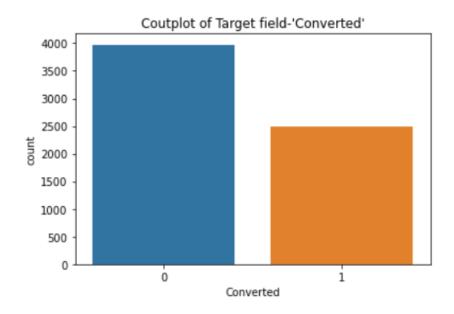
Presentation

• We have ~40% customers that are likely to be converted.

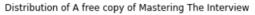


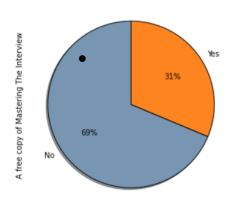
]: train_new['Converted'].value_counts(normalize=True)

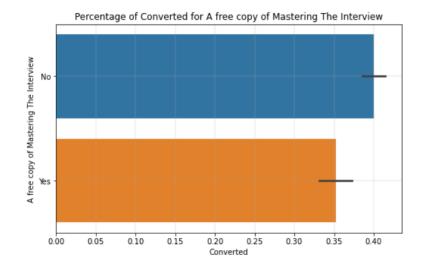
]: 0 0.614564 1 0.385436

Name: Converted, dtype: float64

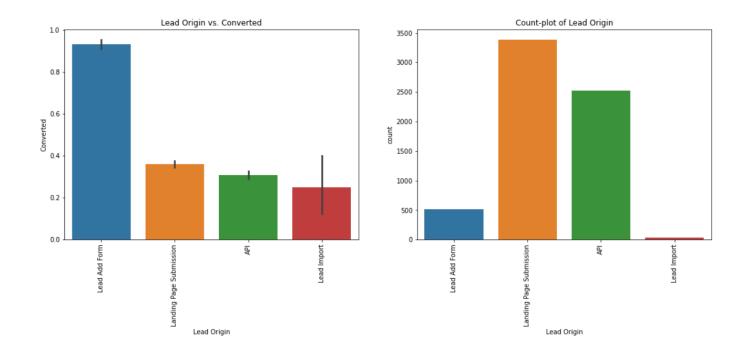
 We can see that the customer who is interested in 'Mastering the interview', they are likely to be converted.



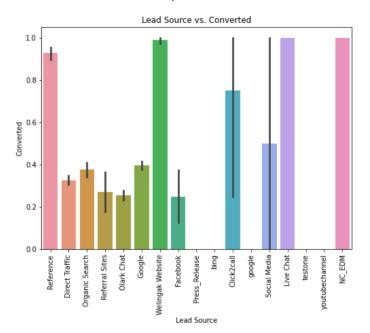


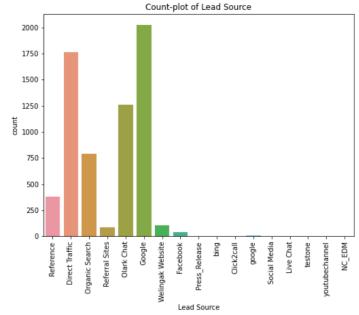


• Customers whose Lead Origin is 'Lead Add Form' are highly likely to be converted. There are approximately 500 'Lead Add Form' records in train dataset which is a good number.

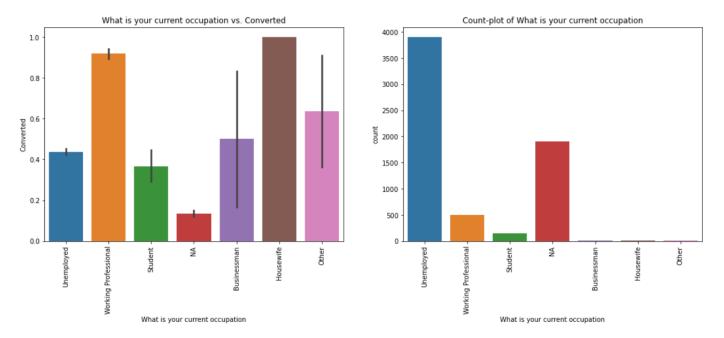


• Similarly, Customers whose Lead Source is 'Reference' are highly likely to be converted.



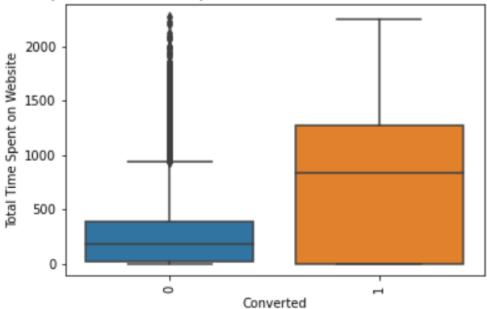


• Targeting working professional would be a good idea as their converted rate is very high.



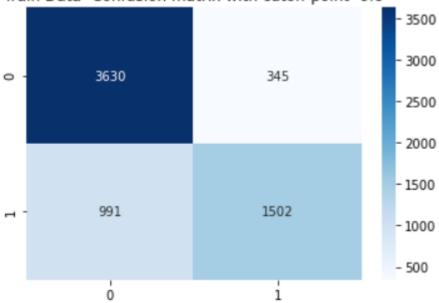
People who are spending more time on the website are likely to be Converted.

Boxplot of 'Total Time Spent on Website' for each 'Converted' class

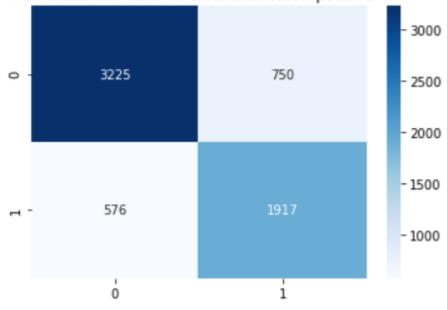


• We can see how sensitivity is increasing by decreasing the cutoff point.

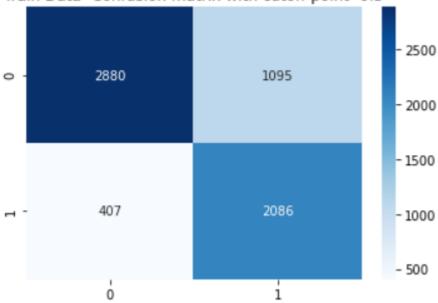
Train Data- Confusion matrix with cutoff point- 0.6



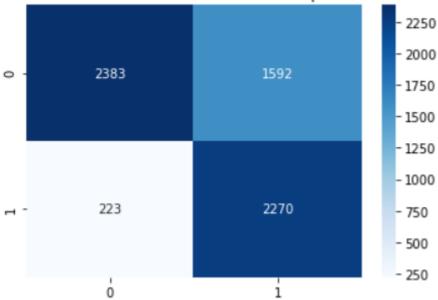
Train Data- Confusion matrix with cutoff point- 0.4

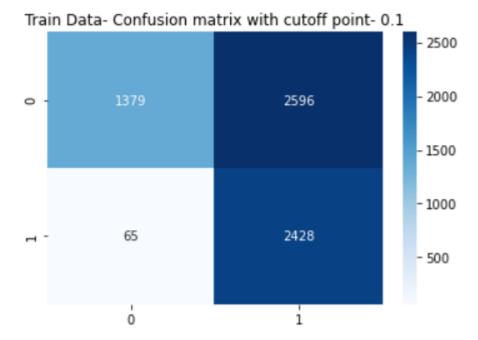


Train Data- Confusion matrix with cutoff point- 0.3



Train Data- Confusion matrix with cutoff point- 0.2





Hence, if we have to make calls aggressively then we can decrease the cutoff point, and if we want to reduce the number of calls then we can increase the cutoff point.