

X EDUCATION-LEAD SCORING CASE STUDY



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Problem Statement:

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. Through this process, some of the leads get converted while most do not. The typical lead conversion rate at X education is around 30%

The CEO has set a target lead conversion rate of 80%. The lead scoring model should help the sales team to prioritize potential leads that have a higher conversion chance and enable them to focus on communicating with them.

Goals of the Case Study:

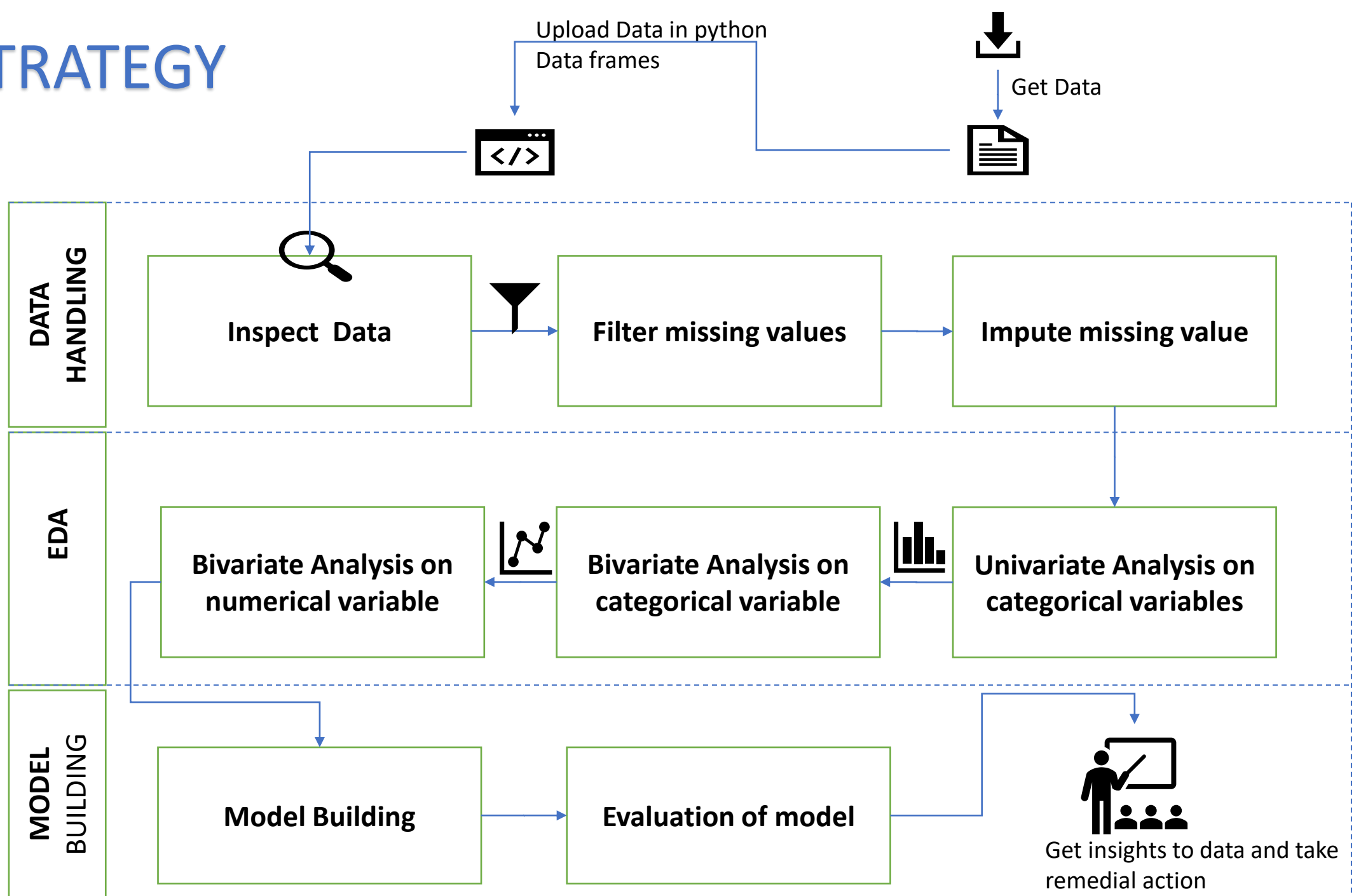
1. Build a logistic regression model to assign a lead score between 0 and 100 to each of the leads which can be used by the company to target potential leads. A higher score would mean that the lead is hot, i.e. is most likely to convert whereas a lower score would mean that the lead is cold and will mostly not get converted.

2. There are some more problems presented by the company which your model should be able to adjust to if the company's requirement changes in the future so need to handle these as well

Content

- Strategy
- Understand Data
- Data Handling
- Univariate Analysis
- Bivariate Analysis
- Correlation Analysis
- Model Building
- Final Model
- Conclusion
- Recommendation

STRATEGY



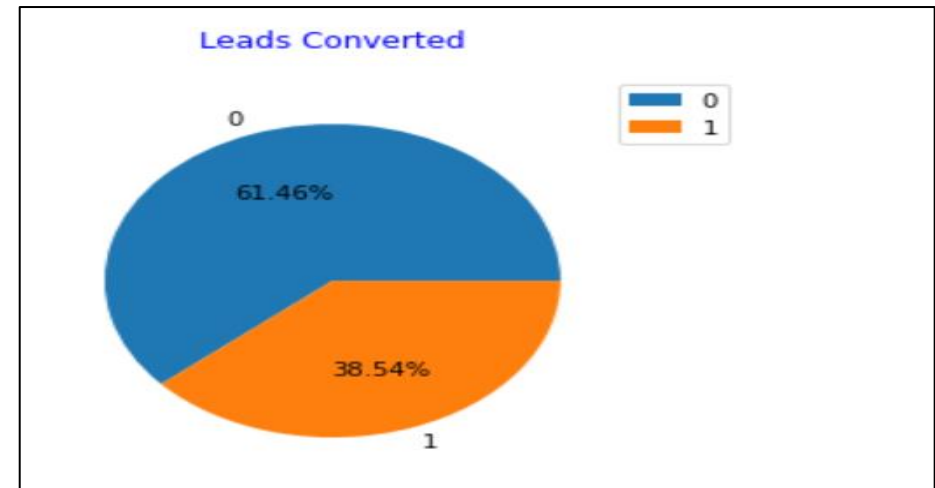
Understand Data

- Dataset used: **Leads.csv**
- Total Data point entries: **9240**
 - Number of column:**37**
 - Target column: **Converted**

```
df.info()
```

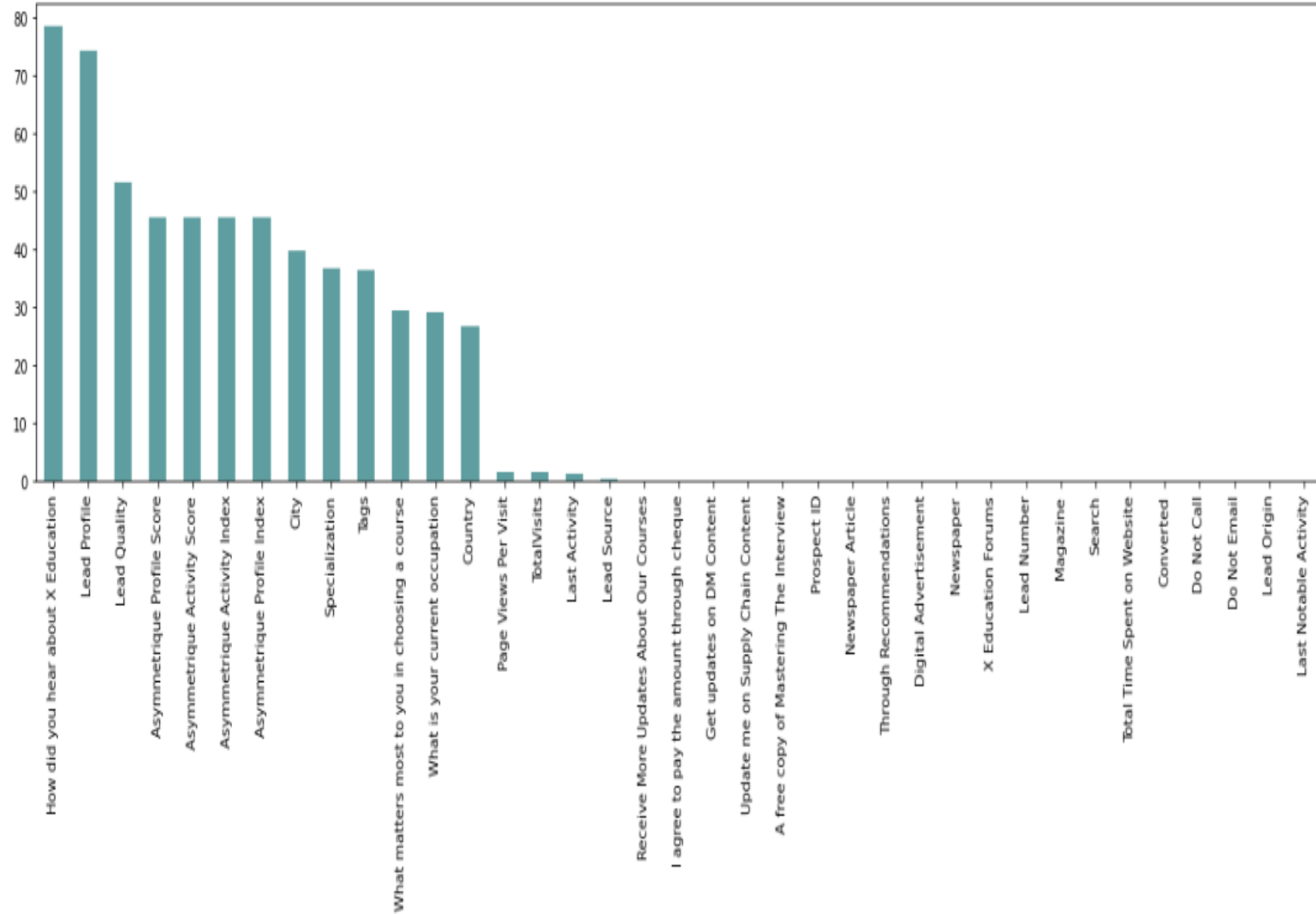
```
<class 'pandas.core.frame.DataFrame'>  
RangeIndex: 9240 entries, 0 to 9239  
Data columns (total 37 columns):
```

- Current Conversion Rate:**38.54%**
 - Values in converted is binary
 - 0 means lead not converted
 - 1 means lead converted



Data Preparation : Data Handling

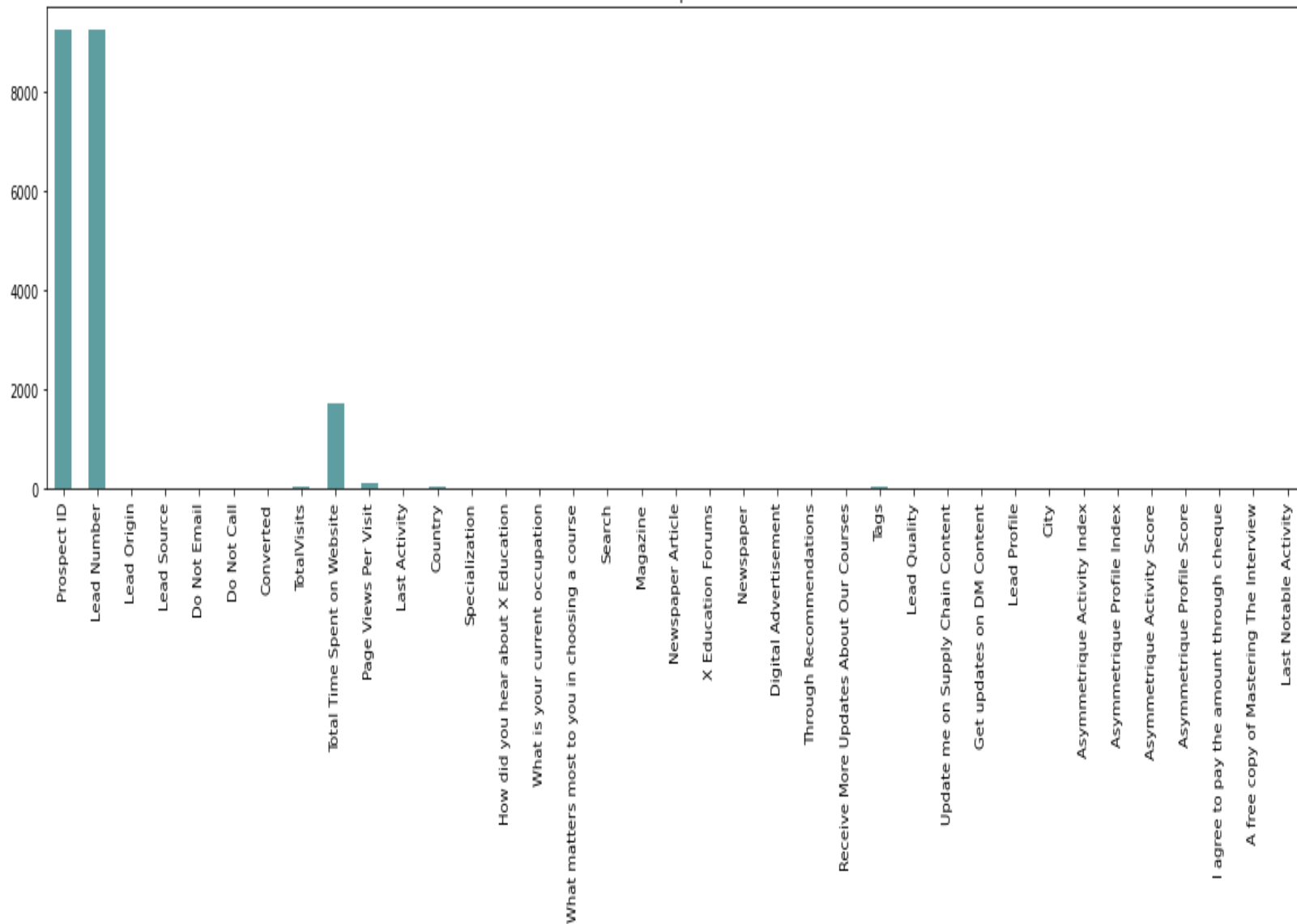
List of Columns and Null counts



| | |
|---|-----------|
| How did you hear about X Education | 78.463203 |
| Lead Profile | 74.188312 |
| Lead Quality | 51.590909 |
| Asymmetrique Profile Score | 45.649351 |
| Asymmetrique Activity Score | 45.649351 |
| Asymmetrique Activity Index | 45.649351 |
| Asymmetrique Profile Index | 45.649351 |
| City | 39.707792 |
| Specialization | 36.580087 |
| Tags | 36.287879 |
| What matters most to you in choosing a course | 29.318182 |
| What is your current occupation | 29.112554 |
| Country | 26.634199 |
| Page Views Per Visit | 1.482684 |
| TotalVisits | 1.482684 |
| Last Activity | 1.114719 |
| Lead Source | 0.389610 |
| Receive More Updates About Our Courses | 0.000000 |
| I agree to pay the amount through cheque | 0.000000 |
| Get updates on DM Content | 0.000000 |
| Update me on Supply Chain Content | 0.000000 |
| A free copy of Mastering The Interview | 0.000000 |
| Prospect ID | 0.000000 |
| Newspaper Article | 0.000000 |
| Through Recommendations | 0.000000 |
| Digital Advertisement | 0.000000 |
| Newspaper | 0.000000 |
| X Education Forums | 0.000000 |
| Lead Number | 0.000000 |
| Magazine | 0.000000 |
| Search | 0.000000 |
| Total Time Spent on Website | 0.000000 |
| Converted | 0.000000 |
| Do Not Call | 0.000000 |
| Do Not Email | 0.000000 |
| Lead Origin | 0.000000 |
| Last Notable Activity | 0.000000 |
| .. | .. |

Data Preparation : Data Handling

List of unique value



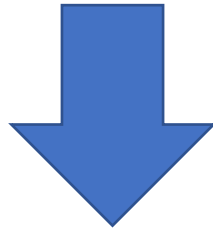
| | |
|---|------|
| Prospect ID | 9240 |
| Lead Number | 9240 |
| Lead Origin | 5 |
| Lead Source | 21 |
| Do Not Email | 2 |
| Do Not Call | 2 |
| Converted | 2 |
| TotalVisits | 41 |
| Total Time Spent on Website | 1731 |
| Page Views Per Visit | 114 |
| Last Activity | 17 |
| Country | 38 |
| Specialization | 19 |
| How did you hear about X Education | 10 |
| What is your current occupation | 6 |
| What matters most to you in choosing a course | 3 |
| Search | 2 |
| Magazine | 1 |
| Newspaper Article | 2 |
| X Education Forums | 2 |
| Newspaper | 2 |
| Digital Advertisement | 2 |
| Through Recommendations | 2 |
| Receive More Updates About Our Courses | 1 |
| Tags | 26 |
| Lead Quality | 5 |
| Update me on Supply Chain Content | 1 |
| Get updates on DM Content | 1 |
| Lead Profile | 6 |
| City | 7 |
| Asymmetrique Activity Index | 3 |
| Asymmetrique Profile Index | 3 |
| Asymmetrique Activity Score | 12 |
| Asymmetrique Profile Score | 10 |
| I agree to pay the amount through cheque | 1 |
| A free copy of Mastering The Interview | 2 |
| Last Notable Activity | 16 |

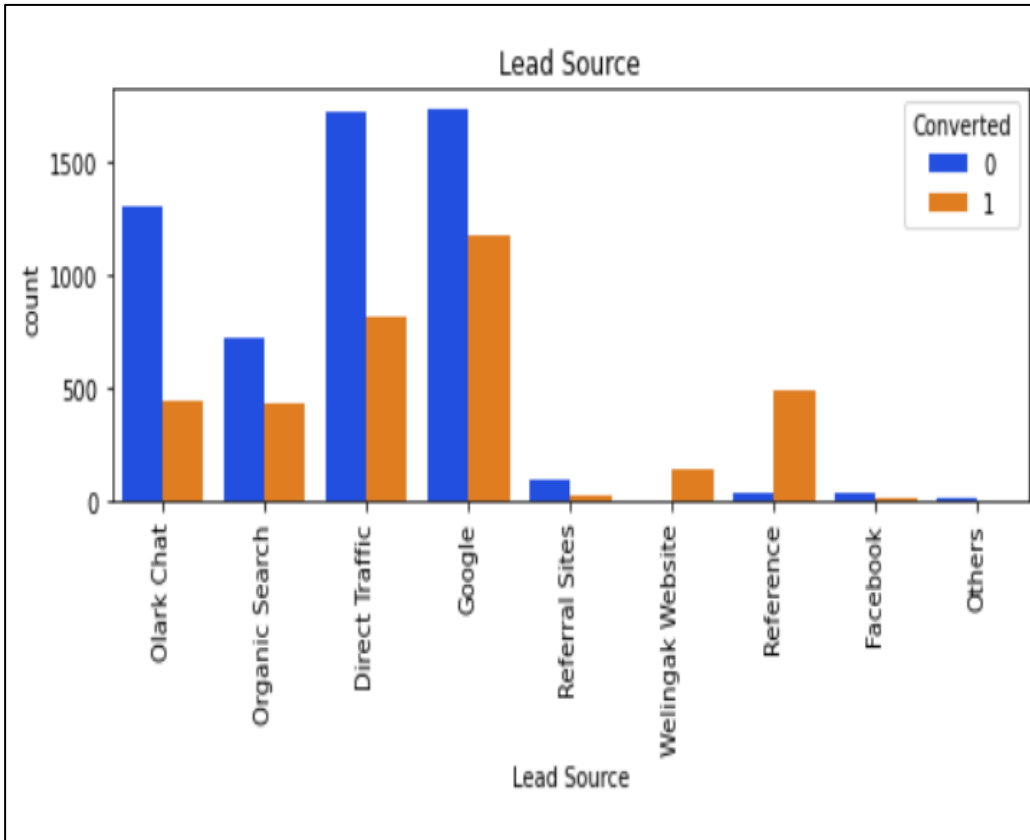
dtype: int64

Data Preparation : Data Handling

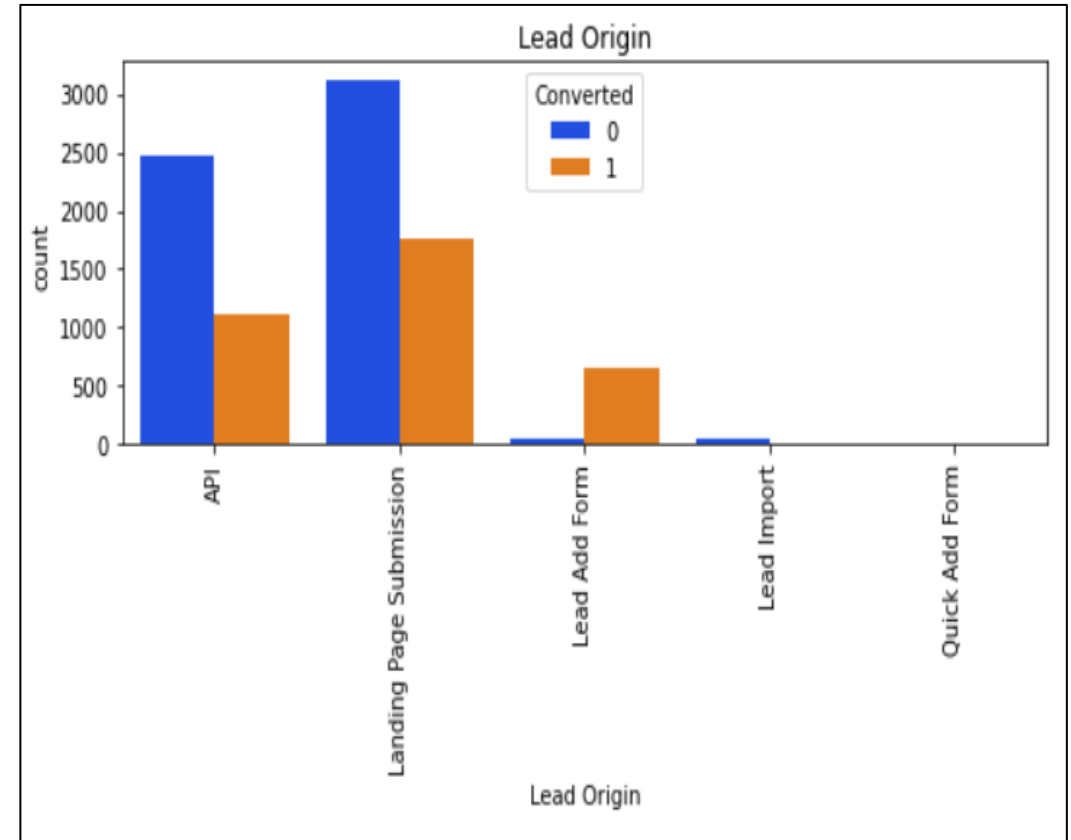
- There are some categorical features having a label as “SELECT”.
 - This means the person might not have selected any value for that field. Hence this is as good as a missing value.
- Identifying all the missing data
 - Dropped columns having more than 40% null values
- Columns have only one category of response (unique) can be dropped
- Highly skewed column can also be dropped

UNIVARIATE ANALYSIS

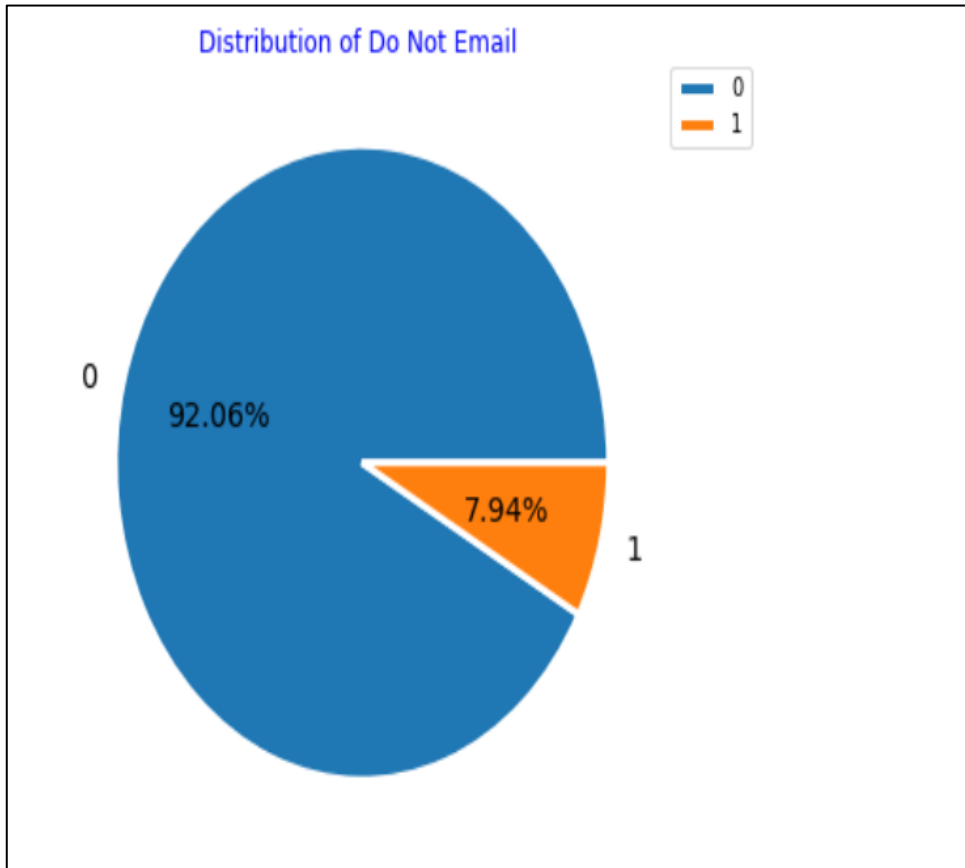




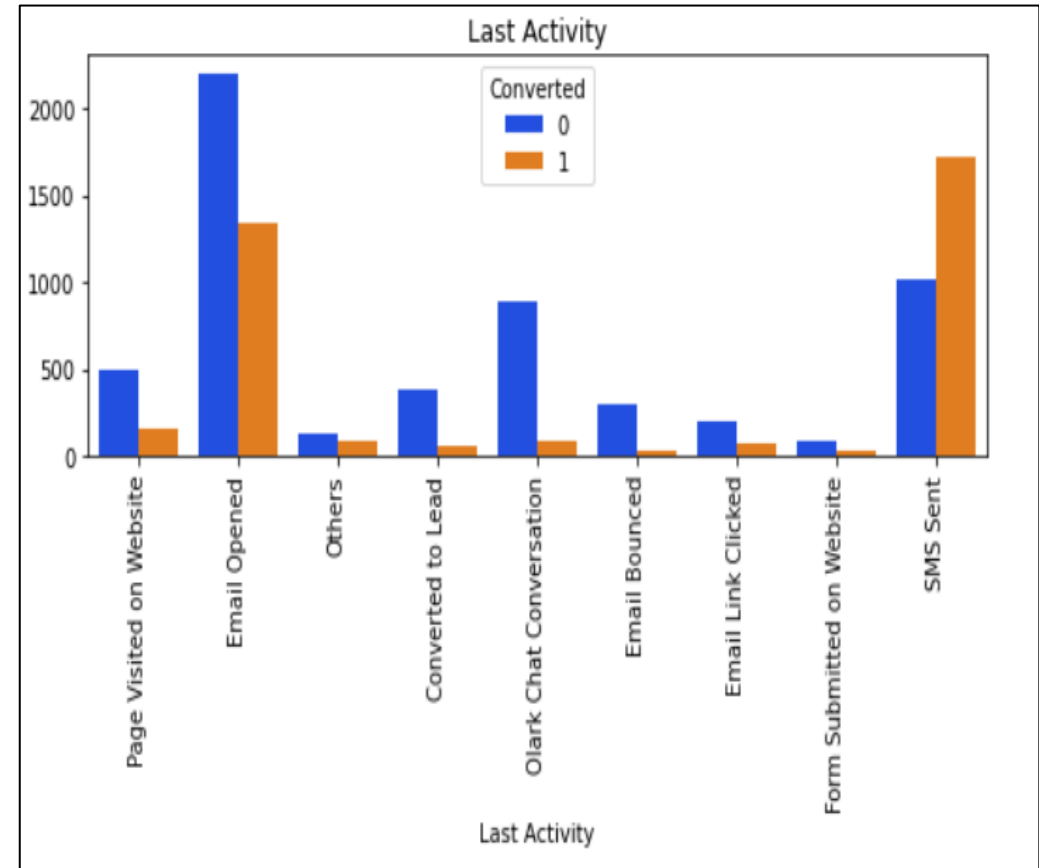
- The count of leads from the Google and Direct Traffic is maximum
- The conversion rate of the leads from Reference and Welingak Website is maximum



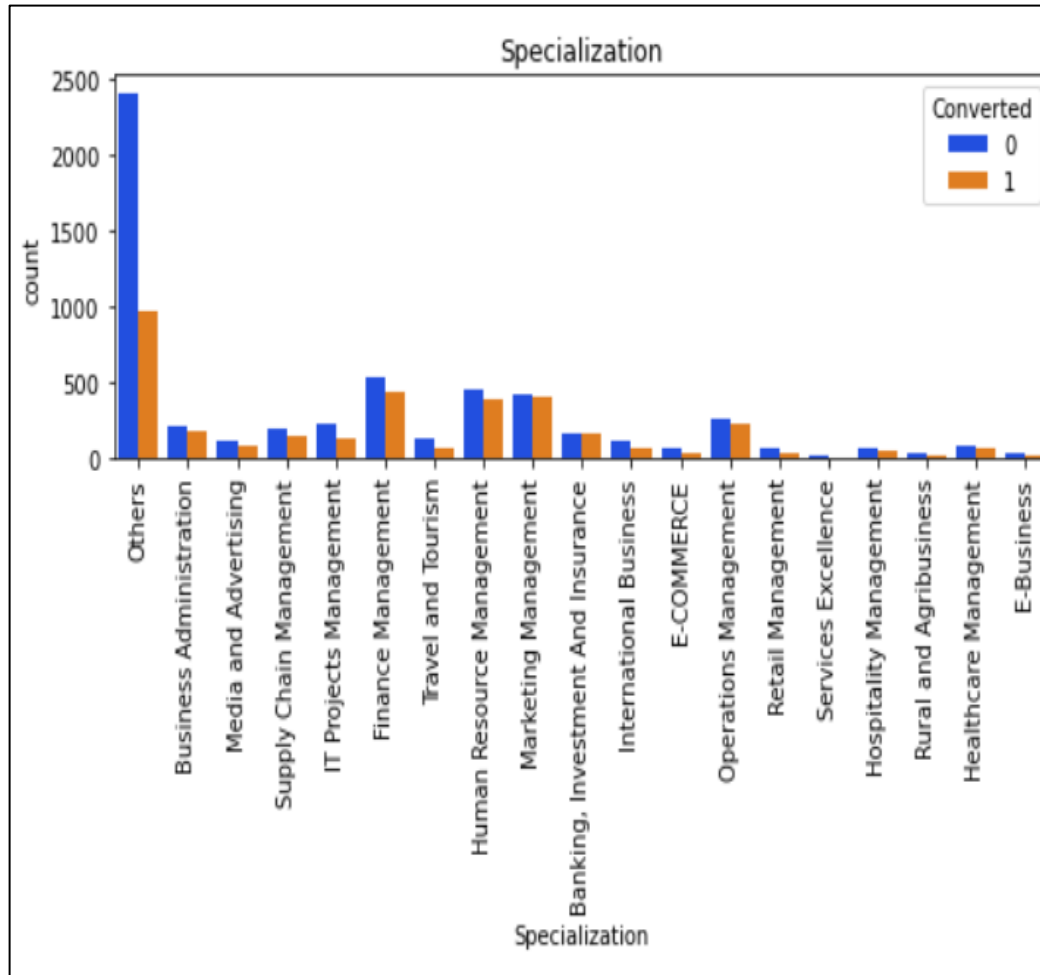
- API and Landing Page Submission has less conversion rate (~30%) but counts of the leads from them are considerable
- The count of leads from the Lead Add Form is pretty low but the conversion rate is very high
- Lead Import has very less count as well as conversion rate and hence can be ignored



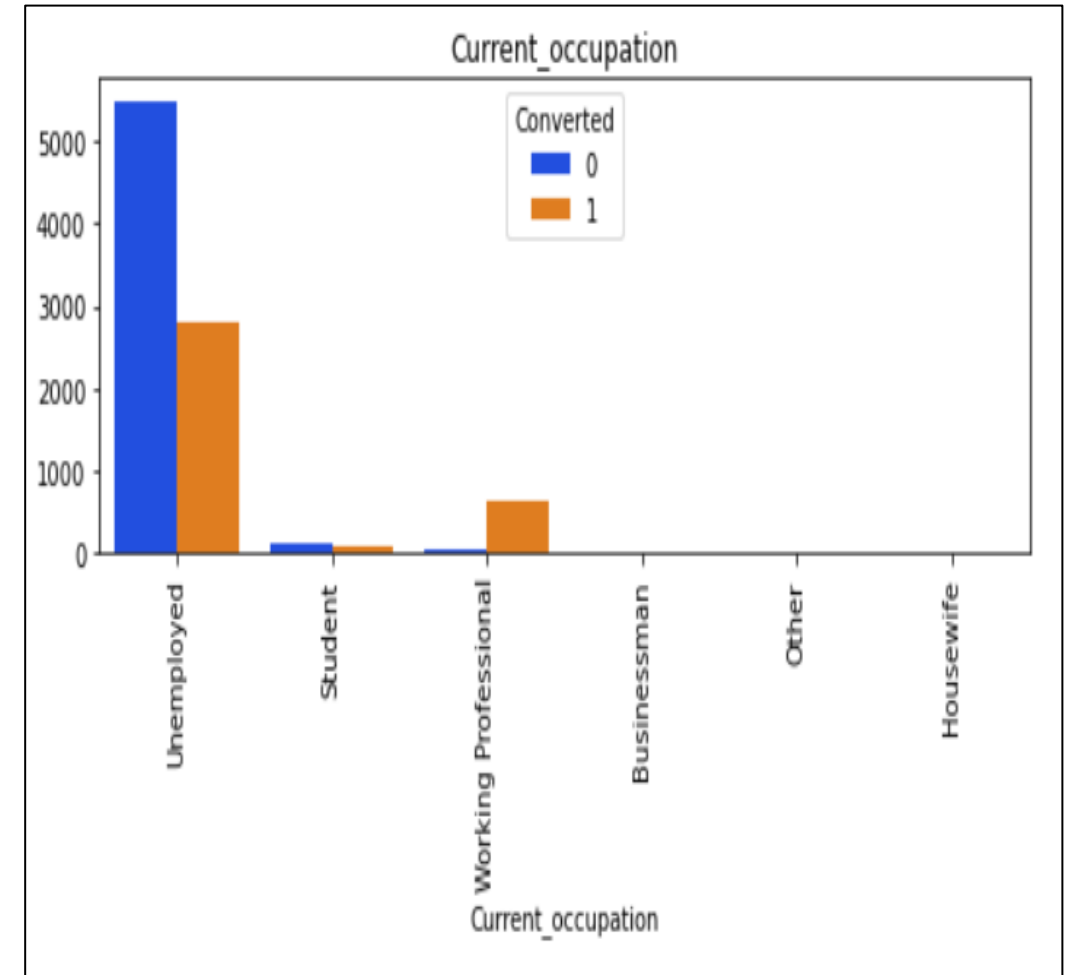
- A large proportion of customers, 92.1%, do not want to receive emails about the course.



- The count of activity as “Email Opened” is max
- The conversion rate of SMS sent as last activity is maximum

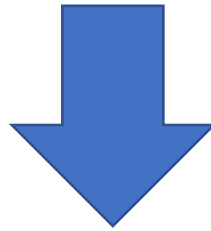


- Looking at above plot, no particular inference can be made for Specialization



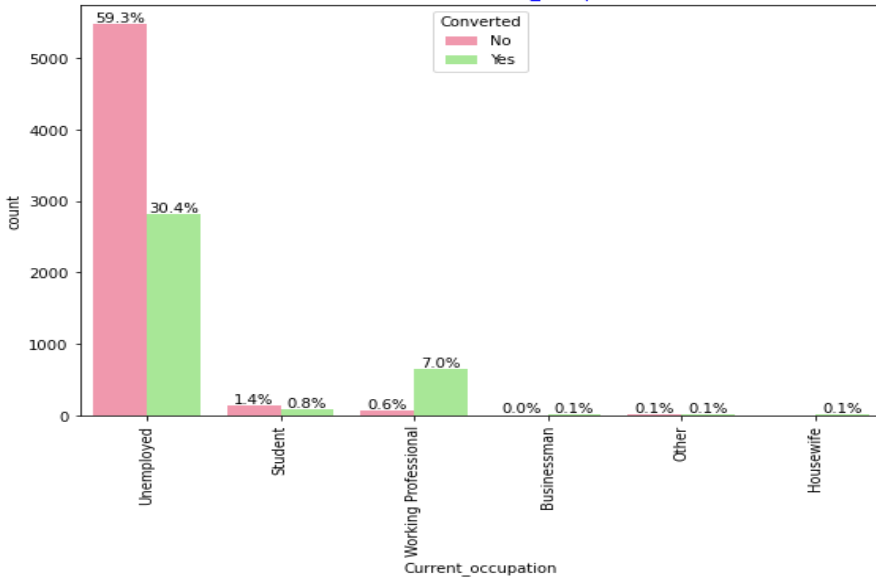
- Looking at above plot, we can say that working professionals have high conversion rate
- Number of Unemployed leads are more than any other category

BIVARIATE ANALYSIS

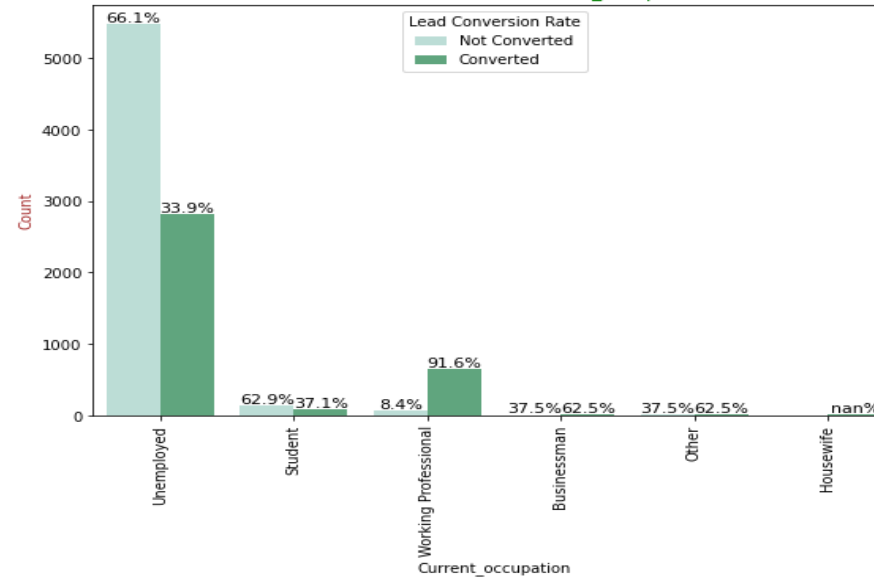


Current_occupation Countplot vs Lead Conversion Rates

Distribution of Current_occupation



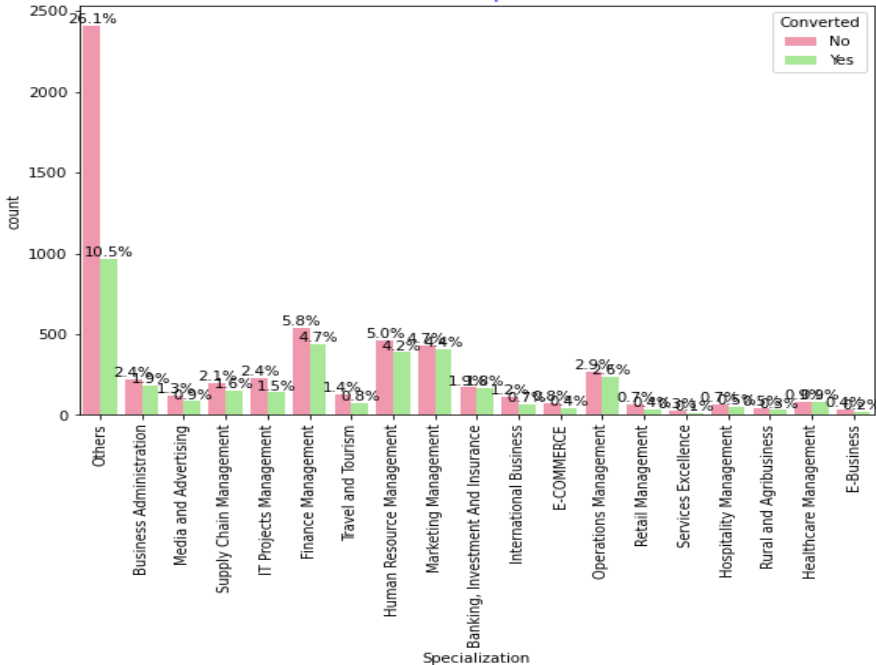
Lead Conversion Rate of Current_occupation



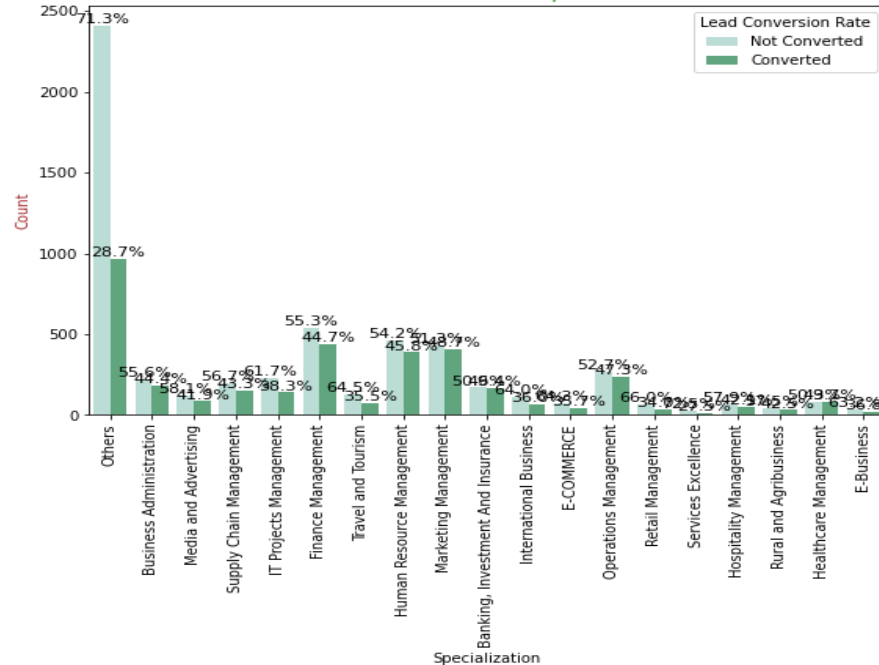
Current_Occupation: Working Professionals have a significantly higher LCR at 91.6% compared to Unemployed people at 33.9%.

Specialization Countplot vs Lead Conversion Rates

Distribution of Specialization

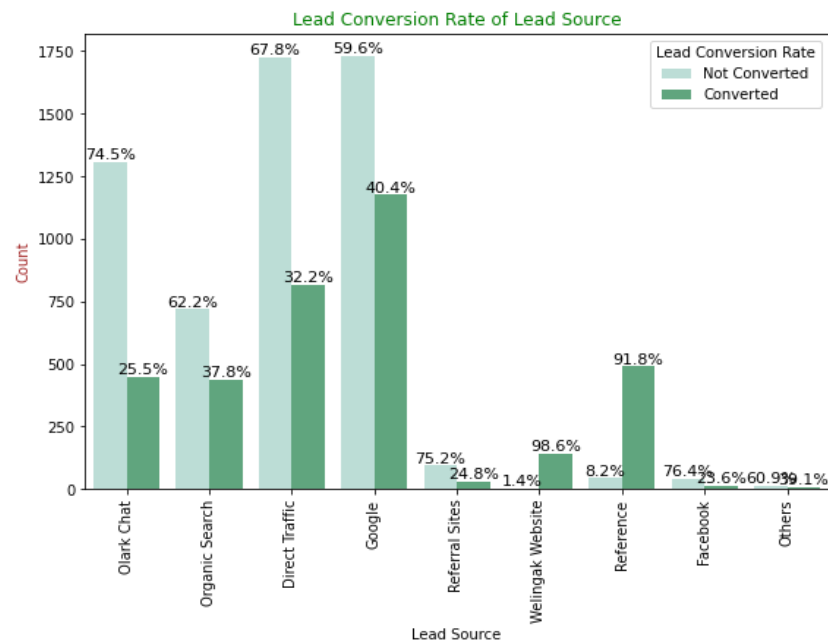
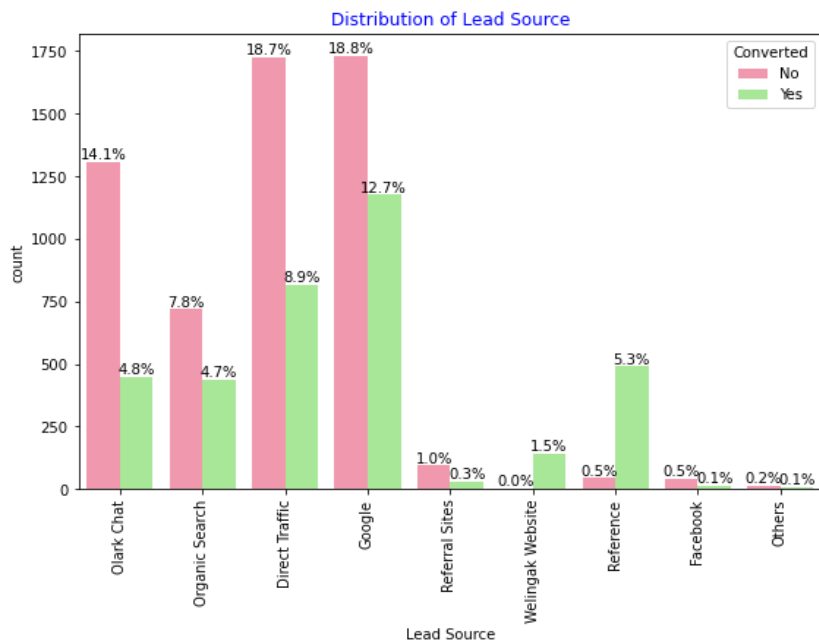


Lead Conversion Rate of Specialization



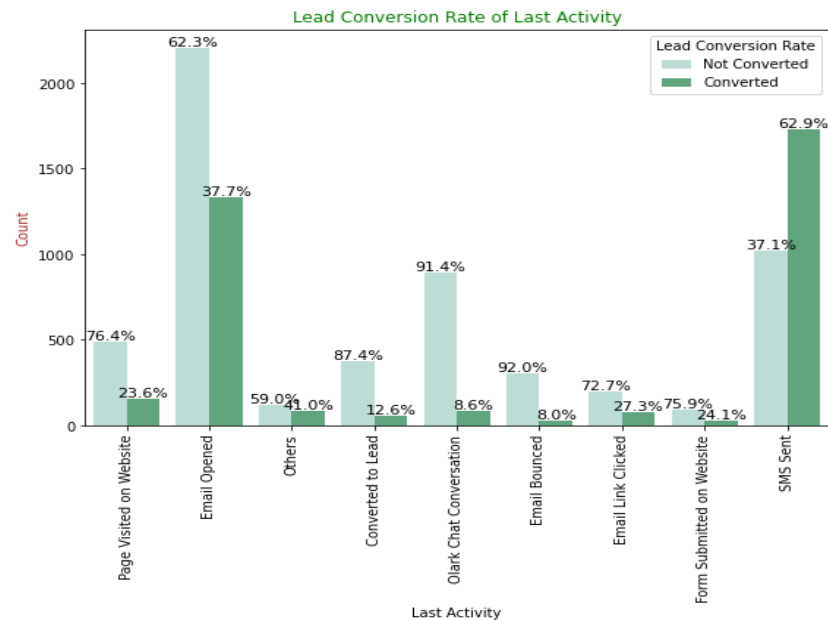
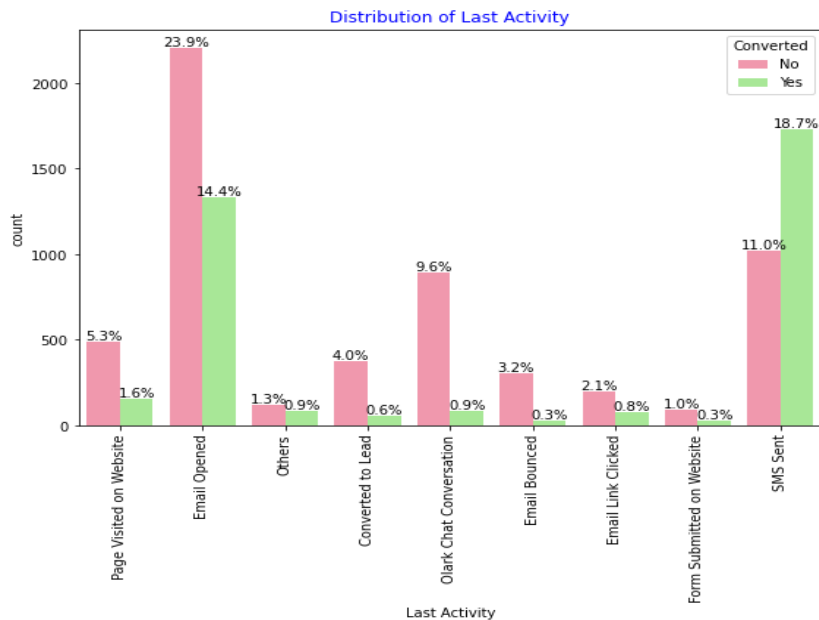
Specialization: Marketing Management, HR Management, Finance Management and Operations Management all show good LCRs, indicating a strong interest among customers in these specializations.

Lead Source Countplot vs Lead Conversion Rates



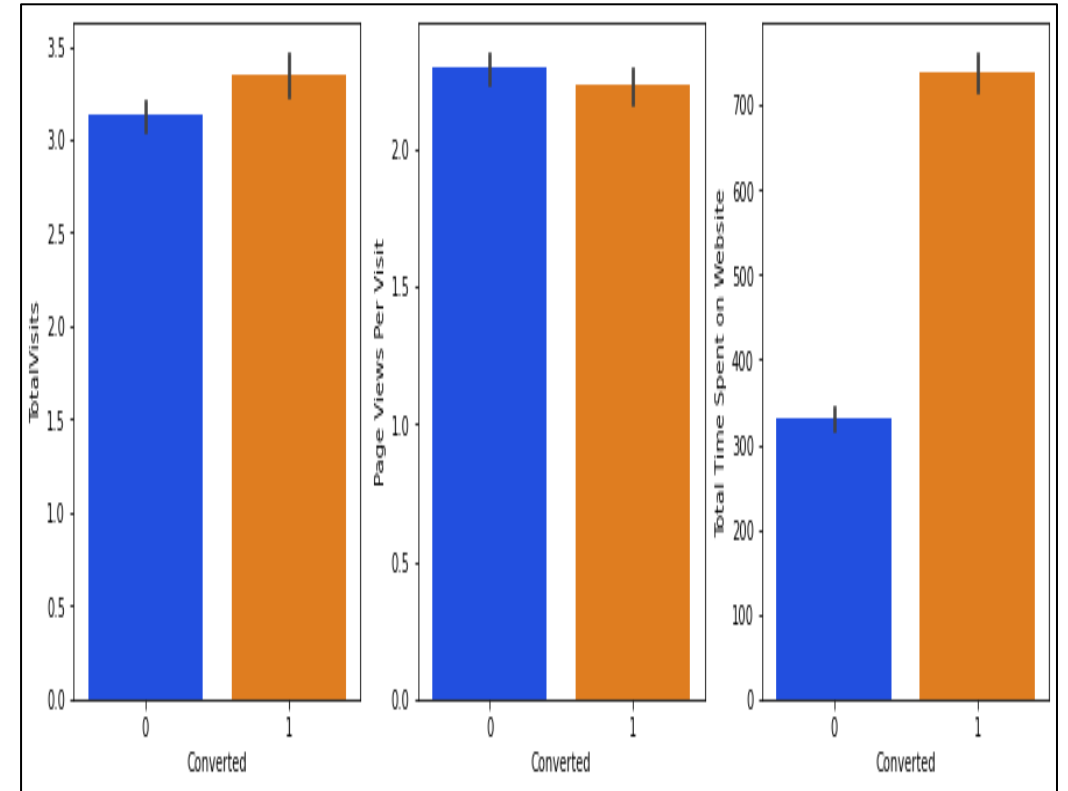
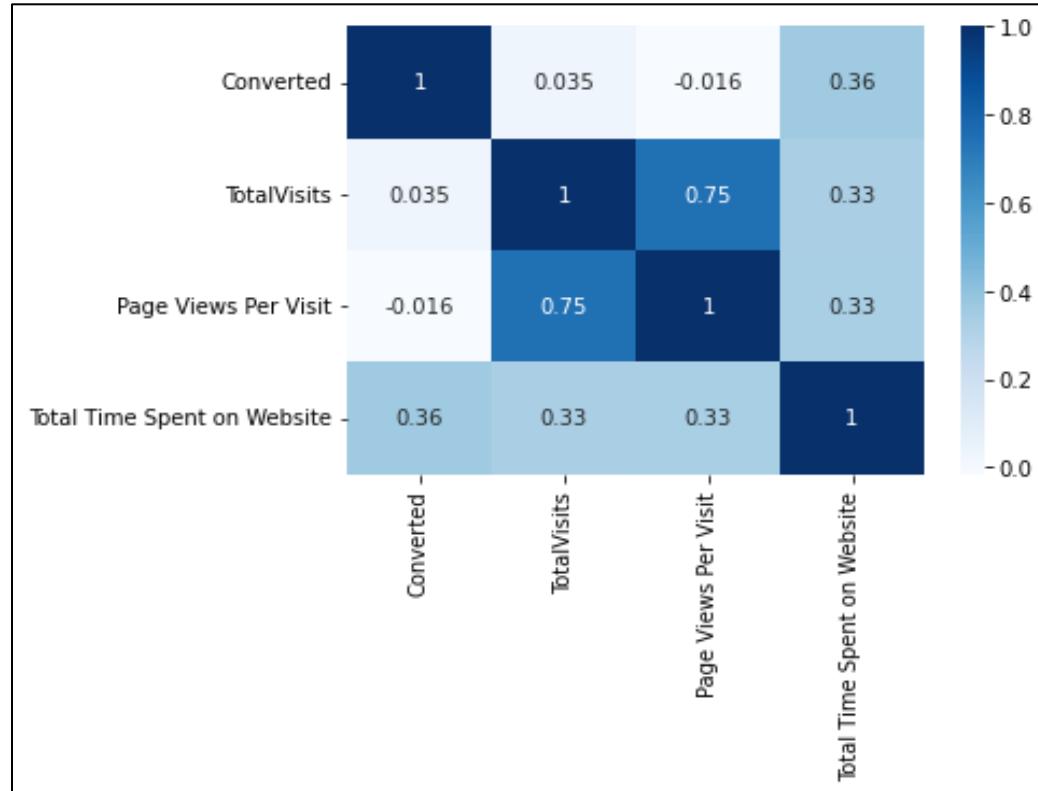
Lead Source: Google is the most effective Lead Source with an LCR of 40.4%

Last Activity Countplot vs Lead Conversion Rates



Last Activity: SMS Sent and Email Opened are the most effective Last Activity types with LCRs of 62.9% and 37.7% respectively.

CORRELATION ANALYSIS



Inference:

- There is a strong positive correlation between 'Total Visits' and 'Page Views per Visit', indicating that customers who visit the website more frequently tend to view more pages per visit.
- Customers who spend more time on the website have a higher LCR, indicating that increasing the time spent on the website can lead to higher conversion rates.

MODEL BUILDING

- Dummy Variable Creation
 - As logistic regression can work with numeric data only, creating dummy variables for the categorical columns.
- Splitting Data into Training and Test set
 - Next, the dataset was split into training and test set, to train model first with a chunk of data and then evaluate its performance on unseen data.
- Feature Scaling
 - Feature Scaling is required before Logistic Regression to bring all the features in same scale, this ensures that features with high magnitude are not given higher importance by Logistic Regression Model.
- Model Building (Feature Selection Using RFE, Improvising the model further VIF and p-values)
- Next, we are using RFE to obtain top 15 features to begin with.
- After that, manually inspecting p-values and VIF to improve the model even further.

Final Model

Logistic Regression Model - 1 is a basic model.

Manual feature reduction process was used in Logistic Regression Model - 2 and 3 to build models by dropping variables with p-value greater than 0.05.

Logistic Regression Model - 4 is stable after four iterations with:

- Significant p-values within the threshold (p-values < 0.05)
- No sign of multicollinearity with VIFs less than 5

Logistic Regression Model - 4 (LRMod4) is the final model used for model evaluation and making predictions.

Model Evaluation

- Here, accuracy is the percentage of correctly predicted labels among all the labels
- Sensitivity = $\text{True Positive} / (\text{True Positive} + \text{False Negative})$
 - Sensitivity gives us the percentage of correctly predicted conversion out of total conversions.
- Specificity = $\text{True Negative} / (\text{True Negative} + \text{False Positive})$
 - Specificity gives us the percentage of correctly predicted non-conversion out of total non-conversions.

| | |
|---------------------------|----------|
| Train Set | |
| ----- | |
| Accuracy | = 0.8057 |
| Sensitivity | = 0.7972 |
| Specificity | = 0.8108 |
| False Positive Rate | = 0.1892 |
| Precision | = 0.722 |
| Recall | = 0.7972 |
| Negative Predictive Value | = 0.8665 |
| None | |
| Test Set | |
| ----- | |
| Accuracy | = 0.8034 |
| Sensitivity | = 0.7927 |
| Specificity | = 0.8104 |
| False Positive Rate | = 0.1896 |
| Precision | = 0.7319 |
| Recall | = 0.7927 |
| Negative Predictive Value | = 0.8569 |
| None | |

Final conclusion:

- As our model predicted leads from **Lead Source_Welingak Website**, **Lead Source_Reference**, **Current_occupation_Working Professional** are likely to be converted more, so should focus on these leads
- Company should focus on leads which our model predicted as 1 and likely focus attention to them as they are potential paying customer.
- Model predicted **last_activity_SMS_sent** and **Activity_Email_opened** are potentials leads and likely to convert.
 - Tailor made SMS or Emails to potentials leads to lure them and increase chance of conversion
- **Total Spent on Website** indicates that consumer checking out website are likely to convert as it implies interest shown by consumer.
 - Enabling customized Ads or contact information on website will help company to lure consumer and increase chance of conversion
- **Lead Source Olak chat** indicates consumers are showing interest and gathering information using chat feature.
- Getting customer details and reaching out to them may turn them to potential lead.

Recommendations

To minimize the rate of useless phone calls we would suggest:

- **Work on website and application** of company so customer can navigate and surf easily. Bad UI interface or badly designed website page are huge turn off for customer
- **Chatbot** to solve minor query of customer on website or app page is added plus
- **Customized SMS and emails** to customer based on their profiles/Bio
- **Referrals and incentives**-based scheme can spread word of mouth about company.
- Also **Marketing** and making consumer aware plays huge part to garner attention of consumer
- Reaching out to existing consumer timely and getting their **feedbacks** should be prioritized

```
Out[116]: Lead_Source_Welingak Website      5.388662
          Lead_Source_Reference              2.925326
          Current_occupation_Working Professional 2.669665
          Last_Activity_SMS Sent             2.051879
          Last_Activity_Others               1.253061
          Total Time Spent on Website         1.049789
          Last_Activity_Email Opened          0.942099
          Lead_Source_Olark Chat              0.907184
          Last_Activity_Olark Chat Conversation -0.555605
          const                              -1.023594
          Specialization_Hospitality Management -1.094445
          Specialization_Others              -1.203333
          Lead_Origin_Landing Page Submission -1.258954
          dtype: float64
```

THANK YOU

