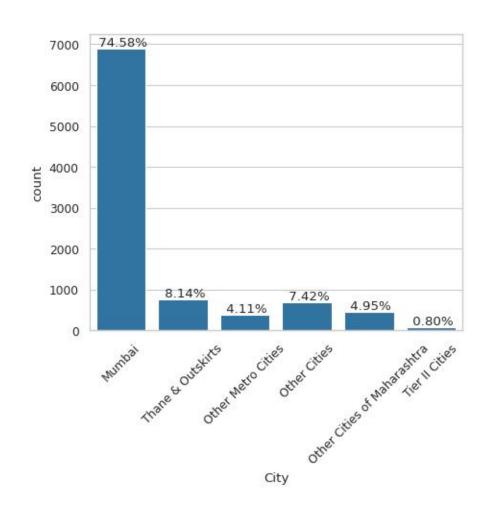
# LEAD SCORING CASE STUDY

**SUBMITTED BY-**

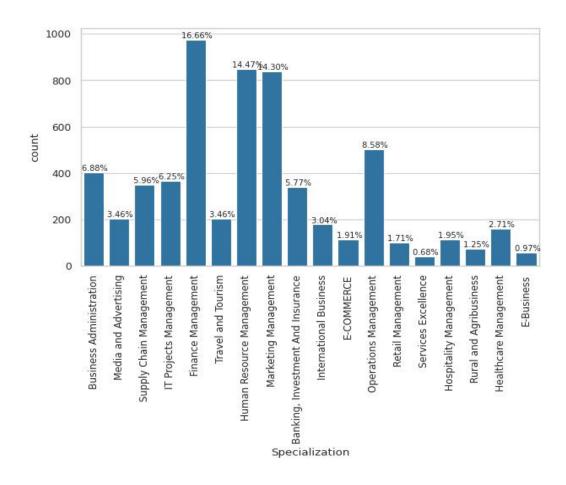
K DHEERAJ KUMAR

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## Data cleaning and EDA

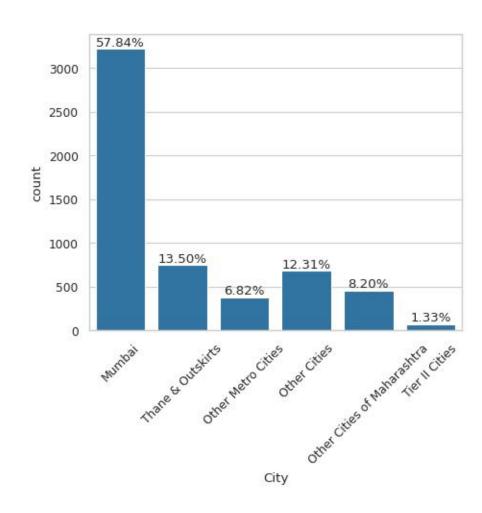


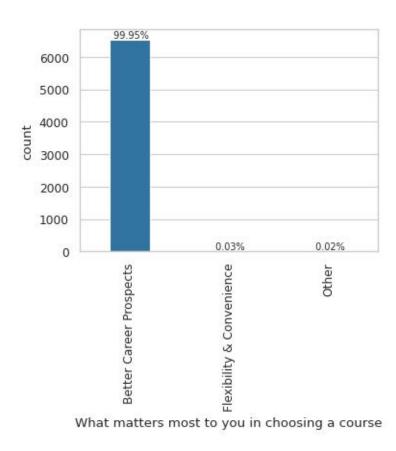
- Note: % values are calculated from total number of nows of the df
- Most values belong to mumbai
- so, we can impute missing values with mumba

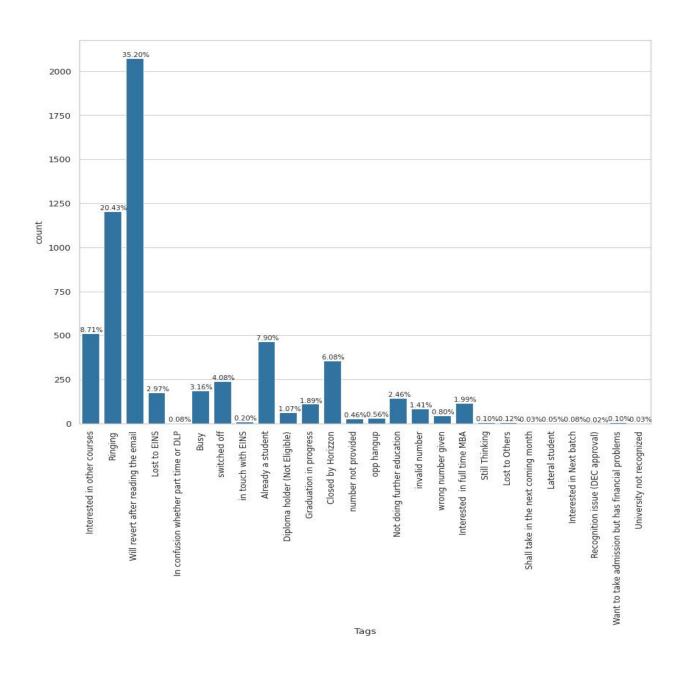


 There are many specializations present here and the missing values are also quite high, therefore we can create another category 'others

## data cleaning

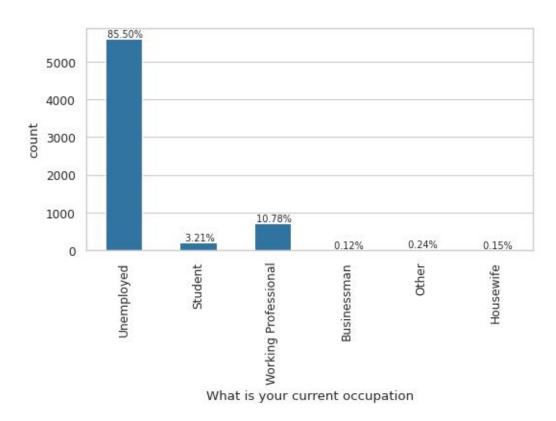




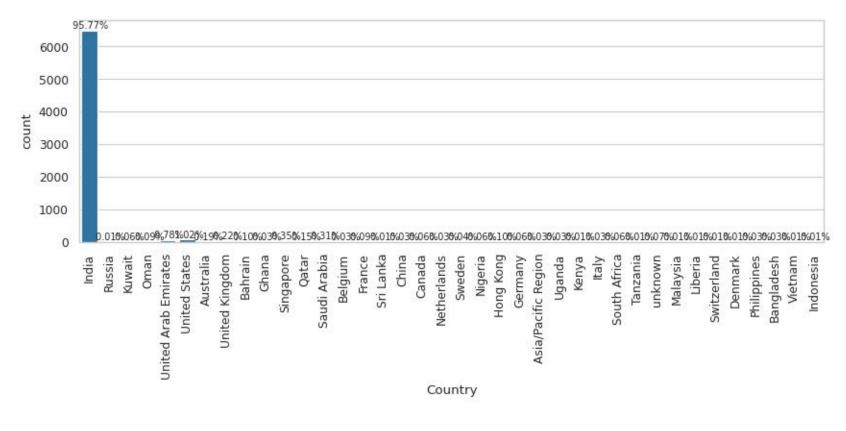


most of the values are 'we will revert after reading the email'

so we can impute the missing values with this

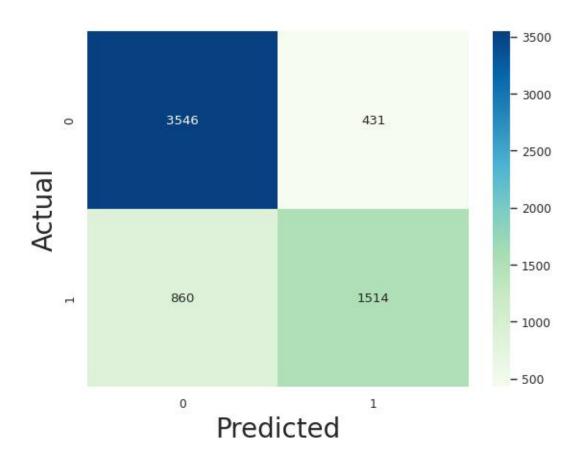


- most of them are unemployed
- so imputing with unemployed



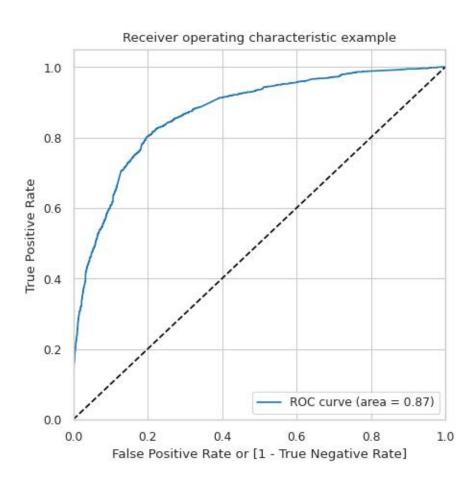
- very skewed plot, but as it is country, it could be an important variable
- so imputing the missing values with india

## Confusion Matrix



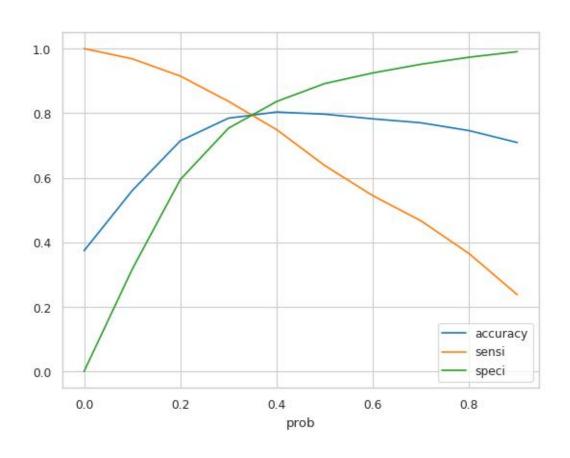
- the model performs well because true positive and true negative values are much greater than false positive and false negative values
- our model stron accuracy but struggles with recall because it misses many actual positive cases

#### **ROC Curve**



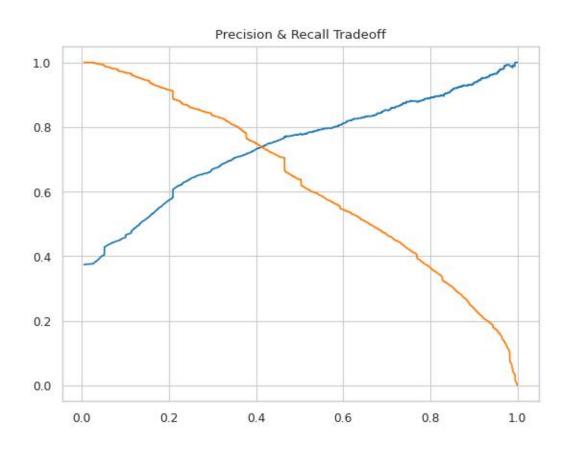
- this shows that the model is able to achieve high true positive rates with relatively low false positive rates
- the area under the curve=0.87 represents that the model has good ability to distinguish between the positive and the negative classes

## Finding optimum cutoff point



- the optimal cutoff point can be seen at around 0.35
- this can be the optimal point for balancing both metrics for optimal model performance

### Precision and Recall tradeoff



- Precision and recall have an inverse relationship.
- At lower thresholds, the model captures more true positives (higher recall) but also makes more false positive errors (lower precision).
- At higher thresholds, precision improves at the cost of recall

## variables and their coefficients of the final model

Lead Source_Welingak Website	5.717363
What is your current occupation_Working Professional	2.795847
Last Activity_SMS Sent	2.050196
Last Activity_Others	1.283260
Last Activity_Email Link Clicked	1.239169
Total Time Spent on Website	0.962912
Last Activity_Email Opened	0.858354
What is your current occupation_Student	0.562399
Lead Source_Olark Chat	0.361796
const	0.014192
Last Notable Activity_Modified	-0.721922
Last Notable Activity_Email Link Clicked	-0.928944
Specialization_Others	-1.706168
Lead Origin_Landing Page Submission	-2.081977

#### Recommendations

#### Call these leads:

- Those from "Welingak Websites" and "Reference" because they are likely to convert.
- Working professionals, as they are also likely to convert.
- Leads who spent more time on the website, since they are more likely to convert.
- Those from "Olark Chat" because they have a higher chance of converting.
- Leads whose last activity was an SMS sent, as they are more likely to convert.
- Do not call these leads:
- Those whose last activity was an "Olark Chat Conversation," as they are unlikely to convert.
- Leads from "Landing Page Submission," since they are also unlikely to convert.
- Leads whose specialization is "Others," as they are less likely to convert.
- Leads who chose "Do not Email" as "yes," since they are unlikely to convert.