Impact of Car Features on Price and Profitability

Link for Excel sheet:

https://docs.google.com/spreadsheets/d/10QceKTy49wcBmaG8W8Pht OBoad8L1MI_/edit?usp=sharing&ouid=107365393175079460343&rt pof=true&sd=true

Project Description:

This project aims to analyze a dataset containing information about various Car Brands, Car models they make and their respective car features along with their prices. The goal is to gain insights about impact of car features on price and profitability, performing various analysis tasks and also build a dashboard to better visualize the insights. The data provided has various missing or null Data, our task is to handle those missing values appropriately, by either deleting or imputing these data. There are various outliers in data, we have to find these outliers. We utilize various excel features such as pivot tables and charts to better represent data. We find trends in car features and their popularities by implementing various methodologies and data analysis techniques such as regression. Thus, by employing statistics and Excel formulas, we will extract meaningful conclusions to help understand the factors that contribute to popularity and profitability of particular cars.

Approach:

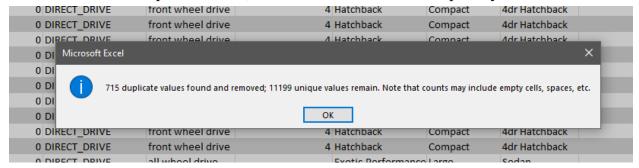
As an individual working on this project, I followed a structured approach to analyze data about Car Brands, models and features. I began by carefully examining the provided database and familiarizing myself with its structure and columns. I tried to find columns which had the most significance in the dataset. I handled missing values by eliminating columns which had most empty cells, and were not significant. And imputed data into cells that were necessary for analysis. Then, I utilized Excel fundamentals to retrieve the necessary information for each task, employing appropriate functions and statistical methods. I focused on data accuracy and quality throughout the project, ensuring reliable results. By leveraging my Excel skills and maintaining a systematic workflow, I successfully executed the project and created a comprehensive report that fulfilled the objectives of providing marketing insights and investor metrics.

Tech-Stack Used:

For this project, I utilized Microsoft Excel as the primary software tool.

Data Cleaning:

Given Data had various missing and duplicate values. For accurate analysis we need to handle this missing data, and eliminate the duplicate data as it is redundant and might skew the results. For Removing the duplicate data we used excel's Remove Duplicates feature in the Data Tools Tab. We had 715 duplicate rows, which were removed completely.



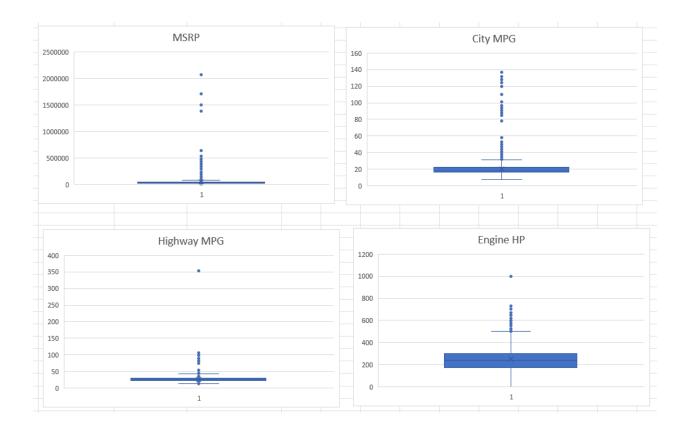
To find missing values we used the COUNTBLANK formula in excel. =COUNTBLANK(A\$2:A\$11160)

| Columns | No. of Null values | Count N/A or Unknown |
|-------------------------|--------------------|----------------------|
| Make | 0 | 0 |
| Model | 0 | 0 |
| Year | 0 | 0 |
| Engine Fuel Type | 3 | 0 |
| Engine HP | 69 | 0 |
| Engine Cylinders | 30 | 0 |
| Transmission Type | 0 | 0 |
| Driven_Wheels | 0 | 0 |
| Number of Doors | 6 | 0 |
| Market Category | 0 | 3376 |
| Vehicle Size | 0 | 0 |
| Vehicle Style | 0 | 0 |
| highway MPG | 0 | 0 |
| city mpg | 0 | 0 |
| Popularity | 0 | 0 |
| MSRP | 0 | 0 |

We removed rows which had less no. of nulls and imputed values in columns such as Engine HP and Engine Cylinders according to the given data.

Data also had some outliers or false values, which needed to be handled. We plotted these outliers using BOX and Whisker chart type.

As seen in the The chart below the features have outliers, some of which are justified but, feature Highway MPG has value which is a bit out of range. So we check with the data of similar Cars and adjust it accordingly.



Insights:

Analysis:

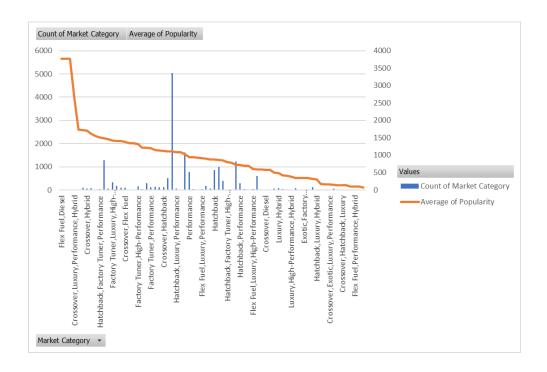
Task 1:

Insight Required: How does the popularity of a car model vary across different market categories?

To perform this task we utilized a pivot table in excel that shows the number of car models in each market category and their corresponding popularity scores.

| 4 | * | U | |
|----|---|--------------------------|-----------------------|
| 1 | Market Category | Count of Market Category | Average of Popularity |
| 2 | Flex Fuel,Diesel | 16 | 5657 |
| 3 | Hatchback,Flex Fuel | 7 | 5657 |
| 4 | Crossover, Flex Fuel, Performance | 6 | 5657 |
| 5 | Crossover,Luxury,Performance,Hybrid | 2 | 3916 |
| 6 | Crossover, Factory Tuner, Luxury, Performance | 5 | 2607.4 |
| 7 | Crossover, Performance | 69 | 2585.956522 |
| 8 | Crossover,Hybrid | 42 | 2563.380952 |
| 9 | Diesel,Luxury | 47 | 2416.106383 |
| 10 | Luxury,Performance,Hybrid | 11 | 2333.181818 |
| 11 | Hatchback, Factory Tuner, Performance | 20 | 2271.9 |
| 12 | Flex Fuel | 855 | 2225.71345 |
| 13 | Crossover,Luxury,Diesel | 33 | 2195.848485 |
| 14 | Factory Tuner,Luxury,High-Performance | 215 | 2133.367442 |
| 15 | Hybrid | 121 | 2116.586777 |
| 16 | Hatchback, Hybrid | 64 | 2111.15625 |
| 17 | Crossover,Flex Fuel | 64 | 2073.75 |
| 18 | Crossover Hatchback Factory Tuner Performance | 6 | 2009 |

This pivot table shows Market Category with its count and average popularity for each. From the above pivot table we plot a combo chart of column-line charts. We select a secondary axis for count to better visualize the chart.

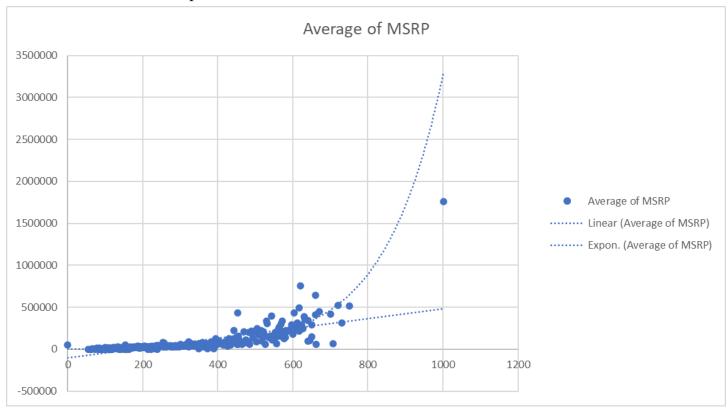


Task 2:

Insight Required: What is the relationship between a car's engine power and its price? To find the relationship between a car's engine power that is Engine HP and its MSRP, we utilize power pivot to find average MSRP for each Engine HP. We then copy this data into a new table and then create a scatter plot of Engine HP vs average MSRP.

| Δ | A | R | C | ט | E | F |
|----------|-----------|-----------------|---|-------------|-------------------|---|
| 1 | Engine HP | Average of MSRP | | Engine HP 🔻 | Average of MSRP 🔻 | |
| 2 | 163 | 2000 | | 163 | 2000 | |
| 3 | 114 | 2000 | | 114 | 2000 | |
| 4 | 102 | 2000 | | 102 | 2000 | |
| 5 | 105 | 2000 | | 105 | 2000 | |
| 6 | 63 | 2000 | | 63 | 2000 | |
| 7 | 113 | 2000 | | 113 | 2000 | |
| 8 | 73 | 2000 | | 73 | 2000 | |
| 9 | 62 | 2000 | | 62 | 2000 | |
| 10 | 96 | 2000 | | 96 | 2000 | |
| 11 | 97 | 2000 | | 97 | 2000 | |
| 12 | 82 | 2000 | | 82 | 2000 | |
| 13 | 81 | 2000 | | 81 | 2000 | |
| 4 | 90 | 2000 | | 90 | 2000 | |
| 15 | 118 | 2000 | | 118 | 2000 | |
| 16 | 92 | 2000 | | 92 | 2000 | |
| 7 | 55 | 2000 | | 55 | 2000 | |
| 18 | 214 | 2000 | | 214 | 2000 | |
| | 1 | 2222 | | | 2000 | |

We now Create a scatter plot for the above table.



We have also added trendlines to understand how MSRP is changing according to the change in Engine HP. Trend seems to increase exponentially rather than linearly, but to predict more accurately we need to have more data available.

Task 3:

Insight Required: Which car features are most important in determining a car's price? To perform this analysis, we need to consider every feature which is correlated with the price of a car. For this we need to perform regression analysis and then plot coefficients of each feature to check which have most impact on MSRP. But for regression analysis we need to have numerical data, so we first convert the data into numerical data by converting categorical data into encoded data.

| I | | |
|---|---------------------|----------|
| 2 | Vehicle Size | Encoding |
| 3 | Compact | 1 |
| 4 | Large | 3 |
| 5 | Midsize | 2 |
| 6 | | |
| 7 | Vehicle Style | Encoding |
| 8 | Coupe | 1 |
| 9 | Sedan | 2 |
| 0 | Convertible | 3 |
| 1 | 4dr SUV | 4 |
| 2 | Wagon | 5 |
| 3 | Crew Cab Pickup | 6 |
| 4 | Extended Cab Pickup | 7 |
| 5 | 4dr Hatchback | 8 |
| 6 | Regular Cab Pickup | 9 |
| | | |

We use this type of conversion to encode data into numerical values.

| D | F | Н | | J | | K | L | N | P | Q |
|---|------------------|------|-----|----------------------------|-------|---------|------------------|-----------------------------|--------------------------|-------------------------|
| | Make (Encoded) ▼ | Year | - | Engine Fuel Type (Encoded) | Engir | ne HP → | Engine Cylinders | Transmission Type (Encoded) | Driven Wheels (Encoded 🔻 | Number of Doors ▼ Vehic |
| | 1 | . 20 | 800 | 1 | | 1001 | 16 | 3 | 3 | 2 |
| | 1 | 20 | 009 | 1 | | 1001 | 16 | 3 | 3 | 2 |
| | 1 | . 20 | 800 | 1 | | 1001 | 16 | 3 | 3 | 2 |
| | 2 | 20 | 16 | 2 | | 1000 | 0 | 1 | 3 | 4 |
| | 2 | 20 | 016 | 2 | | 1000 | 0 | 1 | 3 | 4 |
| | 2 | 20 | 15 | 2 | | 1000 | 0 | 1 | 3 | 4 |
| | 2 | 20 |)14 | 2 | | 1000 | 0 | 1 | 3 | 4 |
| | 2 | 20 | 14 | 2 | | 1000 | 0 | 1 | 1 | 4 |
| | 2 | 20 | 016 | 2 | | 1000 | 0 | 1 | 3 | 4 |
| | 2 | 20 | 15 | 2 | | 1000 | 0 | 1 | 3 | 4 |
| | 2 | 20 | 015 | 2 | | 1000 | 0 | 1 | 1 | 4 |

We get this type of data. But we have to normalize it first. As the parameters have very large differences in their ranges.

To normalize we find maximum and minimum values in each column and then, normalize them using the following formula.

=(Analysis_Task3!\$F2-Analysis_Task3!F\$11199)/Analysis_Task3!F\$11201

Here we subtract minimum values from each value and then divide with the difference between maximum and minimum values, to get normalized values between 0 and 1.

Thus all of the values will get converted into range from 0 to 1.

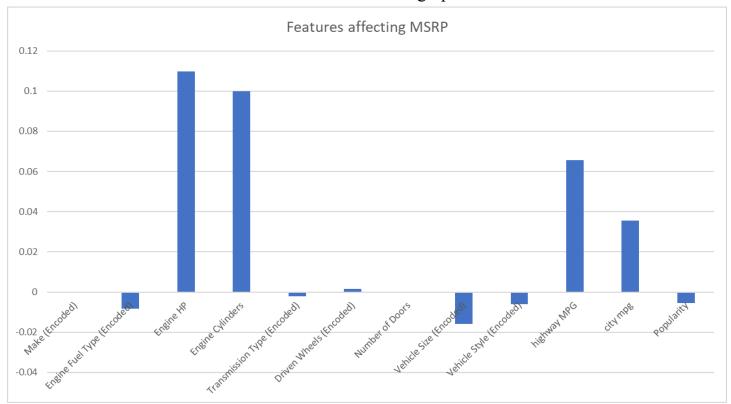
| - 51 | M | 0 | U | L . | 1 | U | - 11 | | , | N. | L | IVI | |
|------|--------------------|---|--------------------|-------------------------------|---------------------------|-------------------|--------------------------|---------------------------|---------------|------------|--------------|----------|--|
| | Make (Encoded) ▼ E | ngine Fuel Type (Encoded) 💌 Engine HP 🔻 | Engine Cylinders 💌 | Transmission Type (Encoded) 🔻 | Driven Wheels (Encoded) 💌 | Number of Doors 💌 | Vehicle Size (Encoded) 🔻 | Vehicle Style (Encoded) 🔻 | highway MPG 🔻 | city mpg 🔻 | Popularity 🔻 | MSRP 💌 | |
| | 0 | 01 | 1 | 0.666666667 | 0.666666667 | 0 | 0 | 0 | 0.005847953 | 0.0076923 | 0.144650752 | 1 | |
| | 0 | 0 1 | 1 | 0.66666667 | 0.66666667 | 0 | 0 | 0 | 0.005847953 | 0.0076923 | 0.144650752 | 0.825509 | |
| | 0 | 01 | . 1 | 0.666666667 | 0.666666667 | 0 | 0 | 0 | 0.005847953 | 0.0076923 | 0.144650752 | 0.72581 | |
| | 0.021276596 | 0.2 0.998942918 | 0 | 0 | 0.666666667 | 1 | 1 | 0.066666667 | 0.271929825 | 0.6538462 | 0.245623342 | 0.064199 | |
| | 0.021276596 | 0.2 0.998942918 | 0 | 0 | 0.666666667 | 1 | 1 | 0.066666667 | 0.257309942 | 0.6461538 | 0.245623342 | 0.053297 | |
| | 0.021276596 | 0.2 0.998942918 | 0 | 0 | 0.66666667 | 1 | 1 | 0.066666667 | 0.251461988 | 0.6307692 | 0.245623342 | 0.049905 | |
| | 0.021276596 | 0.2 0.998942918 | 0 | 0 | 0.66666667 | 1 | 1 | 0.066666667 | 0.239766082 | 0.6076923 | 0.245623342 | 0.049663 | |
| 1 | 0.021276596 | 0.2 0.998942918 | 0 | 0 | 0 | 1 | 1 | 0.066666667 | 0.228070175 | 0.6230769 | 0.245623342 | 0.044285 | |
|) | 0.021276596 | 0.2 0.998942918 | 0 | 0 | 0.666666667 | 1 | 1 | 0.066666667 | 0.277777778 | 0.7230769 | 0.245623342 | 0.042395 | |
| 1 | 0.021276596 | 0.2 0.998942918 | 0 | 0 | 0.666666667 | 1 | 1 | 0.066666667 | 0.274853801 | 0.6769231 | 0.245623342 | 0.040215 | |
| 2 | 0.021276596 | 0.2 0.998942918 | 0 | 0 | 0 | 1 | 1 | 0.066666667 | 0.228070175 | 0.6230769 | 0.245623342 | 0.037792 | |
| 3 | 0.021276596 | 0.2 0.998942918 | 0 | 0 | 0 | 1 | 1 | 0.066666667 | 0.228070175 | 0.6230769 | 0.245623342 | 0.037744 | |
| 4 | 0.021276596 | 0.2 0.998942918 | 0 | 0 | 0.66666667 | 1 | 1 | 0.066666667 | 0.271929825 | 0.7307692 | 0.245623342 | 0.03755 | |
| 5 | 0.021276596 | 0.2 0.998942918 | 0 | 0 | 0.66666667 | 1 | 1 | 0.066666667 | 0.263157895 | 0.7230769 | 0.245623342 | 0.03537 | |
| 5 | 0.021276596 | 0.2 0.998942918 | 0 | 0 | 0.666666667 | 1 | 1 | 0.066666667 | 0.263157895 | 0.7230769 | 0.245623342 | 0.03537 | |
| 7 | 0.021276596 | 0.2 0.998942918 | 0 | 0 | 0 | 1 | 1 | 0.066666667 | 0.257309942 | 0.6923077 | 0.245623342 | 0.035128 | |
| 3 | 0.021276596 | 0.2 0.998942918 | 0 | 0 | 0.666666667 | 1 | 1 | 0.066666667 | 0.27777778 | 0.7230769 | 0.245623342 | 0.033432 | |
| 9 | 0.021276596 | 0.2 0.998942918 | 0 | 0 | 0 | 1 | 1 | 0.066666667 | 0.228070175 | 0.6230769 | 0.245623342 | 0.032947 | |
|) | 0.021276596 | 0.2 0.998942918 | 0 | 0 | 0 | 1 | 1 | 0.066666667 | 0.248538012 | 0.6692308 | 0.245623342 | 0.032899 | |
| 1 | 0.021276596 | 0.2 0.998942918 | 0 | 0 | 0 | 1 | 1 | 0.066666667 | 0.248538012 | 0.6692308 | 0.245623342 | 0.032899 | |
| 2 | 0.021276596 | 0.2 0.998942918 | 0 | 0 | 0.666666667 | 1 | 1 | 0.066666667 | 0.260233918 | 0.7 | 0.245623342 | 0.031009 | |
| 3 | 0.042553191 | 0 0.734672304 | 0.75 | 0.666666667 | 0.666666667 | 0 | 0.5 | 0.133333333 | 0.01754386 | 0.0307692 | 0.204420866 | 0.258491 | |
| 4 | 0.042553191 | 0 0.734672304 | 0.75 | 0.666666667 | 0.666666667 | 0 | 0.5 | 0 | 0.01754386 | 0.0307692 | 0.204420866 | 0.236784 | |
| 5 | 0.063829787 | 0 0.714587738 | 0.75 | 0.666666667 | 0 | 0 | 0.5 | 0 | 0.011695906 | 0.0307692 | 0.490185676 | 0.154075 | |
| 5 | 0.063829787 | 0 0.714587738 | 0.75 | 0.666666667 | 0 | 0 | 0.5 | 0 | 0.011695906 | 0.0307692 | 0.490185676 | 0.152085 | |
| 7 | 0.063829787 | 0 0.714587738 | 0.75 | 0.666666667 | 0 | 0 | 0.5 | 0 | 0.011695906 | 0.0307692 | 0.490185676 | 0.152085 | |
| 3 | 0.042553191 | 0 0.702959831 | 0.75 | 0.666666667 | 0.66666667 | 0 | 0.5 | 0.133333333 | 0.011695906 | 0.0230769 | 0.204420866 | 0.264935 | |
| 9 | 0.042553191 | 0 0.702959831 | 0.75 | 0.666666667 | 0.666666667 | 0 | 0.5 | 0.133333333 | 0.011695906 | 0.0230769 | 0.204420866 | 0.264935 | |
|) | 0.042553191 | 0 0.702959831 | 0.75 | 0.666666667 | 0.666666667 | 0 | 0.5 | 0 | 0.01754386 | 0.0307692 | 0.204420866 | 0.240152 | |
| 1 | 0.042553191 | 0 0.702959831 | 0.75 | 0.666666667 | 0.66666667 | 0 | 0.5 | 0 | 0.01754386 | 0.0307692 | 0.204420866 | 0.240152 | |
| 2 | 0.085106383 | 0 0.689217759 | 0.5 | 0.333333333 | 0 | 0 | 1 | 0 | 0.026315789 | 0.0461538 | 0.326967286 | 0.030983 | |
| 3 | 0.085106383 | 0 0.689217759 | 0.5 | 0.333333333 | 0 | 1 | 1 | 0.066666667 | 0.029239766 | 0.0461538 | 0.326967286 | 0.030983 | |
| 4 | 0.085106383 | 0 0.689217759 | 0.5 | 0.333333333 | 0 | 1 | 1 | 0.066666667 | 0.029239766 | 0.0461538 | 0.326967286 | 0.030983 | |
| 5 | 0.085106383 | 0 0.689217759 | 0.5 | 1 | 0 | 0 | 1 | 0 | 0.026315789 | 0.0461538 | 0.326967286 | 0.029311 | |
| _ | | | | | | | | | | | | | |

We now use this data to perform regression analysis by using the Data Analysis feature in the Data menu. We get following output:

| SUMMARY OUTPUT | | | | | | | | |
|-----------------------------|--------------|----------------|--------------|-------------|----------------|--------------|--------------|-------------|
| Regression Statisti | ics | | | | | | | |
| Multiple R | 0.693768659 | | | | | | | |
| R Square | 0.481314952 | | | | | | | |
| Adjusted R Square | 0.480758373 | | | | | | | |
| Standard Error | 0.021486617 | | | | | | | |
| Observations | 11196 | | | | | | | |
| ANOVA | | | | | | | | |
| | df | SS | MS | F | Significance F | | | |
| Regression | 12 | 4.790932299 | 0.399244358 | 864.7741572 | 0 | | | |
| Residual | 11183 | 5.162908281 | 0.000461675 | | | | | |
| Total | 11195 | 9.95384058 | | | | | | |
| | Coefficients | Standard Error | t Stat | P-value | Lower 95% | Upper 95% | Lower 95.0% | Upper 95.0% |
| Intercept | -0.031589947 | 0.001778814 | -17.75899174 | 1.32073E-69 | -0.035076737 | -0.028103158 | -0.035076737 | -0.02810315 |
| Make (Encoded) | -0.000126829 | 0.000898814 | -0.141107322 | 0.887787701 | -0.001888663 | 0.001635004 | -0.001888663 | 0.00163500 |
| Engine Fuel Type (Encoded) | -0.008170822 | 0.000930401 | -8.782046676 | 1.83872E-18 | -0.009994572 | -0.006347073 | -0.009994572 | -0.00634707 |
| Engine HP | 0.109738052 | 0.003028023 | 36.24082452 | 5.1054E-272 | 0.103802594 | 0.115673511 | 0.103802594 | 0.11567351 |
| Engine Cylinders | 0.099772838 | 0.003461532 | 28.82331736 | 2.7635E-176 | 0.092987626 | 0.106558051 | 0.092987626 | 0.10655805 |
| Transmission Type (Encoded) | -0.002025972 | 0.00084897 | -2.386386485 | 0.017031389 | -0.003690103 | -0.00036184 | -0.003690103 | -0.0003618 |
| Driven Wheels (Encoded) | 0.001653045 | 0.000658457 | 2.510481034 | 0.012070678 | 0.000362352 | 0.002943737 | 0.000362352 | 0.00294373 |
| Number of Doors | -0.000224093 | 0.000559842 | -0.400279344 | 0.688958426 | -0.001321482 | 0.000873295 | -0.001321482 | 0.00087329 |
| Vehicle Size (Encoded) | -0.016006733 | 0.000670551 | -23.87101698 | 7.0368E-123 | -0.017321131 | -0.014692335 | -0.017321131 | -0.01469233 |
| Vehicle Style (Encoded) | -0.006067337 | 0.001019628 | -5.950541158 | 2.75268E-09 | -0.008065987 | -0.004068687 | -0.008065987 | -0.00406868 |
| highway MPG | 0.065650438 | 0.017952962 | 3.65680265 | 0.000256553 | 0.030459471 | 0.100841405 | 0.030459471 | 0.10084140 |
| city mpg | 0.035422796 | 0.006433244 | 5.506210878 | 3.74738E-08 | 0.022812505 | 0.048033086 | 0.022812505 | 0.04803308 |
| | | | | | | | | |

By using this we can plot a bar graph to see which features are affecting the MSRP most.

We select coefficients of each features and create a bar graph as below:

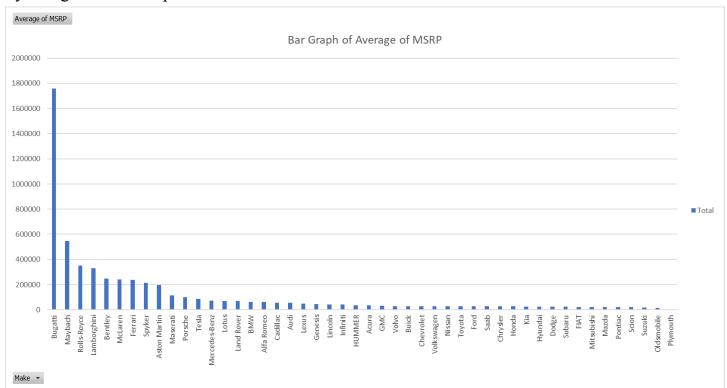


As we can see from above graph that Engine HP, Cylinders and MPG are some of the deciding factors for the MSRP of a car.

Task 4: Insight Required: How does the average price of a car vary across different manufacturers? For doing this task we utilized pivot tables and found the average price of a car for each car manufacturer.

| | A | | ט | |
|---|------------------|----|-----------------|--|
| | Car Manufactures | Ψļ | Average of MSRP | |
| ! | Bugatti | | 1757223.667 | |
| | Maybach | | 546221.875 | |
| Ļ | Rolls-Royce | | 351130.6452 | |
| , | Lamborghini | | 331567.3077 | |
| į | Bentley | | 247169.3243 | |
| • | McLaren | | 239805 | |
| | Ferrari | | 238218.8406 | |
| • | Spyker | | 214990 | |
| 0 | Aston Martin | | 198123.4615 | |
| 1 | Maserati | | 113684.4909 | |
| 2 | Porsche | | 101622.3971 | |
| 3 | Tesla | | 85255.55556 | |
| 4 | Mercedes-Benz | | 72135.02647 | |
| 5 | Lotus | | 68377.14286 | |
| 6 | Land Rover | | 68067.08633 | |

By using this we can plot a column chart to visualize this data.



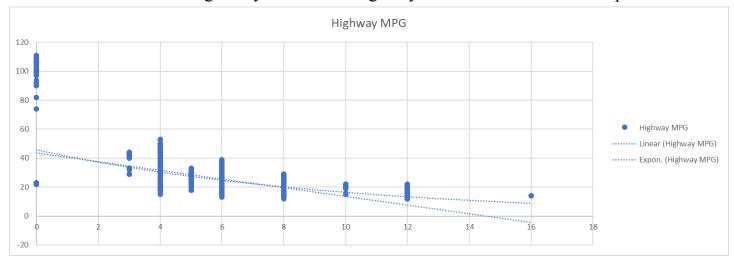
We can see here that brands such as Bugatti have very high average MSRP as they are into high end cars and don't have any cars in lower price segments.

Task 5:

Insight Required: What is the relationship between fuel efficiency and the number of cylinders in a car's engine?

To find relationships between fuel efficiency and number of cylinders in a car's engine we have to create a scatter plot of number of cylinders vs its MPG and see if there exists any trend by plotting a trendline.

We select two columns Engine Cylinders and Highway MPG and create a scatter plot.



We also create a correlation matrix to check if there exist any correlation between them. We create a pivot table of no. of Cylinders and Highway MPG and City MPG. We create a correlation matrix by using the CORREL function in excel and conditional formatting.

| Α | В | С | D | E | F | | G | | H |
|--------------------------|--------------------|---------------------|---|-----------|----------------|------|-----------------------|-------|---------------------|
| No. of Cylinders 🔻 Avera | age of highway MPG | Average of city mpg | | N | No. of Cylinde | rs 🔻 | Average of highway MI | PG | Average of city mpg |
| 0 | 81.6627907 | 90.1744186 | | O |) | | 81.662 | 27907 | 90.174418 |
| 3 | 38.66666667 | 32.03333333 | | 3 | } | | 38.6666 | 66667 | 32.0333333 |
| 1 | 31.50057484 | 23.9029662 | | 4 | l | | 31.5005 | 7484 | 23.902966 |
| i | 26.06508876 | 18.77514793 | | 5 | j | | 26.0650 | 08876 | 18.7751479 |
| i | 24.00679634 | 17.13452074 | | 6 | j | | 24.0067 | 79634 | 17.1345207 |
| } | 20.17278287 | 14.18399592 | | 8 | 3 | | 20.1727 | 78287 | 14.1839959 |
| 0 | 20 | 12.56923077 | | 1 | .0 | | | 20 | 12.5692307 |
| .2 | 17.73684211 | 11.25 | | 1 | .2 | | 17.7368 | 34211 | 11.2 |
| .6 | 14 | 8 | | 1 | .6 | | | 14 | |
| Grand Total | 26.61403352 | 19.73214446 | | | | | | | |
| J | | K | | | | | L | | М |
| _ | Correlati | on Between Cylin | | d Highway | MPG | | | | |
| | No. of Cy | linders | | | Ave | erag | e of highway MPG | Ave | rage of city mpg |
| No. of Cylinders | | | | | 1 | | | | |
| Average of highwa | ay MPG | | | -0.777 | 7122379 | | 1 | | |
| Average of city mp | og | | | -0.729 | 775621 | | 0.996412646 | | 1 |
| | | | | | | | | | |
| | | | | | | | | | |
| | | | | | | | | | |
| | | | | | | | | | |

As we can see there is less correlation between MPG and no. cylinders. But there is high correlation between highway MPG and city MPG.

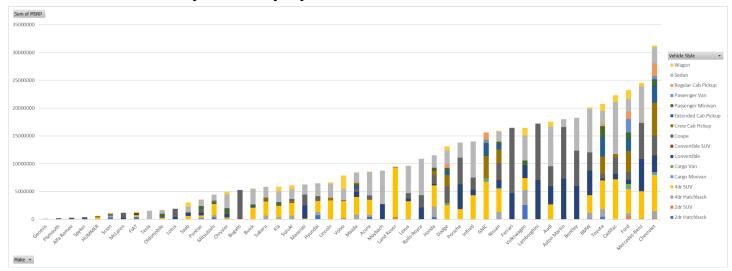
Dashboard:

Task 1: How does the distribution of car prices vary by brand and body style?

We created a stacked column chart of car price for each brand and each body style in that column. To create this chart, we first need to create a pivot table consisting of Sum of MSRP for each category. These categories being, brand of car in row and body style in columns and we get an interactive table consisting of total MSRP for each body style that each car brand makes. Table looks like below:

| A | В | С | D | E | F | G | Н | 1 | J | K | | L | М | | N | | 0 | Р | Q | R |
|----------------|-------------------|-----------|--------------|---------|---------------|-----------|-------------|-----------------|----------|-----------------|----------|--------------|--------------|--------|---------------|---------|------------|---------|---------|--------------------|
| 1 Sum of MSRP | Body Style 🔻 | | | | | | | | | | | | | | | | | | | |
| 2 Car Brand | → 2dr Hatchback 2 | 2dr SUV 4 | dr Hatchback | 4dr SUV | Cargo Minivan | Cargo Van | Convertible | Convertible SUV | Coupe | Crew Cab Pickup | Extended | d Cab Pickup | Passenger Mi | inivan | Passenger Van | Regular | Cab Pickup | Sedan | Wagon | Grand Total |
| 3 Genesis | | | | | | | | | | | | | | | | | | 139850 | | 139850 |
| 4 Plymouth | 40000 | | 14000 | | | | 85631 | | 8000 |) | | | | 31688 | | | | 38759 | 16000 | 234078 |
| 5 Alfa Romeo | | | | | | | 129800 | | 178200 |) | | | | | | | | | | 308000 |
| 6 Spyker | | | | | | | 219990 | | 209990 |) | | | | | | | | | | 429980 |
| 7 HUMMER | | | | 377490 | | | | | | 242405 | | | | | | | | | | 619895 |
| 8 Scion | 366325 | | 282470 | | | | | | 330210 |) | | | | | | | | 32500 | 184445 | 1195950 |
| 9 McLaren | | | | | | | 280225 | | 918800 |) | | | | | | | | | | 1199025 |
| 10 FIAT | 420715 | | | 369305 | | | 327965 | | | | | | | | | | | | 287570 | 1405555 |
| 11 Tesla | | | | | | | | | | | | | | | | | | 1534600 | | 1534600 |
| 12 Oldsmobile | | | | 238150 | | | 2000 | | 274015 | 5 | | | 4 | 492055 | | | | 665161 | 20000 | 1691381 |
| 13 Lotus | | | | | | | 413260 | | 1501300 |) | | | | | | | | | | 1914560 |
| 14 Saab | 12000 | | 34586 | 541905 | | | 632628 | | | | | | | | | | | 1066500 | 751280 | 3038899 |
| 15 Pontiac | 148782 | | 162975 | 401550 | | | 463914 | | 663715 | 5 | | | 5 | 541192 | | | | 1156535 | 20855 | 3559518 |
| 16 Mitsubishi | 370169 | | 403835 | 2009807 | 2000 | | 209893 | | | 240210 | | 134360 |) | 2000 | | | 8000 | 1058563 | | 4438837 |
| 17 Chrysler | 98805 | | | 250545 | | | 628105 | | 112510 |) | | | | 922295 | | | | 2479859 | 501075 | 4993194 |
| 18 Bugatti | | | | | | | | | 527167 | | | | | | | | | | | 5271671 |
| 19 Buick | | | | 2141770 | | | 179325 | | 18534 | 1 | | | | 330065 | | | | 2838590 | 8212 | 5516496 |
| 20 Subaru | 12000 | | 678060 | 2539900 | | | | | 354470 | | | | | | | | | 1833110 | 10000 | 5793521 |
| 21 Kia | | | 406960 | 2049645 | | | | | 142630 | | | | 4 | 494650 | | | | 1976360 | 772405 | 5842650 |
| 22 Suzuki | 44496 | 12000 | 584387 | 2303493 | | | | 12019 | | 304131 | | 259659 | | | | | | 1797070 | 683707 | 6109137 |
| 23 Maserati | | | | 155000 | | | 2342963 | | 1972284 | | | | | | | | | 1782400 | | 6252647 |
| 24 Hyundai | 789650 | | 528880 | 1994390 | | | 20 .2500 | | 685920 | | | | | 133075 | | | | 2323987 | | 6455902 |
| 25 Lincoln | 703000 | | 020000 | 3422570 | | | | | 17342 | | | | | 200010 | | | | 2458245 | 269705 | 6621122 |
| 26 Volvo | 157550 | | | 3131700 | | | 121600 | | 6000 | | | | | | | | | 2072945 | | 7906766 |
| 27 Mazda | 18000 | 12000 | 853180 | 3175515 | | | 870505 | | 541879 | | | 580033 | . 4 | 443130 | | | 265486 | | 33350 | 8411649 |
| 28 Acura | 480917 | 12000 | 357440 | 2663505 | | | 0,0000 | | 793748 | | | 50005. | | | | | 200-101 | 4134552 | 201360 | 8631522 |
| 29 Maybach | 400317 | | 337440 | 2005505 | | | 2762750 | | 755740 | • | | | | | | | | 5976800 | 201300 | 8739550 |
| 30 Land Rover | | 476394 | | 8839200 | | | 2.02.00 | 14573 | ı | | | | | | | | | 0570000 | | 9461325 |
| 31 Lexus | | | 94700 | | | | 472065 | | 101647 | , | | | | | | | | 4837596 | 31105 | 9604912 |
| 32 Rolls-Royce | | | 3.700 | -102574 | | | 2141365 | | 2204675 | | | | | | | | | 6539010 | 52205 | 10885050 |
| 33 Honda | 413200 | | 1919260 | 3800589 | | | 252135 | | 1588705 | | | | | 553185 | | | | 2264390 | | 11541679 |
| 34 Dodge | 38000 | 12000 | | | 60520 | 338497 | 6000 | | 2973842 | | | 684682 | | 557425 | 7070 | 8 | 651408 | | 793055 | 13147377 |
| 35 Porsche | 28827 | 12000 | 10000 | 1815200 | 55520 | 550-57 | 4504586 | | 4758533 | | | 00-1002 | | | ,070 | - | 001400 | 2713500 | , 55055 | 13820646 |
| 36 Infiniti | 20027 | | | 4340200 | | | 980050 | | 2175750 | | | | | | | | | 6490009 | | 13986009 |
| 37 GMC | | 118835 | | 6633919 | 142750 | 460085 | 200030 | | 21/3/30 | 4062482 | | 2175866 | | 150630 | 59967 | n | 1284328 | | | 15628565 |
| 38 Nissan | 14683 | 110000 | 1347320 | | 128620 | 400083 | 1406552 | 12107 | 2937632 | | | 1026379 | | 413320 | 33907 | • | 19914 | | 175000 | 15935555 |
| | 14083 | | 134/320 | 4149030 | 128020 | | 4723811 | | 11713289 | | | 1020375 | , ' | +15320 | | | 19914 | 1/03130 | 1/3000 | 16437100 |
| 39 Ferrari | | | | | | | 4/23811 | | 11/13289 | 7 | | | | | | | | | | 10437100 |

We can plot a stacked column chart from this table where each column in the chart for a particular brand would have the total sum of MSRP of all body styles and sections would have different colors to identify each body style. Chart looks like below:



We can also change chart features by changing the filter in the pivot table.

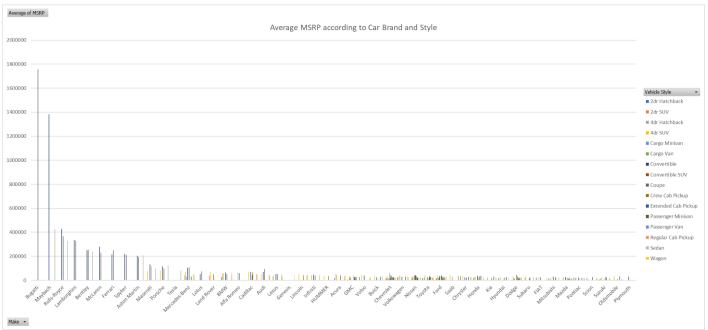
Task 2: Which car brands have the highest and lowest average MSRPs, and how does this vary by body style?

We created a clustered column chart of average car price for each brand and each body style in that column. To create this chart, we first need to create a pivot table consisting of the average of MSRP for each category. These categories being, brand of car in row and body style in columns and we get an interactive table consisting of average MSRP for each body style that each car brand makes.

Pivot table looks like below:

| Average of MSI | RP Body Style 💌 | | | | | | | | | | | | | | | |
|----------------|-----------------|-------------|-----------------|-------------|-------------------------|-------------|-----------------|-------------|-----------------|---------------------|-------------------|---------------|--------------------|-------------|-------------|--------------------|
| Car Brands | → 2dr Hatchback | 2dr SUV | 4dr Hatchback 4 | 4dr SUV | Cargo Minivan Cargo Van | Convertible | Convertible SUV | Coupe | Crew Cab Pickup | Extended Cab Pickup | Passenger Minivan | Passenger Van | Regular Cab Pickup | Sedan | Wagon | Grand Total |
| Bugatti | | | | | | | | 1757223.667 | | | | | | | | 1757223.66 |
| Maybach | | | | | | 1381375 | | | | | | | | 426914.2857 | | 546221.87 |
| Rolls-Royce | | | | | | 428273 | | 367445.8333 | | | | | | 326950.5 | | 351130.645 |
| Lamborghini | | | | | | 336402.381 | | 328291.9355 | | | | | | | | 331567.307 |
| Bentley | | | | | | 250536.25 | | 254270.4 | | | | | | 236836 | | 247169.324 |
| McLaren | | | | | | 280225 | | 229700 | | | | | | | | 23980 |
| Ferrari | | | | | | 214718.6818 | | 249218.9149 | | | | | | | | 238218.840 |
| Spyker | | | | | | 219990 | | 209990 | | | | | | | | 21499 |
| Aston Martin | | | | | | 203379.3056 | | 192892.6042 | | | | | | 206962.1429 | | 198123.461 |
| Maserati | | | | 77500 | | 130164.6111 | | 116016.7059 | | | | | | 99022.22222 | | 113684.490 |
| Porsche | 5765.4 | | | 82509.09091 | | 115502.2051 | | 99136.10417 | | | | | | 123340.9091 | | 101622.397 |
| Tesla | | | | | | | | | | | | | | 85255.55556 | | 85255.55556 |
| Mercedes-Benz | 2 | | 40933.33333 | 68400.13889 | 28950 | 104617.5273 | | 109713.678 | | | 32500 | | | 48833.90299 | 43069 | 72135.02647 |
| Lotus | | | | | | 51657.5 | | 75065 | | | | | | | | 68377.1428 |
| Land Rover | | 39699.5 | | 71283.87097 | | | 48577 | | | | | | | | | 68067.0863 |
| BMW | 26699 | | 55155 | 58536.11111 | | 63814.07246 | | 52445.25397 | | | | | | 71832.11009 | 43266.66667 | 62162.5586 |
| Alfa Romeo | | | | | | 64900 | | 59400 | | | | | | | | 6160 |
| Cadillac | | | | 72551.06061 | | 70400.5 | | 45439.6 | 66572.22222 | | | | | 51178.5163 | 47364 | 56368.2651 |
| Audi | 2000 | | | 48634.54545 | | 70029.89362 | | 93586.57895 | | | | | | 46391.87013 | 33894 | 54574.121 |
| Lexus | | | 31566.66667 | 45042.48571 | | 52451.66667 | | 50823.6 | | | | | | 48864.60606 | 31105 | 47549.0693 |
| Genesis | | | | | | | | | | | | | | 46616.66667 | | 46616.6666 |
| Lincoln | | | | 50331.91176 | | | | 2167.75 | 41205.45455 | | | | | 41665.16949 | 44950.83333 | 43560.0131 |
| Infiniti | | | | 45686.31579 | | 46669.04762 | | 40291.66667 | | | | | | 41076.00633 | | 42640.2713 |
| HUMMER | | | | 37749 | | | | | 34629.28571 | | | | | | | 36464.4117 |
| Acura | 17175.60714 | | 51062.85714 | 42959.75806 | | | | 39687.4 | | | | | | 33614.2439 | 33560 | 35087.487 |
| GMC | | 8488.214286 | | 37479.76836 | 23791.66667 21908.80952 | | | | 39062.32692 | 27895.7179 | 5 25105 | 28555.71429 | 25182.90196 | | | 32695.7426 |
| Volvo | 26258.33333 | | | 45386.95652 | | 40533.33333 | | 2000 | | | | | | 22289.73118 | 26271.42391 | 29724.6842 |
| Buick | | | | 33996.34921 | | 25617.85714 | | 2059.333333 | | | 30005.90909 | | | 29568.64583 | 2053 | 29034.1894 |
| Chevrolet | 2000 | 13807.85714 | 18930.29412 | 33553.95876 | 20007.14286 8298.666667 | 62835 | 17716.66667 | 38939.16667 | 39255.74172 | 24170.1627 | 9 24934.28571 | 28555.71429 | 19824.84211 | 19882.64865 | 15825 | 29018.3500 |
| Volkswagen | 24134.62963 | | 28416.21053 | 41699.1 | | 27673.68675 | | 2000 | | | 29239.67742 | | | 30795.79861 | 26385.64815 | 28978.5228 |
| Nissan | 2097.571429 | | 24059.28571 | 34294.46281 | 21436.66667 | 39070.88889 | 43691.66667 | 35393.15663 | 32733.78378 | 20527.5 | 8 22962.22222 | | 2212.666667 | 22604.23077 | 17500 | 28921.1524 |
| Toyota | 18950 | | 22186.50794 | 40851.6 | | 25777.86667 | | 15615.28846 | 36845.82353 | 26251.3082 | 7 30038.73846 | | 17592.66667 | 24800.27083 | 31742.4359 | 28846.560 |
| Ford | 2000 | 16133.55172 | 19572.93103 | 42027.60577 | 19700 20605.59259 | 34762.2381 | | 34101.07317 | 41566.13187 | 23808.1666 | 7 22587.17391 | 32836.45946 | 17797.80822 | 23258.65306 | 30066.01852 | 28525.1828 |
| Saab | 2000 | | 2034.470588 | 41685 | | 28755.81818 | | | | | | | | 36775.86207 | 34149.09091 | 27879.8073 |
| Chrysler | 32935 | | | 35792.14286 | | 25124.2 | | 22502 | | | 29751.45161 | | | 26103.77895 | 26372.36842 | 26990.2378 |
| Honda | 17216.66667 | | 26656.38889 | 28575.85714 | | 36019.28571 | | 21763.08219 | 34100.68182 | | 36879 | | | 26027.47126 | | 26655.1478 |
| Kia | | | 19379.04762 | 31533 | | | | 20375.71429 | | | 32976.66667 | | | 23811.56627 | 20326.44737 | 25513.7554 |

We can plot a clustered column chart from this table where each cluster in the chart for a particular brand would have the column of average MSRP of all body styles and sections would have different colors to identify each body style. Chart looks like below:

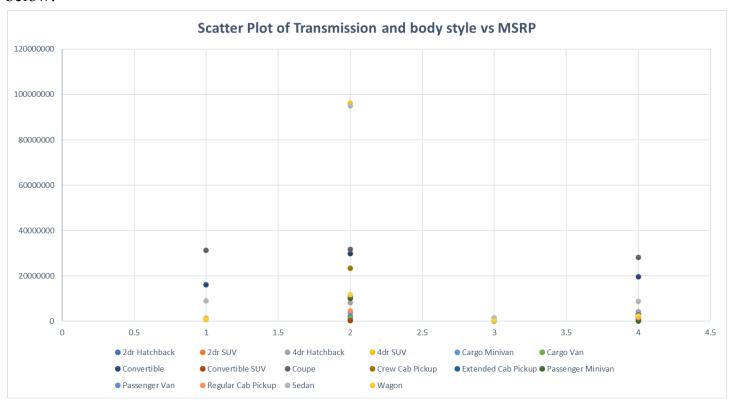


Task 3: How do the different features such as transmission type affect the MSRP, and how does this vary by body style?

To find the effect of transmission type on MSRP we have to create a pivot table and add body style as column and transmission type as row. We consider the average of MSRP to better visualize the data. We get the following pivot table.

| - | | J** | | | | | | | |
|----|-------------------|---------------|-------------|---------------|-------------|---------------|-------------|-------------|-------|
| 4 | Α | В | С | D | Е | F | G | Н | |
| | Average of MSRP | Body Style 🔻 | | | | | | | |
| 2 | Transmission Type | 2dr Hatchback | 2dr SUV | 4dr Hatchback | 4dr SUV | Cargo Minivan | Cargo Van | Convertible | Conve |
| 3 | AUTOMATED_MANUAL | 27470.41667 | | 29347.04545 | 40451.15385 | | | 129082.2339 | |
| Į. | AUTOMATIC | 20784.09901 | 24153.60606 | 23888.73529 | 41658.40017 | 20292.93103 | 17019.29762 | 95153.3131 | |
| 5 | DIRECT_DRIVE | 31800 | | 32799.72973 | 49800 | | | | |
| 5 | MANUAL | 12840.65556 | 9173.018519 | 17500.36364 | 17422.08791 | | | 64794.34437 | 1 |
| 7 | Grand Total | 16220.74634 | 14855.31034 | 22416.46757 | 40747.54467 | 20292.93103 | 17019.29762 | 88439.88633 | |
| 3 | | | | | | | | | |
|) | | | | | | | | | |

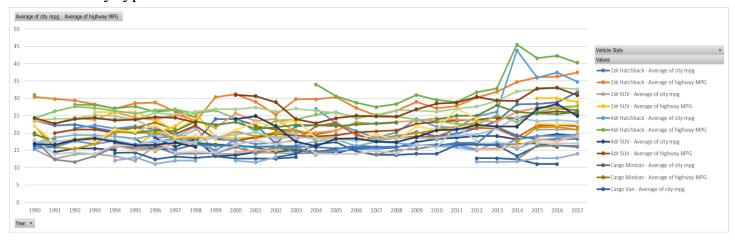
We copy down the contents of the pivot table to create a scatter plot. Scatter plot look like below:



Task 4: How does the fuel efficiency of cars vary across different body styles and model years? To find how fuel efficiency of a car varies across different body styles across different years we create a pivot table consisting of average highway MPG and City MPG across body styles as columns and years as rows. We get following table:

| , | | 11 | Ü | • | - | | | | _ |
|-------------------|------------------------|---------------------|------------------------|---------------------|------------------------|---------------------|------------------------|---------------------------------------|---------|
| | | | | | | | | Body Style 🔻 | |
| Cargo Minivan | | 4dr SUV | | 4dr Hatchback | | 2dr SUV | | 2dr Hatchback | |
| Average of city m | Average of highway MPG | Average of city mpg | Average of highway MPG | Average of city mpg | Average of highway MPG | Average of city mpg | Average of highway MPG | Average of city mpg | Year |
| | | | 31 | 22 | 20 | 15.25 | 30.4 | 23.6 | 1990 |
| | 20 | 14.5 | | | 16.25 | 12.5 | 29.83333333 | 22.16666667 | 1991 |
| | 21 | 15.5 | 28.16666667 | 21.33333333 | 18.28571429 | 13.85714286 | 29.39285714 | 22.39285714 | 1992 |
| | 21 | 15.5 | 28.125 | 22.25 | 18.85714286 | 14 | 28.25925926 | 21.48148148 | 1993 |
| | 20 | 15 | 27.14285714 | 21.28571429 | 17.625 | 13.25 | 27.05263158 | 20.42105263 | 1994 |
| 1 | | | 27.66666667 | 22 | 16 | 12 | 28.6 | 21.6 | 1995 |
| 1 | 21.25 | 18.5 | 26.125 | 18.625 | 20 | 16.2 | 28.8 | 21.2 | 1996 |
| | 19.7 | 16 | 26.66666667 | 18.88888889 | 22 | 18.66666667 | 26.25 | 19.5 | 1997 |
| | 22.11111111 | 18.22222222 | 24.5 | 18 | 26 | 22 | 23.2 | 17.2 | 1998 |
| | 18.3 | 13.3 | | | 18.5 | 14 | 30.33333333 | 24 | 1999 |
| | 17.73333333 | 13.6 | | | 18.5 | 14 | 31.22222222 | 24 | 2000 |
| | 18.72727273 | 14.45454545 | | | 18.66666667 | 14.33333333 | 29 | 22.28571429 | 2001 |
| | 19.79411765 | 15.73529412 | | | 19 | 14.25 | 25.25 | 17 | 2002 |
| 15.166666 | 19.22857143 | 14.97142857 | | | 18.75 | 14.08333333 | 29.75 | 22 | 2003 |
| 1 | 19.04081633 | 14.65306122 | 34 | 27 | 18.75 | 14.25 | 29.71428571 | 22.28571429 | 2004 |
| 15.333333 | 19.33333333 | 14.19047619 | 30.6 | 22.8 | 18.66666667 | 14.33333333 | 30.33333333 | 22.55555556 | 2005 |
| 16.333333 | 20.19444444 | 15.58333333 | 28.75 | 20.58333333 | | | 27.25 | 19.66666667 | 2006 |
| , | 20.46296296 | 15.38888889 | 27.45454545 | 18.54545455 | | | 25.09090909 | 17.72727273 | 2007 |
| | 20.765625 | 15.78125 | 28.33333333 | 20.16666667 | | | 26.42857143 | 18.85714286 | 2008 |
| | 22.59139785 | 17.39784946 | 31 | 24 | | | 29 | 20.25 | 2009 |
| | 23.25454545 | 18.21818182 | 29.5 | 21.8125 | | | 27.125 | 19 | 2010 |
| | 23.58333333 | 18.68055556 | 28.93103448 | 21.44827586 | | | 27.83333333 | 19.83333333 | 2011 |
| | 23.8444444 | 19.15555556 | 31.76190476 | 24.78571429 | | | 30.21428571 | 21.35714286 | 2012 |
| | 24.47368421 | 19.12280702 | 32.8627451 | 26.11764706 | | | 31.90909091 | 23.45454545 | 2013 |
| | 24.2231405 | 18.15702479 | 45.46808511 | 43.82978723 | | | 34.75 | 28.25 | 2014 |
| 2 | 25.76350093 | 19.04283054 | 41.57638889 | 35.95138889 | 30 | 21 | 36.10294118 | 28.41176471 | 2015 |
| 22.333333 | 26.1965812 | 19.61025641 | 42.28 | 37.456 | 30 | 21 | 36.26530612 | 28.85714286 | 2016 |
| , | 25.70974576 | 19.36016949 | 40.29411765 | 34.75630252 | 29 | 21 | 37.4375 | 31.75 | 2017 |
| 18.517241 | 24.508028 | 18.48456155 | 37.81146305 | 32.08898944 | 19.55172414 | 14.85057471 | 31.37804878 | otal 24.0804878 | Grand T |

We use this table to create a line plot with markers to visualize the data as a timeline and across different body types.

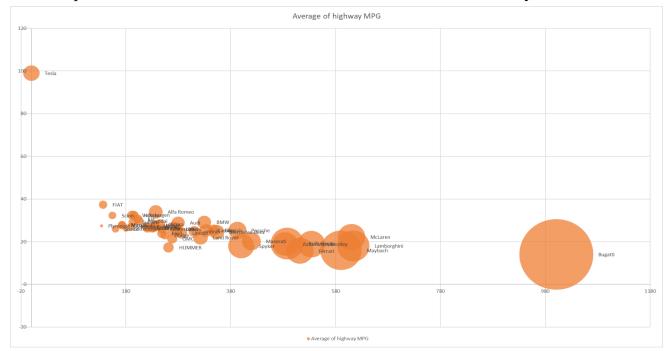


Task 5: How does the car's horsepower, MPG, and price vary across different Brands? To find relationships between a car's horsepower and MPG and price across different brands we can create a bubble plot to better visualize, for this we find average MPG, price and horsepower across each brand and create a pivot table. This pivot table looks like below:

| 4 | Α | | В | С | D | E | |
|---|--------------|------------|----------------------|---------------------|------------------------|-----------------|--|
| | Row Labels | ▼ / | Average of Engine HP | Average of city mpg | Average of highway MPG | Average of MSRP | |
| | Acura | | 244.9634146 | 20.00406504 | 28.2195122 | 35087.4878 | |
| | Alfa Romeo | | 237 | 24 | 34 | 61600 | |
| | Aston Martin | | 483.7582418 | 12.56043956 | 18.93406593 | 198123.4615 | |
| | Audi | | 280 | 19.63551402 | 28.92834891 | 54574.1215 | |
| | Bentley | | 533.8513514 | 11.55405405 | 18.90540541 | 247169.3243 | |
| | BMW | | 329.6203704 | 20.70061728 | 29.12654321 | 62162.55864 | |
| | Bugatti | | 1001 | 8 | 14 | 1757223.667 | |
| | Buick | | 220.0105263 | 18.78421053 | 27.01052632 | 29034.18947 | |
| 0 | Cadillac | | 332.7954545 | 17.36111111 | 25.24494949 | 56368.26515 | |
| 1 | Chevrolet | | 249.4837512 | 19.12070566 | 25.93221913 | 29018.35005 | |
| 2 | Chrysler | | 230.5351351 | 17.74054054 | 26.38378378 | 26990.23784 | |
| 3 | Dodge | | 254.5984848 | 16.45643939 | 22.99810606 | 24900.33523 | |
| 4 | Ferrari | | 511.9565217 | 10.56521739 | 15.72463768 | 238218.8406 | |
| 5 | FIAT | | 136.6129032 | 30.64516129 | 37.33870968 | 22670.24194 | |
| 5 | Ford | | 248.7730061 | 17.89815951 | 23.87730061 | 28525.18282 | |
| 7 | Genesis | | 347.3333333 | 16.33333333 | 25.33333333 | 46616.66667 | |
| 3 | GMC | | 268.2949791 | 15.79916318 | 21.47698745 | 32695.74268 | |
| 9 | Honda | | 195.8637413 | 25.2147806 | 32.39953811 | 26655.14781 | |
|) | HUMMER | | 261.2352941 | 13.52941176 | 17.29411765 | 36464.41176 | |

We copy this contents to create a table to create a bubble plot.

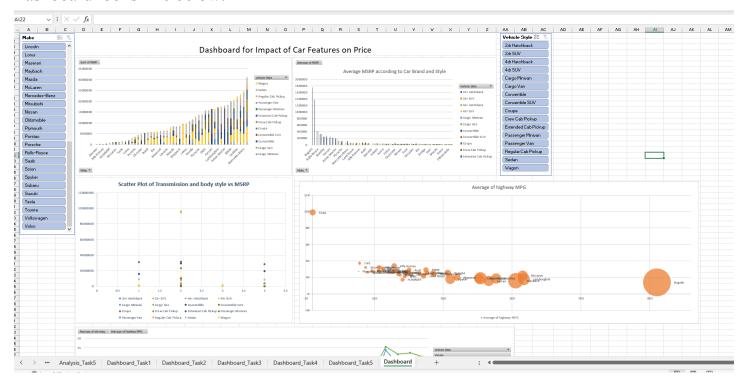
On the X axis there would be average horsepower and on the Y axis average MPG. Each bubble would represent each car brand and would be labeled to better identify the brands



Making Dashboard:

We have created each chart to visualize different parameters and relationships and trends, we can now create a dashboard by combining all these charts into one single worksheet to have a better understanding of the data. We copy all these charts into one worksheet and add slicers to change parameters which are shown the data point for. We add two slicers in the worksheet, make, and vehicle style, which represent car brand and body style. We then make connections with these slicers with all the charts to make the dashboard functional. Now we can easily find different trends and relationships between price and parameters.

Dashboard looks like below:



Results:

While working on this project, I have gained a better understanding of Impact of Car Features on Price and Profitability as well as popularity of the Car. I have improved my understanding of Advanced Excel methodologies. By analyzing Car features Data, I was able to provide insights on various aspects such as Features most affecting MSRP, Outliers in the Data, relation between Engine HP and MSRP, Regression Analysis, average MSRP across different brands and relation between no. of cylinders and fuel efficiency. I was also able to create different visualizations to improve data understanding and create a dashboard for ease of understanding between various parameters in Car Features.

This project has helped me enhance my Excel skills, particularly in data visualization and creating pivot tables and charts to derive meaningful insights. It has also improved my ability to interpret data and provide actionable recommendations based on the analysis.