

# LEAD SCORING CASE STUDY

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# Problem Statement

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An education company named X Education sells online courses to industry professionals. On any given day, many professionals who are interested in the courses land on their website and browse for courses.

When these people fill up a form providing their email address or phone number, they are classified to be a lead. Moreover, the company also gets leads through past referrals. Once these leads are acquired, employees from the sales team start making calls, writing emails, etc.

Now, although X Education gets a lot of leads, its lead conversion rate is very poor. If they successfully identify this set of leads, the lead conversion rate should go up.

The company requires you to build a model to identify the leads of higher conversion rate and lower conversion rate.

The CEO, in particular, has given a ballpark of the target lead conversion rate to be around 80%.

# Problem Approach

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As a Data Scientist, our job is to analyze the data available and help out the CEO of the company to achieve profits via targeting which leads are to be concentrated to increase conversion percentage.

Since this is a Machine Learning Model.

Data is readily available.

Following are the steps to approach:

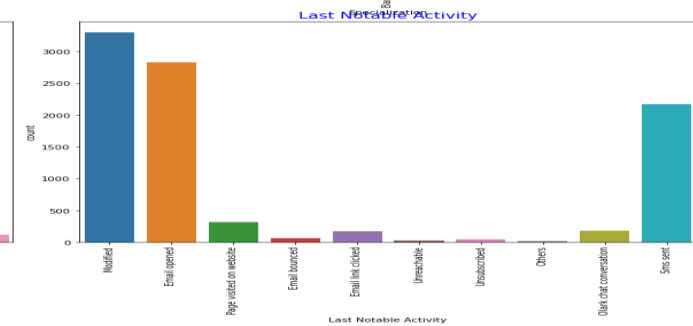
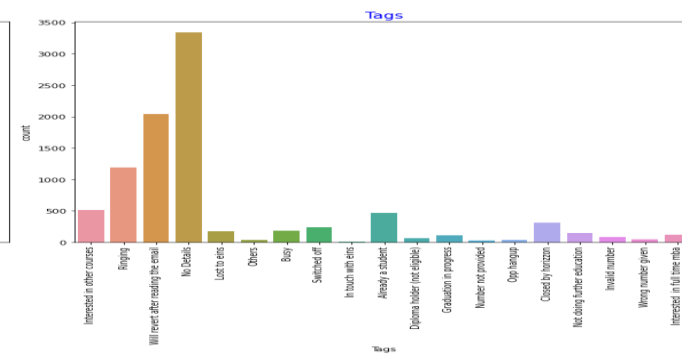
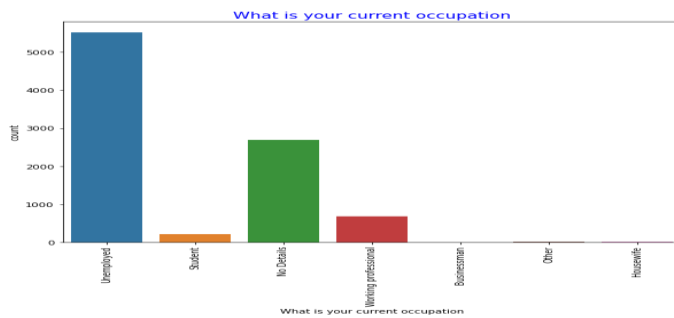
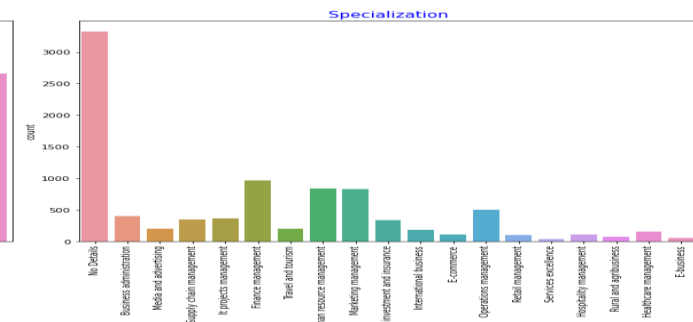
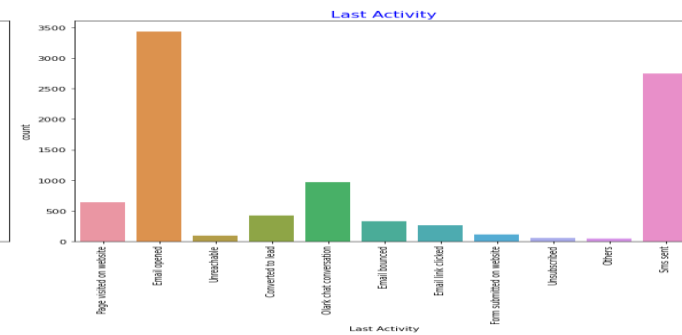
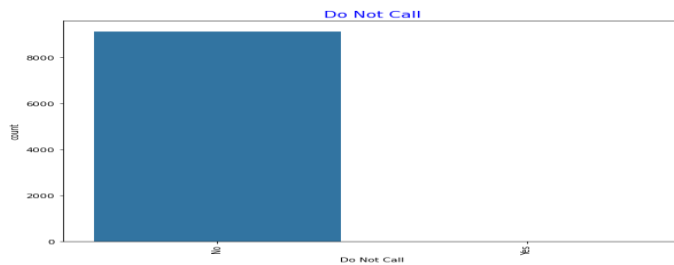
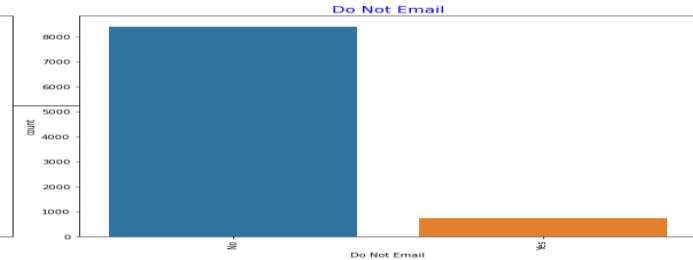
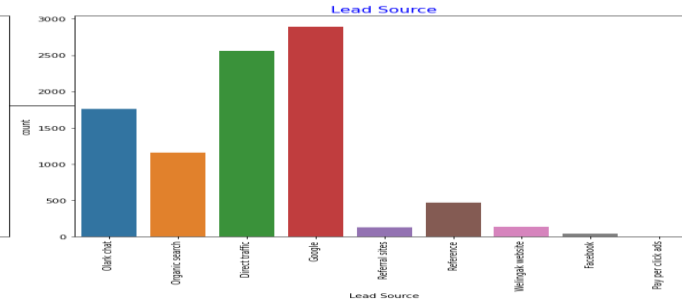
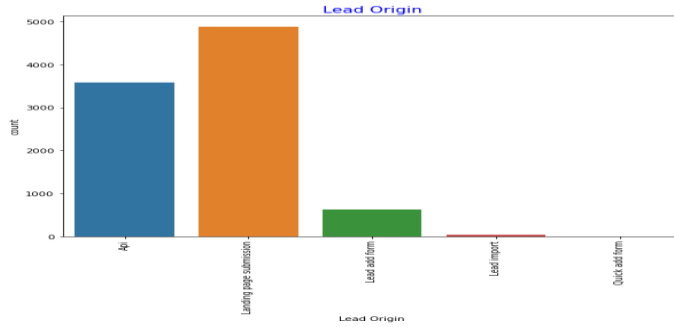
1. Read and Clean the data
2. Visualization and Outlier treatment
3. Data Modelling
4. Data Training
5. Evaluate the model
6. Inference

# EDA

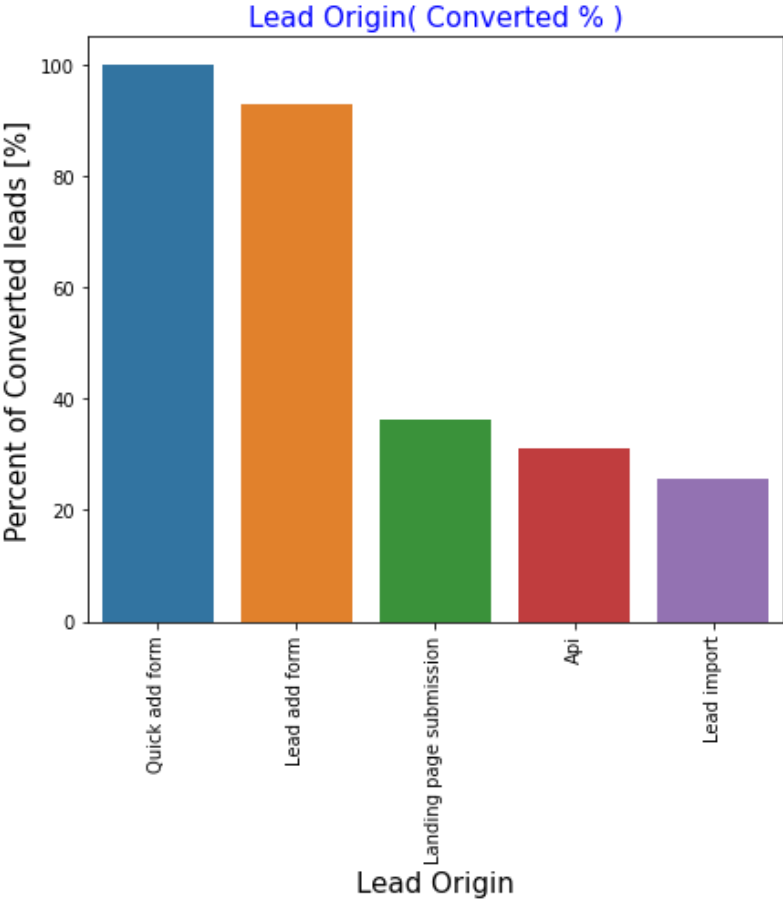
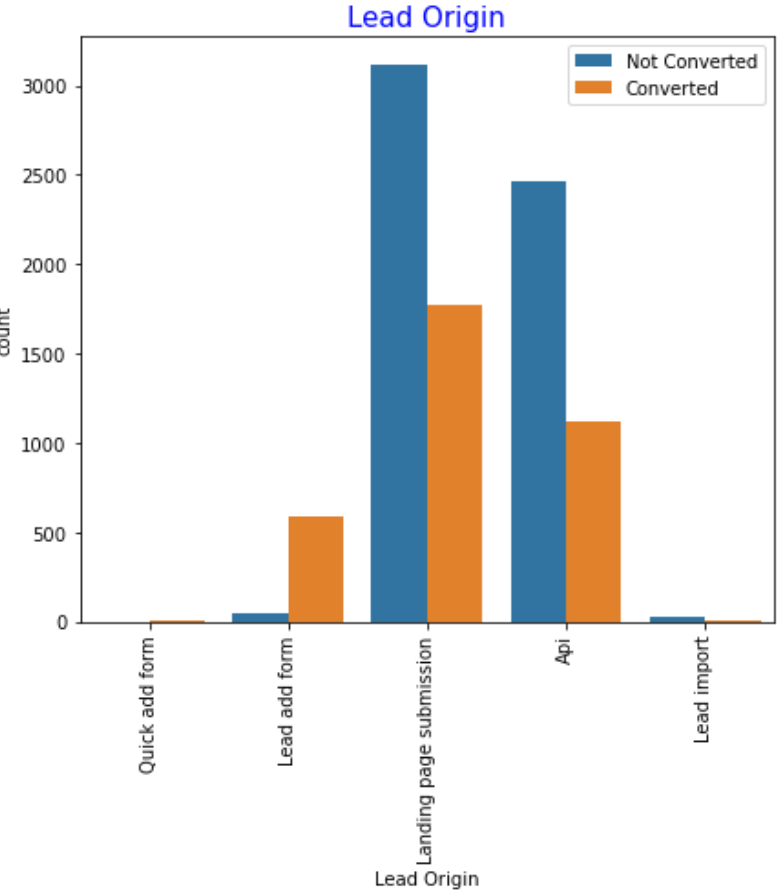
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- ✓ The Data set is huge and has more variables to be looked at.
- ✓ After understanding the variables, Deleted the variables which have high null values(>37%).
- ✓ Visualized the Variables based on type of data(Categorical and Numerical).
- ✓ Since, we have only 3 numerical variables (after dropping) which contains the outliers and outlier treatment in this data set could lead to loss of data which can harm the further analysis. So skipped outlier treatment.

# Count vs Categorical Columns

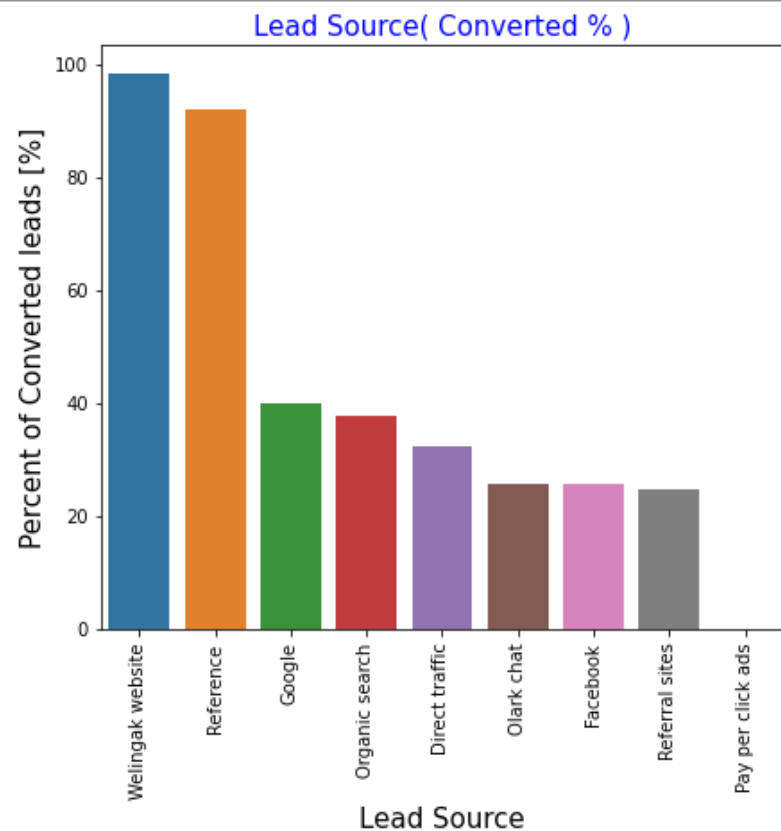
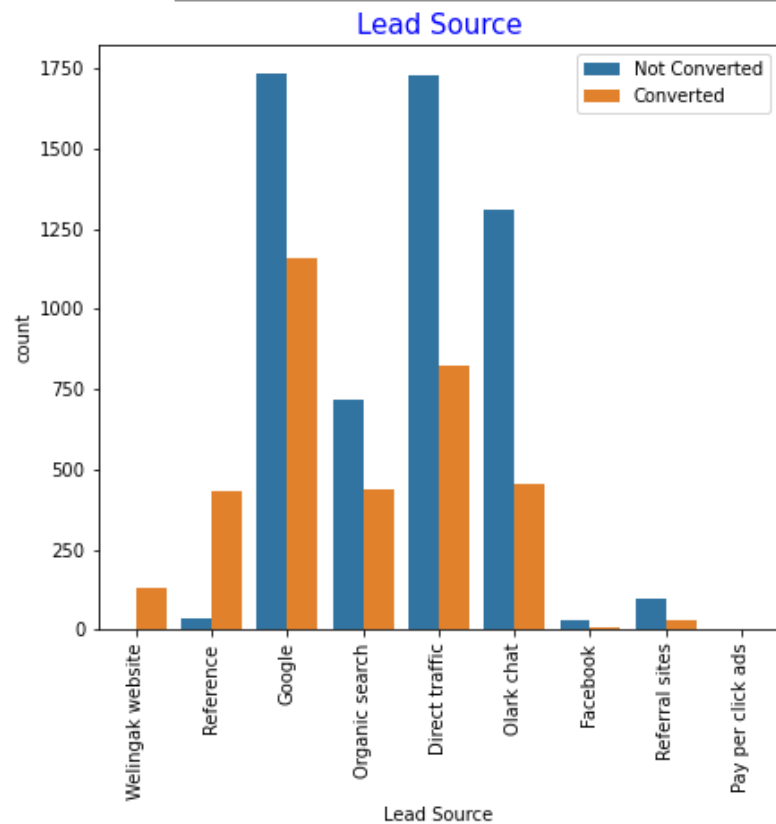


# Lead Origin (converted vs not-converted)



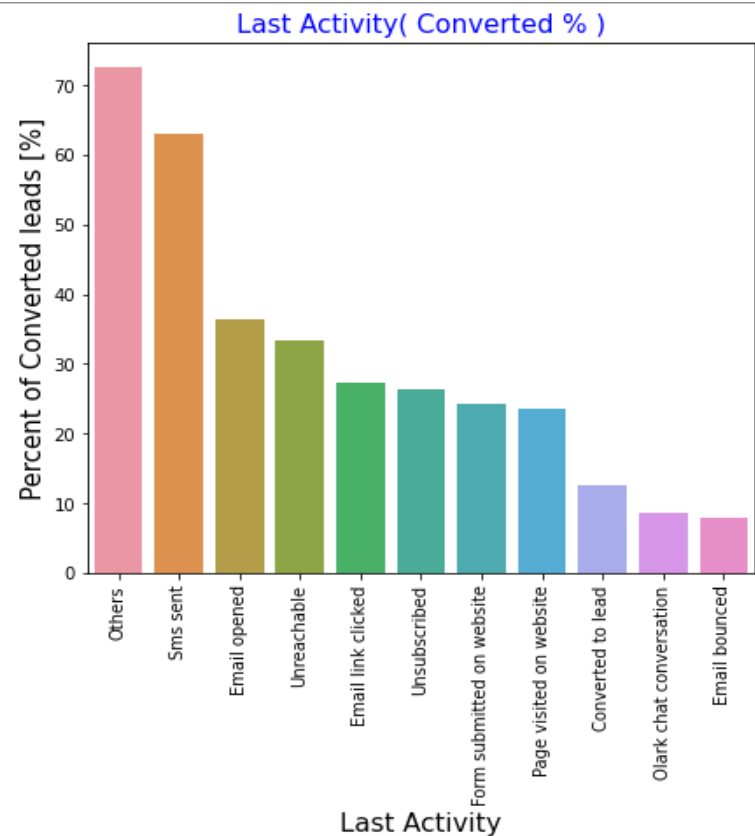
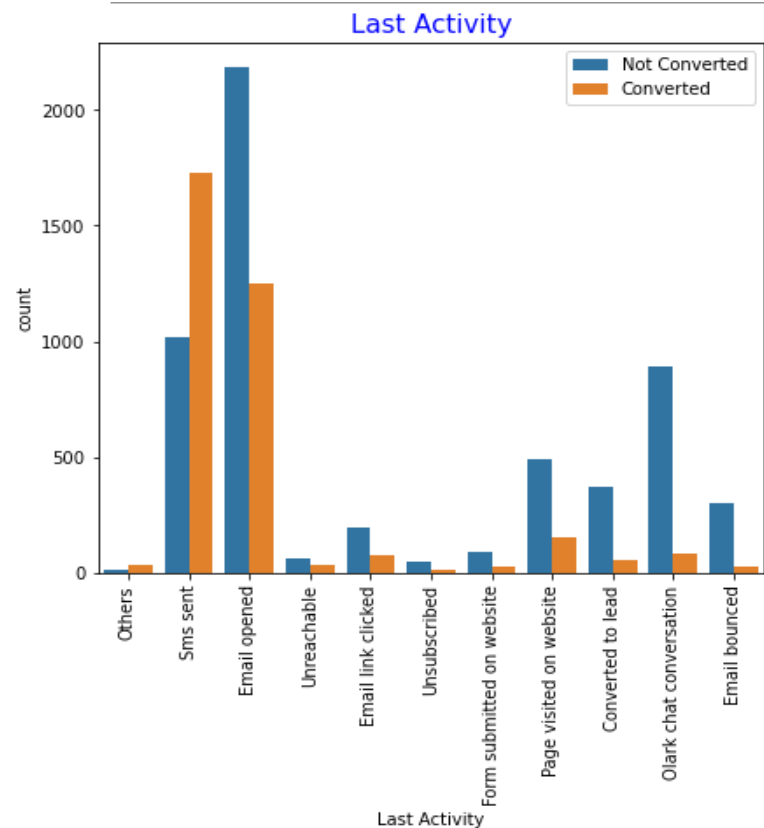
Highest leads	Highest lead conversion rate
1. Landing page submissions	1. Quick lead form
2. API	2. Lead add form
3. Lead add	3. Landing on page

# Lead Source (converted vs not-converted)



Highest leads	Highest lead conversion rate
1. Google	1. Wellington website
2. Direct traffic	2. Reference
3. Olark chat	3. Google

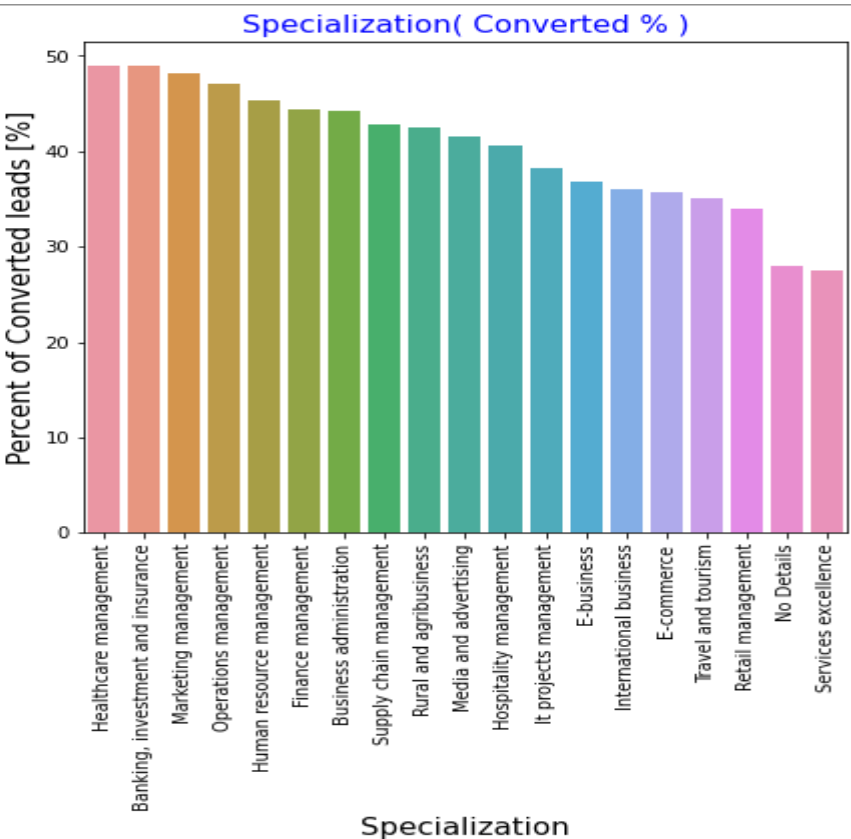
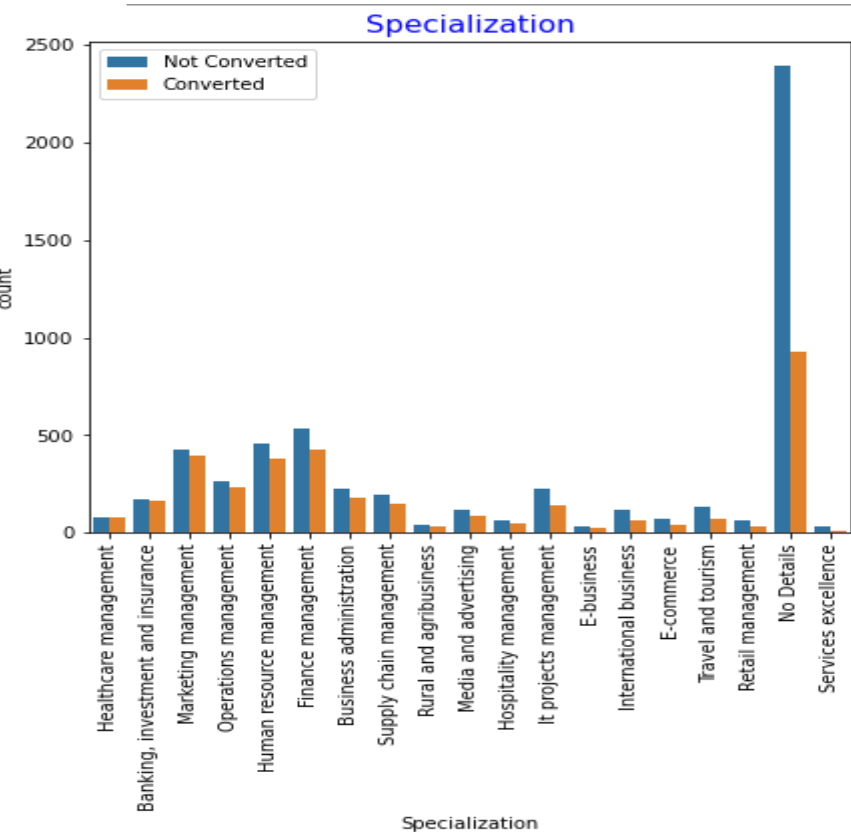
# Last Activity (converted vs not-converted)



Highest leads	Highest lead conversion rate
1. Email Opened	1. Sms sent
2. Sms sent	2. Email sent
3. Olark chat	

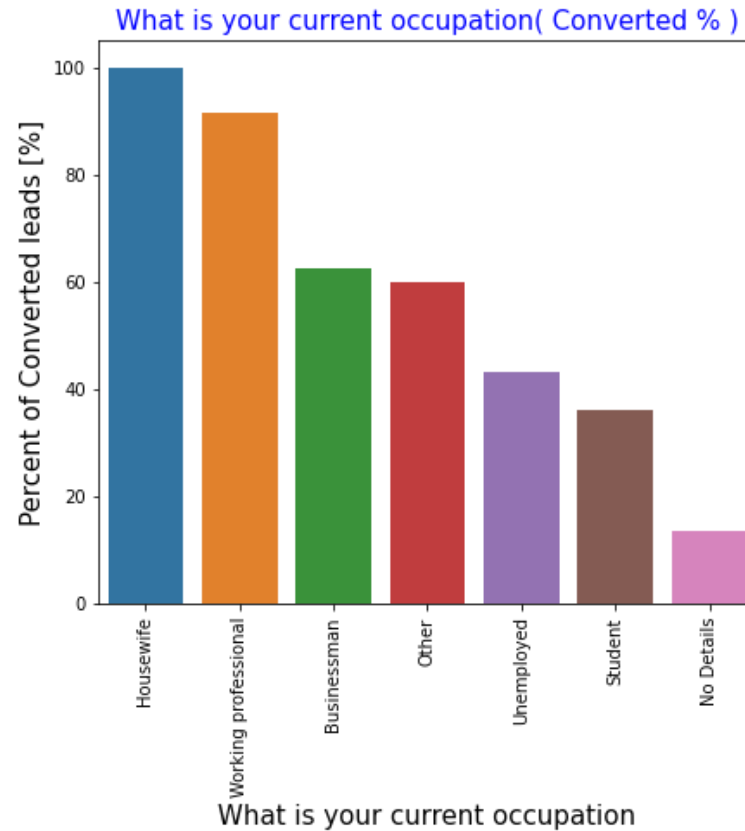
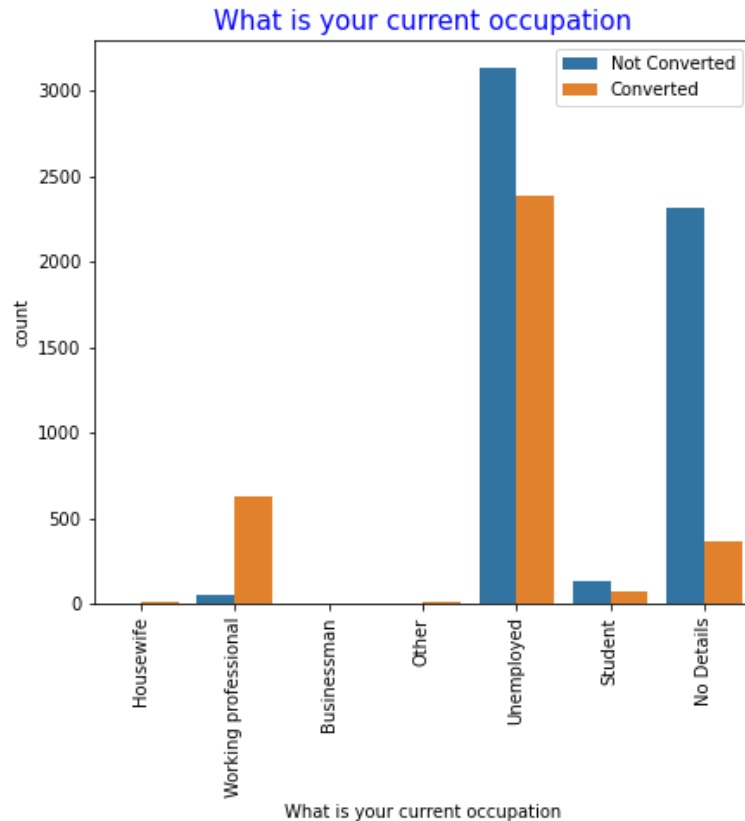


# Specialization (converted vs not-converted)



Highest leads	Highest lead conversion rate
1. Finance	1. Health-care
2. Marketing	2. Banking, Investment and Insurance
3. Operations	3. Marketing

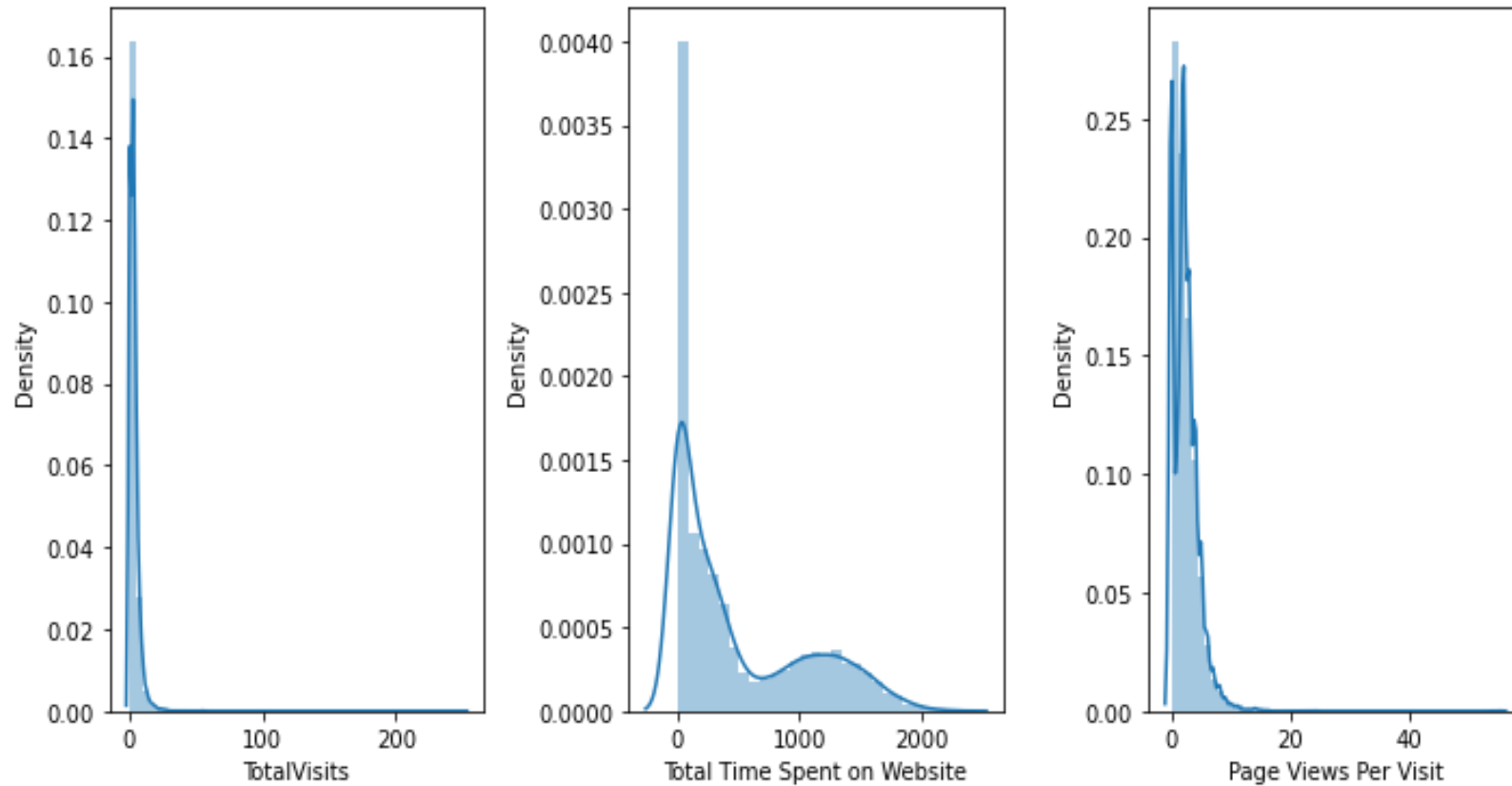
# Occupation (converted vs not-converted)



Highest leads	Highest lead conversion rate
1. Unemployed	1. Housewife
2. Working professionals	2. Working Professional
3. Students	3. Businessman

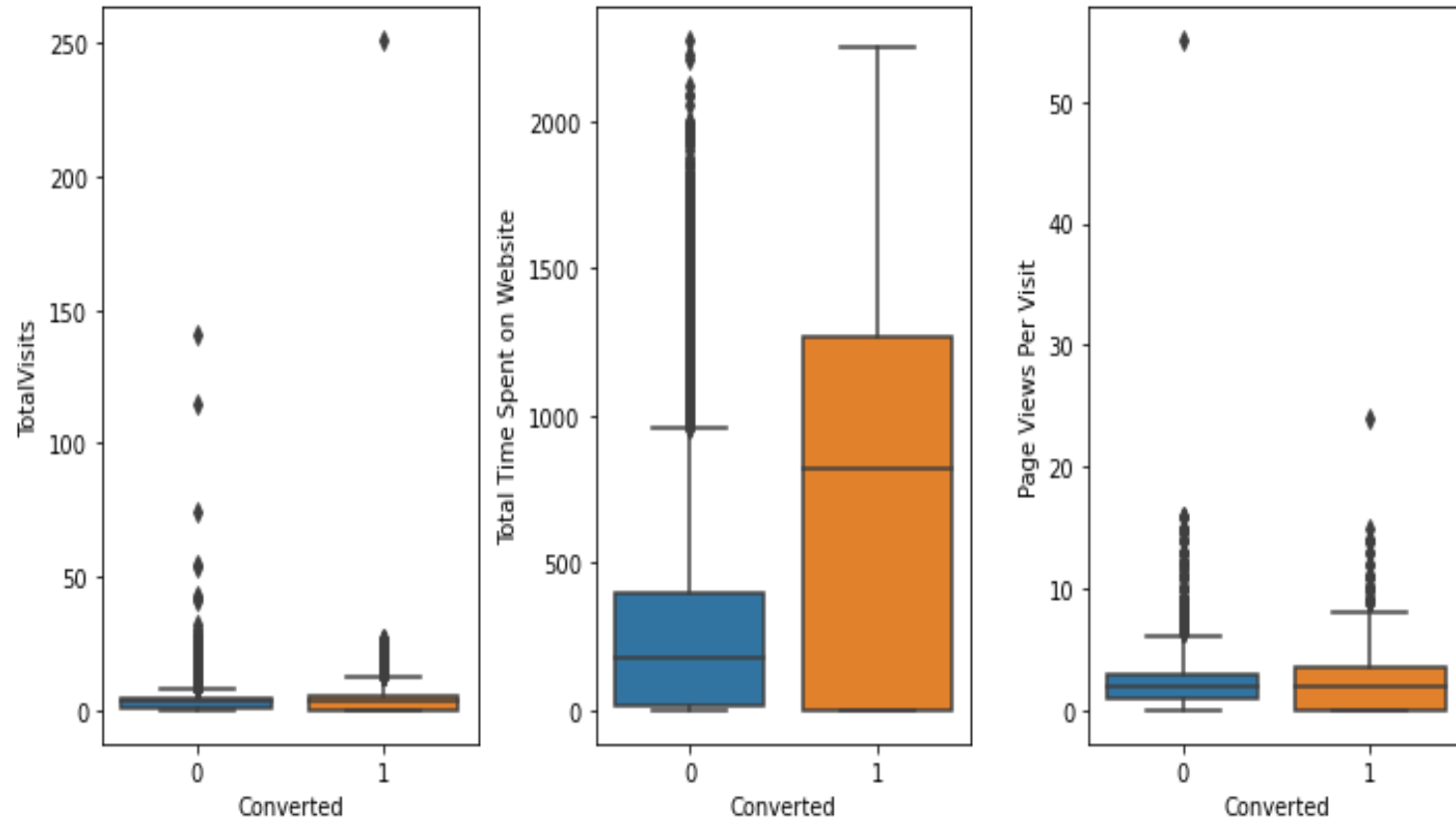
# Numerical columns vs Density

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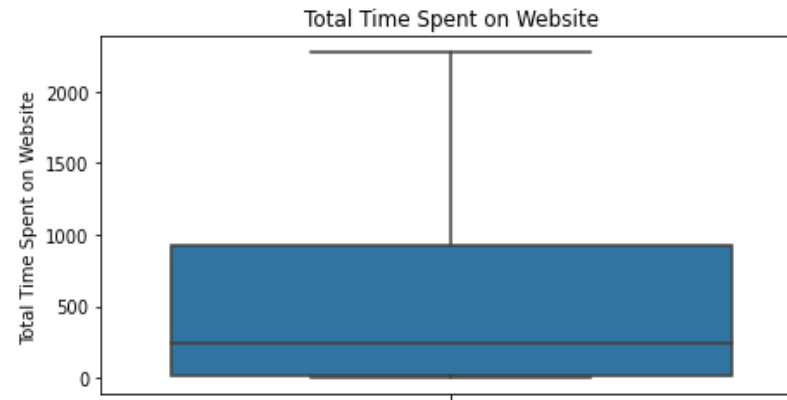
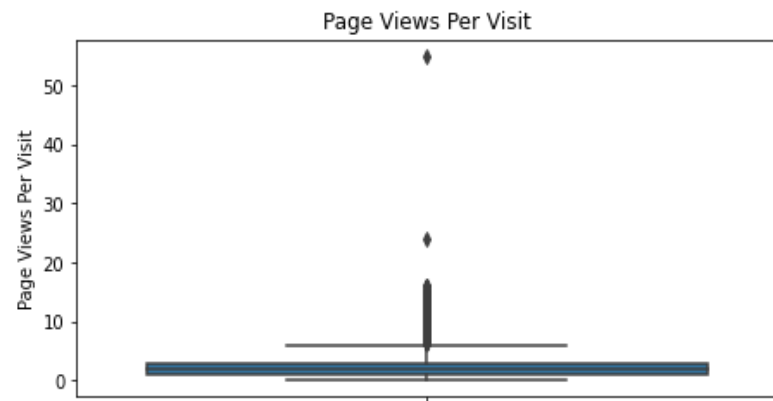
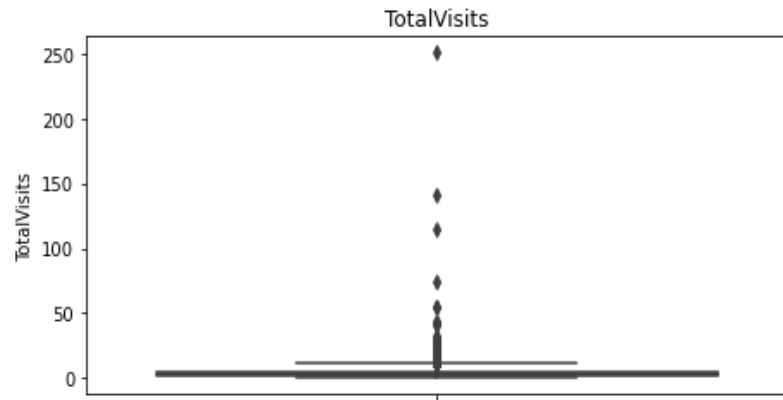
- **Since data is normalized and is clearly skewed to left side**

# Numerical Columns vs Conversion



- **Conversion rate is more in Total time spent on website.**

# Outliers for numerical columns



- **Since there are only 3 numerical columns left.**
- **Outliers are present in the variables and are important for the analysis**

# Model Obtained after RFE

Generalized Linear Model Regression Results

Dep. Variable:	Converted	No. Observations:	6351
Model:	GLM	Df Residuals:	6335
Model Family:	Binomial	Df Model:	15
Link Function:	logit	Scale:	1.0000
Method:	IRLS	Log-Likelihood:	-1550.8
Date:	Sun, 07 Feb 2021	Deviance:	3101.6
Time:	12:25:56	Pearson chi2:	6.13e+03
No. Iterations:	23		
Covariance Type:	nonrobust		

	coef	std err	z	P> z	[0.025	0.975]
const	-1.9942	0.070	-28.446	0.000	-2.132	-1.857
Total Time Spent on Website	3.7442	0.190	19.737	0.000	3.372	4.116
Tags_Already a student	-3.9735	0.715	-5.554	0.000	-5.376	-2.571
Tags_Closed by horizzon	6.0075	0.715	8.403	0.000	4.606	7.409
Tags_Diploma holder (not eligible)	-3.0007	1.023	-2.932	0.003	-5.006	-0.995
Tags_Interested in full time mba	-2.7293	0.730	-3.740	0.000	-4.160	-1.299
Tags_Interested in other courses	-2.4577	0.318	-7.730	0.000	-3.081	-1.835
Tags_Invalid number	-23.2232	1.67e+04	-0.001	0.999	-3.27e+04	3.26e+04
Tags_Lost to eins	5.0873	0.721	7.058	0.000	3.675	6.500
Tags_Not doing further education	-3.5775	1.015	-3.524	0.000	-5.567	-1.588
Tags_Number not provided	-23.9428	2.79e+04	-0.001	0.999	-5.48e+04	5.47e+04
Tags_Ringing	-2.7678	0.238	-11.629	0.000	-3.234	-2.301
Tags_Switched off	-2.6284	0.519	-5.065	0.000	-3.646	-1.611
Tags_Will revert after reading the email	4.8588	0.182	26.739	0.000	4.503	5.215
Tags_Wrong number given	-23.5944	2.22e+04	-0.001	0.999	-4.36e+04	4.35e+04
Lead Source_Welingak website	5.5876	0.720	7.760	0.000	4.176	6.999

This is the Model-1 which have a few unwanted and invalid parameters(high p-value and VIF) which are to be reduced manually.

# Final Model after RFE and Manual Reduction

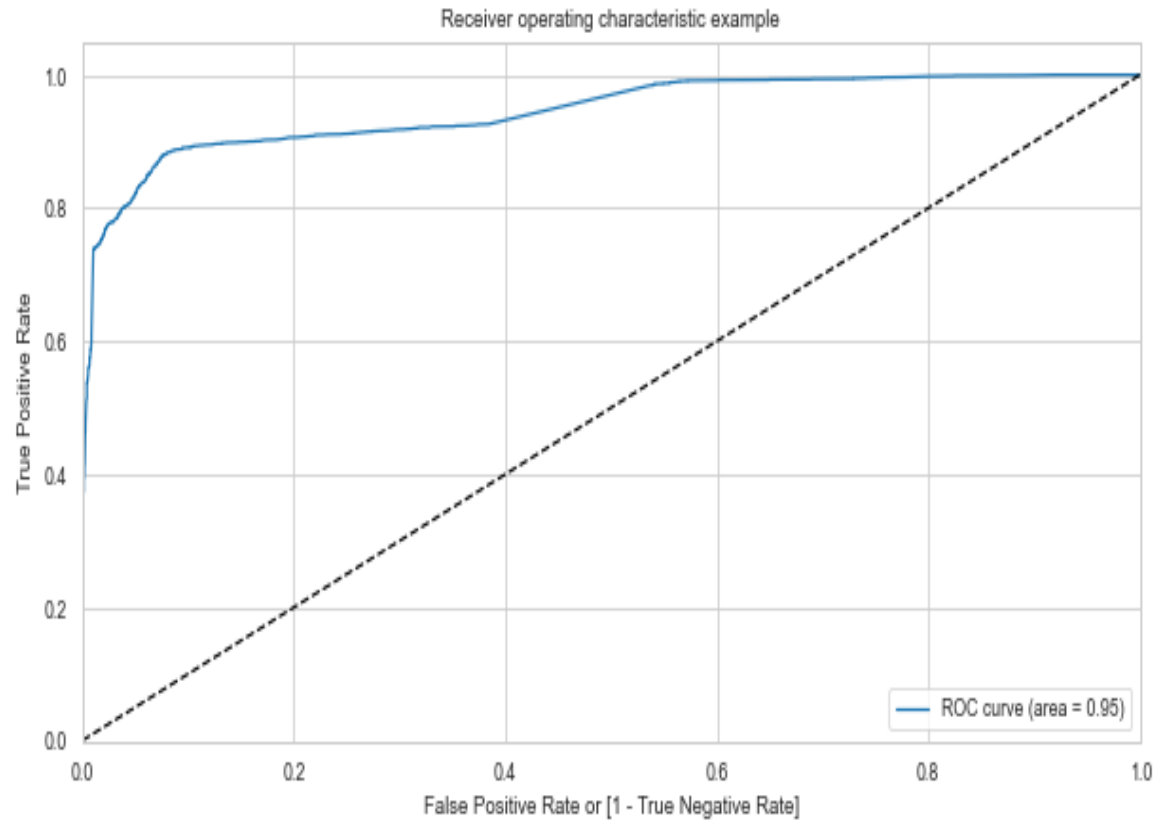
## Generalized Linear Model Regression Results

Dep. Variable:	Converted	No. Observations:	6351
Model:	GLM	Df Residuals:	6338
Model Family:	Binomial	Df Model:	12
Link Function:	logit	Scale:	1.0000
Method:	IRLS	Log-Likelihood:	-1580.5
Date:	Sun, 07 Feb 2021	Deviance:	3161.0
Time:	12:25:59	Pearson chi2:	6.22e+03
No. Iterations:	8		
Covariance Type:	nonrobust		

	coef	std err	z	P> z	[0.025	0.975]
const	-2.0463	0.070	-29.290	0.000	-2.183	-1.909
Total Time Spent on Website	3.7058	0.187	19.779	0.000	3.339	4.073
Tags_Already a student	-3.9070	0.715	-5.463	0.000	-5.309	-2.505
Tags_Closed by horizzon	6.0628	0.715	8.481	0.000	4.662	7.464
Tags_Diploma holder (not eligible)	-2.9341	1.023	-2.868	0.004	-4.939	-0.929
Tags_Interested in full time mba	-2.6627	0.729	-3.650	0.000	-4.092	-1.233
Tags_Interested in other courses	-2.3915	0.318	-7.531	0.000	-3.014	-1.769
Tags_Lost to eins	5.1447	0.721	7.139	0.000	3.732	6.557
Tags_Not doing further education	-3.5103	1.015	-3.459	0.001	-5.499	-1.521
Tags_Ringing	-2.7002	0.238	-11.364	0.000	-3.166	-2.234
Tags_Switched off	-2.5625	0.519	-4.941	0.000	-3.579	-1.546
Tags_Will revert after reading the email	4.9156	0.182	27.071	0.000	4.560	5.271
Lead Source_Welingak website	5.6397	0.720	7.833	0.000	4.229	7.051

- The Model-4 after all necessary reductions.
- This is our final model with significant parameters for business development

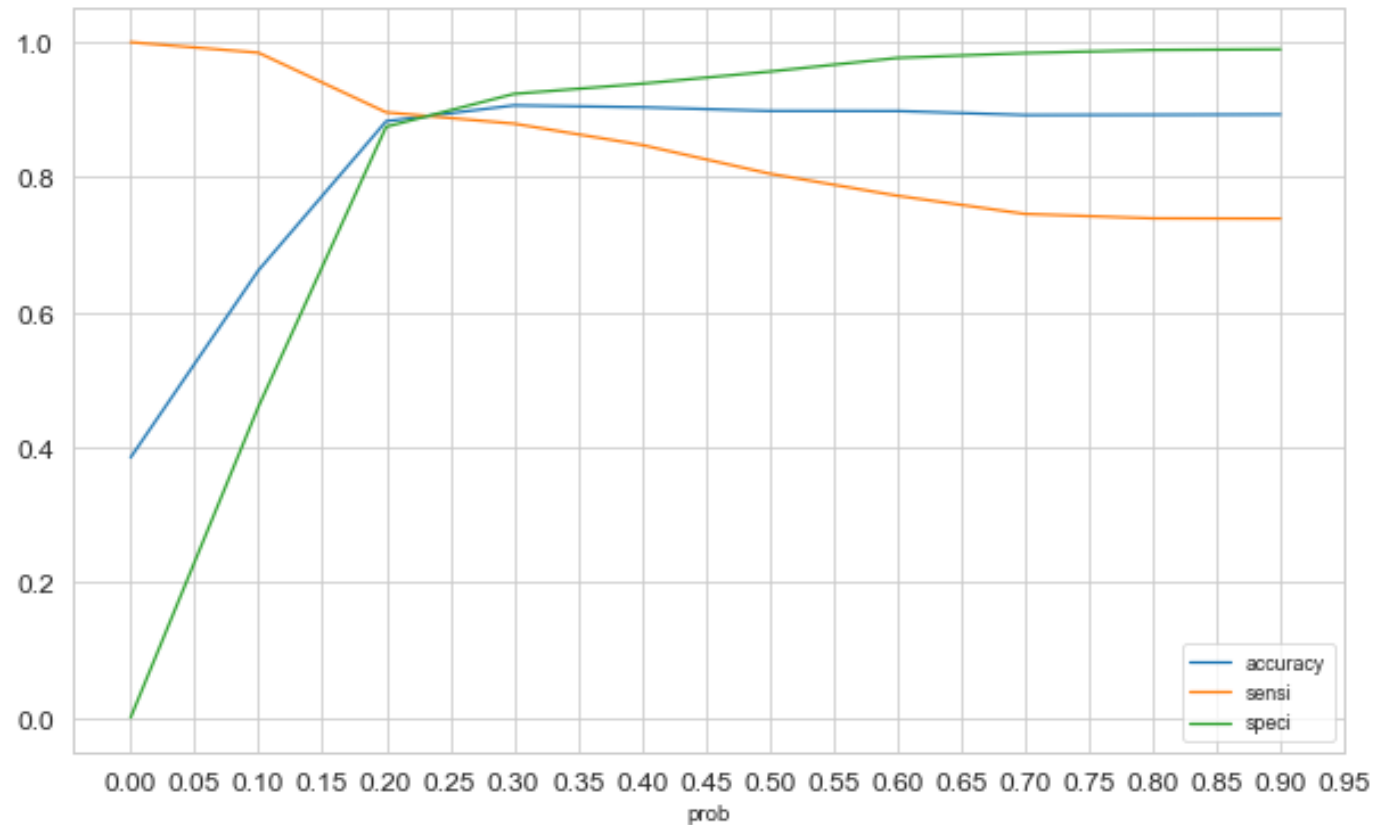
# ROC curve train-set



**The ROC curve gives area =0.95 which is exceptional value w.r.t model obtained.**



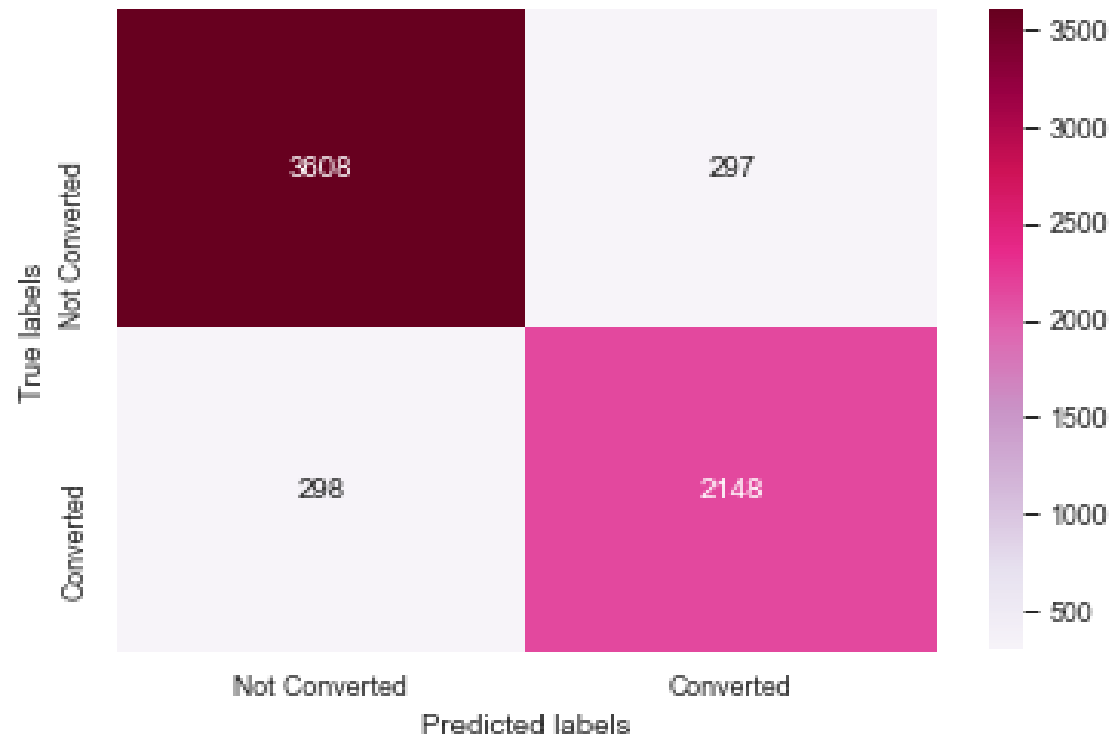
# Trade-off curve



- **Trade-off point is 0.24, since it is very low conversion rate**
- **We tried doing the analysis with 0.33 trade-off which come out to be a good. So considered the later for the same.**

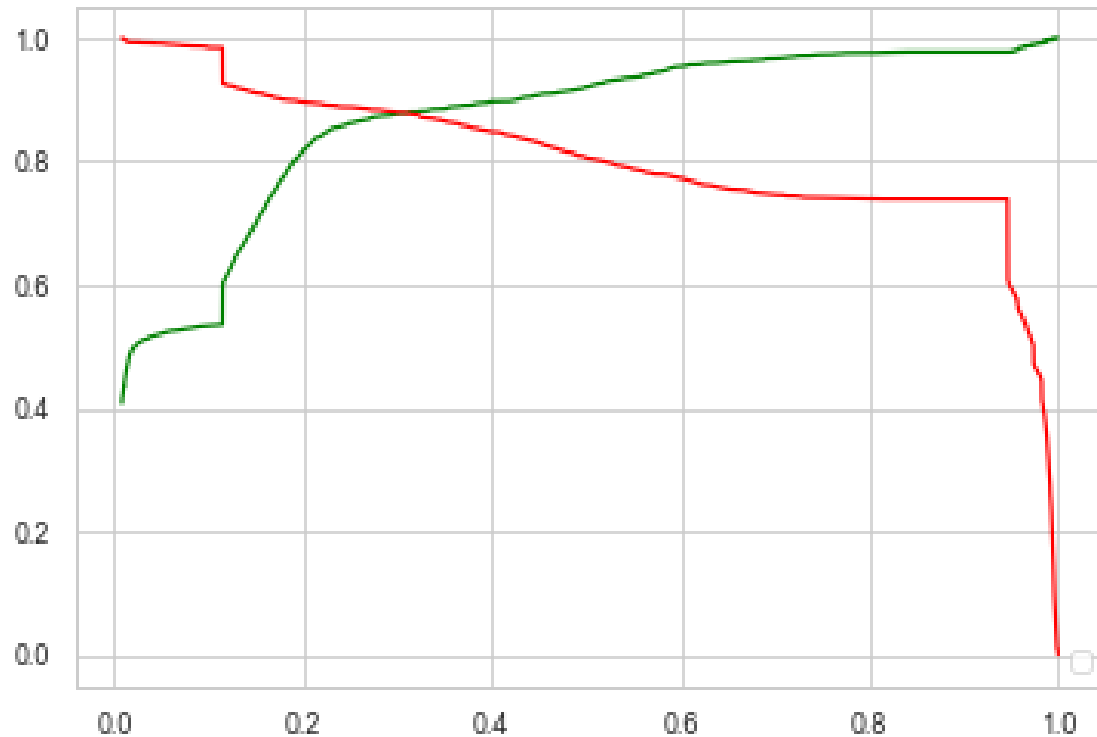
# Confusion Matrix (after trade-off)

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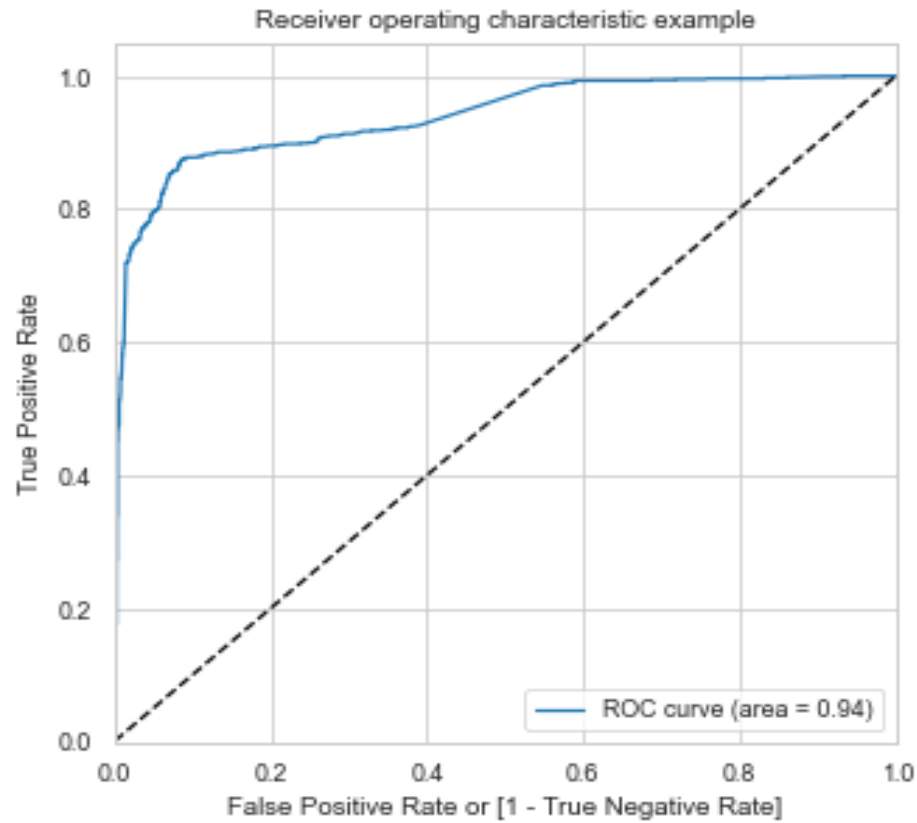
# Precision recall curve

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- The precision-recall curve gives an exceptional accuracy of around 0.88.
- Which indicates the model is Good

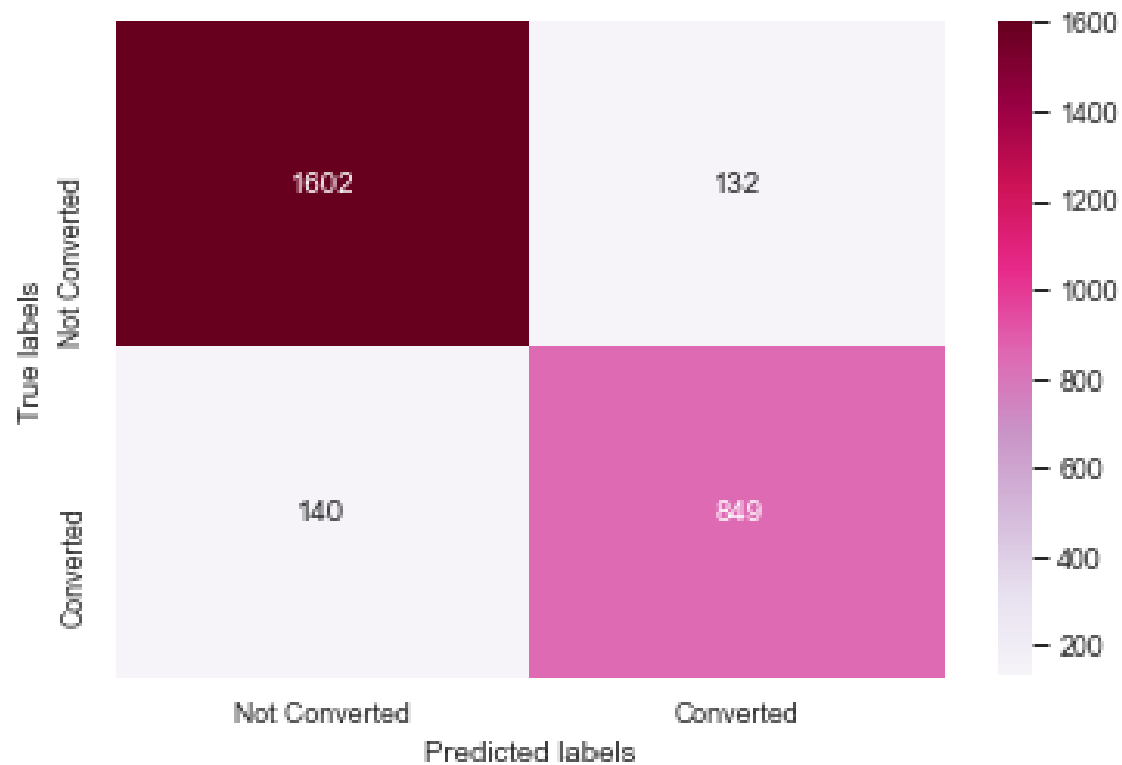
# ROC curve test-set



**The ROC curve gives area =0.94 which is exceptional value w.r.t model obtained.**

# Confusion matrix on test-set

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# Metrics train-set vs test-set

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## **Metrics after trade off point on train-set:**

Accuracy : 0.905

Sensitivity : 0.805

Specificity : 0.956

## **Metrics for test-set:**

Accuracy : 0.900

Sensitivity : 0.858

Specificity : 0.923

- ❖ **The metrics of in two sets are almost similar, so the model is good.**
- ❖ **The Metrics are above the expected business perspective**

# Conclusion

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The CEO of the company expected the model to be 80% accurate. As expected model is more than 80% accurate. So model is good to go for the business development.

Here are the top five parameters which need to be concentrated more by the sales to increase conversion rate.

1. Total Time Spent on Website
2. Tags\_Will revert after reading the email
3. Tags\_Ringing
4. Tags\_Closed by horizzon
5. Tags\_Lost to eins