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Capstone Project- Classification Insurance cross sell prediction

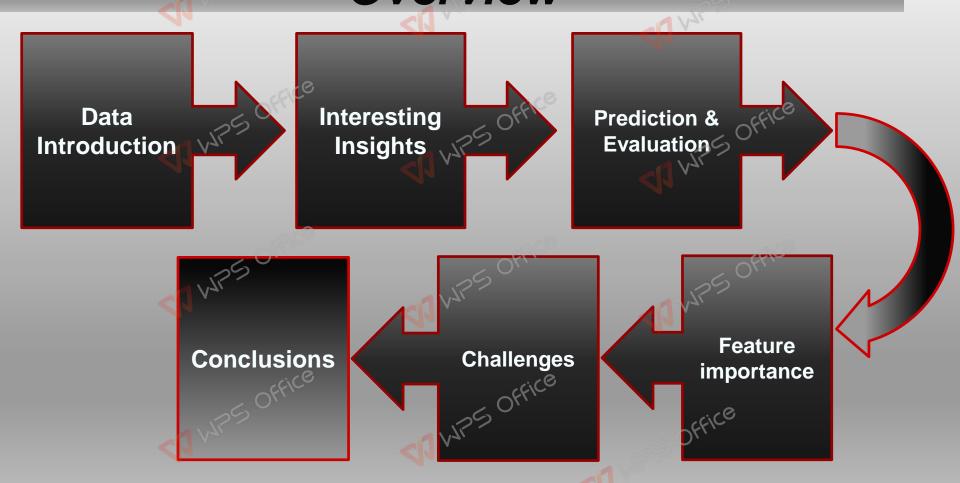
**Presented By :ABDUL QUADIR

Problem Statement

Al

Our client is an Insurance company that has provided Health Insurance to its customers now they need your help in building a model to predict whether the policyholders (customers) from past year will also be interested in Vehicle Insurance provided by the company.





vehicle

Data Introduction

Column

Driving License

Region Code

Previously Insured

Vehicle Age

Policy Sales

Channel

Vintage

Response

Vehicle Damage

Annual Premium

id

Gender

Age

Unique ID for the customer

Gender of the customer

Age of the customer

Age of the Vehicle

damaged in the past.

Mail, Over Phone, In Person, etc.

Unique code for the region of the customer

0 : Customer does not have DL, 1 : Customer already has DL

The amount customer needs to pay as premium in the year

1: Customer is interested, 0: Customer is not interested

Number of Days, Customer has been associated with the company

1 : Customer already has Vehicle Insurance, 0 : Customer doesn't have one.

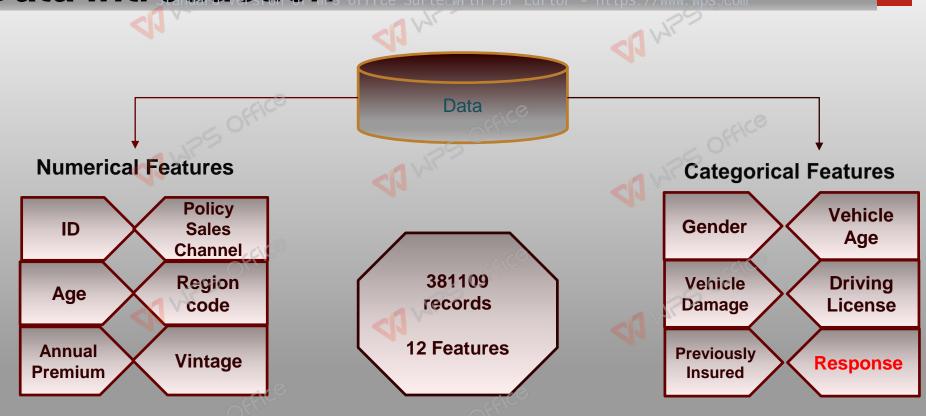
Customer got his/her vehicle damaged in the past. 0 : Customer didn't get his/her

Anonymized Code for the channel of outreaching to the customer i.e. Different Agents, Over

Description

Data Introduction









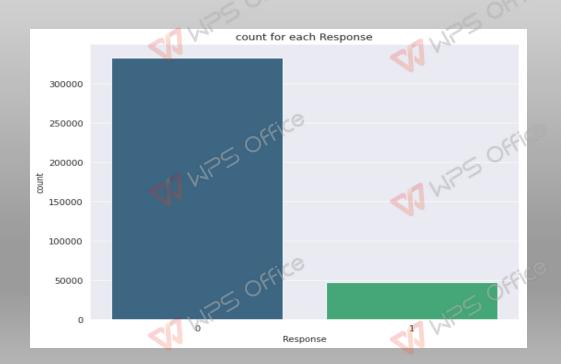
Interesting insights:

Draw insights and take **Business decisions**

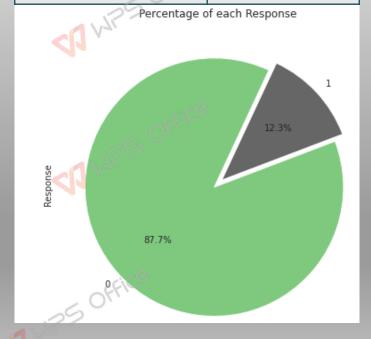
Overall conversion rate is 12.3%



381109 people are contacted/approached and 46582 are interested in vehicle insurance.

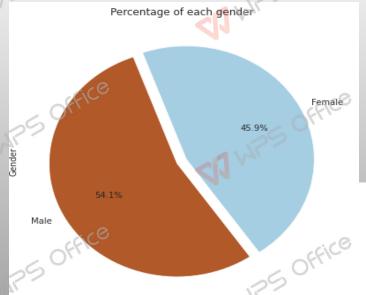


Response	Counts	
0	333107	
1 OFFICE	46582	



Men's give higher conversion

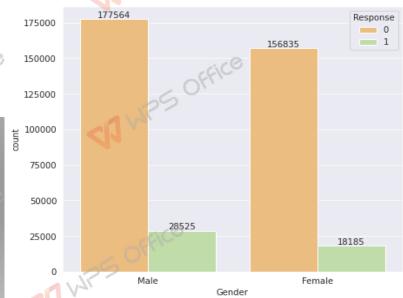




18	Male		13.84%	
		Female	10.39%	
		CO VIP .		
	175000	177564		

Gender

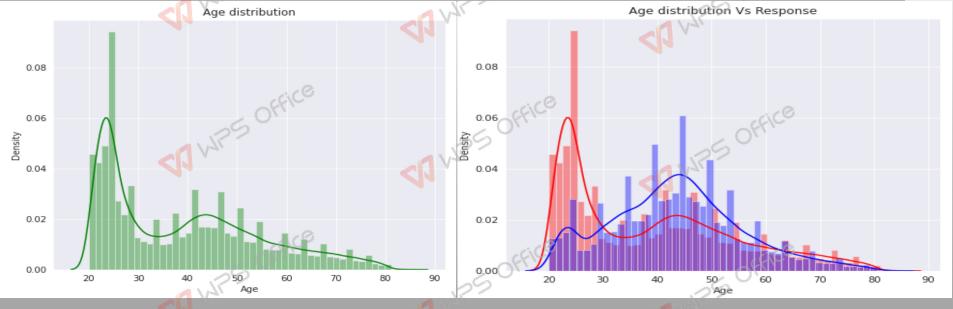
Observing the given data shows that interest rate is comparatively high for Male gender



Percent conversion

Target more people having age range 30-55 years Al



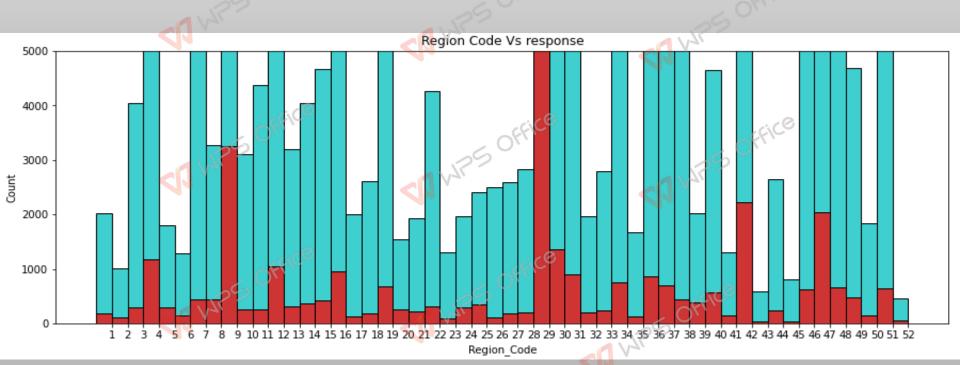


It has been observed that more customers are targeted having age range 23-30 years but we should target customers having age 30-55 years.

Region 29 has given us highest conversions



We have seen that people are more interested which are from Region having code 29 followed by region 9 and Region 42

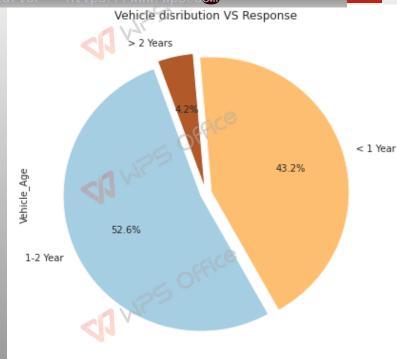


Poor conversion from "< 1 year" vehicle age



We analyzed that although 43% of records are coming from vehicle age "< 1year" but the interest rate is not very satisfying



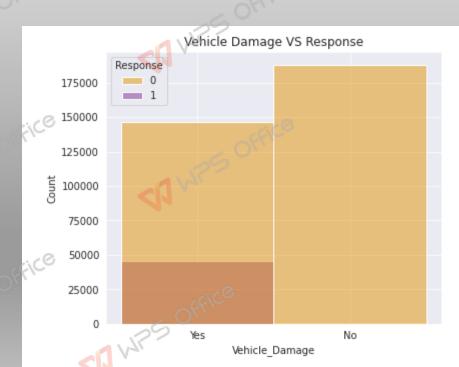


Attract people having no damage vehicle



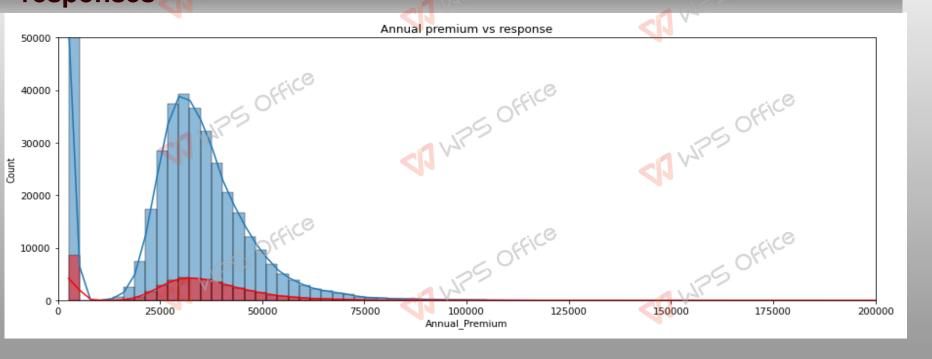
	Vehicle_Damage		
Yes	192413		
No	188696		

As per the data we have seen that if a customer has no vehicle damage then they are not interested. Hence organization should make some strategy to attract no-damage vehicle



Less records having > 75000 premium and hence less +ve responses





We Observed that there are very less people who were offered higher premium insurance and hence no. of positive responses also less. Same can be seen in the adjacent plot.

Top 10 Sales channels



It has been observed that channel no. 26 comes out to be best channel as it gave us 15891 interested people count followed by channel 124 which gave us 13996 conversion.

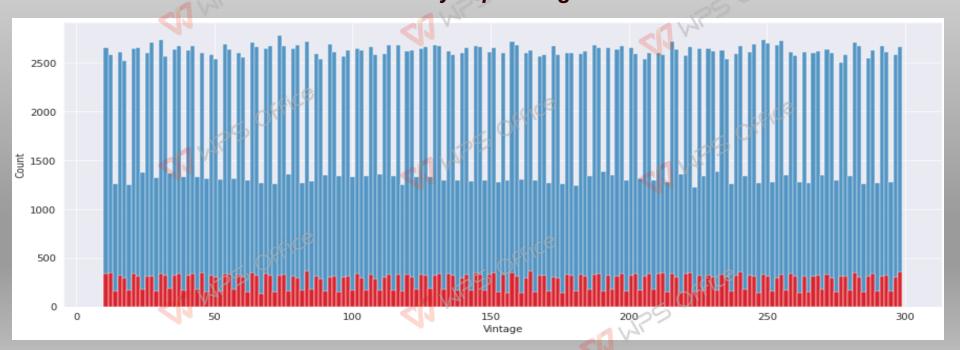


	Policy_Sales_Cha	nnel	Counts	Response_1
0	CD W	152	134784	3858
1		26	79700	15891
2		124	73995	13996
3		160°C	21779	475
4	1429	156	10661	2297
5		122	9930	1720
6		157	6684	1794
7		154	5993	1474
8	OFFICE	151	3885	122
9	25 OFT.	163	2893	880

Vintage is directly proportional to insurance sell

Al

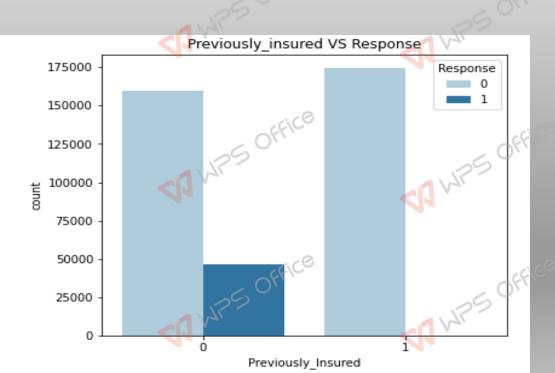
We have seen that vintage is directly proportional to interested customers. Which means more we target to particular range we get more interested customers but the conversion is not really impressing.



Not insured means good opportunity



If a customer is not previously insured then such customers are more likely to buy.



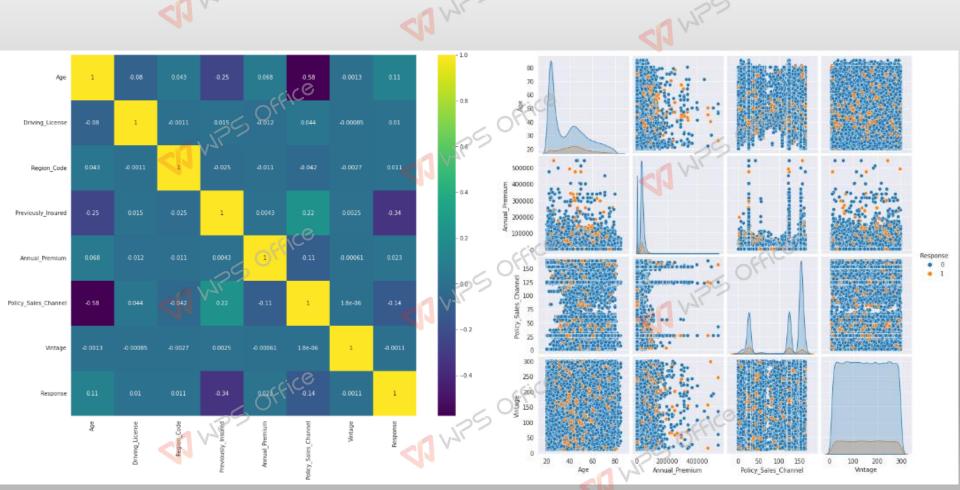


Prev	iously_Insured
0	206481
1 OFFICE	174628

Relation between features







Precision score Version of WPS Office Suitewith PDF Editor - https://www.wps.com



Predictions & 100 EVALUATION

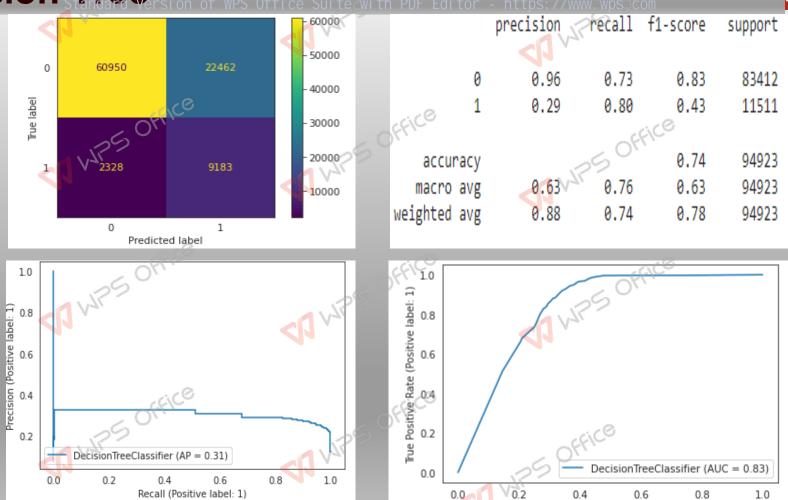
Confusion matrix

F1 Score

AUGIPS OFFICE

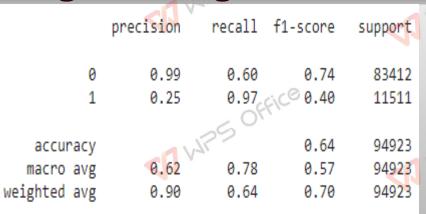
Decision Tree

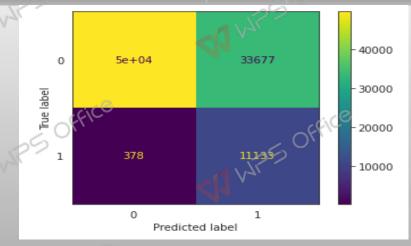


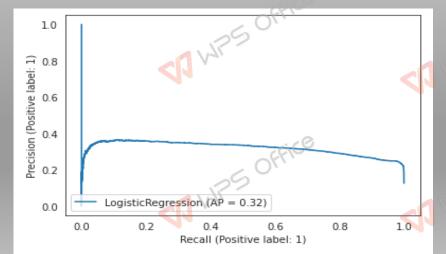


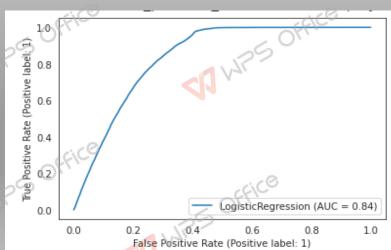
Logistic Regression





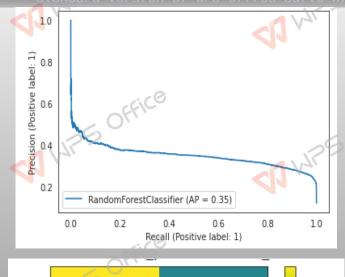


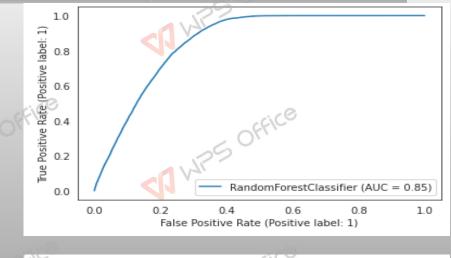


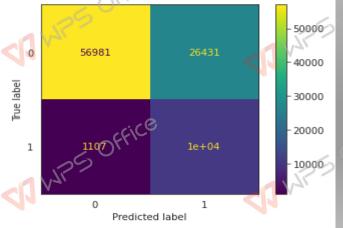






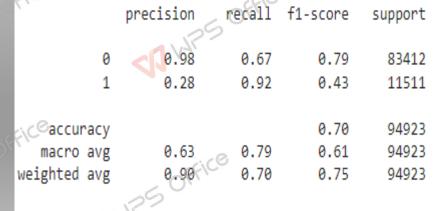






([[56981, 26431],

[1107, 10404]])



Feature Importance office suite with PDF Editor - https://www.wps.com

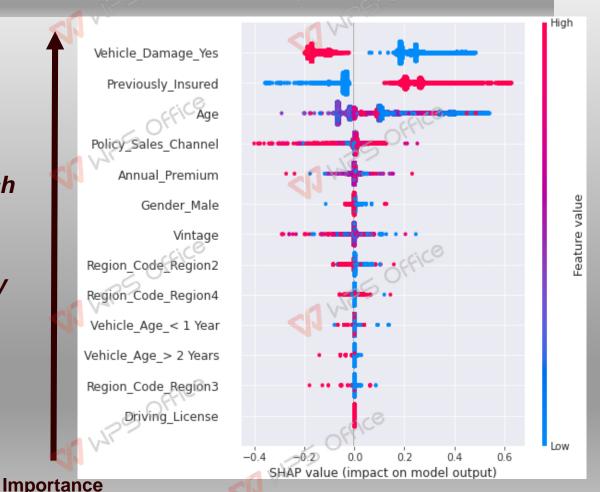




Global Feature importance based on SHAP value

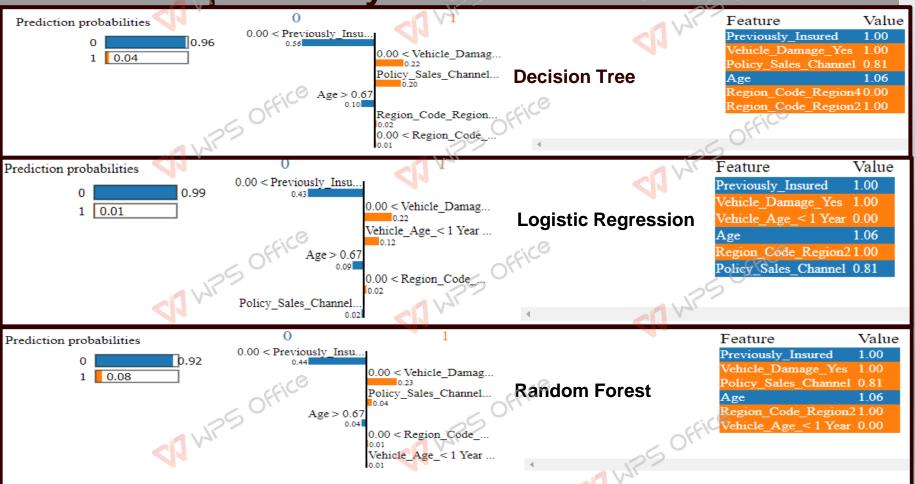
AI

In this plot we can see which features are contributing more to the prediction whether a customer will buy insurance or not



Local interpretability







- As the number of records is very high hence it was very time consuming to run a model even in some code/model it take a lot of time to run the code.
- > We observed that the dataset that we are given is imbalance dataset and thus we had to use some synthetic technique to create a balanced dataset.
- > Evaluation metric selection was challenge as choosing the incorrect metric can lead poor results.
- > Due to very large dataset we were not able to run complex models even when we tried it took approx. six hours but still there was no output and this might be the case because the system configuration is not that much efficient.
- AS the dataset has mixed datatypes hence we faced challenges while we have transformed data to feed our model.



- Men's are more interested compare to women.
- ❖ We can target more people having age range 30-55 years.
- * Region 29 has given us highest is giving us the most possible customers.
- Less people are interested having "< 1year" vehicle age while the count of ">1year" is comparatively good. Hence organization should make some strategy to attract no-damage vehicle.
- **❖** Less records having > 75000 premium and hence less +ve responses.
- * We saw that channel no. 26 comes out to be best channel as it gave us 15891 interested customers followed by channel 124 which gave us 13996 conversion.

- ❖ If a customer is not previously insured then such customers are more likely to buy.
- If we look from the perspective of explainability so we can choose decision tree but Random forest results are slightly better however the model is bit complex.
- Vehicle damage is the most important feature that drive predictions followed by previously insured.

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THANK YOU

Presented By:

ABDUL QUADIR