



Lead Score Assignment

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Problem Statement



An education company named X Education sells online courses to industry professionals.



Many professionals who are interested in the courses land on their website and browse for courses.



Once these people land on the website, they might browse the courses or fill up a form for the course or watch some videos.



Once these leads are acquired, employees from the sales team start making calls, writing emails, etc.



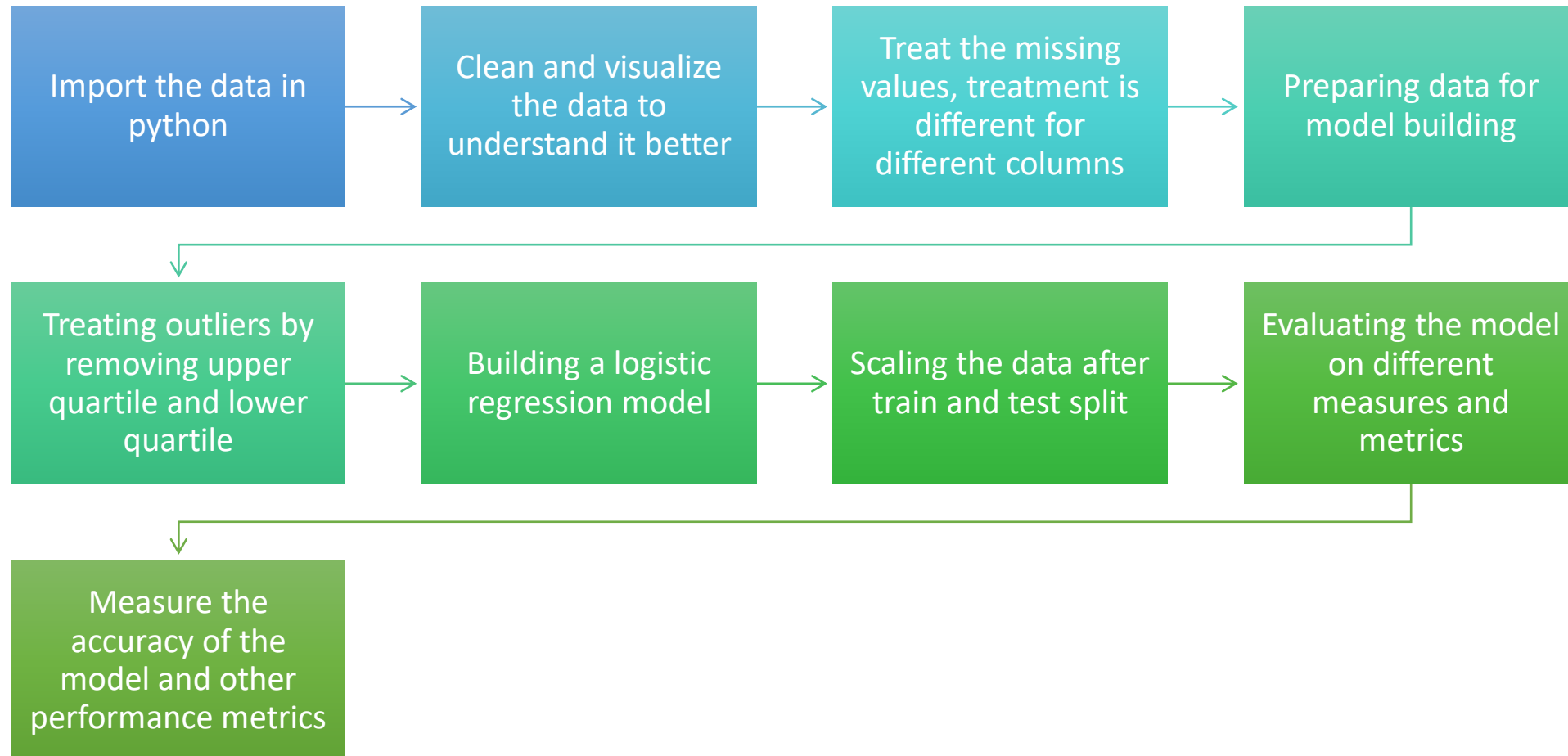
Through this process, some of the leads get converted while most do not.

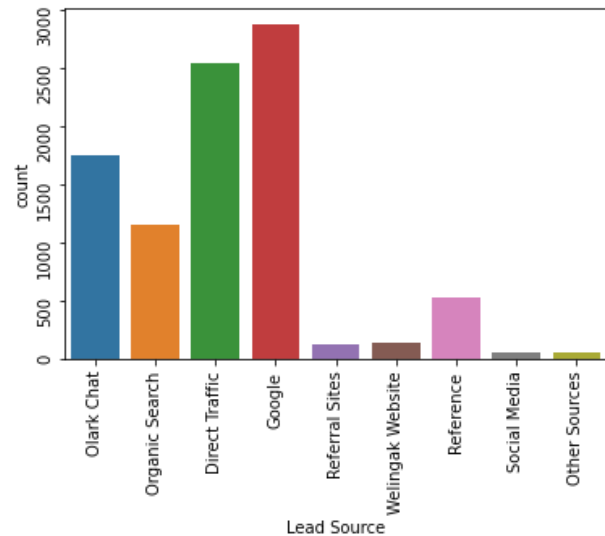
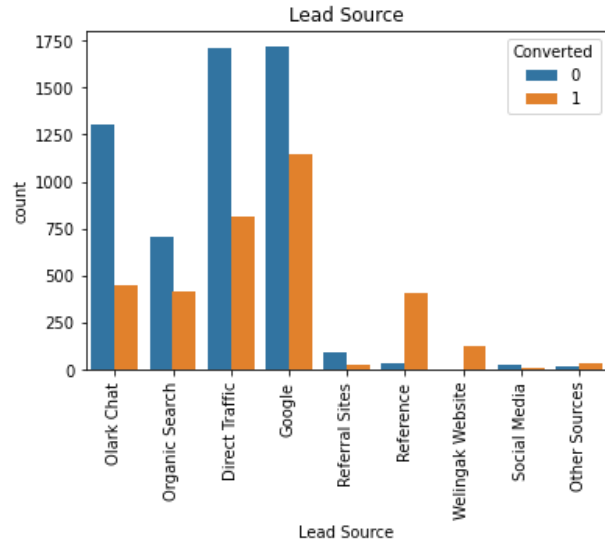


OBJECTIVE

To make this process more efficient, the company wishes to identify the most potential leads, also known as 'Hot Leads'. The company requires to build a model wherein you need to assign a lead score to each of the leads

Approach



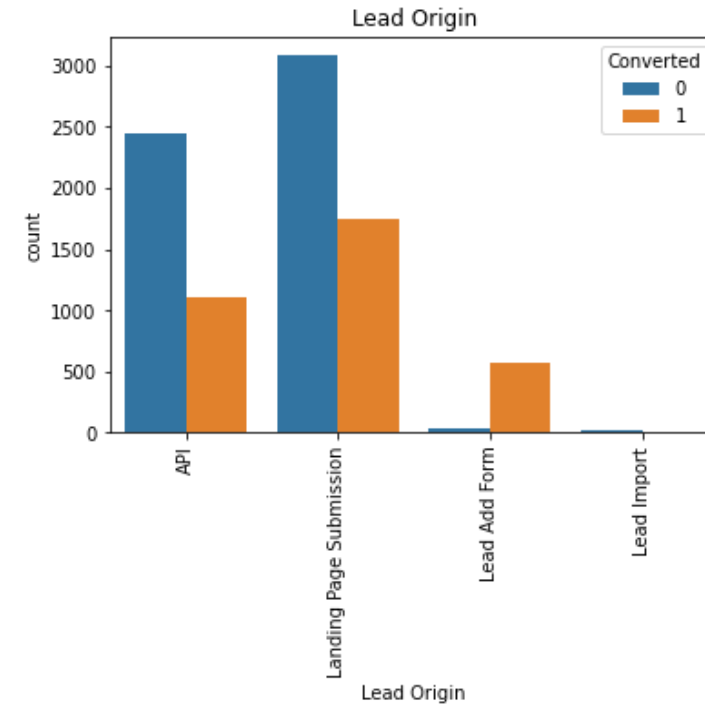
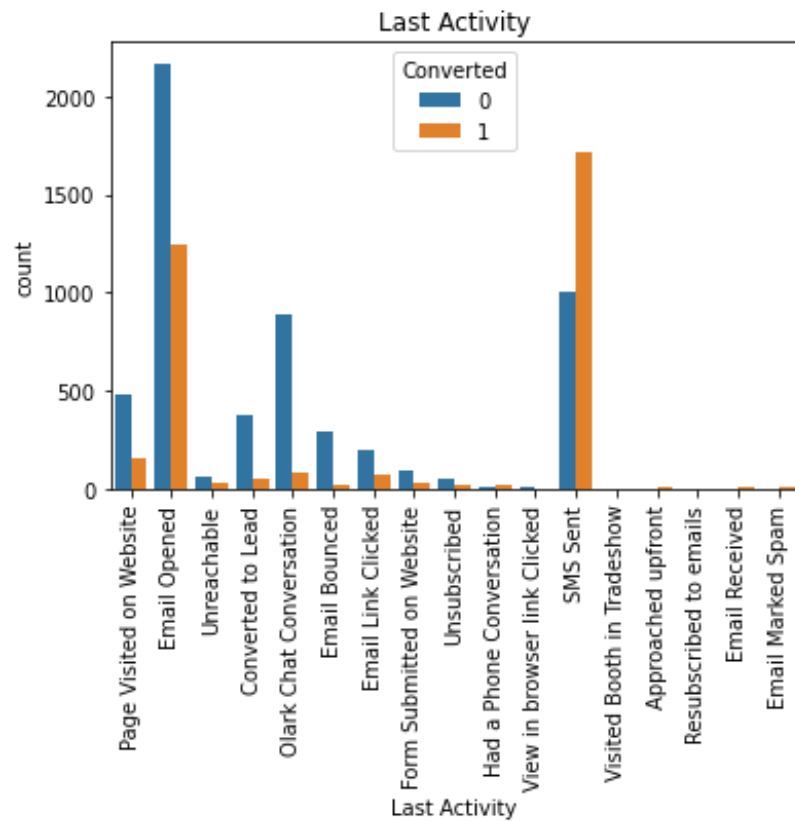


Univariant and Bivariant analysis

- Google searches have the highest conversion rate as compared to other options

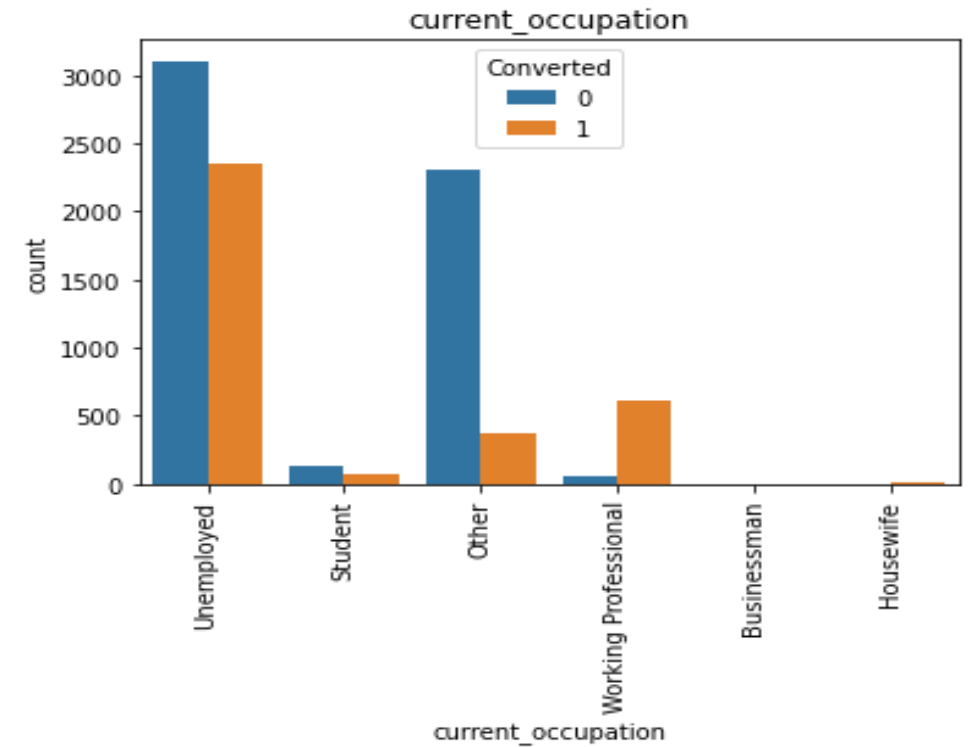
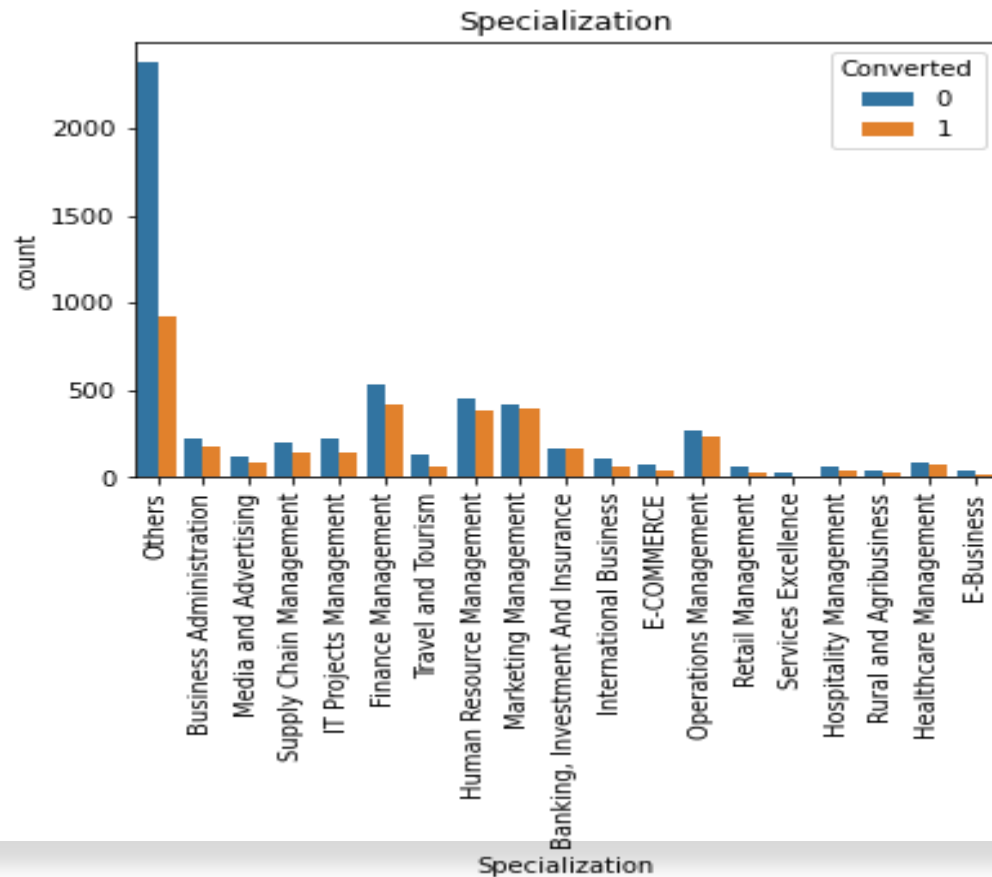
Direct traffic is next and then organic and Olark chat

Social media and reference sites fall behind



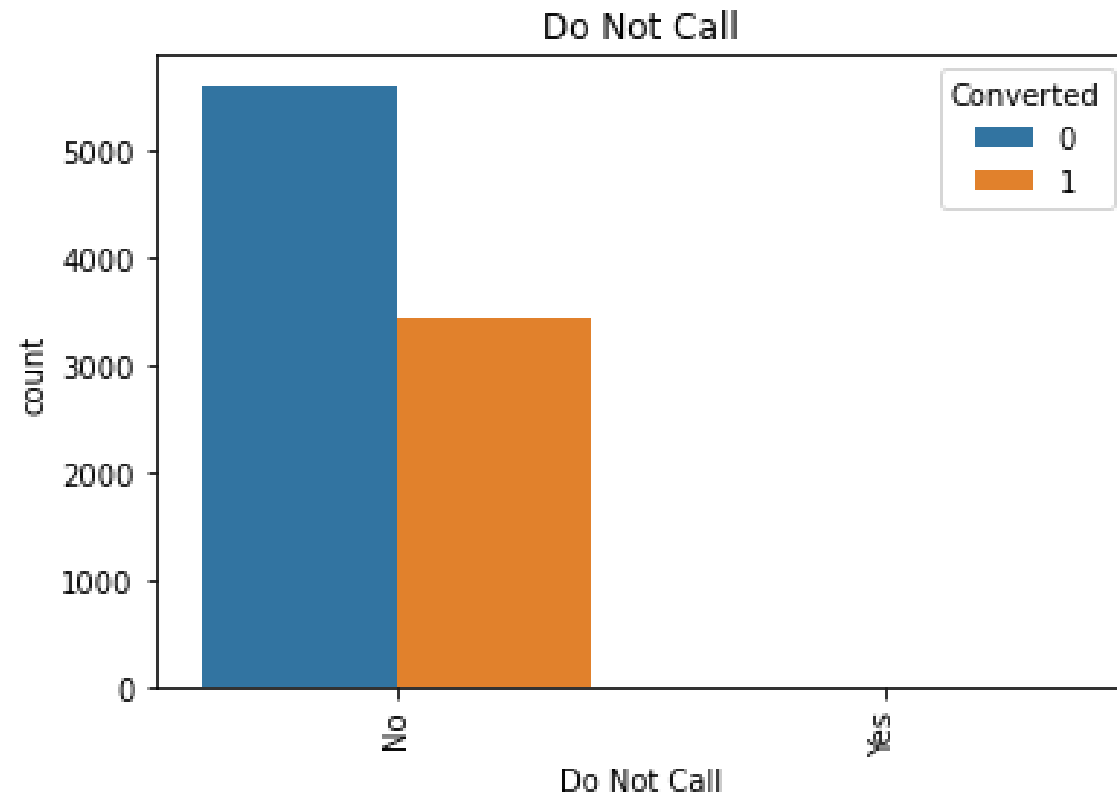
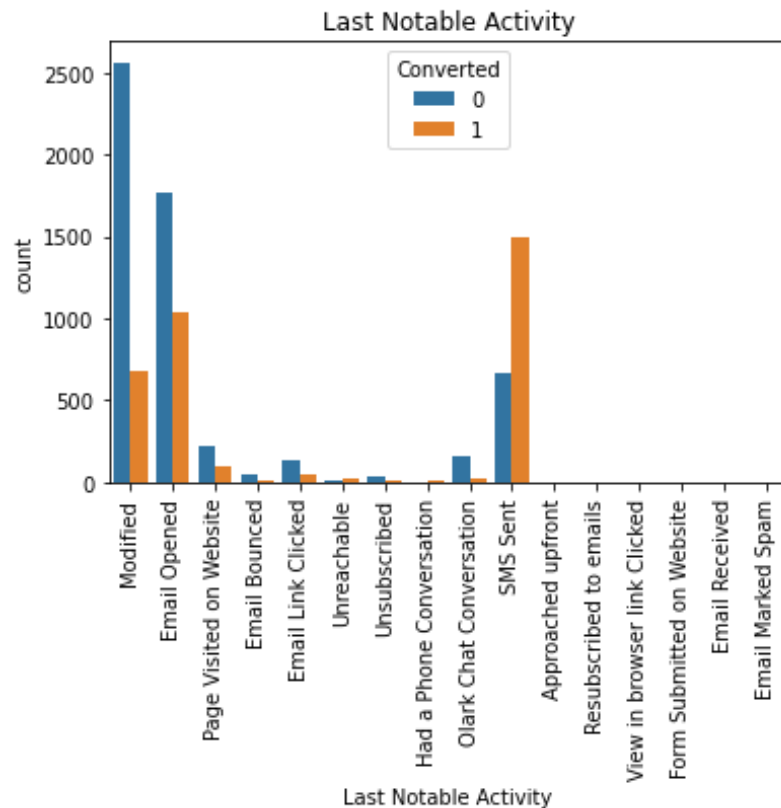
Last Activity and Lead Origin

- Customers who communicated via SMS have highest conversion rates
- Customers who end up submitting landing page and add form end up converting
- Lead import has less conversion rate



Specialization and Current Occupation

- Specialization of majority is not known so not a very clear picture, however
- Finance management , Marketing management specialization stream customers have high conversion rate
- Unemployed are more likely to convert to upskill themselves for the job market

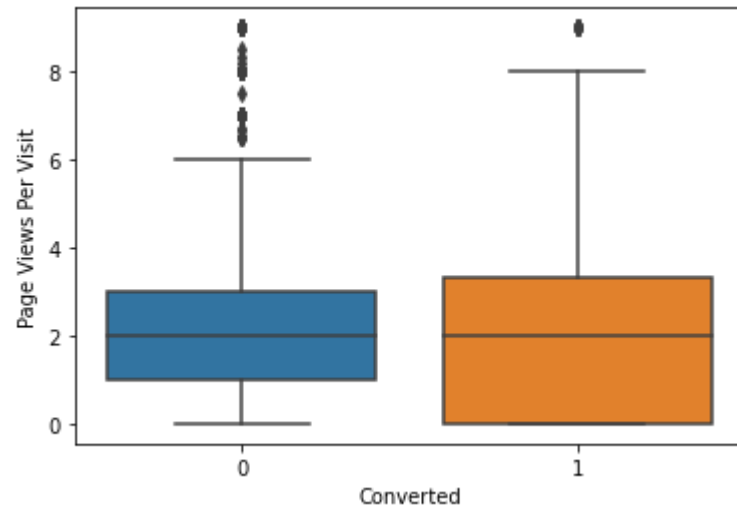


Last Notable Activity and Do not Call

- SMS sent as the last communication with the customers have high lead score conversion rate
- Most converted customers did not like to be called

Steps followed during data cleaning

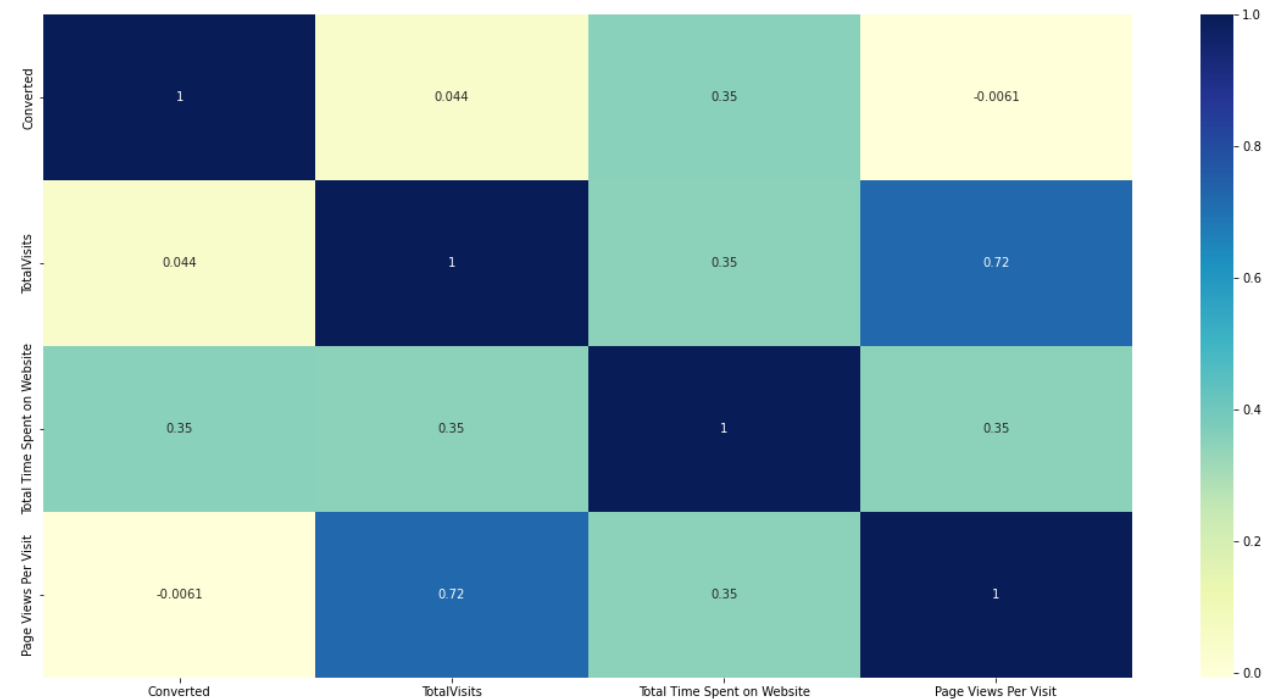
- Duplicate rows were checked
- % Null values for each column were calculated
- Those columns having more than 45% null values were dropped
- Lead Source null values were saved as other sources
- Country column was reassigned as India, Other Countries, USA, UAE
- Specialization column null values were assigned as not known
- Now no of unique values were calculated for each categorical column
- The ones with only one unique value are dropped as they are not contributing to the model building
- Reason for course nan values reassigned as not known
- Id and Lead number column are dropped as they are not adding value
- Source of information column nan values are categorized as not known
- Various columns have Select options also they are converted to not known or others in some cases
- Remaining columns with less than 1% missing values the respective rows were dropped
- City and country columns were also taken care of by categorizing them to not known category



More time spent on website leads to the Conversion of customers

Mean is same for both the cases however large no of ppl who visited the page / no of views, get enrolled to the program

Outliers were treated by removing top and bottom 1 % of the data



Model Building



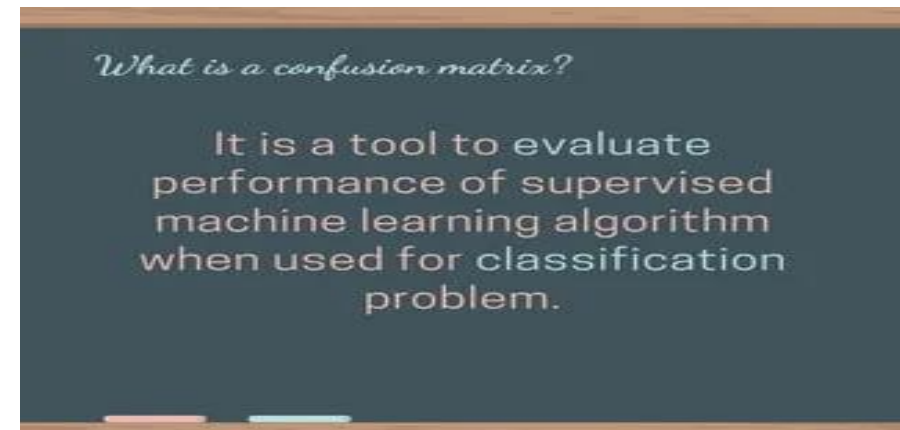
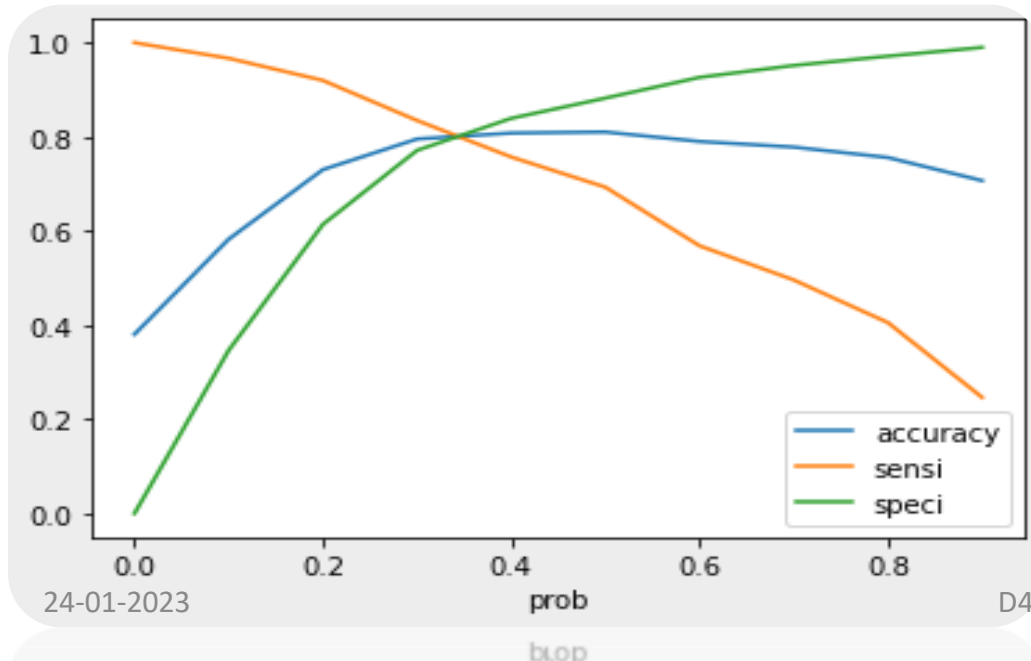
Final Features selected

Features		VIF
8	lead_quality_not_sure	3.24
0	totalvisits	2.86
1	total_time_spent	2.15
7	lead_quality_might_be	1.90
2	lead_source_olark_chat	1.74
12	last_notable_activity_sms_sent	1.52
6	lead_quality_low_in_relevance	1.24
9	lead_quality_worst	1.19
3	lead_source_reference	1.13
5	do_not_email_yes	1.12
4	lead_source_welingak_website	1.06
10	asymmetrique_activity_index_low	1.06
11	last_notable_activity_olark_chat_conversation	1.05
13	last_notable_activity_unreachable	1.01

Model 1 at threshold of 0.5

- Accuracy: 84.89%
- Sensitivity(Recall): 74.52%
- Specificity: 66.98%
- Precision: 83.52%
- F-Score: 78.77%

Actual\Predicted	No	Yes
No	3578	348
Yes	603	1764



Threshold choosen as 0.3

Model 2: Threshold as 0.3

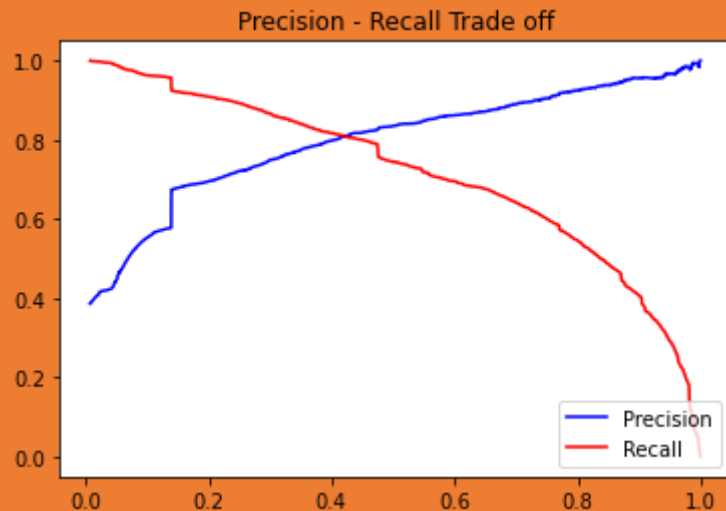
Actual\Predicted	No	Yes
No	3384	542
Yes	405	1962

Clearly, at the cut-off of 0.37, we have achieved an accuracy of 85%, sensitivity(Recall) of 83%, specificity of 63%, precision of 78.3%, and a F-score of 0.8055.

- Accuracy:84.95%
- Sensitivity(Recall): 82.89%
- Specificity: 63.30%
- Precision: 78.35%
- F-Score: 80.56%

Test dataframe
model performance
evaluation matrix

Actual\Predicted	No	Yes
No	1415	251
Yes	169	863



24-01-2023

Accuracy: 84.43%

Sensitivity(Recall): 83.62%

Specificity: 62.11%

Precision: 77.47%

F-Score: 80.43%

Clearly, with the probability cut-off of 0.37, we have achieved an accuracy of around 84.6%, sensitivity(Recall) of 83.6%, specificity of around 62%, precision of around 78% and F-score of 0.806

D46 July 2022 batch

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Final Model: Cut off probability at 0.41

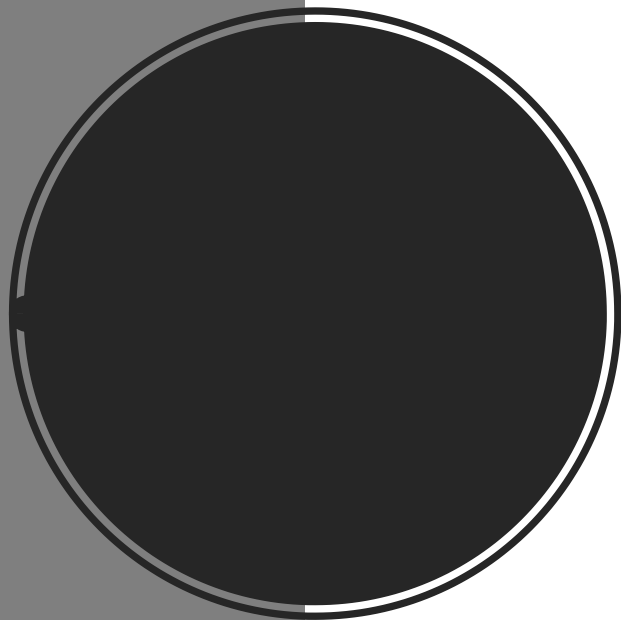
Train Data

- Accuracy: 85.44%
- Sensitivity(Recall): 81.37%
- Specificity: 64.18%
- Precision: 80.22%
- F-Score: 80.79%

With cutoff probability at 0.41, we have an accuracy of 85%, sensitivity(recall) of around 81%, specificity of 64.2%, precision of 80.3% and F-score of 0.808

- Test Data
- Accuracy: 84.95%
- Sensitivity(Recall): 81.98%
- Specificity: 63.09%
- Precision: 79.36%
- F-Score: 80.65%

At a threshold probability of 0.41, we have achieved an accuracy of around 84.9%, Sensitivity(recall) of around 82%, Specificity of 63%, Precision of 79.4%, and F-score of around 0.807



No.	Actual outcome	Predicted outcome	lead_score
0	0	0	31
1	1	1	98
2	1	0	9
3	1	1	59
4	0	0	0

Clearly, the model is able to predict the Conversion rate successfully. An overall accuracy of about 85% on our logistic regression model suggests there is an 85% chance that our predicted leads will be converted. This meets our CEO's target of atleast 80% lead conversion.

Conclusion

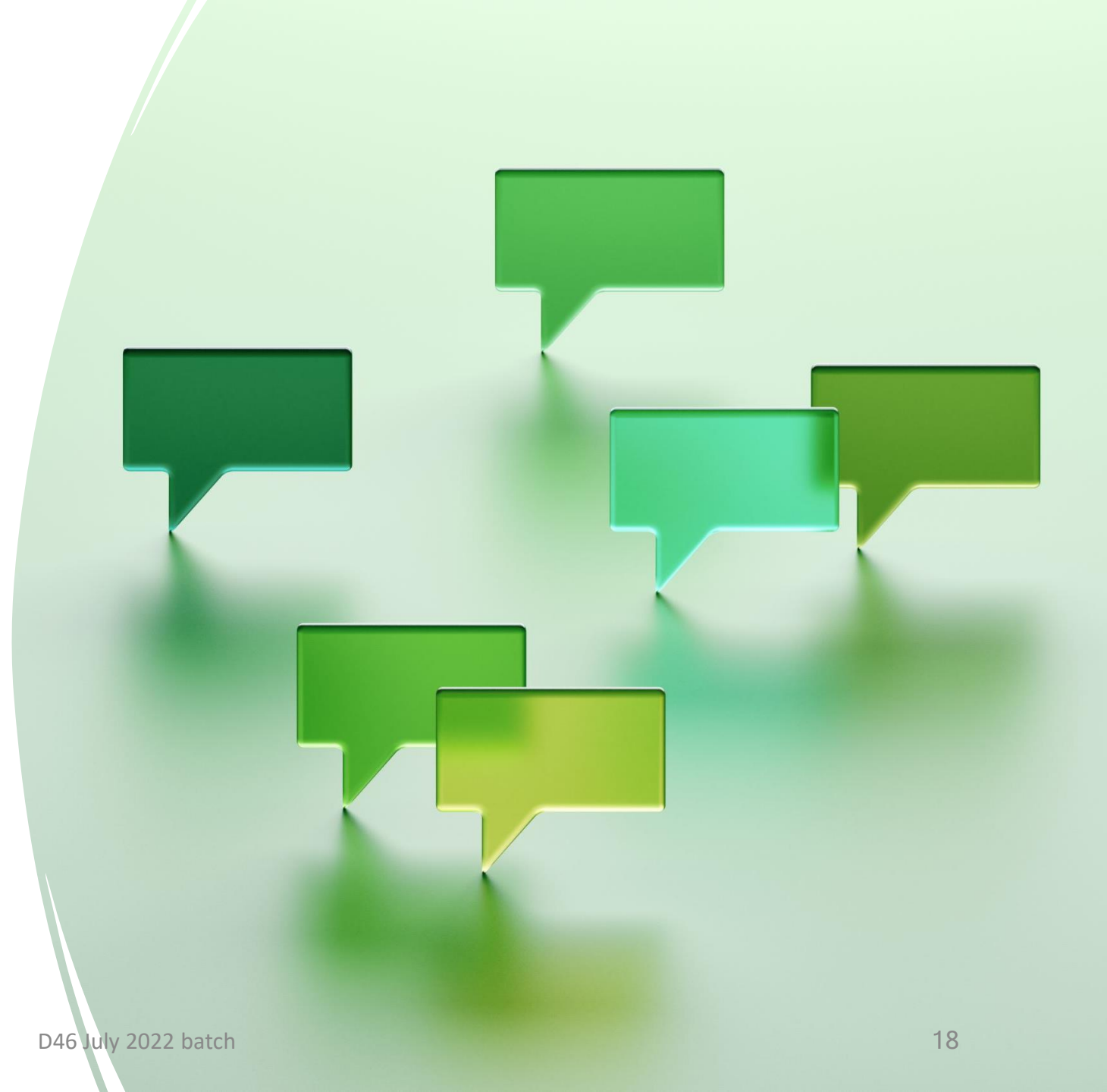
The top three variables are:

- ✚ Total time spent on website
 - i. The total time spent by the customer on the website.
 - ii. The more the time spent means more likely to convert.
- ✚ Lead Origin
 - i. The origin identifier with which the customer was identified to be a lead. Includes API, Landing Page Submission, etc.
 - ii. If the lead fills out a lead add form more likely to convert
- ✚ Lead source
 - i. The source of the lead. Includes Google, Organic Search, Olark Chat, etc.
 - ii. Of these ppl contacting through Olark chat are more likely to convert

Conclusion

Top dummy variables are:

- + Lead Origin- Lead Add form :
Customers filling out add forms are more likely
- + Lead Source – Olark Chat :
Customers engaging in an Olark chat window are more likely to convert
- + Last activity – SMS sent: If customer has sent an SMS, more likely to convert



Suggestions

- ❖ First and foremost, we should arrange the customers based on the amount of time spent by them on the website. The more the time spent, strong likelihood of them converting to avail the course, so continuous follow up and messages should be sent.
- ❖ Targeting the customers who engage on the chat or have come through Welingak Website. Those customers who have last communicated through SMS are also most likely to get converted so need to be followed up regressively. The queries resolution should be quick via SMS and a quick call-in case some more information needs to be communicated.
- ❖ The students at some school, and those customers who have marked do not contact via email, are very unlikely to be converted. So, resources need not be wasted on following up, however basis information asked should be communicated promptly.

The customers spending long time, messaging via SMS, filling out the ADD form and chatting on Olark chat are most likely to get enrolled to the course. So, the website experience should be good and smooth.

Suggestions for better customer enrolment rate



Make groups and follow up based on the categories of customers like strong leads, not so strong and weak leads. Make email draft to be sent to these categories



Keep the draft emails short and engaging referring to the alumni achievements or where are they placed.



Catch the attention by keeping some free engaging sessions on the subject matter.



Consider using an informal and personal touch by sighting the same case scenario.



Never assume whether its worth deleting cold leads, let the customer tell that themselves.



Keep prerecorded voicemails and email templates for various scenarios that can be sent with little tweak.