

# **BUSINESS REPORT**

<u>Of</u>

**Austo Motor Company** 

**PDS Coded Project** 

**PGP - DSBA** 

**Major Analysis Covers: Demand of Customer** 

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#### 1.) About Company:

Austo Motor Company is a leading car manufacturer specializing in SUV, Sedan, and Hatchback models. In its recent board meeting, concerns were raised by the members on the efficiency of the marketing campaign currently being used. The board decides to rope in an analytics professional to improve the existing campaign.

#### 2.) Scope of Work/Objective:

Analyze the data to get a fair idea about the demand of customers which will help them in enhancing their customer experience. Suppose you are a Data Scientist at the company and the Data Science team has shared some of the key questions that need to be answered. Perform the data analysis to find answers to these questions that will help the company to improve the business.

#### **Data Overview**

#### 3.) Import the Libraries

For Analyzing the Data set using Python, we have imported all the required libraries as NumPy, Pandas, Metaplotlib and seaborn

#### 4.) Loading the Data

Loading the data is the initial step to analysis in any Project.

#### 5.) Checking the Structure of Data

Checking the Structure of data is important step for understanding of format and features using df.head(), df.info() and df.shape

This shows top 5 records and this Data has 1581 rows and 14 columns

|   | Age | Gender | Profession | Marital_status | Education        | No_of_Dependents | Personal_loan | House_loan | Partner_working | Salary | Partner_salary |
|---|-----|--------|------------|----------------|------------------|------------------|---------------|------------|-----------------|--------|----------------|
| 0 | 53  | Male   | Business   | Married        | Post<br>Graduate | 4                | No            | No         | Yes             | 99300  | 70700.0        |
| 1 | 53  | Femal  | Salaried   | Married        | Post<br>Graduate | 4                | Yes           | No         | Yes             | 95500  | 70300.0        |
| 2 | 53  | Female | Salaried   | Married        | Post<br>Graduate | 3                | No            | No         | Yes             | 97300  | 60700.0        |
| 3 | 53  | Female | Salaried   | Married        | Graduate         | 2                | Yes           | No         | Yes             | 72500  | 70300.0        |
| 4 | 53  | Male   | Salaried   | Married        | Post<br>Graduate | 3                | No            | No         | Yes             | 79700  | 60200.0        |

#### 6.) Check the Types of Data

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1581 entries, 0 to 1580
Data columns (total 14 columns):
# Column Non-Null Count Dtype

0 Age 1581 non-null int64
1 Gender 1528 non-null object
2 Profession 1581 non-null object
--- -----
3 Marital_status 1581 non-null object
4 Education 1581 non-null object
    No of Dependents 1581 non-null int64
   Personal_loan 1581 non-null object
House_loan 1581 non-null object
 6
 7
    Partner_working 1581 non-null object
8
                         1581 non-null int64
 9
     Salary
 10 Partner_salary 1475 non-null float64
 11 Total_salary 1581 non-null int64
 12 Price
                       1581 non-null int64
13 Make
                         1581 non-null object
dtypes: float64(1), int64(5), object(8)
memory usage: 173.0+ KB
```

Data types of Columns are float64 = 1, int64 = 5, object = 8 Numerical = 6 Object = 8

#### 7.) Treating Missing Values



dtype: int64

Missing Values affects the analysis, So identify and treating them is very important, In this Data Set we have Missing values mentioned below:

Gender = 53

Partner Salary = 106

Right way of treating the Missing values in done by Mean, Mode and Median Gender is replaced with Mode and Partner salary with Mean.

#### 8.) Check the Statistical Summary

|                  | count  | unique | top           | freq | mean         | std          | min     | 25%     | 50%     | 75%     | max      |
|------------------|--------|--------|---------------|------|--------------|--------------|---------|---------|---------|---------|----------|
| Age              | 1581.0 | NaN    | NaN           | NaN  | 31.922201    | 8.425978     | 22.0    | 25.0    | 29.0    | 38.0    | 54.0     |
| Gender           | 1528   | 4      | Male          | 1199 | NaN          | NaN          | NaN     | NaN     | NaN     | NaN     | NaN      |
| Profession       | 1581   | 2      | Salaried      | 896  | NaN          | NaN          | NaN     | NaN     | NaN     | NaN     | NaN      |
| Marital_status   | 1581   | 2      | Married       | 1443 | NaN          | NaN          | NaN     | NaN     | NaN     | NaN     | NaN      |
| Education        | 1581   | 2      | Post Graduate | 985  | NaN          | NaN          | NaN     | NaN     | NaN     | NaN     | NaN      |
| No_of_Dependents | 1581.0 | NaN    | NaN           | NaN  | 2.457938     | 0.943483     | 0.0     | 2.0     | 2.0     | 3.0     | 4.0      |
| Personal_loan    | 1581   | 2      | Yes           | 792  | NaN          | NaN          | NaN     | NaN     | NaN     | NaN     | NaN      |
| House_loan       | 1581   | 2      | No            | 1054 | NaN          | NaN          | NaN     | NaN     | NaN     | NaN     | NaN      |
| Partner_working  | 1581   | 2      | Yes           | 868  | NaN          | NaN          | NaN     | NaN     | NaN     | NaN     | NaN      |
| Salary           | 1581.0 | NaN    | NaN           | NaN  | 60392.220114 | 14674.825044 | 30000.0 | 51900.0 | 59500.0 | 71800.0 | 99300.0  |
| Partner_salary   | 1475.0 | NaN    | NaN           | NaN  | 20225.559322 | 19573.149277 | 0.0     | 0.0     | 25600.0 | 38300.0 | 80500.0  |
| Total_salary     | 1581.0 | NaN    | NaN           | NaN  | 79625.996205 | 25545.857768 | 30000.0 | 60500.0 | 78000.0 | 95900.0 | 171000.0 |
| Price            | 1581.0 | NaN    | NaN           | NaN  | 35597.72296  | 13633.636545 | 18000.0 | 25000.0 | 31000.0 | 47000.0 | 70000.0  |
| Make             | 1581   | 3      | Sedan         | 702  | NaN          | NaN          | NaN     | NaN     | NaN     | NaN     | NaN      |

By using describe() Function in Pandas, we have generated statistical summary of this Dataset.

#### 9.) Check for and treat(if needed) data irregularities

Upon Reviewing the data, it has been observed that there are some variables that has multiple and invalid names in Gender value, which has been corrected by using replace function

|      | Age | Gender | Profession | Marital_status | Education        | No_of_Dependents | Personal_loan | House_loan | Partner_working | Salary | Partner_salary | Total_salary | Price |
|------|-----|--------|------------|----------------|------------------|------------------|---------------|------------|-----------------|--------|----------------|--------------|-------|
| 0    | 53  | Male   | Business   | Married        | Post<br>Graduate | 4                | No            | No         | Yes             | 99300  | 70700.0        | 170000       | 61000 |
| 1    | 53  | Female | Salaried   | Married        | Post<br>Graduate | 4                | Yes           | No         | Yes             | 95500  | 70300.0        | 165800       | 61000 |
| 2    | 53  | Female | Salaried   | Married        | Post<br>Graduate | 3                | No            | No         | Yes             | 97300  | 60700.0        | 158000       | 57000 |
| 3    | 53  | Female | Salaried   | Married        | Graduate         | 2                | Yes           | No         | Yes             | 72500  | 70300.0        | 142800       | 6100  |
| 4    | 53  | Male   | Salaried   | Married        | Post<br>Graduate | 3                | No            | No         | Yes             | 79700  | 60200.0        | 139900       | 5700  |
|      | 100 | ***    | rec.       | 940            | (heat)           | 3600             |               | 344        | 1948            | 50     | ěs.            | 144          | - 69  |
| 576  | 22  | Male   | Salaried   | Single         | Graduate         | 2                | No            | Yes        | No              | 33300  | 0.0            | 33300        | 2700  |
| 1577 | 22  | Male   | Business   | Married        | Graduate         | 4                | No            | No         | No              | 32000  | NaN            | 32000        | 3100  |
| 1578 | 22  | Male   | Business   | Single         | Graduate         | 2                | No            | Yes        | No              | 32900  | 0.0            | 32900        | 3000  |
| 1579 | 22  | Male   | Business   | Married        | Graduate         | 3                | Yes           | Yes        | No              | 32200  | NaN            | 32200        | 2400  |
| 1580 | 22  | Male   | Salaried   | Married        | Graduate         | 4                | No            | No         | No              | 31600  | 0.0            | 31600        | 3100  |

#### 10.) Observation and Insights

Now, the Data Looks Perfect and Fine for analyzing the demand of Customers according to requirements, Budget and Need.

By using this Data we can Provide Information for Increasing sales and Individual experience.

## **Univariate Analysis**

#### 11.) Exploring Categorical and Numerical variables in the data

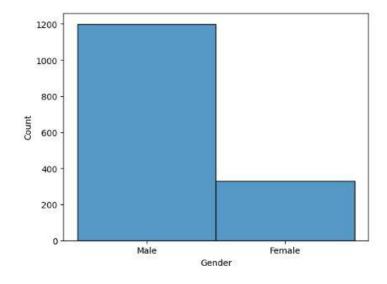
We have few categorical columns mentioned below

- 1. Gender
- 2. Profession
- 3. Marital status
- 4. Education
- 5. Personal loan
- 6. House loan
- 7. Partner working
- 8. Make

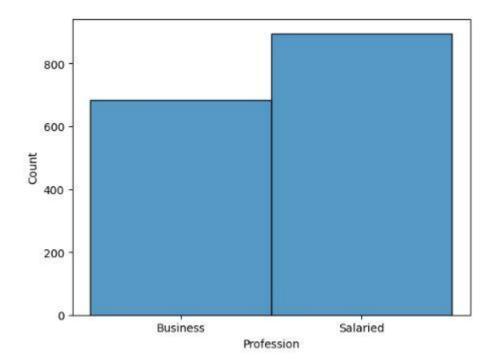
By using Plotting techniques, we will show the categorical distribution of data by using Bar Chart

Gender: Graph clearly shows Dominance of Male car owners as compared to Female.

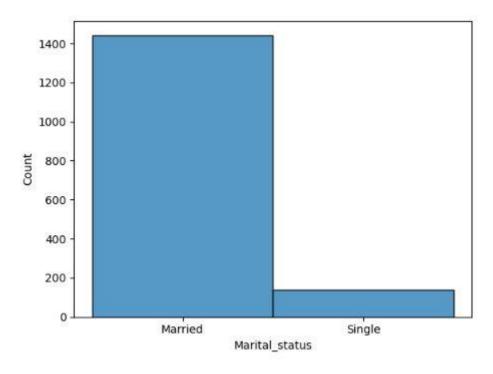
**Bar Chart of Gender:** According to this male Prefer buying cars more than Females



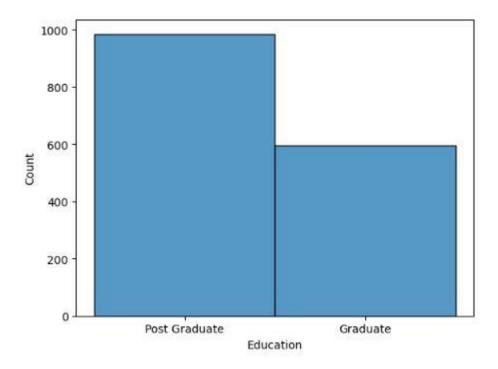
<u>**Profession :** </u> Data Suggests cars are more preferred among salaried person as compared to Business



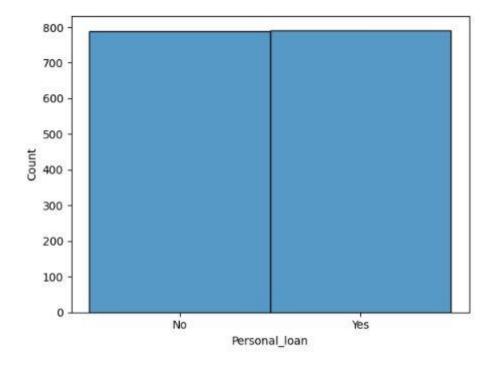
<u>Marital Status</u>: Married People are considering more into buying cars as compared to Single.



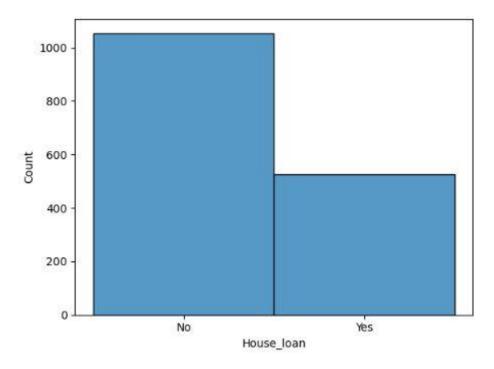
**Education:** Data Shows Post Graduates preferring buying Cars with comparison to Gradute.



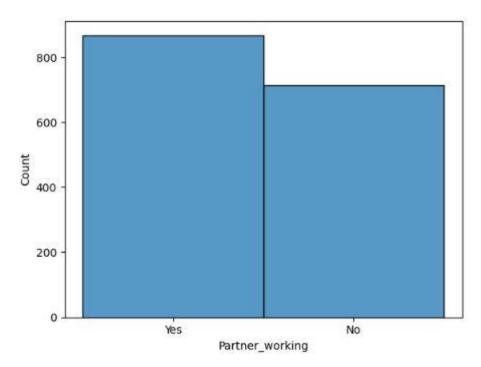
**Personal Load:** There is no impact of Personal Loan in buying of Cars



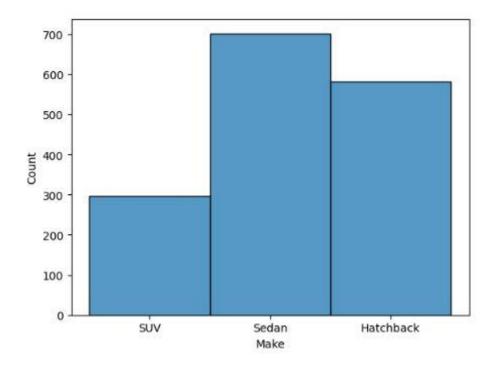
<u>House Loan:</u> Data Shows that People with No House loan are buying more cars, it looks like House loan disbalances the budget.



<u>Partner Working</u>: This data shows that, whose Partner is working preferring buying more cars as compared to Partner not working, that means having money in hand is easy decision of buying cars compared to the person with Non-working partners.



<u>Make:</u> This Graph shows that People love to buy and travel from Sedan, Their may be a reason of Budget, as Sedan comes in the middle class budget and SUV is costly and only preferred by those whose salary and income is high. Hatchback comes in Between both.



#### Exploring Numerical Variables mentioned below, It can be both int and float

- Age
- No of Dependents
- Salary
- Partner Salary
- Total Salary
- Price

For Numerical variable and distribution, box plot will be best for understanding

Box Plot is a graphical representation that displays the distribution of datasets and provides summary. It consists of 5 major key values that represent important characteristics of the data.

Minimum(min) = Smallest value in the dataset

First Quartile(Q1) = The Median of the lower half of Data set

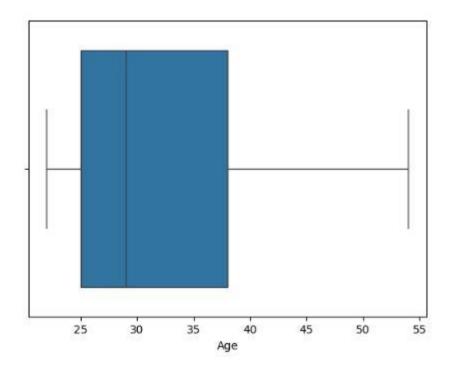
Median = The Middle value of the Dataset.

Third Quartile(Q3) = Median of Upper half in Dataset

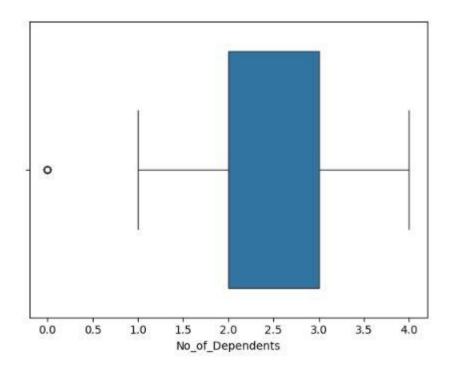
Maximum(Max) = Largest Value in Dataset

### Here Shows Graphical Representation

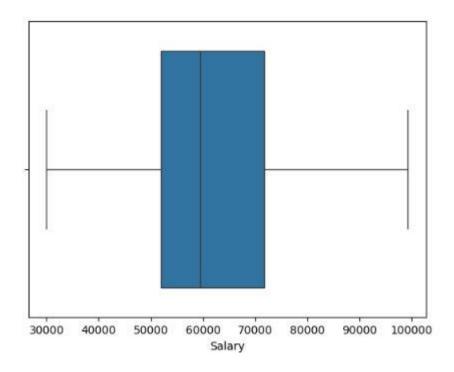
**Box Plot of Age** 



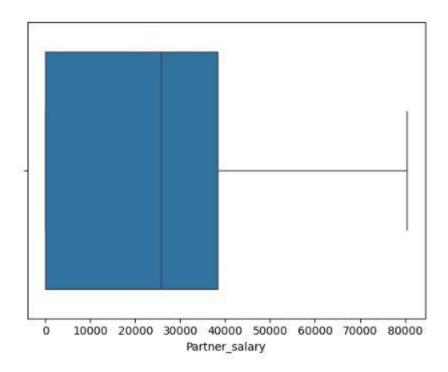
**Box Plot of No of Dependents** 



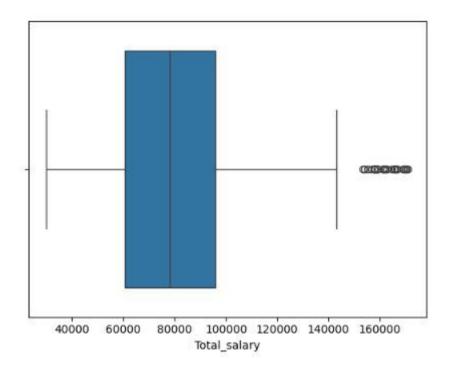
#### **Box Plot of Salary**



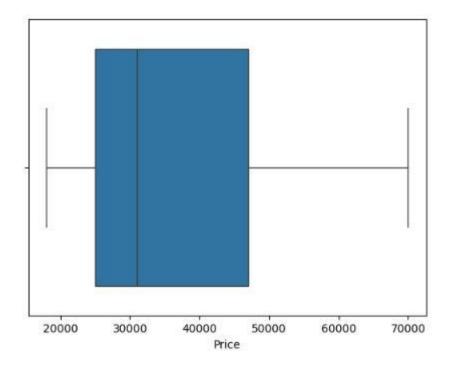
**Box Plot of Partner Salary** 



#### **Box Plot of Total Salary**



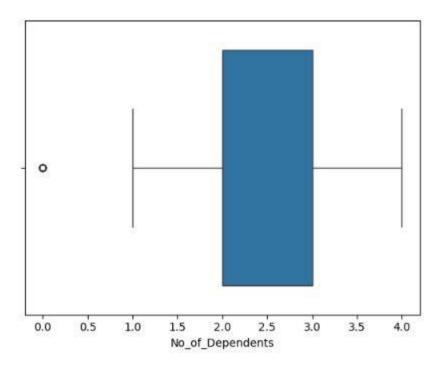
**Box Plot of Price** 



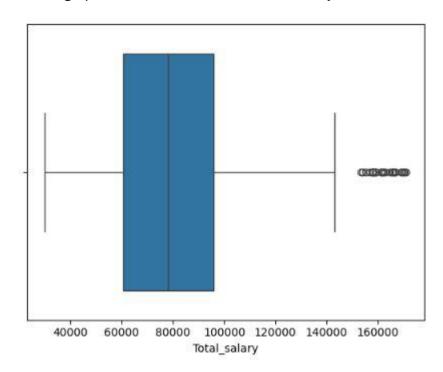
#### 12.) Identification and treatment of outliers

We can see that there are some outliers present in few variables of this Data, which can impact the analysis of Data, This leads to have bad effects on data analysis.

Treating outliers appropriately is most important for Stable results in Data Analysis Below graph shows the variable of No of Dependents



Below graph shows the variable of Total Salary



#### Treatment of outliers:

By using Median technique, treatment of outliers can be done, In this Outliers has been replaced by Median Value.

#### 13.) Observations & Insights

Data had outliers which has been treated now data has been fully ready for analysis.

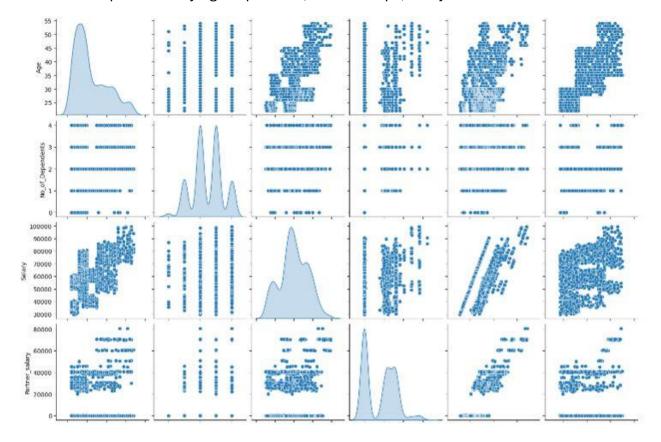
- Analysis shows dominance of Salaried male who is married.
- Personal loan doesn't really make any difference in buying a car as per data.
- Sedan is the most preferred choice as per car.

## **Bivariate Analysis**

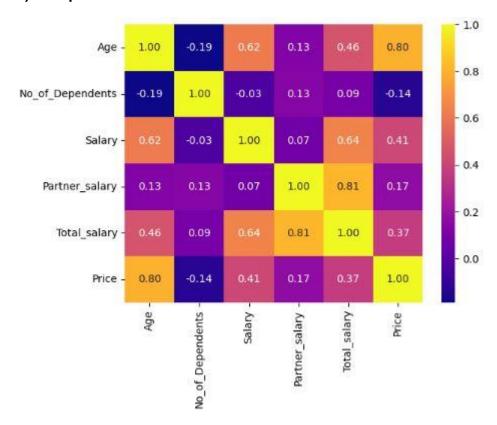
#### 14.) Explore the relationship between all numerical variables

By using Pair Plot to draw relationship between all the numerical values. This chart shows and represents the relationship between all numerical variables.

This Chart helps in identifying the patterns, relationships, analysis.



#### 15.) Explore the correlation between all numerical variables

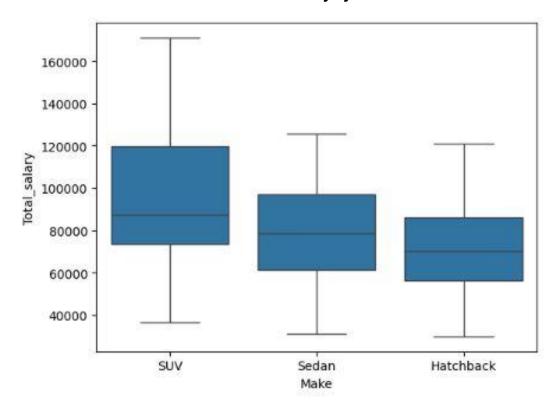


#### 16.) Explore the relationship between categorical vs numerical variables

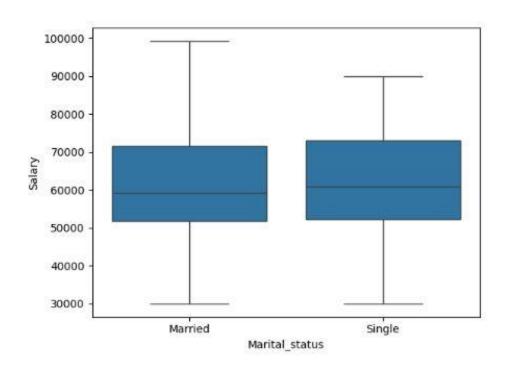
Co-related a few important Categorical and numerical variables by using Box Plot to explore these relationships

- Total Salary Vs Make
- Marital Status Vs Salary
- Education Vs Salary
- Personal Loan Vs Total Salary
- House Loan Vs Price
- Gender Vs Age
- Car Choice Across Age Groups

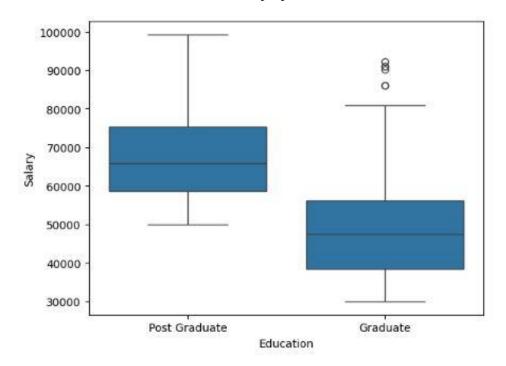
#### **Box Plot of Total Salary by Make**



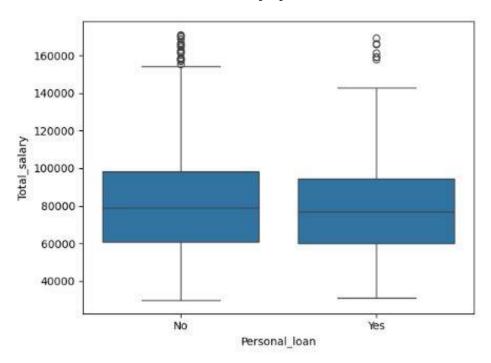
### **Box Plot of Salary by Marital Status**



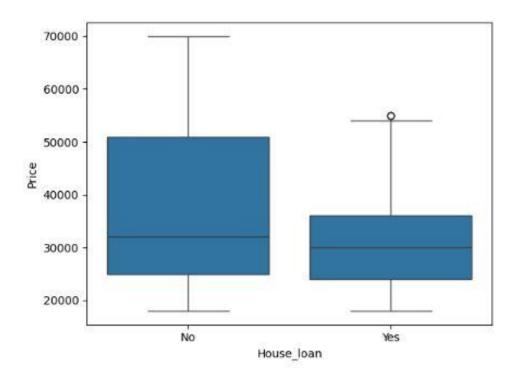
#### **Box Plot of Salary by Education**



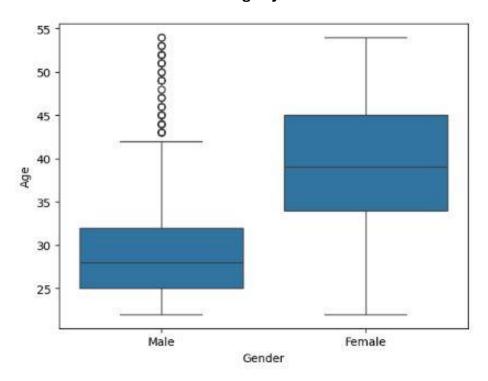
#### **Box Plot of Total Salary by Personal Loan**



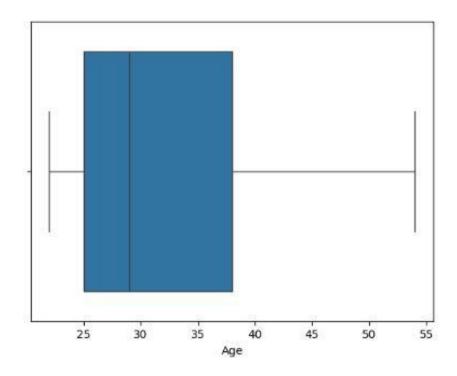
#### **Box Plot of Price by House Loan**

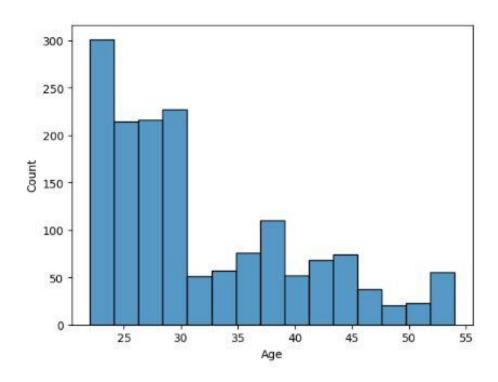


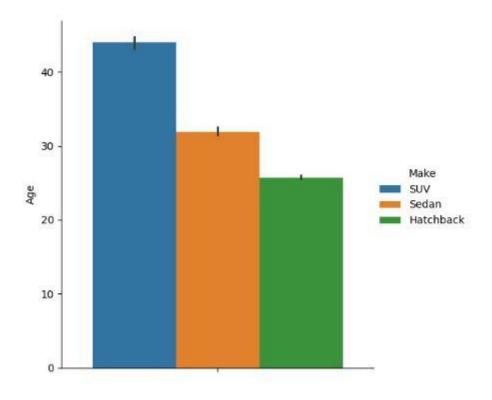
## Box Plot of Age by Gender



## Car Choices across Age Groups



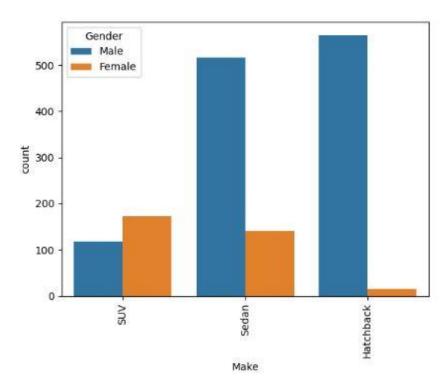




## **Key Questions**

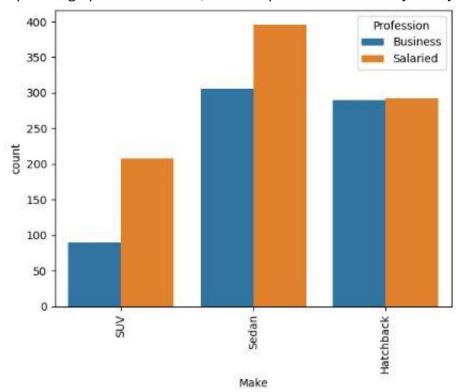
#### 17.) Do men tend to prefer SUVs more compared to women?

No, As per the Graph shown below, Female tend to prefer SUVs more as compared to men



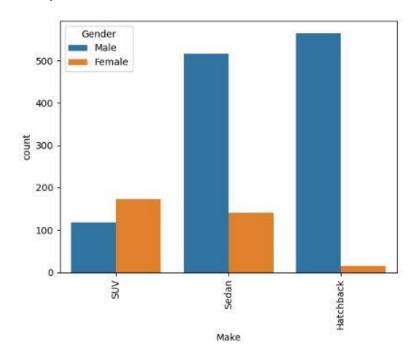
#### 18.) What is the likelihood of a salaried person buying a Sedan?

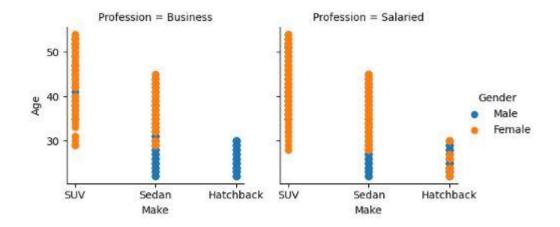
As per the graph shown Below, Salaried person is more likely to buy Sedan.



## 19.) What Evidence or data supports Sheldon Cooper's Claim that a salaried male is an easier target for a SUV sale over a Sedan sale?

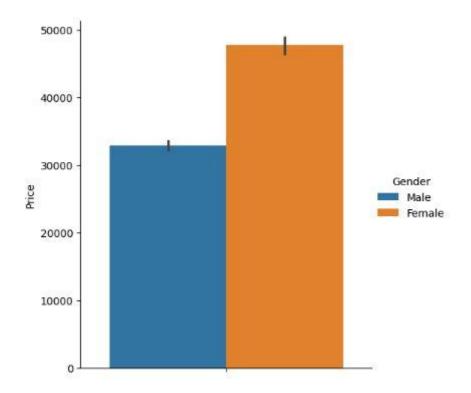
As per the graph shown below, SUV's are more preferred by Salaried women as compared to Salaried Men.





#### 20.) How does the amount spend on purchasing automobiles vary by gender?

As per below mentioned graph, it suggests that Female are spending more money on purchasing Automobiles as compared to Male



**21.)** How much money was spent on purchasing automobiles by individuals who took a personal loan?

Below attached data shows the average amount of money spent by individuals who took Personal loan.

| Descend loss  | Price        |
|---------------|--------------|
| Personal_loan |              |
| No            | 36742.712294 |
| Yes           | 34457.070707 |
| thme: float64 |              |

dtype: float64

**22.)** How does having a working partner influence the purchase of higher-priced cars ?

Data Shows that there is a moderate influence for purchase of higher-priced cars

|               | Price        |
|---------------|--------------|
| Partner_worki | ing          |
| No            | 36000.000000 |
| Yes           | 35267.281106 |

dtype: float64

## **Actionable Insights & Recommendations**

#### 23.) Actionable Insights & Business Recommendations

• Demographic Analysis

Insight: This Dataset is Majorly in favour as a Male Salaried dominant and married people, Most of them have Post Graduate Degree.

Recommendation: Need to make a plan to reach out to more single people for buying a car and Hatchback can be promoted for salary Professionals and Sedan can be promoted more to increase more revenue.

#### • Income and Affordability

Insight: Most people have working partners with partner earning less. Data shows people have taken Personal loans more than House Loans

Recommendation: We have seen from the data that people with working partners are not spending much on buying higher prices cars. Hatchbacks with higher prices can be recommended to them considering their high income, Dependents and martial status.

#### Loan Analysis

Insight: Many of the people in this data set have taken personal loans. Data shows that people have taken Personal loans more as compared to House loans

Recommendation: collaboration with financial organizations to offer attractive loan packages with less interest or any other facilities for car buyers with lesser loan liabilities.

#### • Gender Analysis

Insight: It has been observed that Female prefers less buying cars as compared to male

Recommendation: We can target Salaried male or female, depend upon the place of their organization, if they can communicate by car.

#### Make Preference Analysis

Insight: We have seen that sedan is the most preferred choice followed by hatchback and SUV is the least Preferred.

Recommendation: Need to target audience for SUV's Sales by creating Digital marketing strategies by customized offers and other Discounted Liabilities.

#### Professional Analysis

Insight: It has been observed that Business person are Less preferring buying any of the cars including SUV and Sedan as compared to Salaried person, But in terms of Hatchback its almost same.

Recommendation: Need to target the Business owners with Affordable Hatchback in the budget or according to their need and preference including some business benefits, if they buy with their company name.

#### • Choice Analysis

Insight: According to Analysis, it has been observed that Male prefers buying Sedan and Hathback more & Female prefers buying SUV more as compared to Men.

Recommendation: When we taget the audience we should kept in mind while doing online campaigns, We can taget more SUV to female, and target Sedan, Hatchback to male, in order to get more revenue and sales.

**THANK YOU**