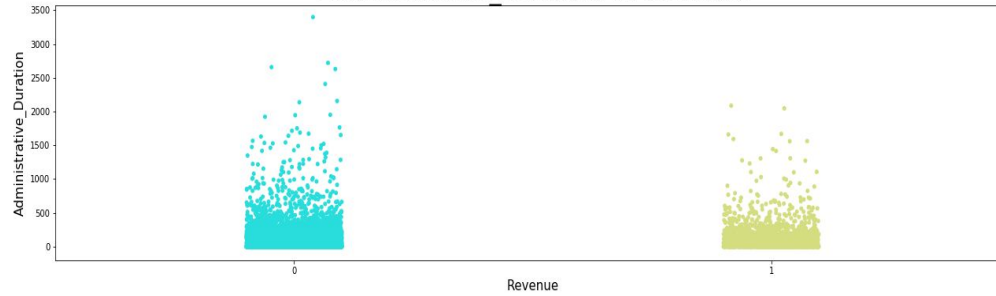


- More the number of visitors more is the sale.
- We have products that fulfill the needs of the customers.

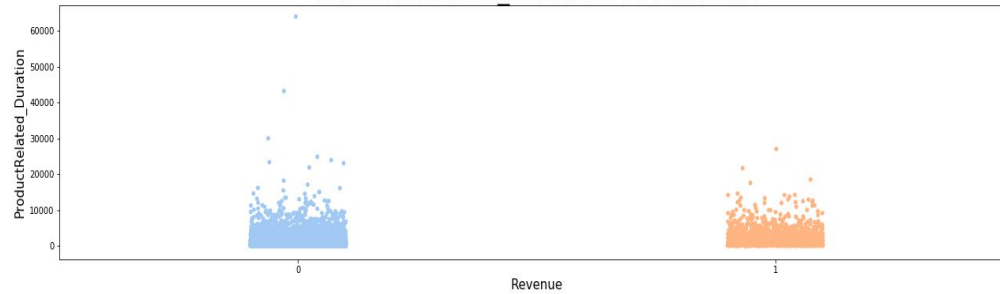
Informational\_Duration vs Revenue



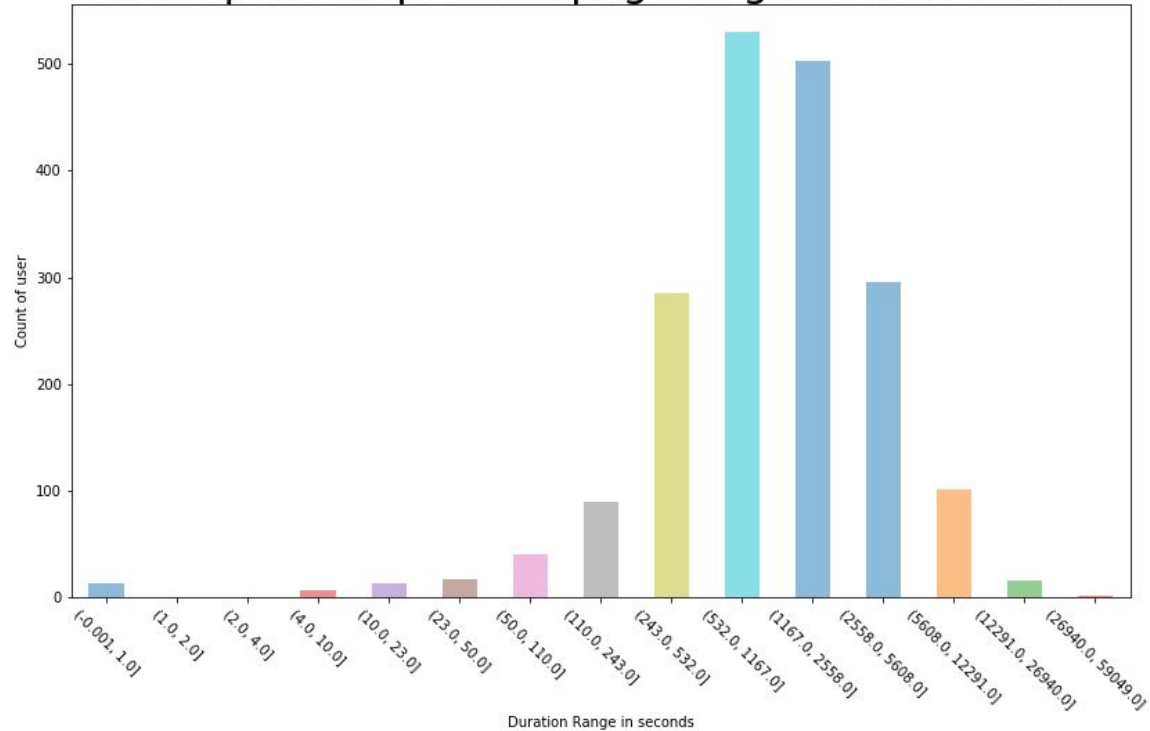
Administrative\_Duration vs Revenue



ProductRelated\_Duration vs Revenue

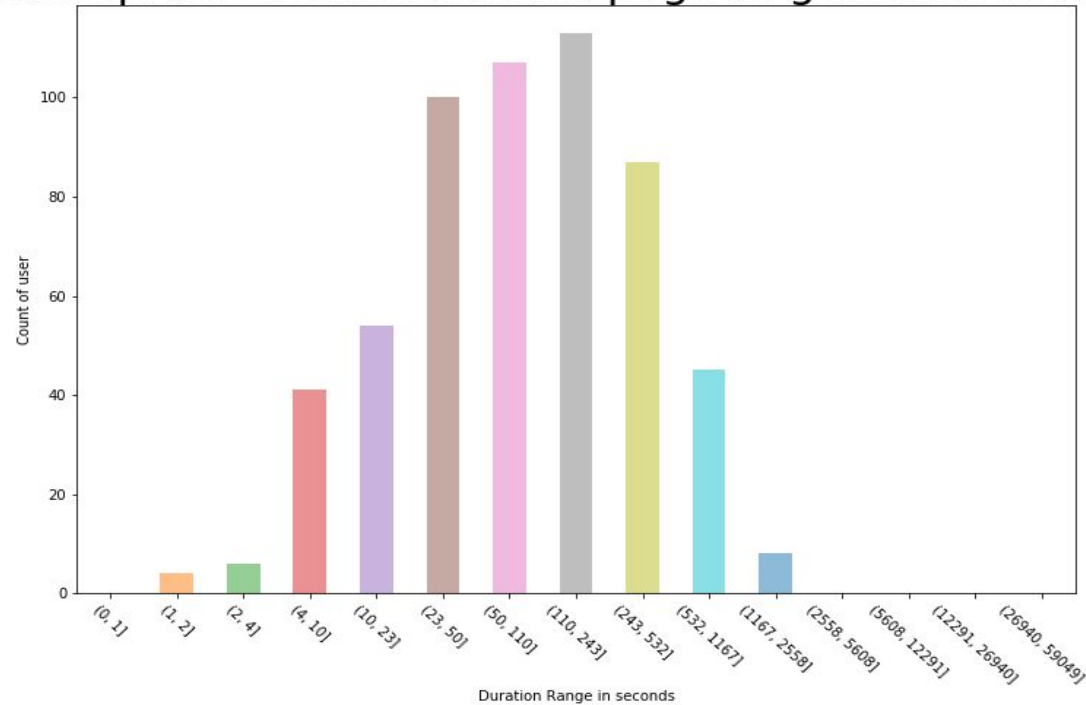


## Time spent on product page to generate revenue



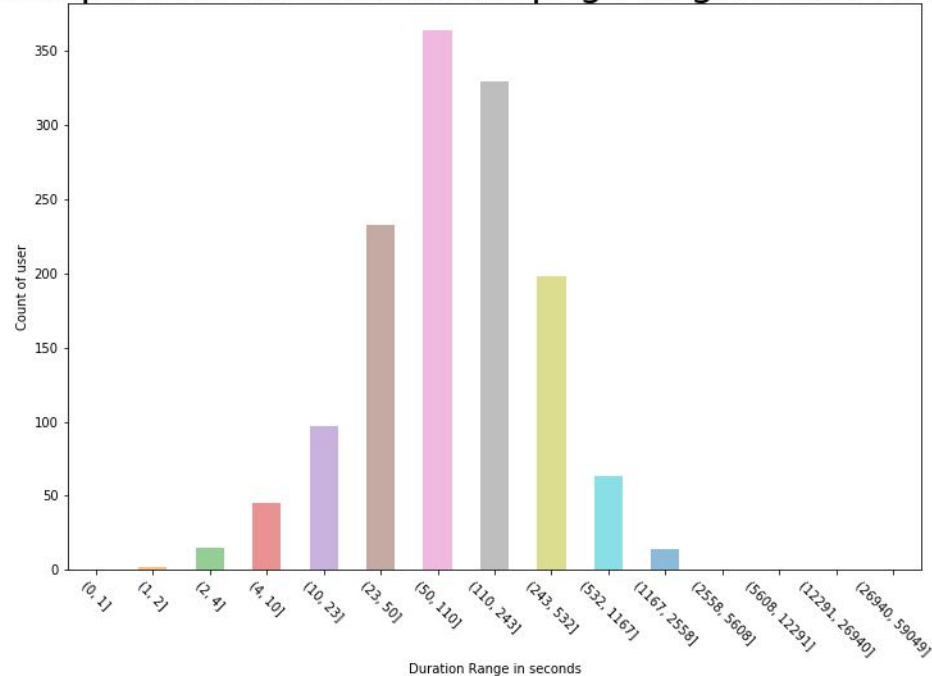
- We need to improve on the overall search engine of the website and cater them with the right product when they try to find one.

## Time spent on informational page to generate revenue



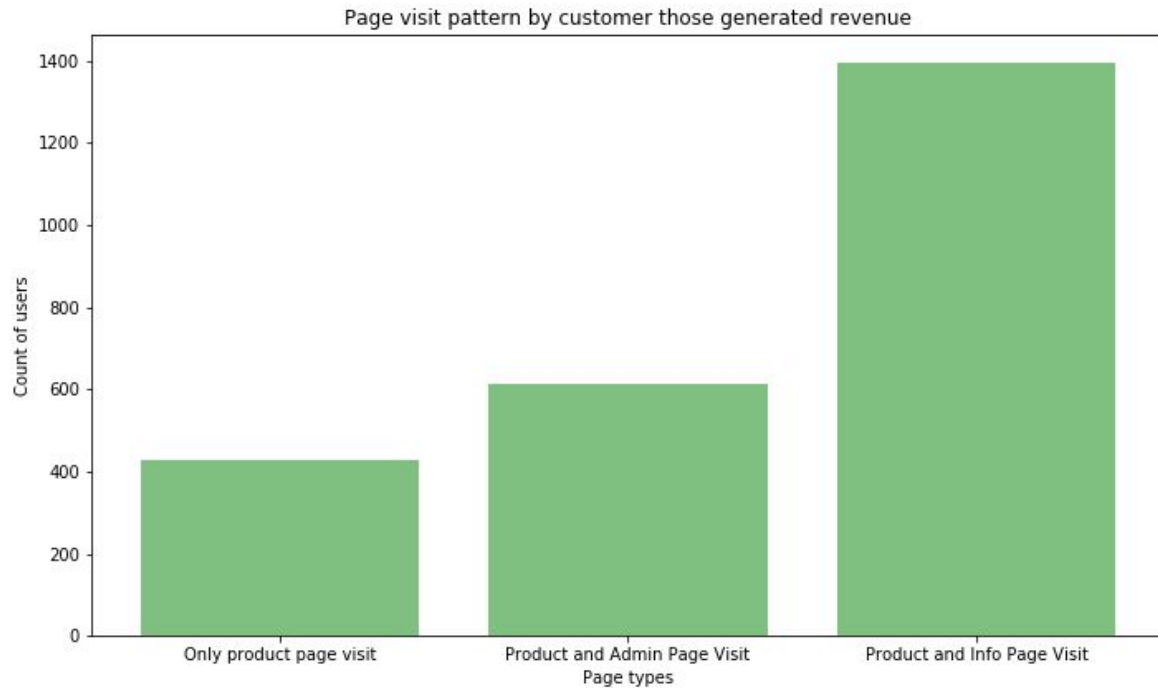
- Many users has to visit the info page to be sure of the product they are going to buy.

## Time spent on administrative page to generate revenue



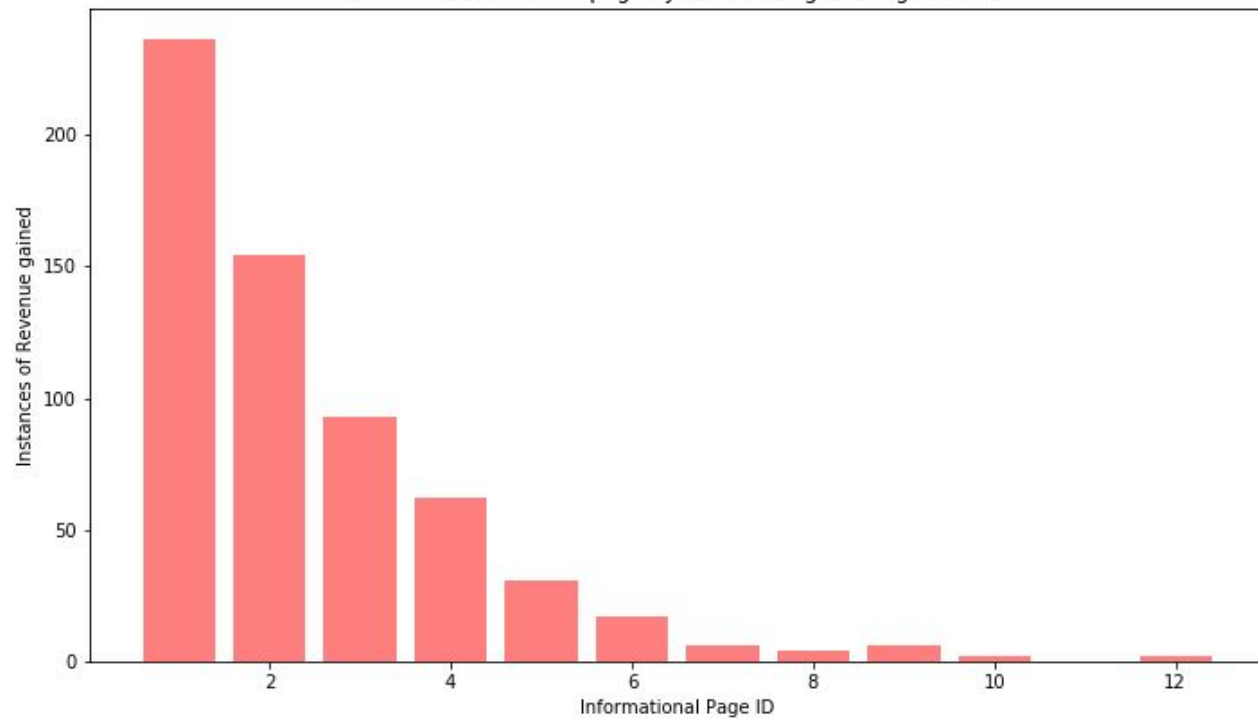
- We have around 2000 unique users who have given us the revenue, and we can see that more than 70% of the users have to visit the administrative page in order to buy the product, also around 50% customers have to spend more than a minute on the administrative pages.

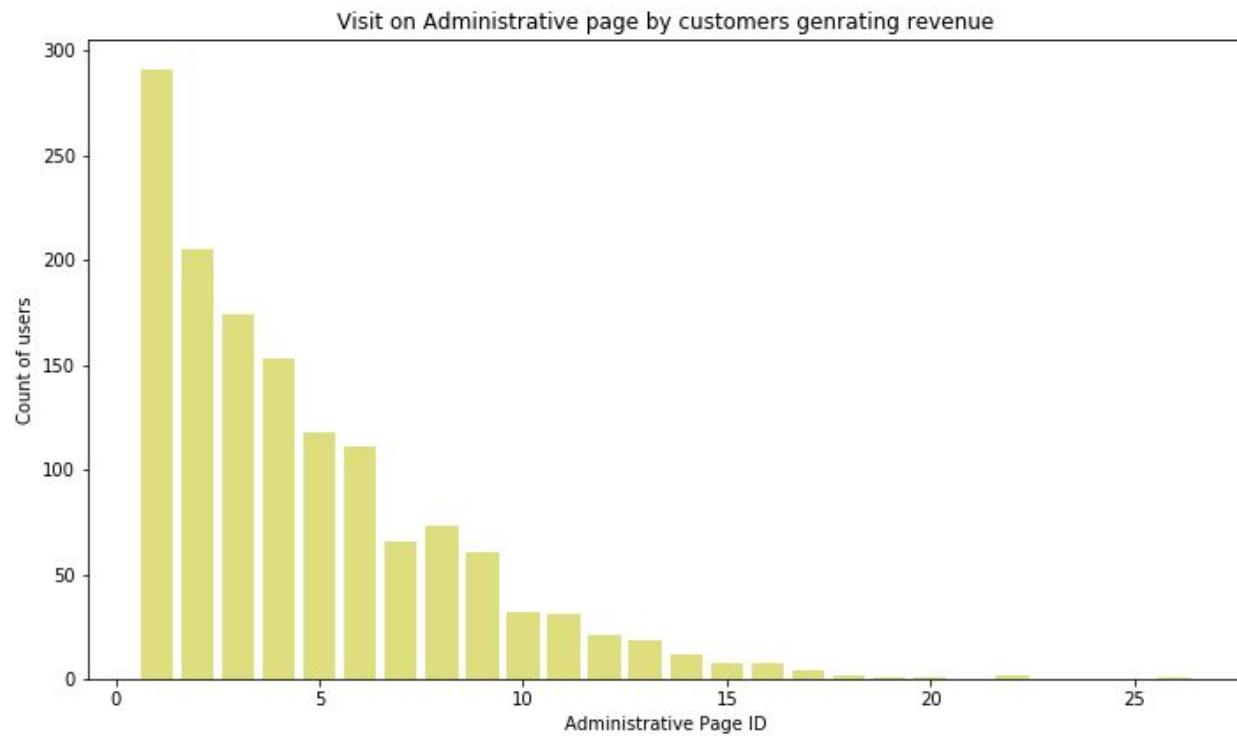




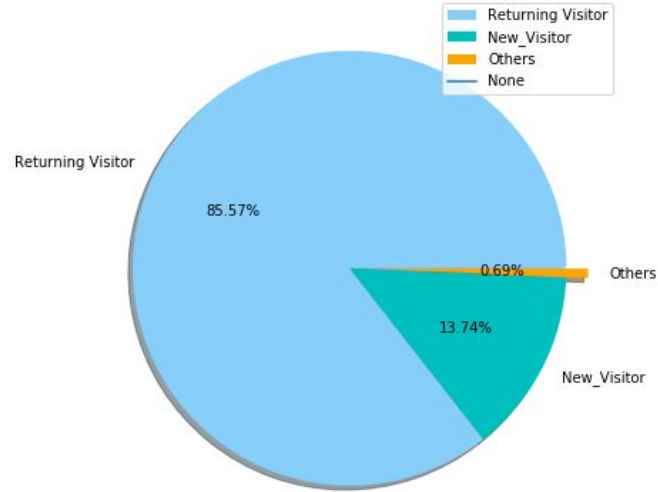
- We have data of users visiting different pages on website.
- Prioritizing the task to retain the potential customers.

Visit on Informational page by customers genrating revenue

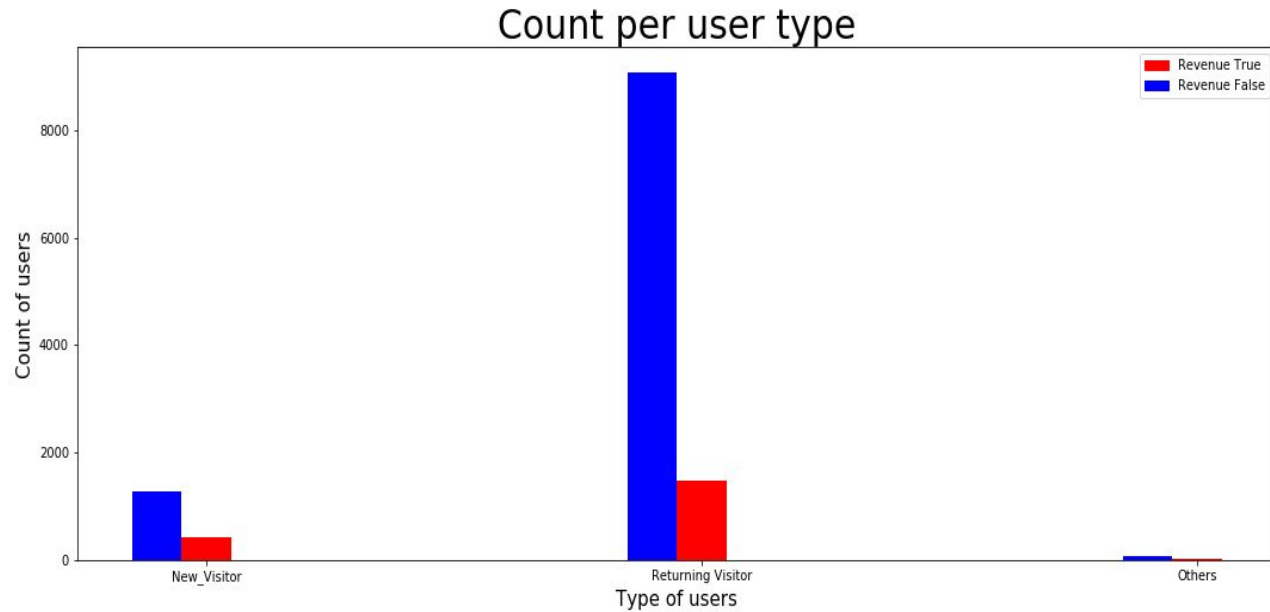




## Different Visitor Types



- We have 3 categories of users, the new users, returning users and others.

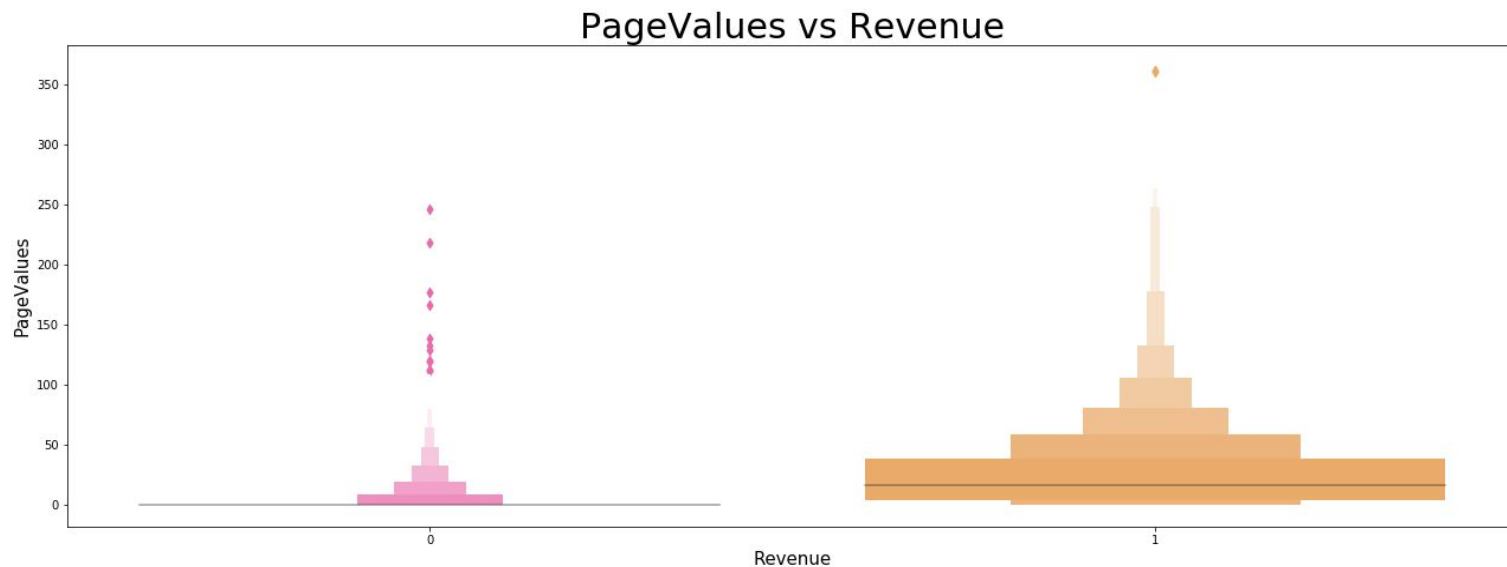


- Distribution of the user in each type of user.
- Returning users give us the most of the revenue.



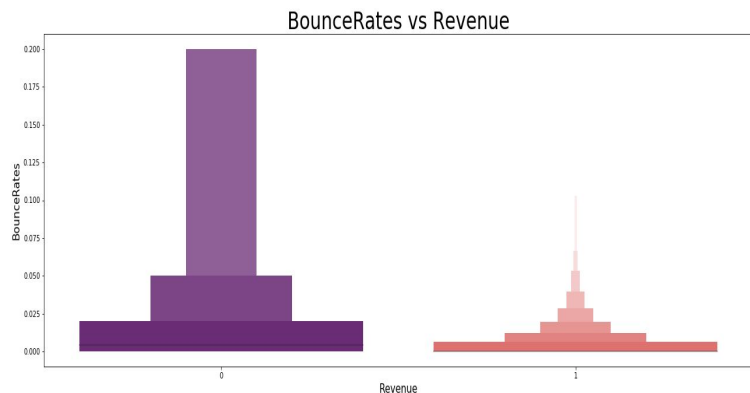
- Most of the purchases are on the weekdays.
- We should come up with schemes and offers that will also attract customers on the weekends.

# Bivariate Analysis: Page Value vs Revenue

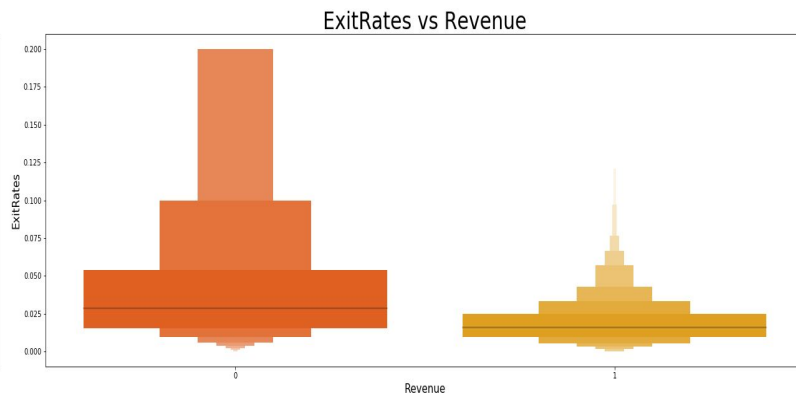


# Bivariate Analysis: Bounce Rate and Exit Rate vs Revenue

**Bounce Rate** : Avg time between a user opening a page on the site and exiting without triggering any other requests.

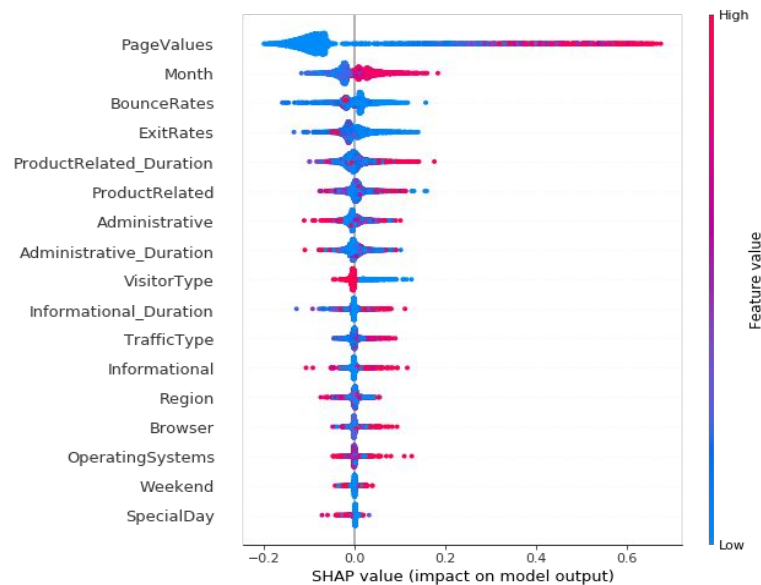
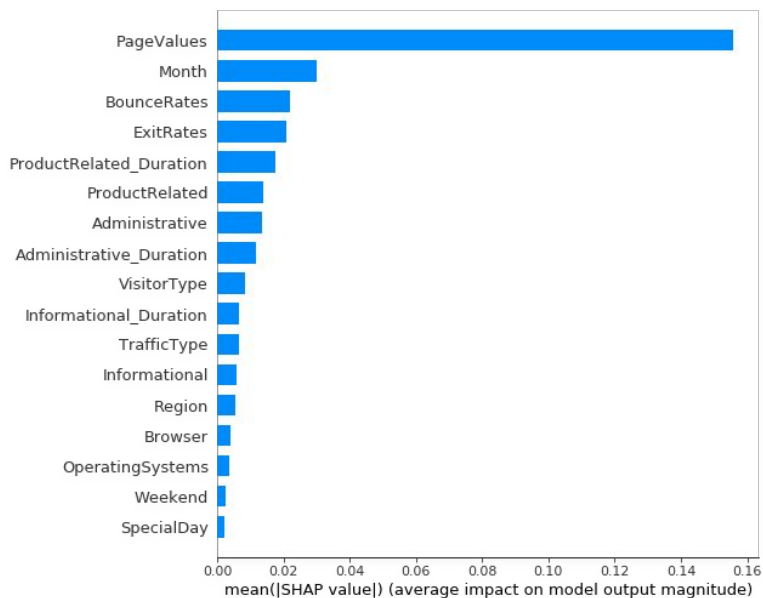


**Exit Rate** : Exit Rate is the percentage of users who exit the page and close out the session.





# SHapley Additive exPlanation (SHAP) Analysis

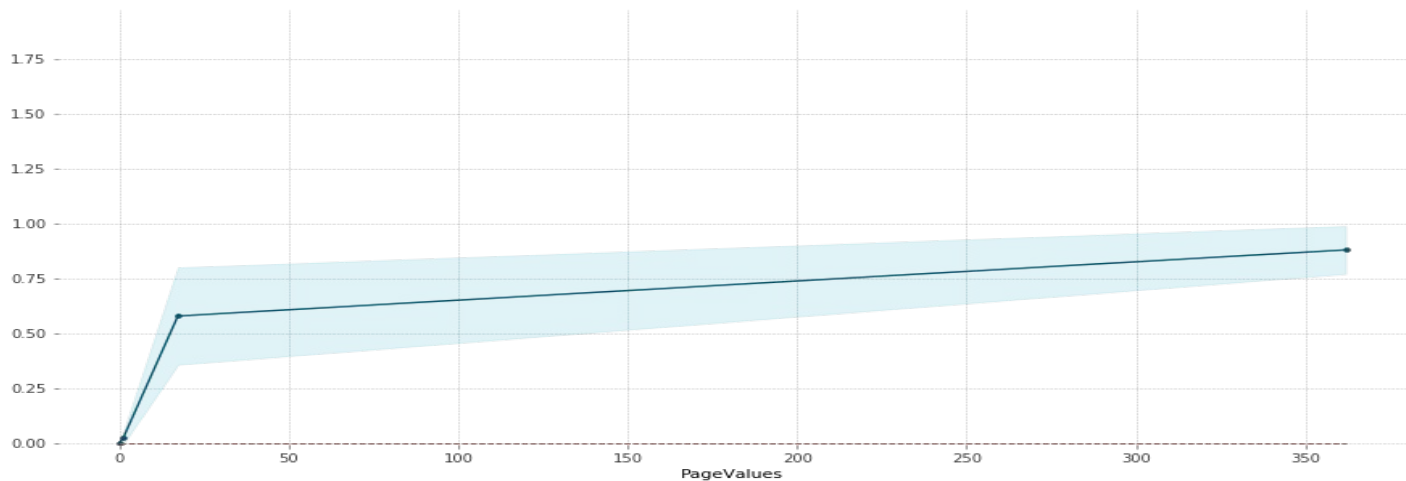


# Partial Dependence Plot

The partial dependence plot (PDP or PD plot) shows the marginal effect one or two features have on the predicted outcome of a machine learning model. The plot can show whether the relationship between the target and a feature is linear, monotonic or more complex

**PDP for feature "PageValues"**

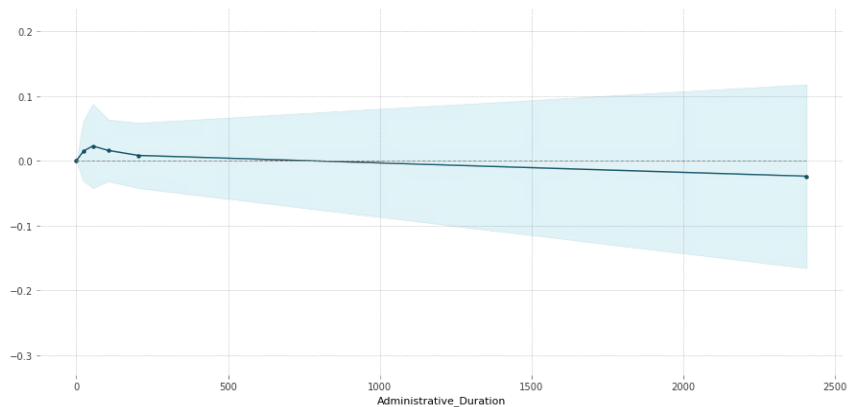
Number of unique grid points: 4



# Partial Dependence Plot

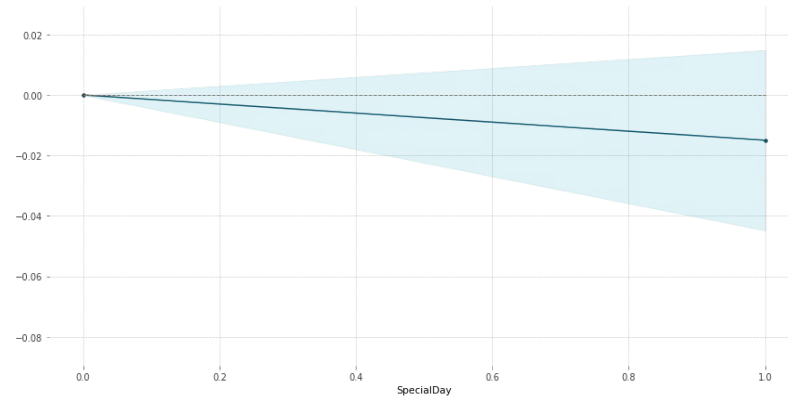
PDP for feature "Administrative\_Duration"

Number of unique grid points: 6



PDP for feature "SpecialDay"

Number of unique grid points: 2

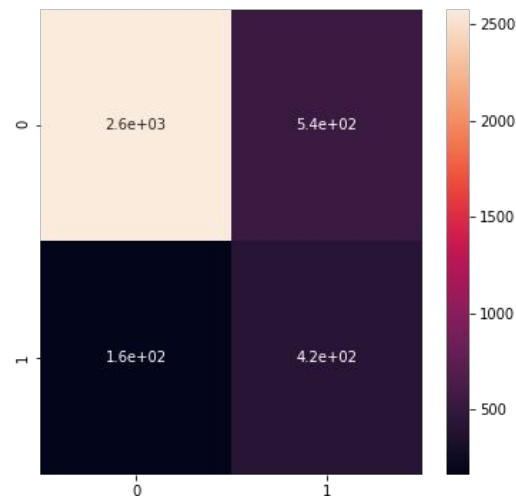


# Naive Bayes

Naive Bayes Classifier is probabilistic classifier which uses Bayes' theorem with strong (naive) independence assumptions between the features

## Classification Report :

	Precision	Recall	f1-score	support
<b>0</b>	0.94	0.83	0.88	3114
<b>1</b>	0.44	0.72	0.55	585
<b>accuracy</b>			0.81	3699
<b>macro avg</b>	0.69	0.78	0.71	3699
<b>weighted avg</b>	0.86	0.81	0.83	3699



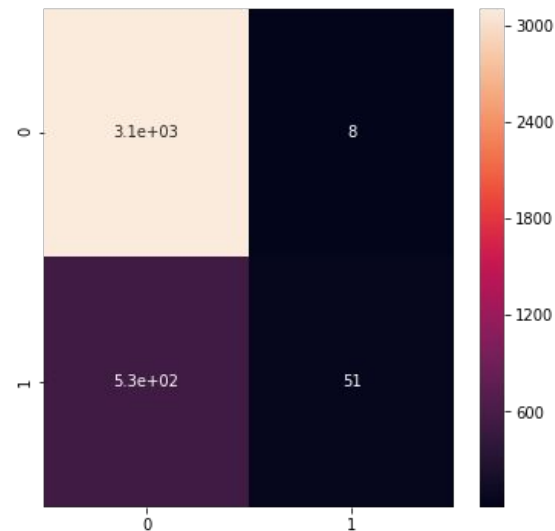
# Support Vector Machine (SVM)

Supervised non-probabilistic binary classifier algorithm, when given labeled training data, outputs an optimal hyperplane which categorizes new examples.

**Classification Report :**

Without SMOTE	Precision	Recall	f1-score	support
0	0.85	1.00	0.92	3114
1	0.86	0.09	0.16	585
accuracy			0.85	3699
macro avg	0.86	0.54	0.54	3699
weighted avg	0.86	0.85	0.80	3699

**Confusion Matrix:**

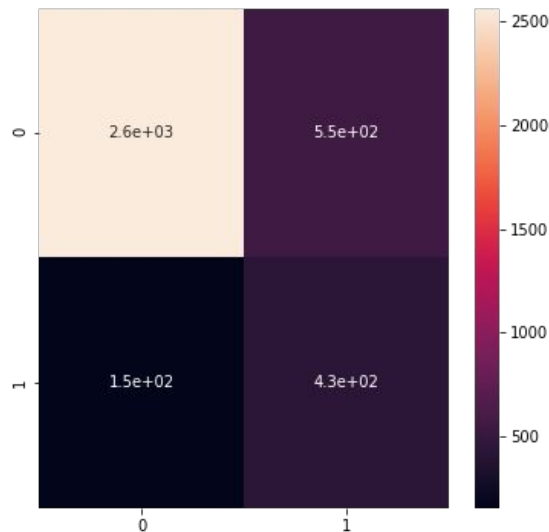


# Support Vector Machine (SVM)

Classification Report :

With SMOTE	Precision	Recall	f1-score	support
0	0.94	0.82	0.88	3114
1	0.44	0.74	0.55	585
accuracy			0.81	3699
macro avg	0.69	0.78	0.71	3699
weighted avg	0.86	0.81	0.83	3699

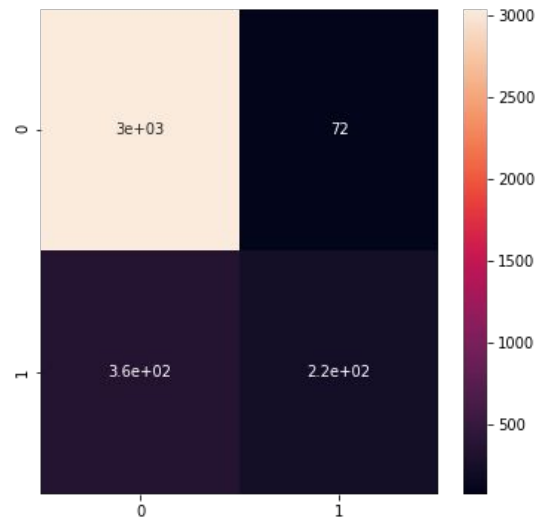
Confusion Matrix:



# Logistic Regression

Classification Report :

	Precision	Recall	f1-score	support
<b>0</b>	0.89	0.98	0.93	3114
<b>1</b>	0.76	0.38	0.51	585
<b>accuracy</b>			0.88	3699
<b>macro avg</b>	0.82	0.68	0.72	3699
<b>weighted avg</b>	0.87	0.88	0.87	3699



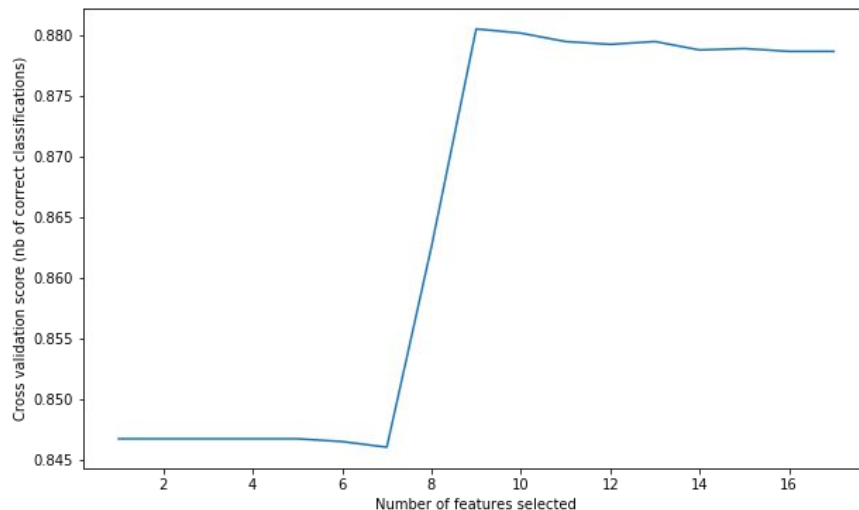
# Logistic Regression

Next we reduced the dimensionality by using Recursive Feature Elimination (RFE).

With Logistic Regression as the model, RFE selected the following **9** Features :

**Selected features:**

['Informational', 'BounceRates',  
'ExitRates', 'PageValues', 'SpecialDay',  
'Month', 'OperatingSystems',  
'VisitorType', 'Weekend']

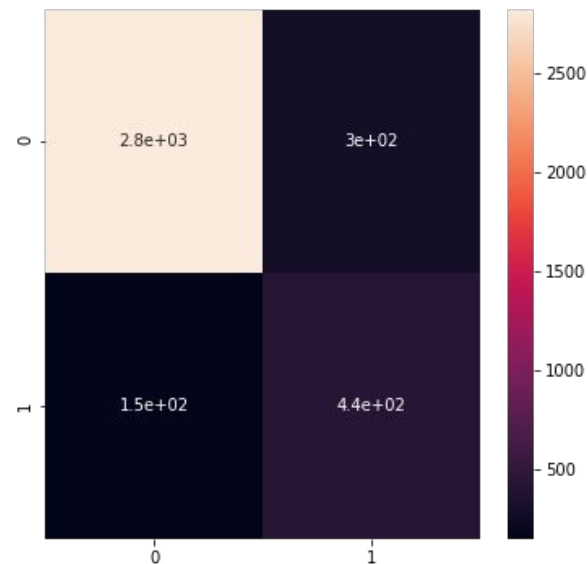




# Logistic Regression

Classification Report : with SMOTE and RFE

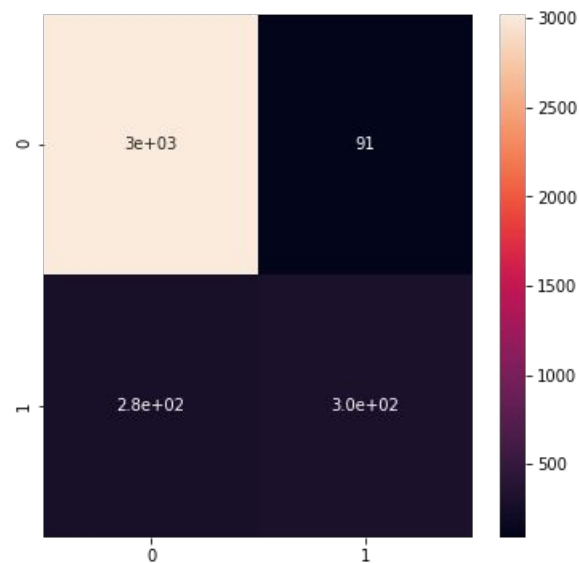
	Precision	Recall	f1-score	Support
<b>0</b>	0.95	0.91	0.93	3114
<b>1</b>	0.60	0.75	0.66	585
<b>Accuracy</b>			0.88	3699
<b>Macro avg</b>	0.77	0.83	0.79	3699
<b>Weighted avg</b>	0.89	0.88	0.89	3699



# Random Forest Classifier

Classification Report :

	Precision	Recall	f1-score	Support
<b>0</b>	0.92	0.97	0.94	3114
<b>1</b>	0.77	0.52	0.62	585
<b>Accuracy</b>			0.90	3699
<b>Macro avg</b>	0.84	0.75	0.78	3699
<b>Weighted avg</b>	0.89	0.90	0.89	3699



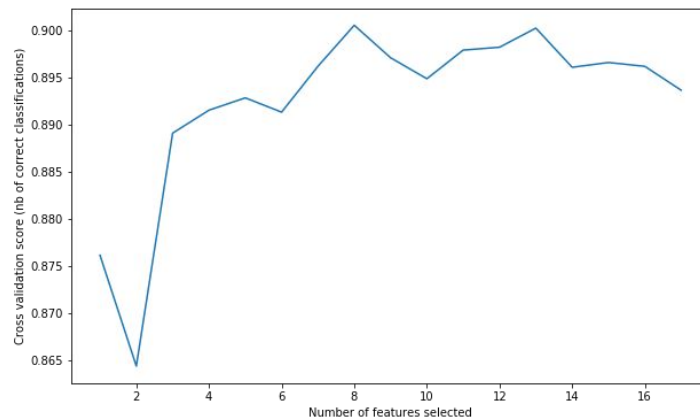
# Random Forest Classifier

Next we reduced the dimensionality by using Recursive Feature Elimination (RFE).

With Random Forest Classifier as the model, RFE selected the following **12** Features :

**Selected features:**

['Administrative', 'Administrative\_Duration', 'Informational\_Duration', 'ProductRelated', 'ProductRelated\_Duration', 'BounceRates', 'ExitRates', 'PageValues', 'Month', 'Browser', 'Region', 'TrafficType']

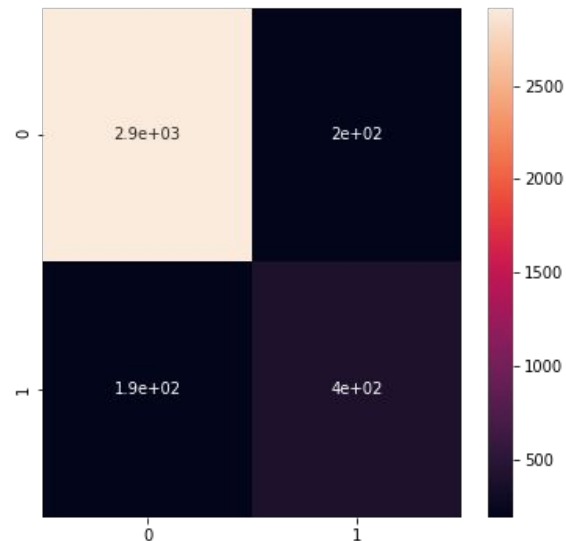


# Random Forest Classifier

We again evaluated our classifier and obtained the following results:

**Classification Report : With SMOTE and RFE**

	Precision	Recall	f1-score	Support
<b>0</b>	0.94	0.94	0.94	3114
<b>1</b>	0.67	0.68	0.67	585
<b>Accuracy</b>			0.89	3699
<b>Macro avg</b>	0.80	0.81	0.80	3699
<b>Weighted avg</b>	0.90	0.90	0.90	3699



# Neural Network

```
model = keras.Sequential([
    keras.layers.Dense(60, input_shape=(x_train.shape[1],), activation=tf.nn.relu),
    keras.layers.Dense(units=1, activation=tf.nn.sigmoid)
])
```

Model: "sequential"

Layer (type)	Output Shape	Param #
=====		
dense (Dense)	(None, 60)	1080
=====		
dense_1 (Dense)	(None, 1)	61
=====		

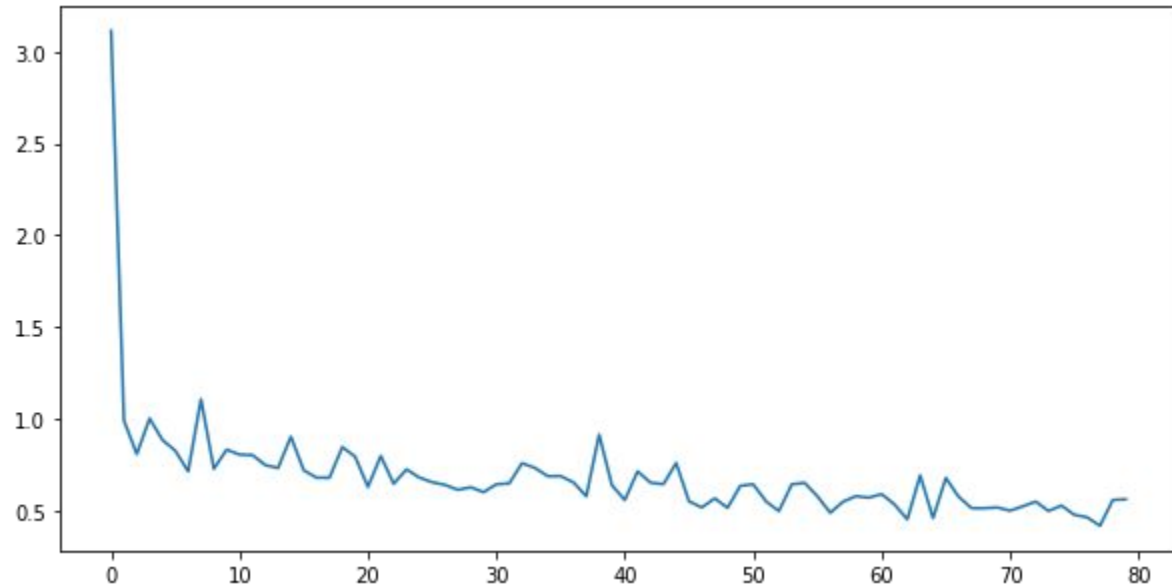
Total params: 1,141

Trainable params: 1,141

Non-trainable params: 0

# Neural Network

Training period : 80 Epochs



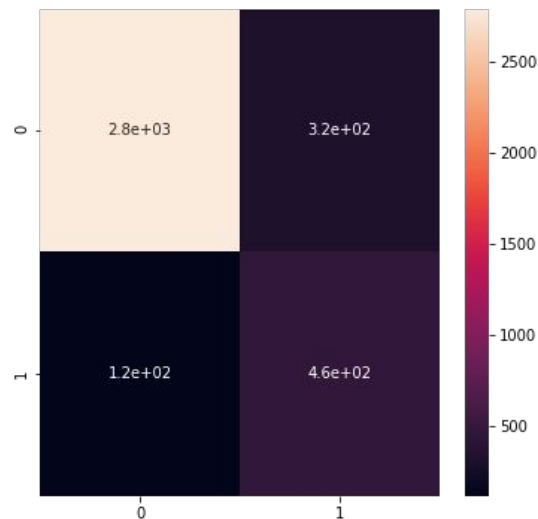
# Neural Network

Classification Report : With SMOTE

	Precision	Recall	f1-score	Support
<b>0</b>	0.96	0.90	0.93	3114
<b>1</b>	0.59	0.79	0.68	585
<b>Accuracy</b>			0.88	3699
<b>Macro avg</b>	0.77	0.85	0.80	3699
<b>Weighted avg</b>	0.90	0.88	0.89	3699

Training accuracy: 0.8807786

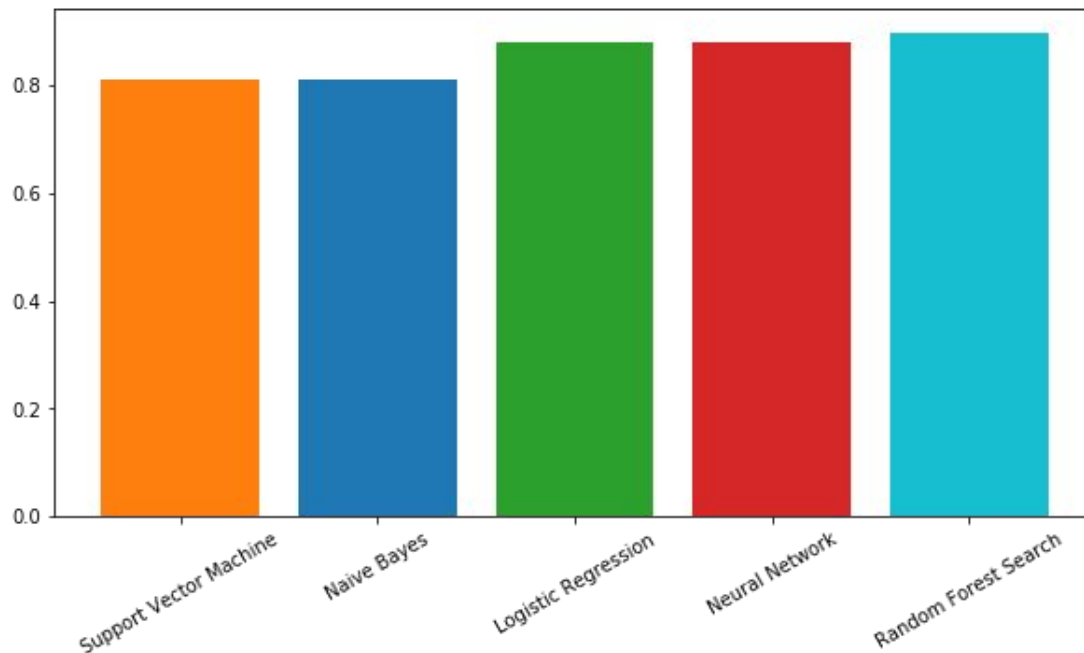
Testing accuracy: 0.88050824



# Comparison of Models

Accuracy

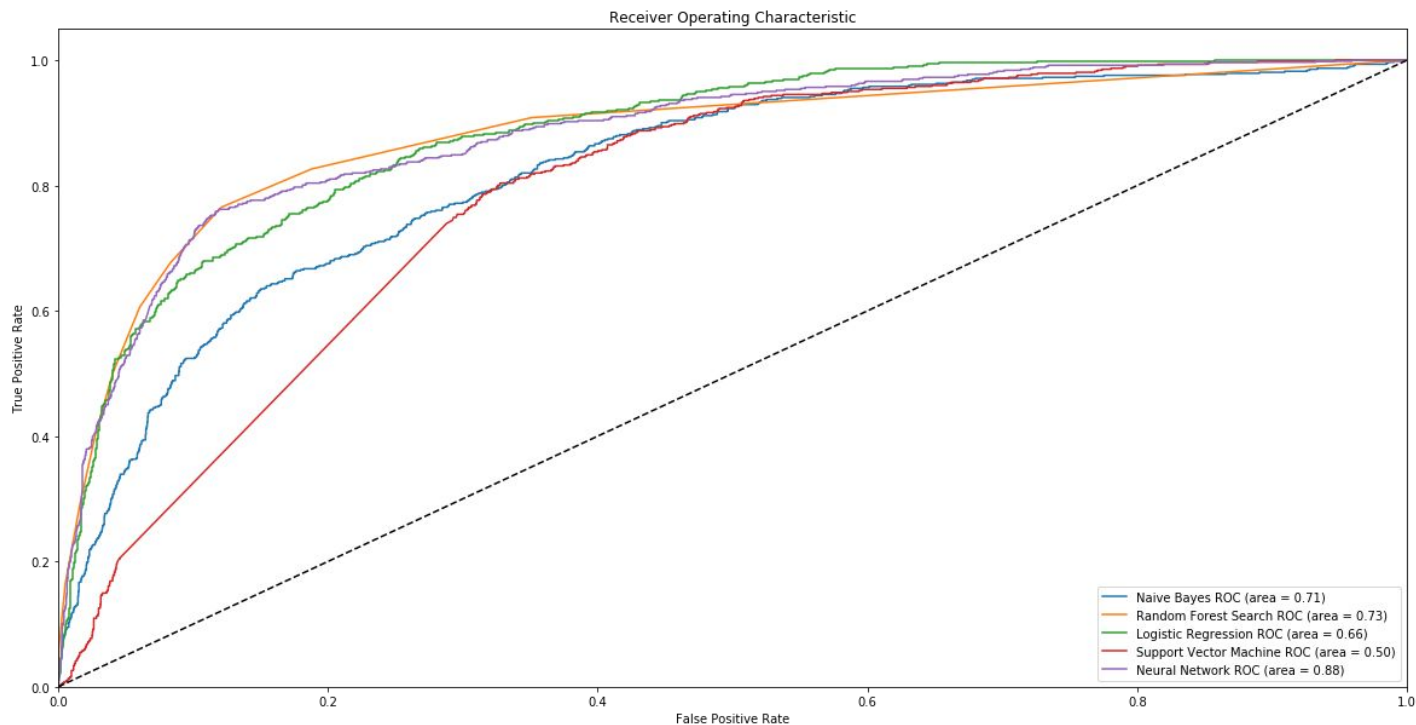
:





# Comparison of Models

## ROC Curves (Before optimization)



# Comparison of Models

## ROC Curves (After optimization)

