

# Car Feature Analysis Report

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Jupyter Notebook:- Click Here to Download

Video Presentation:- Click Here to Watch Video Presentation

## Project Description!

Car companies want to make the most money while giving people the cars they want. But it's not easy with lots of competition and different kinds of cars like electric and regular ones. A Data Analyst can help by looking at data about cars and prices. They can figure out what things people like in cars and what kinds of cars make the most money. By doing this, the car company can decide how much to charge for their cars and what features to put in new cars. This helps the company do well in the market and make more money in the long run.

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Tech Stack Used

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Described what I have achieved through the project

"Unlocking the power of data can drive car manufacturers to thrive in a changing world, satisfying consumers while boosting profits."

#### **Tech Stack Used:-**

- -> Jupyter Notebook
- -> Tableau Public
- -> PowerPoint
- -> Github
- -> Google Drive

Importing and Exploring Dataset



#### **Importing and Exploring Dataset**

- First, I created a new notebook and imported libraries and dataset.
- application data = pd.read csv("car data.csv")
- Then I explored the dataset using the following programs.
- > To see the five rows from top and bottom I used head() and tail() respectively.
- application\_data.head()
- application data.tail()
- > Then by Using the column function see the names of all the columns/variables.
- application\_data.columns
- By using the describe function obtained all descriptive stats for the numeric variables in the DataSet.
- application data.describe(
- > By using isnull() method of pandas data frame calculated total null values in each variable.
- car\_data.isnull().sum(
- Then filled those null values with mean, median or mode. (According to variables)
- car\_data['Engine HP'] = car\_data['Engine HP'].fillna(car\_data['Engine HP'].mean())
- car\_data['Engine Cylinders'] = car\_data['Engine Cylinders'].fillna(car\_data['Engine Cylinders'].median())
- car\_data['Engine Fuel Type'] = car\_data['Engine Fuel Type'].fillna(car\_data['Engine Fuel Type'].mode())

```
car_data.columns
Make
Model
                                          Index(['Make', 'Model', 'Year', 'Engine Fuel Type', 'Engine HP',
Year
                                                  'Engine Cylinders', 'Transmission Type', 'Driven Wheels',
Engine Fuel Type
                                                  'Number of Doors', 'Market Category', 'Vehicle Size', 'Vehicle Style',
                        69
Engine HP
                                                  'highway MPG', 'city mpg', 'Popularity', 'MSRP'],
                        30
Engine Cylinders
                                                 dtype='object')
Transmission Type
Driven Wheels
Number of Doors
                                               car data.describe()
Market Category
                     3742
Vehicle Size
Vehicle Style
                                                                                Engine
                                                                                          Number of
                                                                                                        highway
                                                         Year
                                                                Engine HP
highway MPG
                                                                              Cylinders
                                                                                             Doors
                                                                                                           MPG
city mpg
Popularity
                                                 11914.000000
                                                              11914.000000
                                                                           11914.000000
                                                                                       11908.000000
                                                                                                    11914.000000
                                           count
MSRP
dtype: int64
                                                  2010.384338
                                                                249.386070
                                                                              5.629763
                                                                                           3.436093
                                                                                                       26.637485
                                           mean
       car data.duplicated()
                                                     7.579740
                                                                108.875192
                                                                              1.778413
                                                                                           0.881315
                                                                                                        8.863001
                                             std
 0
             False
                                                                              0.000000
                                                                                           2.000000
                                                                                                       12.000000
                                                  1990.000000
                                                                 55.000000
                                             min
             False
 1
             False
                                            25%
                                                  2007.000000
                                                                170.000000
                                                                              4.000000
                                                                                           2.000000
                                                                                                       22.000000
  3
             False
  4
             False
                                            50%
                                                                                           4.000000
                                                                                                       26.000000
                                                  2015.000000
                                                                227.000000
                                                                              6.000000
             . . .
 11909
             False
                                            75%
                                                  2016.000000
                                                                300.000000
                                                                              6.000000
                                                                                           4.000000
                                                                                                       30.000000
             False
 11910
                                                  2017.000000
                                                               1001.000000
                                                                              16.000000
                                                                                           4.000000
                                                                                                      354.000000
                                            max
             False
 11911
 11912
             False
```

city mpg

11914.000000

19.733255

8.987798

7.000000

16.000000

18.000000

22.000000

137.000000

15

144

5

13

20

56

car data.isnull().sum()

False

Length: 11914, dtype: bool

11913

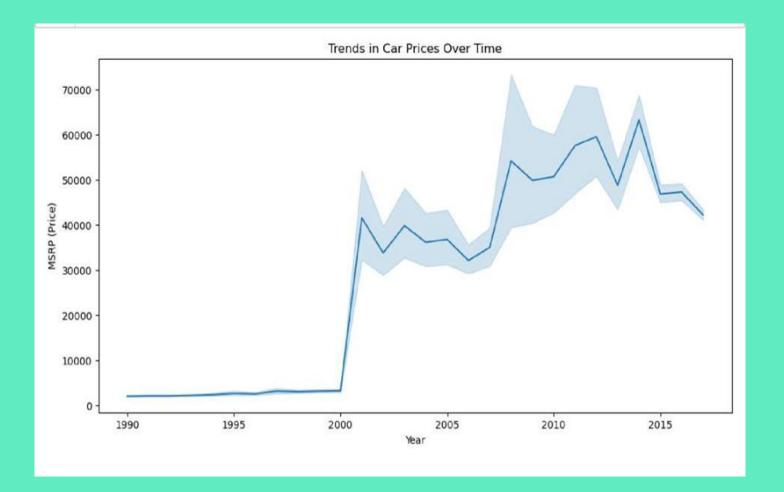
## Understanding the Dataset

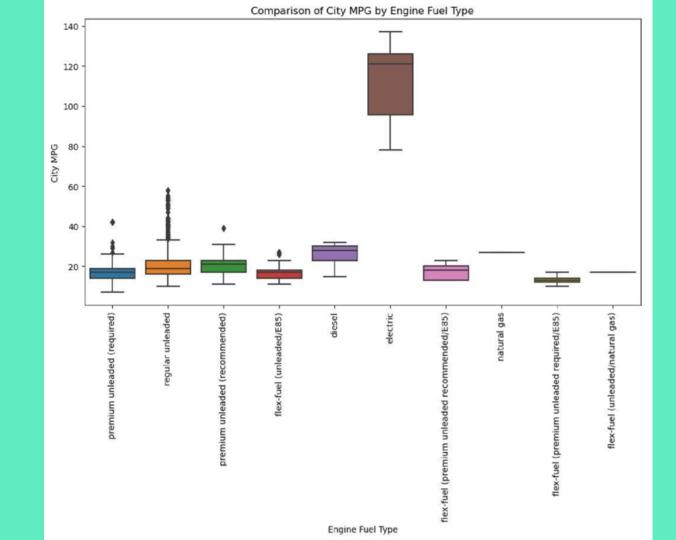
Gain insights into various aspects of the automotive industry.



#### **Understanding the Dataset**

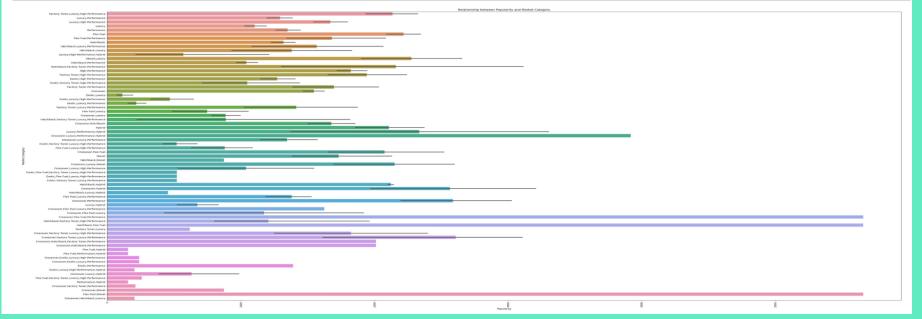
- Analyzing trends in the car features and pricing over time.
- plt.figure(figsize=(12, 6))
- sns.lineplot(x='Year', y='MSRP', data=car data'
- plt.title('Trends in Car Prices Over Time')
- plt.xlabel('Year')
- plt.vlabel('MSRP (Price)'
- > plt.show()
- Compare fuel efficiency of different types of cars
- plt.figure(figsize=(12, 6))
- sns.boxplot(x='Engine Fuel Type', y='city mpg', data=car\_data
- plt.title('Comparison of City MPG by Engine Fuel Type'
- plt.xlabel('Engine Fuel Type')
- plt.ylabel('City MPG'
- plt.xticks(rotation=90)
- plt.show()





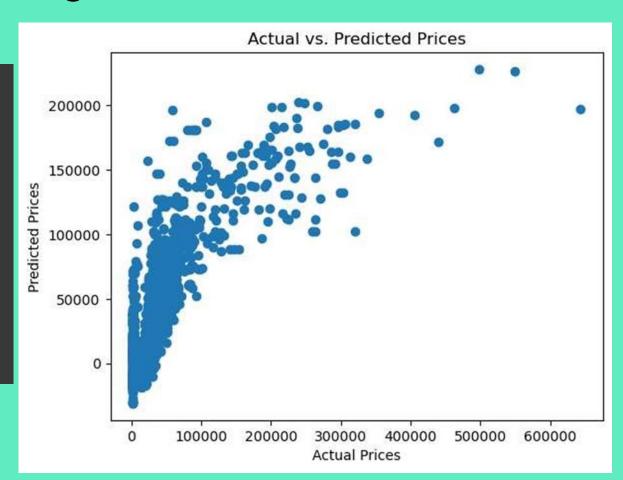
#### Investigating the relationship between a car's features and its popularity

```
plt.figure(figsize=(50, 25))
sns.barplot(x='Popularity', y='Market Category', data=car_data)
plt.title('Relationship between Popularity and Market Category')
plt.xlabel('Popularity')
plt.ylabel('Market Category')
plt.xticks(rotation=90)
plt.show()
```



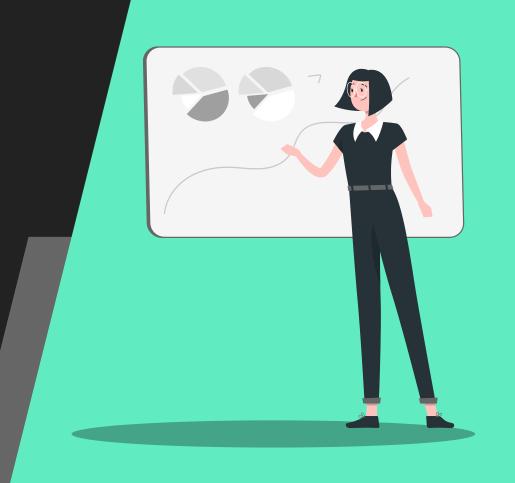
#### **Understanding the Dataset**

Predicting the price of a car based on its features and market category: By using the various features and market category variables in the dataset. I have developed a model to predict the price of a car. This could help manufacturers and consumers understand how different features affect the price of a car and make informed decisions about pricing and purchasing.



## Analysis of Dataset

Finding valuable insights using data analysis techniques such as regression analysis, pivot tables, sensitivity analysis, optimization, and time series analysis.

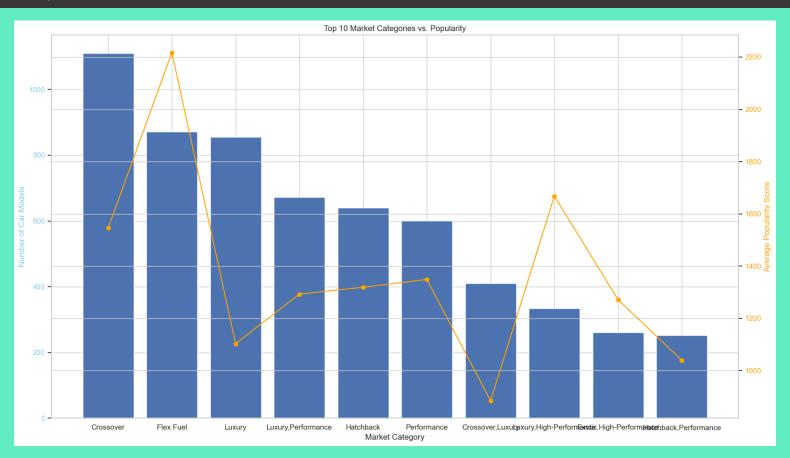


### How does the popularity of a car model vary across different market categories?

Task 1.A: Creating a pivot table that shows the number of car models in each market category and their corresponding popularity scores.

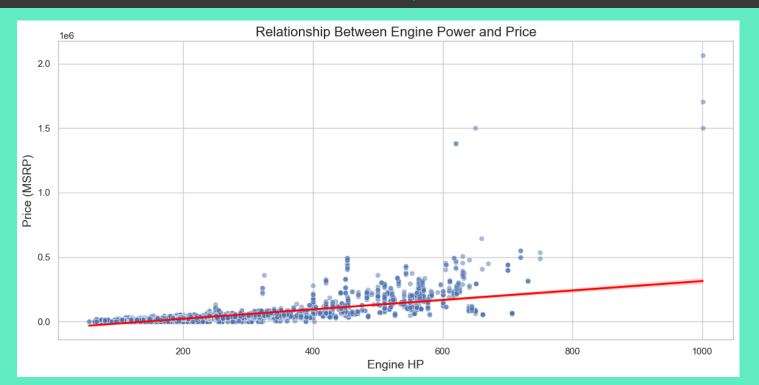
	Number of Car Models	Average Popularity Score
Market Category		
Crossover	1110	1545.263063
Flex Fuel	872	2217.302752
Luxury	855	1102.657310
Luxury,Performance	673	1292.615156
Hatchback	641	1318.865835
Exotic,Luxury,High-Performance,Hybrid	1	204.000000
Flex Fuel,Factory Tuner,Luxury,High-Performance	1	258.000000
Crossover, Exotic, Luxury, Performance	1	238.000000
${\bf Crossover, Exotic, Luxury, High-Performance}$	1	238.000000
Performance, Hybrid	1	155.000000
71 rows × 2 columns		

Task 1.B: Creating a combo chart that visualizes the relationship between market category and popularity.



#### What is the relationship between a car's engine power and its price?

Task 2: Creating a scatter chart that plots engine power on the x-axis and price on the y-axis. Add a trendline to the chart to visualize the relationship between these variables.



#### Which car features are most important in determining a car's price?

Task 3: Using regression analysis to identify the variables that have the strongest relationship with a car's price.

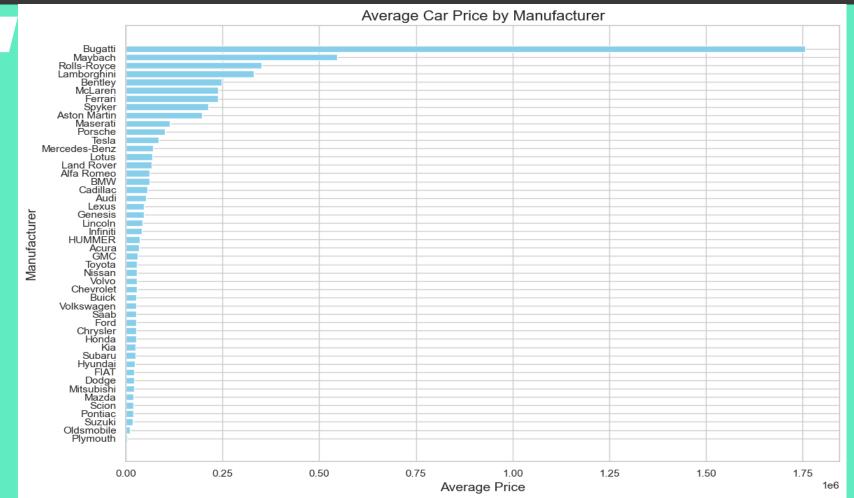


How does the average price of a car vary across different manufacturers?

Task 4.A: Creating a pivot table that shows the average price of cars for each manufacturer

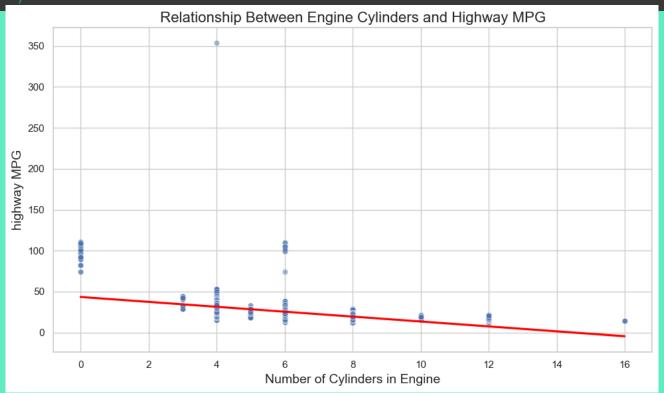
MSRP Make Plymouth 3.122902e+03 Oldsmobile 1,154254e+04 Suzuki 1.790721e+04 Pontiac 1.932155e+04 Scion 1.993250e+04 Mazda 2.003938e+04 Mitsubishi 2.124054e+04 Dodge 2.239006e+04 FIAT 2.267024e+04 Hyundai 2.459704e+04 Subaru 2.482750e+04 Kia 2.531017e+04 Honda 2.667434e+04 Chrvsler 2.672296e+04 Ford 2.739927e+04 Saab 2.741350e+04 Volkswagen 2.810238e+04 Buick 2.820661e+04 Chevrolet 2.835039e+04 VOIVO2.854116e+04 Nissan 2.858343e+04 2.903002e+04 Tovota GMC 3.049330e+04 3.488759e+04 Acura HUMMER 3.646441e+04 Infiniti 4.239421e+04 Lincoln 4.283983e+04 Genesis 4.661667e+04 Lexus 4.754907e+04 Audi 5.345211e+04 Cadillac 5.623132e+04 BMW 6.154676e+04 Alfa Romeo 6.160000e+04 Land Rover 6.782322e+04 6.918828e+04 Lotus Mercedes-Benz 7.147623e+04 Tesla 8.525556e+04 Porsche 1.016224e+05 Maserati 1.142077e+05 Aston Martin 1.979104e+05 Spyker 2.133233e+05 Ferrari 2.382188e+05 McLaren 2.398050e+05 Bentlev 2.471693e+05 Lamborghini 3.315673e+05 Rolls-Rovce 3.511306e+05 Maybach 5.462219e+05 Bugatti 1.757224e+06

#### Task 4.B: Creating a bar chart to visualize the relationship between manufacturer & average price



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

Task 5. A: Creating a scatter plot and trend line with the number of cylinders on the x-axis and highway MPG on the y-axis.



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

Task 5.B: Calculate the correlation coefficient between the number of cylinders and highway MPG

```
# Calculate the correlation coefficient
correlation = car_data['Engine Cylinders'].corr(car_data['highway MPG'])
# Print the correlation coefficient
print(f"Correlation coefficient between Engine Cylinders and Highway MPG:
Correlation coefficient between Engine Cylinders and Highway MPG: -0.60
```

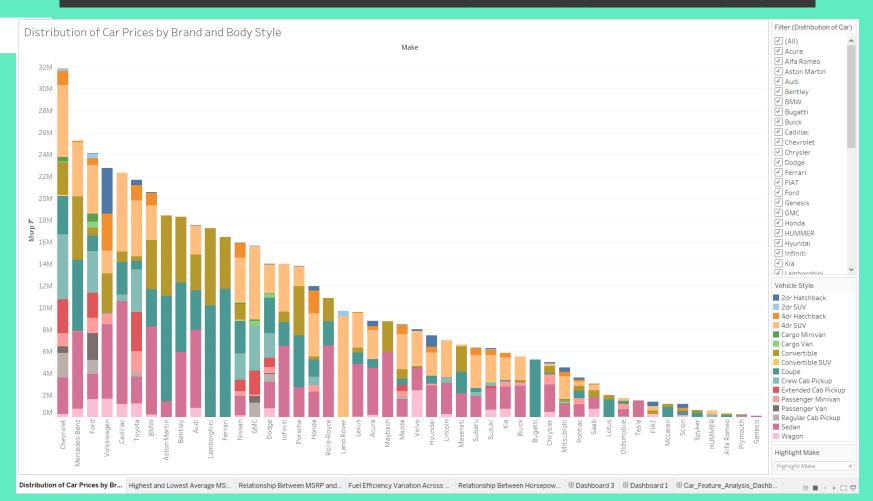
Inference: A negative value indicates a negative correlation. i.e., as number of cylinders increases, highway MPG decreases

## Building the Dashboard

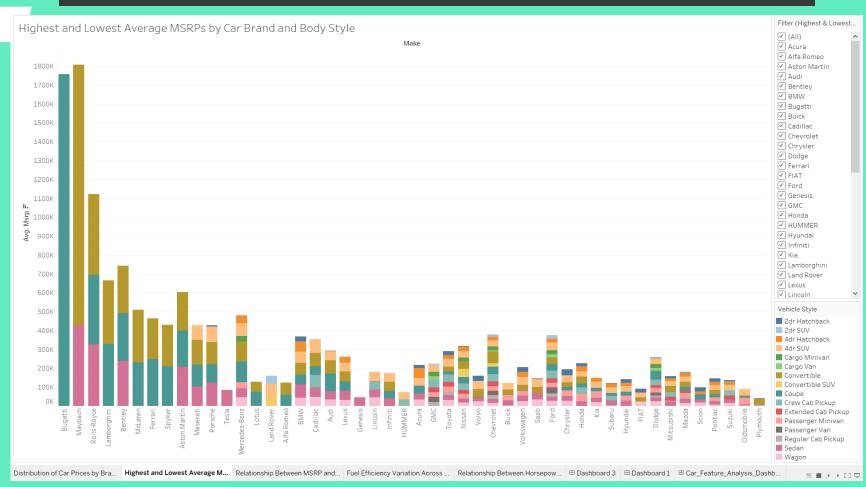
Creating the Interactive Dashboard using filters and slicers.



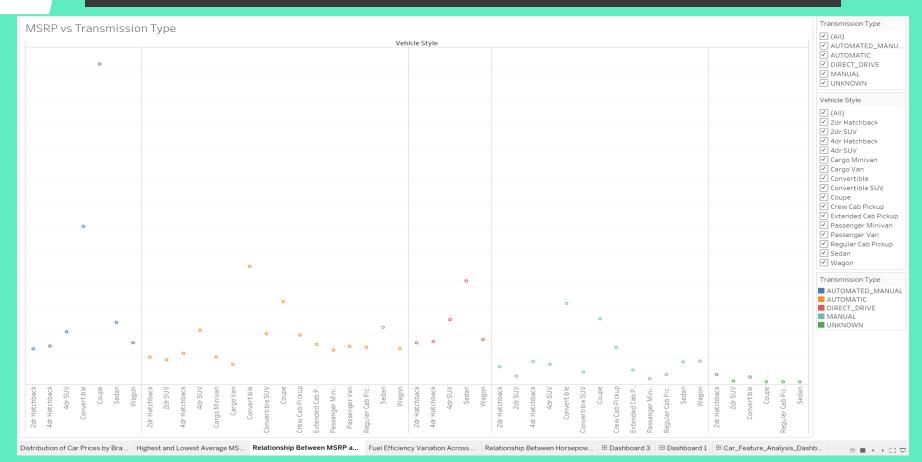
#### Task 1: How does the distribution of car prices vary by brand and body style?



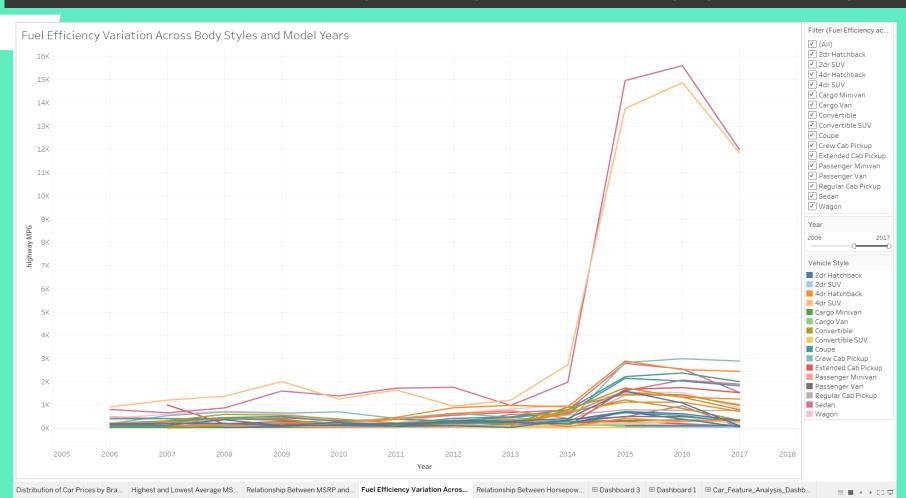
#### Task 2: Highest and Lowest Average MSRPs by Car Brand and Body Style



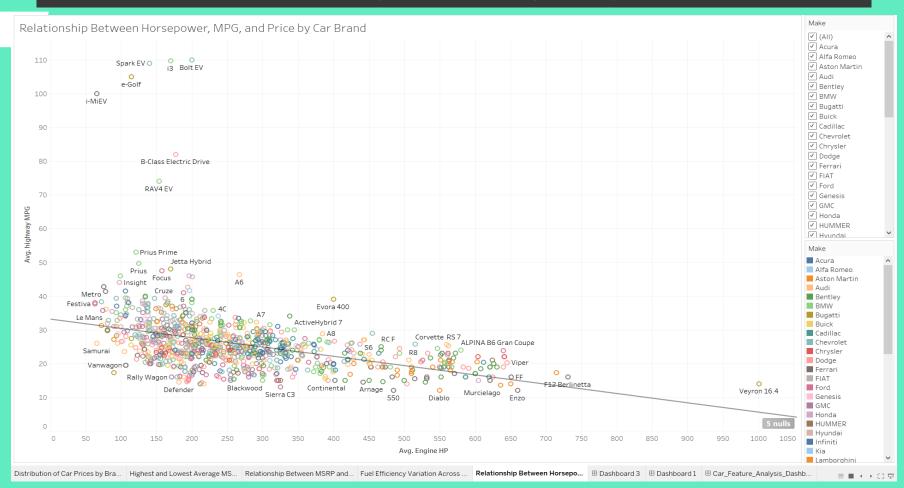
#### Task 3: How do the different features such as transmission type affect the MSRP, and how does this vary by body style?



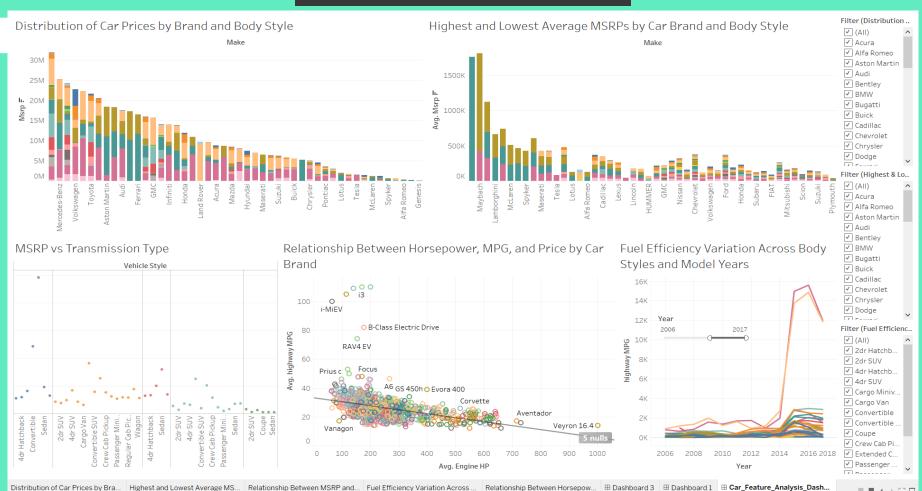
#### Task 4: How does the fuel efficiency of cars vary across different body styles and model years?



#### > Task 5: How does the car's horsepower, MPG, and price vary across different Brands?



#### Car\_Feature\_Analysis\_Dashboard



### Conclusion

- Thus, I have completed a Car Feature Analysis.
- Given key findings and all meaningful trends or patterns I have discovered.
- I have learned to use Python Libraries (pandas, matplotlib, numpy and seaborn) to analyze the dataset.
- I have learned how to create a dashboard using Tableau Public.
- I have learned to use Jupyter Notebook for Data Analysis.
- GitHub Repository and drive links are given as follows.

GitHub Repository:- https://github.com/ShindeYash/Impact\_of\_Car\_Features\_Analysis

Jupyter Notebook:- <a href="https://drive.google.com/file/d/10ebLWE6TFEC1Yix7CrDRP-FvPtNPxEVj/view?usp=sharing">https://drive.google.com/file/d/10ebLWE6TFEC1Yix7CrDRP-FvPtNPxEVj/view?usp=sharing</a>

Drive Link:- <a href="https://drive.google.com/drive/folders/1UE-rOaYAqR\_Qnp4i-erhj4NEAJKw8vyw?usp=sharing">https://drive.google.com/drive/folders/1UE-rOaYAqR\_Qnp4i-erhj4NEAJKw8vyw?usp=sharing</a>

Video Presentation:-

https://www.loom.com/share/fd5d07602fa04530a0209a6bb0d357e2?sid=c97d33e54b25-412d-bb78-68d6d9d48c61



### Thanks!

Do you have any questions? yashpradeepshinde@gmail.com Yash Shinde

