

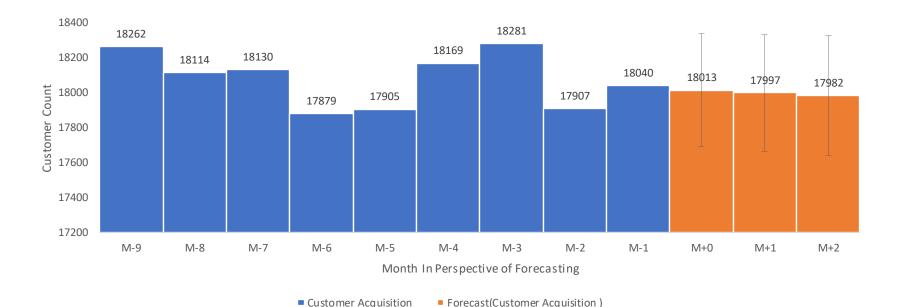
### Prodyogíkí - AnIT and Analytics Based Case Study Competition





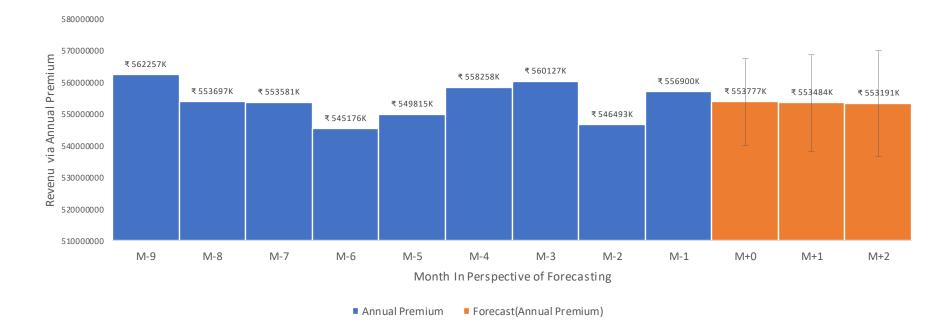
### Predictive Modeling To Forecast Future Sales

- □ Forecasting the **Number of Customer** and **Revenue Generated** based on **Customer Association** only for customers who have taken the insurance (insured = 1)
- □ Using Customer Association dates a Customer is associated to M-XX (if a customer is associated for 30 days; he is assigned to M-1). Customer Association date < 271 is used to forecast M-0,M+1,M+2). Assumptions: Considering the data for 270 days we have divided it into bins of 30 days each. Thus 9 months data has been used for the projection. The past twenty days i.e. data points from 270 to 290 haven't been included in forecasting as this would have had the least significance on future trends.
- ☐ The m-o-m growth rate obtained has been close to -0.14% which clearly points to the stagnant (slightly declining) customers being acquired by the company. This is a major issue of concern for the company.



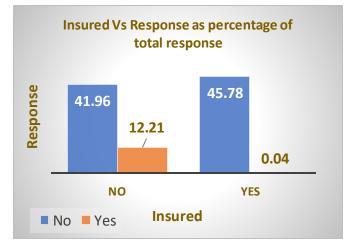
### Predictive Modeling To Forecast Future Revenue

- ☐ The forecasting for the revenues has been done similar to number of customers acquired. Thus 9 months duration have been used for the future projection of monthly premiums.
- □ There is no **seasonality** present in the data which is the reason for no deviation in the trendline. The projected revenues for the upcoming 3 months (M+0, M+1, M+2) is almost stagnant. The m-o-m growth rate has been slight negative at around -0.11%.
- On an average the revenue generated per customer is around **Rs. 30,649.67.** The number of customers associated needs to be increased in order to increase the total revenue generation of the company.



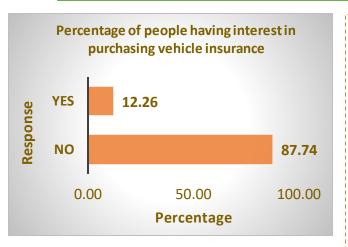
#### **Insured Vs Response**

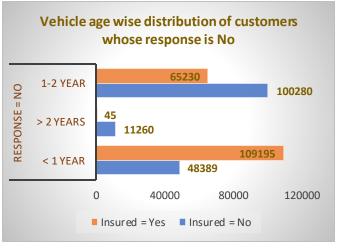
	Response		
Insured		No	Yes
	No	159929	46552
	Yes	174470	158



It has also been observed that only 0.090% of the people having past history of purchasing vehicle insurance has shown interest in purchasing further which is a quite low number

### **Exploratory Data Analysis**

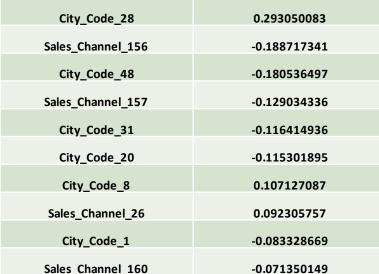


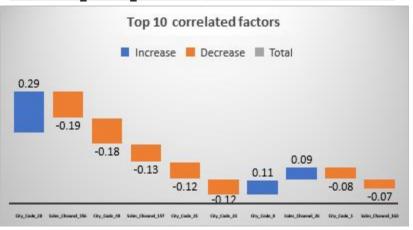


### **Insights:**

- As the number of people having vehicle insurance in past and not interested in purchasing is high as 174470 i.e. 99.90% of the people having past purchasing history, it can be inferred that people either having ongoing insurance or are not satisfied with the scheme.
- Further, it is also observed that out of 174470 people, 109257 people are having vehicle age less than 1 years which means their vehicle insurance is ongoing (assuming they purchase insurance on same day as vehicle purchase)

## Factors Affecting Annual Premium





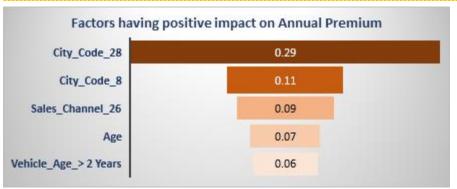
### Factors affecting Annual Premium

To calculate the major factors (Top 10) affecting on annual premium, first we converted the categorical variable such as "Gender", "Vehicle\_Age", "Vehicle\_Damage", "City\_Code", "Sales\_Channel" into binary variables leading to 216 columns.

The correlation between each independent variable and Annual premium is calculated. The top 10 affecting factors on annual premium is shown in the table.

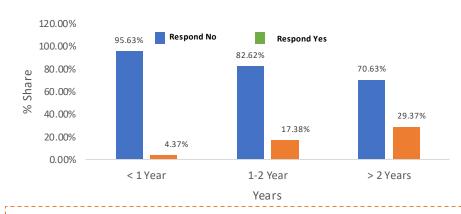
The positive correlation indicates the increase in annual premium while negative correlation indicates inverse relationship with actual premium.

Hence, to increase the revenue, the factors having positive impact on Annual Premium can be targeted to increase the sales. Hence, factors having positive impact on the Annual premium are as follows:





### Customer's Reponses Based On Vehicles Age



**Insights:** Respondents having vehicles aged > 2yrs are more inclined to buy a vehicle insurance are compared to people with vehicles aged <1 year. This might be due to in-house purchase of the insurance when the car is purchased.

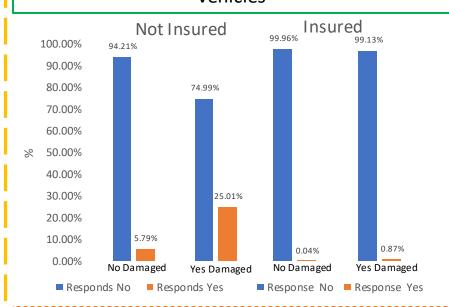
### Not Insured Customers' Premium vs Response

Response	113	NO
Mean	31620	30170
Median	33012	32065
Std Dev	18654	18194

### **Insights:**

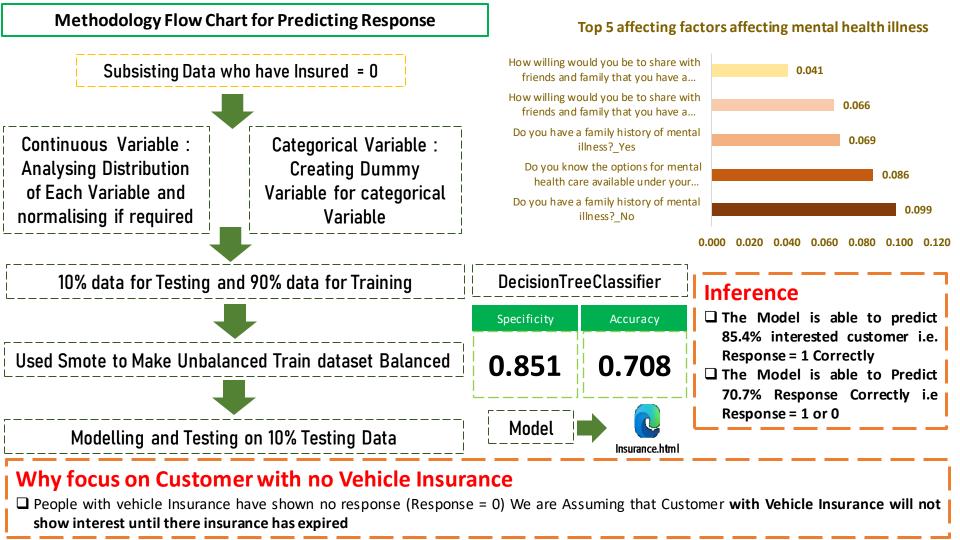
Customer response to Vehicle insurance (Insurance = 0) does not depend on premium to be paid as the mean ,median and Std Dev are nearly same.

# Customer's Response Based On Damage Of Vehicles

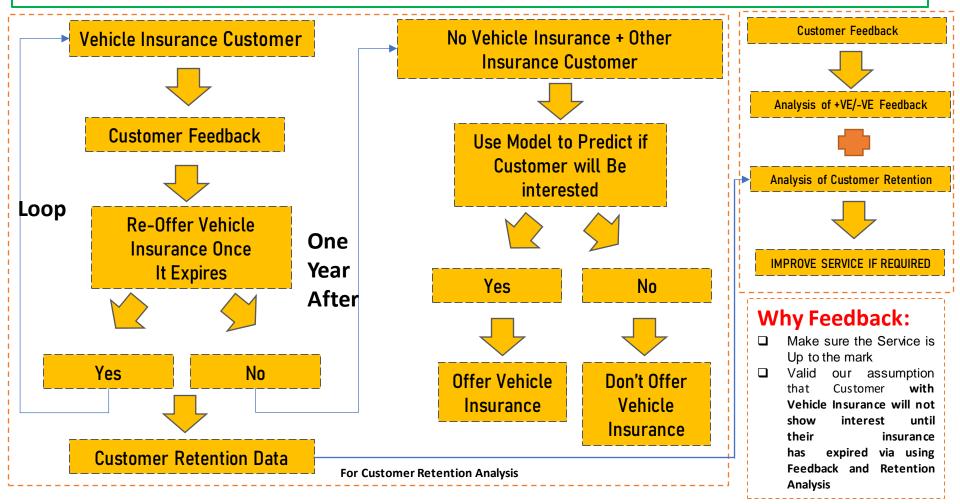


### **Insights:**

□ It is evident from the above graph that the % of people who were interested in buying the vehicle insurance were higher in the case of damaged vehicle which are not insured similar trend can be seen in Insured Vehicles even though the yes response is small



### Strategy for Company to Apply



### **Cost Benefit Analysis of the Strategy**

### **Benefit**

☐ Model can help in **Identifying 85.4%** interested customer Correctly help in **increasing the Revenue** by adding **new**Customers for Vehicle Insurance

Feedback and Customer retention analysis can help **in improving service** regarding Vehicle Insurance

The strategy will help company to add new customer by both **improving Customer Retention** and **Increasing** 

Customer acquisition improving the company's stagnate growth observed during the forecasting

### **Projected Benefit of the Model**

- ☐ Number of Days = 289 (Customer Association Ranges From 10 to 299)
- Expected Annual Premium from not insured interested customer (**Response = 1 and Insured = 0**) = Rs 1,471,978,440
- □ Expected Annual Premium for a day = Rs 1471978440 \* (1/289) = Rs 5093351
   □ According to model 85% can be identify Correctly = 0.85\*5093351 = Rs 4329348
- ☐ Per Day Actual Revenue = Rs18516785 (Total Premium from Insured = 1 / 289 Days )
- Expected Increase in Customer Via Model Per day = **137**
- ☐ Expected Increase in Revenue Via Model = Rs 4329348/Rs18516785 = 23.38%



THANK YOU