Mahindra First Choice Services



Presented by:

Vishnu Sreenivasan Quid Morbiwala Karan Pande Hemant Gupta



company.

Business Problem

| Geol | location | Based | Analysis: |
|------|----------|--------------|-----------|
|------|----------|--------------|-----------|

| | Extract insights regarding the ownership pattern of cars based on location features of the customers. Find out how the different type of services and service time vary with location. Analyse the revenue generated at different locations throughout the country |
|----------|--|
| Ma | arket Segmentation: |
| - | Divide the customer base into different segments based on the activeness, number of visits to garage, service type and revenue generated. This will help to better understand the customer behavioural patterns and have targeted marketing campaigns and rewards based on the segments |
| Cu | stomer Lifetime Value Prediction: |
| _ _ | Predict the value given by the customer over his lifetime based on the average spend over a year and number of visits. Analysing CLV would help Mahindra First Choice focus on the customers who are going to generate good value to the |



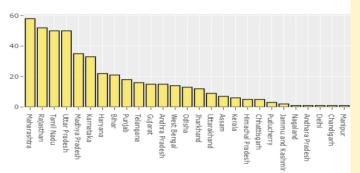
Exploratory Data Analysis

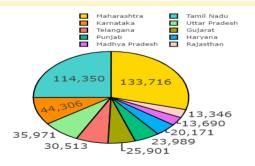
Lets Start with the first steps of Visualizations



Geolocation Analysis

Plant Count in various state



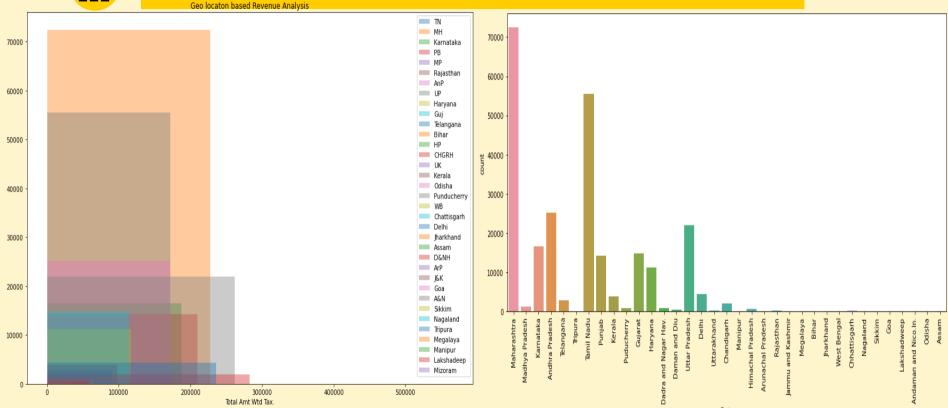


Plant count vs car count

- ☐ Maharashtra, Rajasthan, Tamil Nadu, UP,MP tops in number of plants
- ☐ Maha,TN, Karnataka, UP, Telangana tops in car count
- ☐ Plants vs car count is very less in states like Karnataka, Telangana
- □ Possible reasons could be car owners in Karnataka ,Telangana are more as compared to other states
- ☐ On the other hand Rajasthan is 2nd in terms of count but the car count is very less.
- □ Possible reasons could be customers shift to other service provider or car owners are less in this state as compared to other states



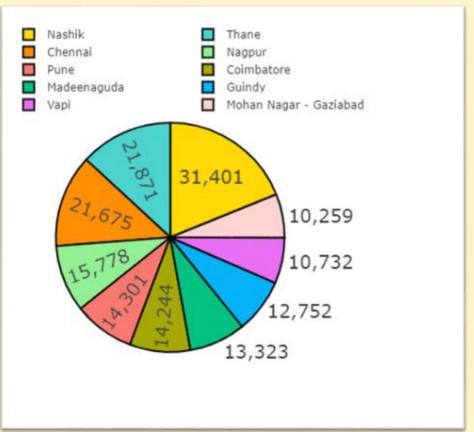
Which state has most revenue?



- Maharashtra has highest customer base followed by Tamil Nadu.
- Top 3 revenue generating states are MH TN AP



Which city has most revenue

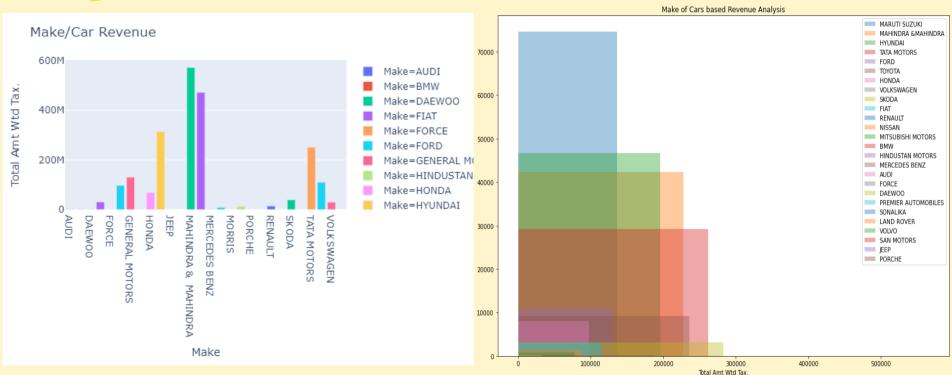


Top 5 cities given above top the list, this is obvious as they are top major metropolitan cities in India

- Nashik(31,401)
- Thane(21,871)
- Chennai(21,675)
- Nagpur(15,778)
- Pune(14,301)



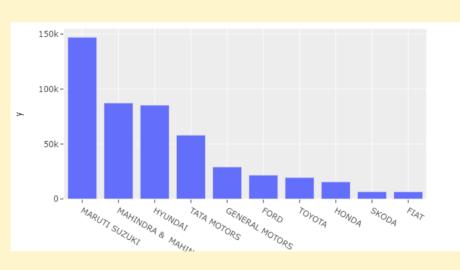
Which make/car is more popular?



- M&M contributed the most towards revenue ,followed by Maruti Suzuki, Hyundai, Tata Motors
- Meanwhile, Maruti Suzuki tops on the basis of count but the revenue produced is less as compared to Mahindra and Mahindra
- Make of Cars based Revenue Analysis



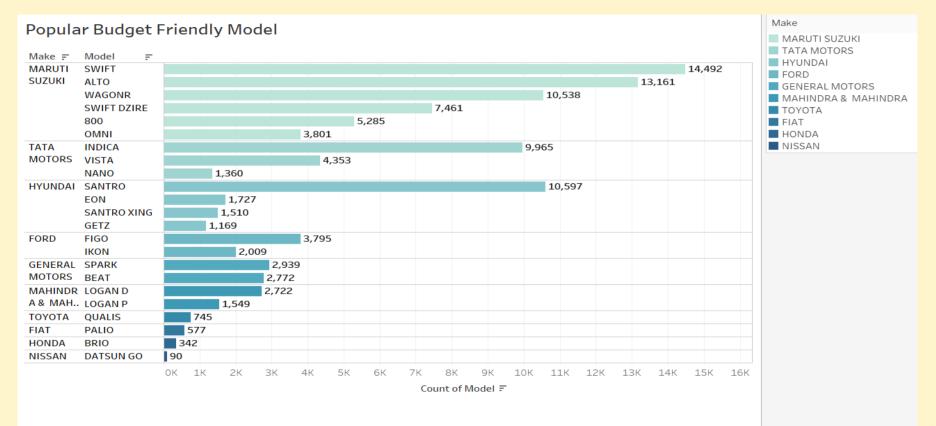
Which make/car is more popular?





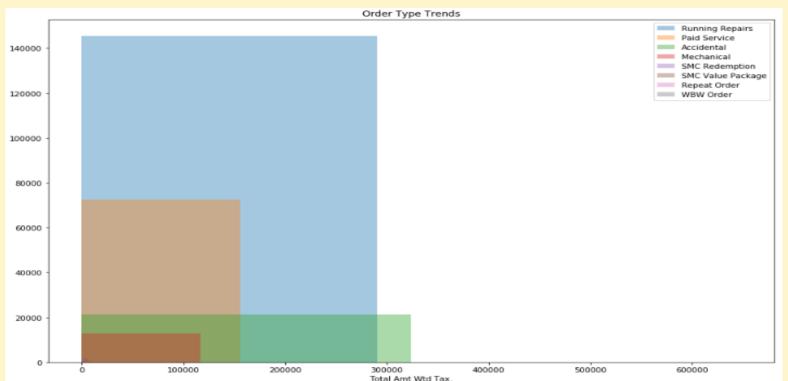


Popular Budget Friendly Car Models





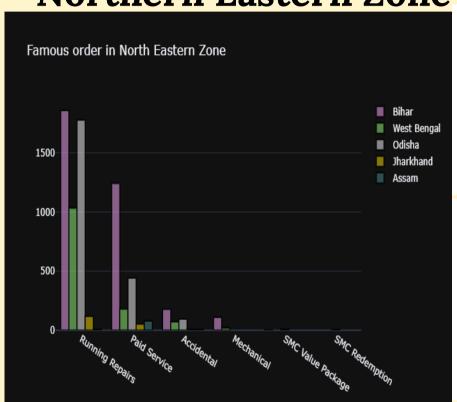
What are the most popular services?

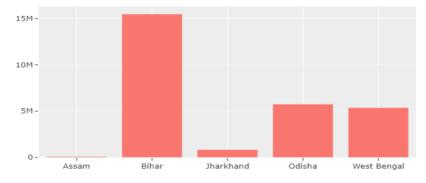


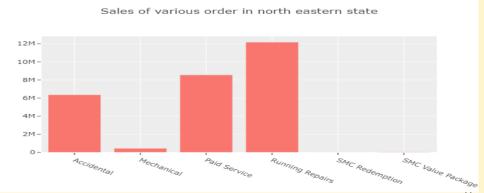
- Frequency of Running services customers is highest.
- Accidental service order generates highest revenue.



Northern Eastern Zone Service Count

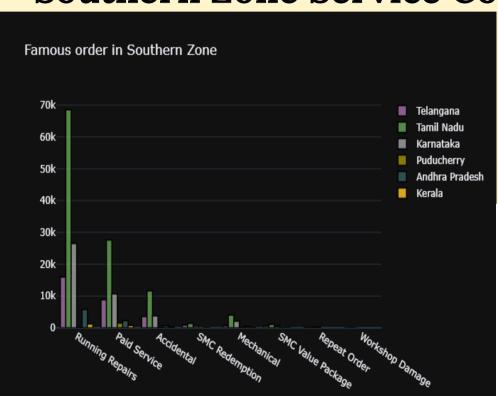




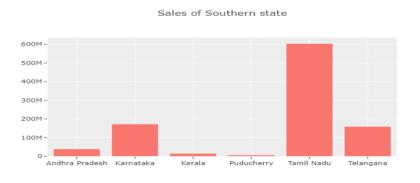




Southern Zone Service Count

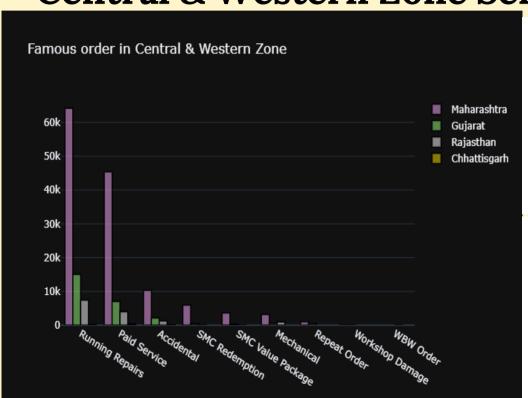






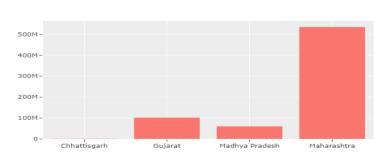


Central & Western Zone Service Count



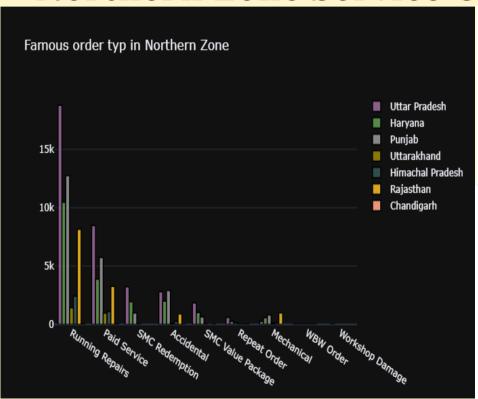


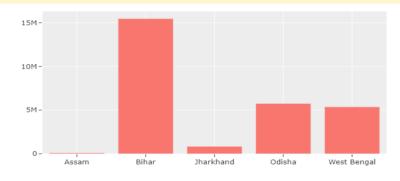
Sales of north eastern state

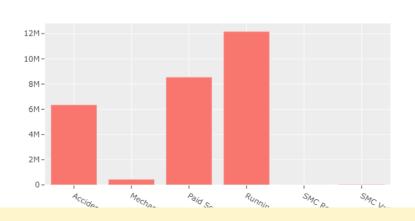




Northern Zone Service Count

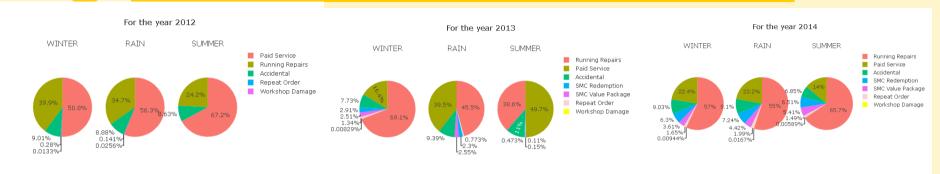


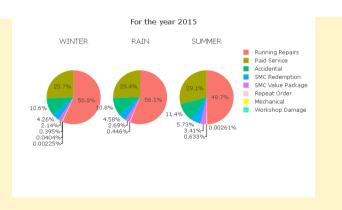


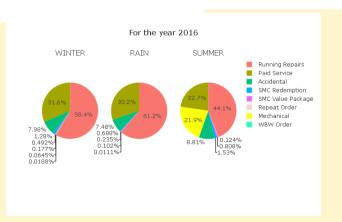




Seasonal Analysis of order types





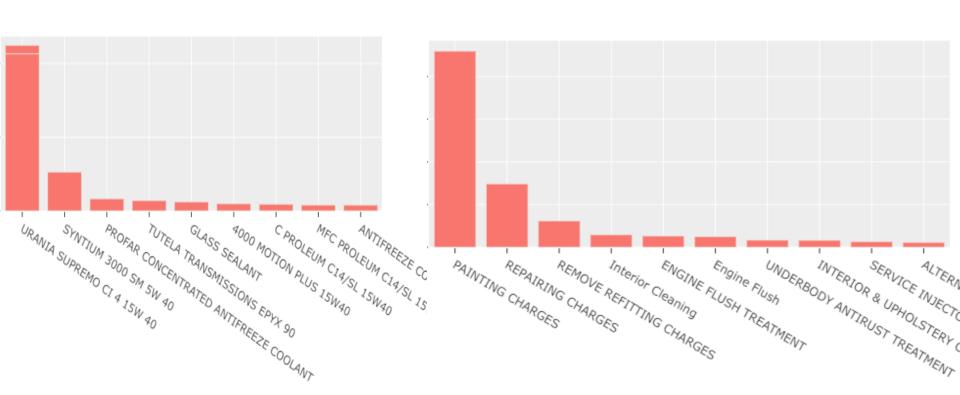




Inventory management analysis for revenue based top 10 parts used and services

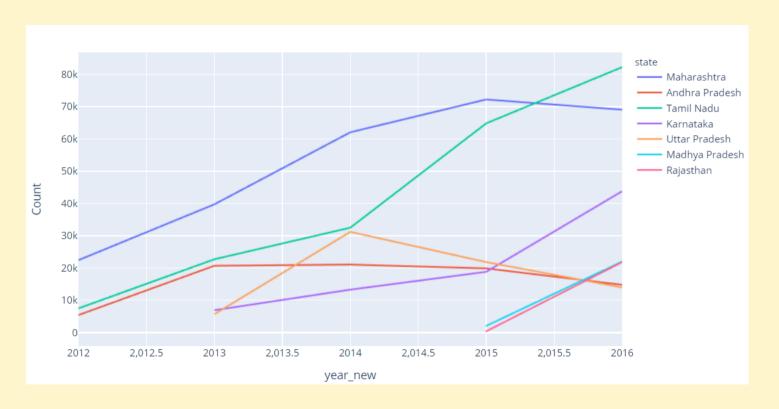
Top 10 most sold parts according to revenue

Revenue earned from external services





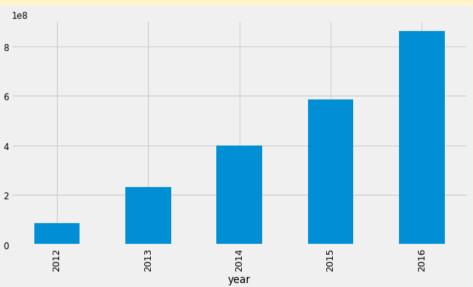
Revenue Analysis State wise Invoice count



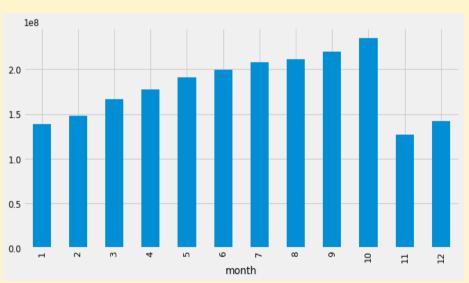


Revenue Analysis trends based on year and months

Year Wise Revenue

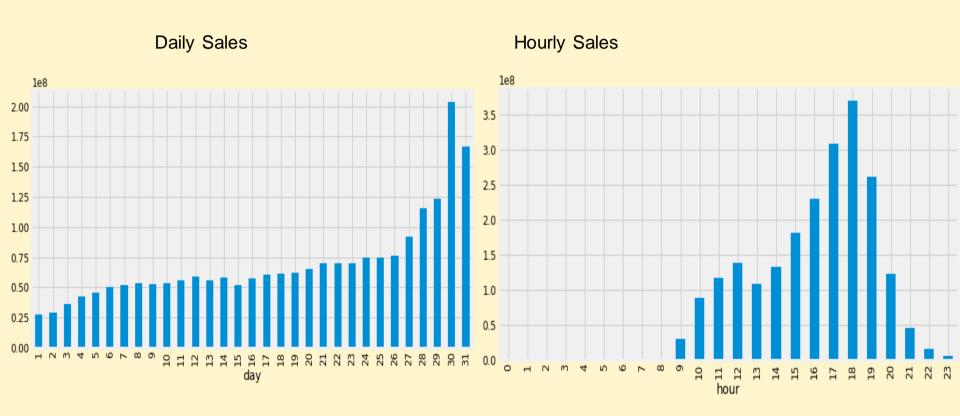


Month wise Revenue





Revenue Analysis hourly and daily sales

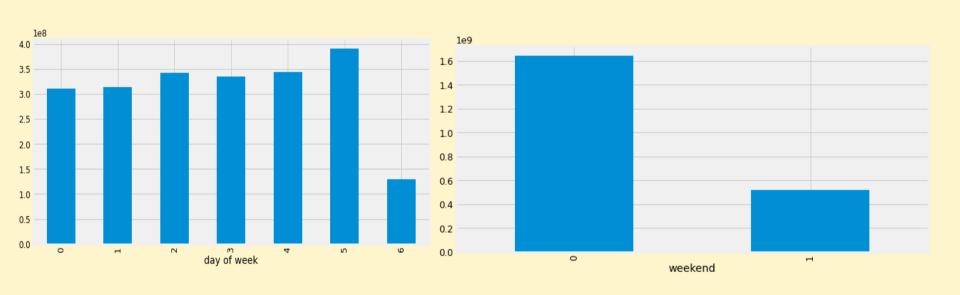




Revenue Analysis daily and weekends

Daily Sales Analysis

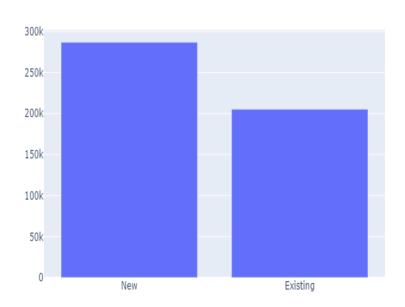
Week day vs Weekend Sales





Customer segmentation Analysis Sales contribution from new vs existing customer

New Vs Exsisting customers count



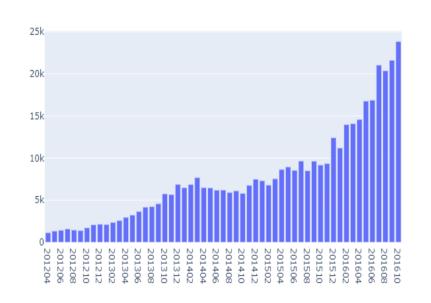
New vs Existing



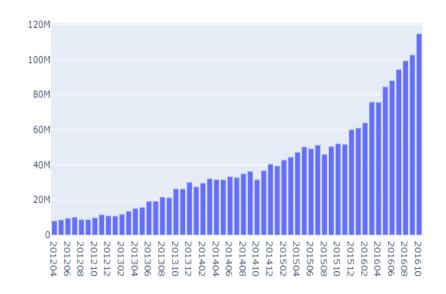


Customer segment analysis Revenue Trends and active customers

Monthly Active Customers

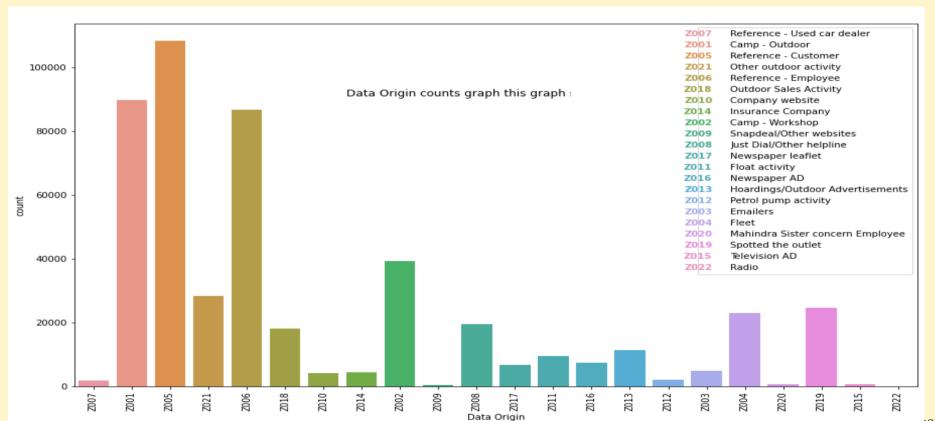


Total Monthly revenue





Custer reference sources

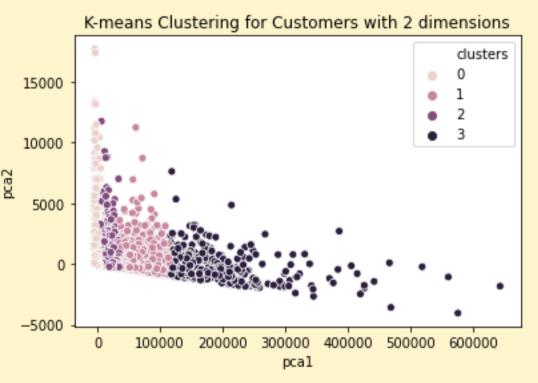




Lets Start with clustering of data



Customer clustering based on revenue

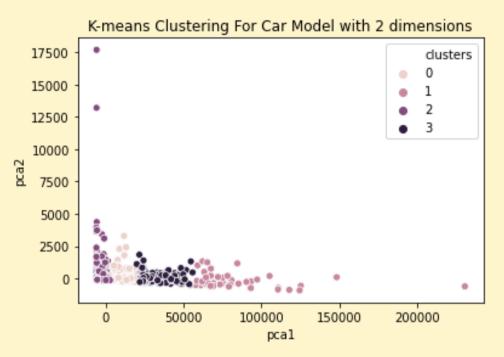


| | Service Hour | Total Amt Wtd Tax. |
|----------|--------------|--------------------|
| clusters | | |
| 0 | 67.075737 | 2198.219239 |
| 1 | 562.379392 | 61303.102984 |
| 2 | 209.657500 | 17274.993898 |
| 3 | 1211.764723 | 181336.634630 |

0-low revenue earning cluster of customer1-high revenue earning cluster of customer2-medium revenue earning cluster of customer3-very high revenue earning cluster of customer



Customer clustering based on Clustering for Revenue generated from different Car Models

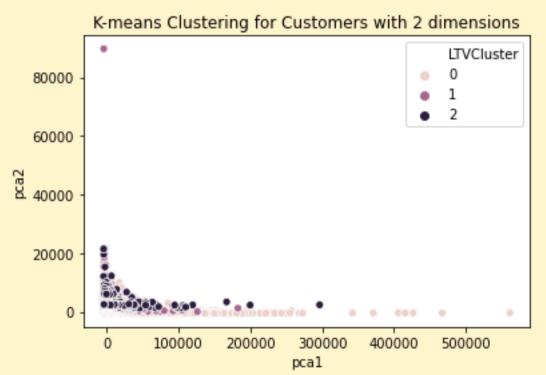


| | Car_Model | Car_Age | order_Accidental |
|----------|-----------|----------|------------------|
| clusters | | | |
| 0 | 916 | 4.411572 | 0.701965 |
| 1 | 40 | 5.550000 | 0.800000 |
| 2 | 5272 | 4.438164 | 0.044765 |
| 3 | 216 | 4.912037 | 0.865741 |

- 0 Medium Revenue Earning Cluster of Car Models
- 1 Very High Revenue Earning Cluster of Car Models
- 2 Low Revenue Earning Cluster of Car Models
- 3 High Revenue Earning Cluster of Car Models



Customer LTV custer



| | LifetimeValue | Recency | Frequency |
|------------|---------------|------------|-----------|
| LTVCluster | | | |
| 0 | 1487.849225 | 148.199911 | 1.506938 |
| 1 | 1983.534525 | 97.450118 | 2.008983 |
| 2 | 2034.003130 | 94.080370 | 2.060100 |

0-low lifetime value of a customer1-medium lifetime value customer2-high lifetime value of a customer



Data Preprocessing

Lets Start with preprocessing of data



Data Preprocessing

| ☐ Dropped columns |
|---|
| ☐ Clean state and city names |
| ☐ Treated null values in model, data origin and Partner type column |
| ☐ Bucketed states in zones |
| ☐ Created year, month and season features |
| ☐ Calculated total service hours |
| ☐ Calculated age of the cars |
| ☐ Scraped car prices for bucketing models into Categories |



Customer Lifetime Value

- ✓ Lifetime value of the customer is a metric that represents the total amount of a money a customer is expected to spend over the lifetime of the car.
- ✓ CLTV = Customer Frequeny * Total amount Spent / Number of years
- ✓ We have calculated Customer LTV for the year 2016.



Modelling

Lets Start with Model Building



Modelling

| Models | Accuracy on train | Accuracy on test |
|---|-------------------|------------------|
| Decision Tree Classifier | 0.995 | 0.857 |
| Random Forest Classifier | 0.9222 | 0.9187 |
| Logistic Regression | 0.921 | 0.918 |
| Decision Tree Classifier with Grid search | 0.895 | 0.857 |

Classification Output:
Low LTV group (<1700)
Mid LTV group (1700 to 2000)
High LTV group (>2000)



Cross Valdiation

We perform cross validation based using voting classifier

| Models used for estimators Decision Tree Classifier + Logistic Regression | Accuracy of Voting(Hard) | Accuracy of Voting(Soft) |
|---|--------------------------|--------------------------|
| Train set | 0.922 | 0.995 |
| Test Set | 0.918819 | 0.859333 |



Actionable Insights

- ✓ States like Karnataka ,Tamil Nadu and Madhya Pradesh had huge spike in customer base every year. Implement their marketing strategies in states with low customer count
- ✓ Investigate the reasons for dip in revenue of Uttar Pradesh and Telangana
- ✓ Eastern states have a significantly lower share of the revenue. This is due to low customer count from these states. More plants could be opened up in this states and marketing strategies similar to top states could help increase the customer base in eastern states.
- ✓ Majority of cars visiting the workshop are of these 4 brands Maruti Suzuki, Hyundai, M&M and Tata Motors. Inventory should be stocked up accordingly.
- Though invoice of Accidental cases are less, they contribute the largest share to the revenue. Average service time is also the highest. To reduce the same, spare parts of top brands stated above could be stocked up in advance.
- The top marketing source has been Customer Reference. We could arrange rewards programs for customers sharing positive testimonials to increase the customer base.

