```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

#### **Upload our Dataset**

```
In [2]: cars_data = pd.read_csv("Car_sales.csv")
         cars_data.head()
                            Model Sales in thousands
Out[2]:
             Manufacturer
                                                        __year_resale_value Vehicle_type Price_in_tho
         0
                           Integra
                                                16.919
                                                                    16.360
                    Acura
                                                                               Passenger
         1
                                TL
                                                39.384
                                                                    19.875
                    Acura
                                                                               Passenger
         2
                                CL
                                                                    18.225
                    Acura
                                                14.114
                                                                               Passenger
         3
                                RL
                                                 8.588
                                                                    29.725
                    Acura
                                                                               Passenger
         4
                     Audi
                               Α4
                                                20.397
                                                                    22.255
                                                                               Passenger
         cars_data.shape
Out[3]: (157, 16)
```

#### **Data Cleaning**

```
In [4]: cars_data.isna().sum()
Out[4]: Manufacturer
                                 0
        Model
                                 0
        Sales_in_thousands
                                 0
         __year_resale_value
                                36
        Vehicle_type
                                 2
        Price_in_thousands
        Engine_size
        Horsepower
                                 1
        Wheelbase
                                 1
        Width
                                 1
        Length
                                 2
        Curb_weight
                                 1
        Fuel_capacity
                                 3
        Fuel_efficiency
        Latest_Launch
                                 0
        Power_perf_factor
        dtype: int64
In [5]: print("mean : ",cars_data['__year_resale_value'].mean())
        print("median : ",cars_data['__year_resale_value'].median())
```

median : 14.18 In [6]: cars\_data['\_\_year\_resale\_value'].fillna(cars\_data['\_\_year\_resale\_value'].median(),i cars\_data.isna().sum() Out[6]: Manufacturer Model 0 Sales\_in\_thousands 0 \_\_year\_resale\_value Vehicle\_type Price\_in\_thousands Engine\_size 1 Horsepower 1 Wheelbase 1 Width 1 Length 1 Curb\_weight 2 Fuel\_capacity 1 Fuel\_efficiency 3 Latest\_Launch 0 2 Power\_perf\_factor dtype: int64 In [7]: threhold = len(cars\_data) \* 0.05 print("threhold : ",threhold) cols\_to\_drop = cars\_data.columns[cars\_data.isna().sum() <= threhold ]</pre> cars\_data.dropna(subset=cols\_to\_drop,inplace=True) cars\_data.isna().sum() threhold: 7.85000000000000005 Out[7]: Manufacturer Model 0 Sales\_in\_thousands \_\_year\_resale\_value Vehicle\_type Price\_in\_thousands 0 Engine\_size 0 Horsepower 0 Wheelbase Width 0 Length 0 Curb\_weight 0 Fuel\_capacity Fuel\_efficiency Latest\_Launch Power\_perf\_factor dtype: int64

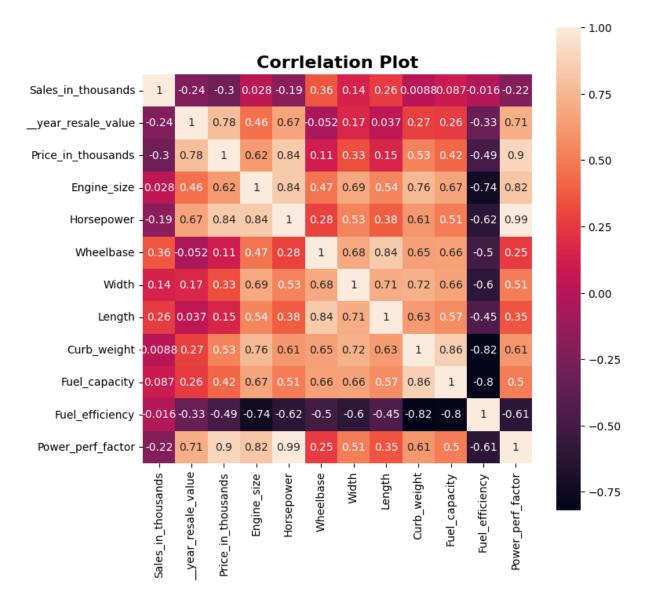
mean: 18.07297520661157

In [8]: cars\_data.describe()

Out[8]:		Sales_in_thousands	year_resale_value	Price_in_thousands	Engine_size	Horsepower
	count	152.000000	152.000000	152.000000	152.000000	152.000000
	mean	53.359072	17.144671	27.331822	3.049342	184.809211
	std	68.938380	10.301344	14.418669	1.049818	56.823152
	min	0.110000	5.160000	9.235000	1.000000	55.000000
	25%	13.714000	12.527500	17.888750	2.300000	147.500000
	50%	29.213000	14.180000	22.747000	3.000000	175.000000
	75%	68.069750	17.806250	31.938750	3.575000	211.250000
	max	540.561000	67.550000	85.500000	8.000000	450.000000

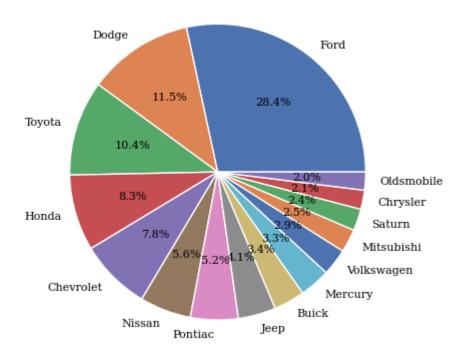
# Finging Realtionship in data

```
In [9]: corr = cars_data.corr(numeric_only=True)
    plt.figure(figsize=(8,8))
    sns.heatmap(corr,annot=True,square=True)
    plt.title("Corrlelation Plot",fontsize=16,fontweight="bold")
    plt.show()
```



### **Analysing The Manufacturer's sales**

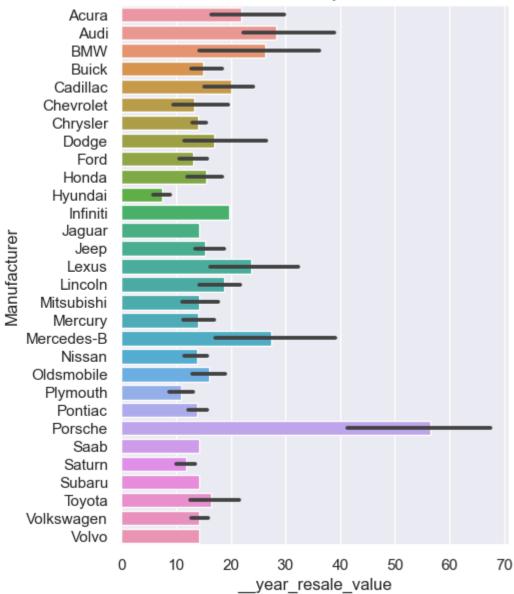
The top 15 Manufacturer in sales



# Year Resale Value by Manufacturer

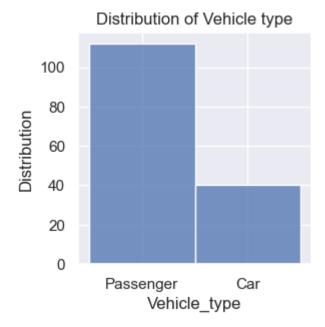
```
In [11]: plt.figure(figsize=(5, 7))
    sns.barplot(data=cars_data, x="__year_resale_value",y="Manufacturer")
    plt.title('Year Resale Value by Manufacturer ')
    plt.ylabel("Manufacturer")
    plt.show()
```





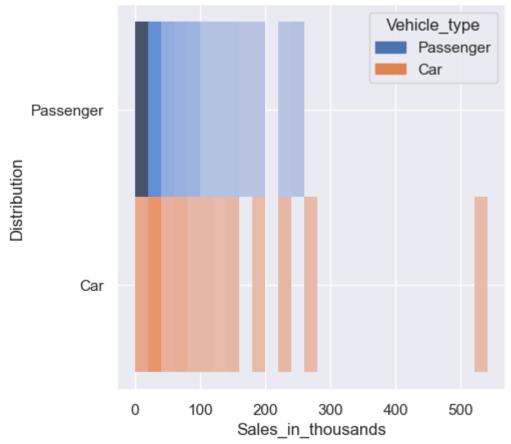
# Analysing the Vehicle type

```
In [12]: plt.figure(figsize=(3, 3))
    sns.histplot(data=cars_data, x="Vehicle_type")
    plt.title('Distribution of Vehicle type')
    plt.ylabel("Distribution")
    plt.show()
```



```
In [13]: plt.figure(figsize=(5, 5))
    sns.histplot(data=cars_data, x="Sales_in_thousands",y="Vehicle_type",hue="Vehicle_t
    plt.title('Distribution of Sales by Vehicle Type ')
    plt.ylabel("Distribution")
    plt.show()
```

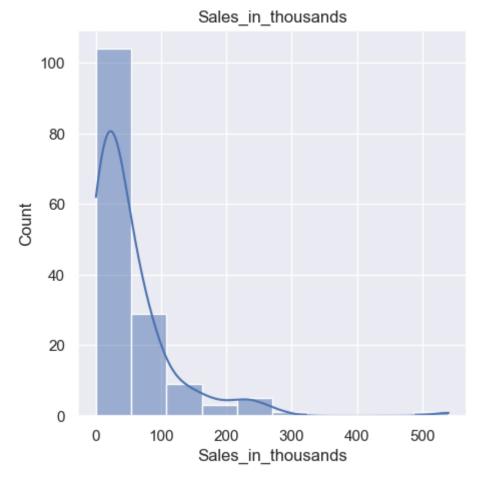


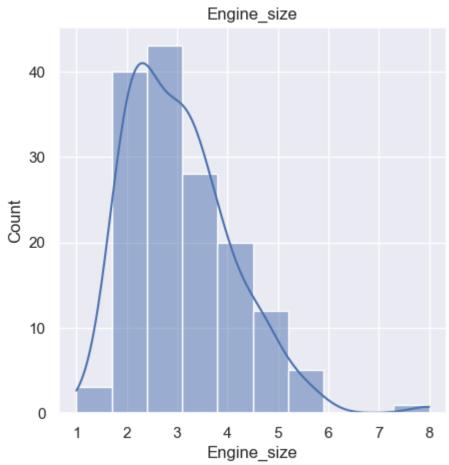


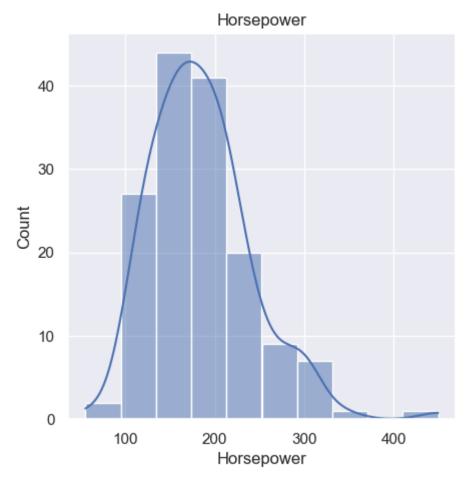
# finding the feature That has more impact on car sales

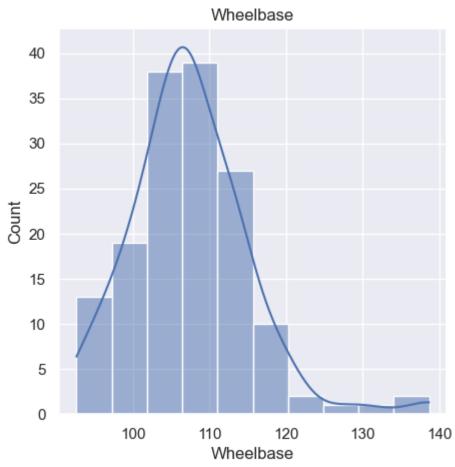
#### **Distribution Analysis**

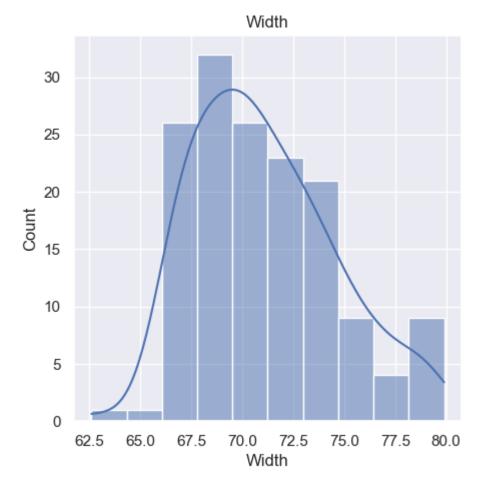
```
In [14]: df_num = cars_data.select_dtypes(include = ['float64', 'int64'])
         print(df_num.columns)
         df_num = df_num.drop(columns=["__year_resale_value", "Price_in_thousands"])
         print(df_num.columns)
       Index(['Sales_in_thousands', '__year_resale_value', 'Price_in_thousands',
               'Engine_size', 'Horsepower', 'Wheelbase', 'Width', 'Length',
               'Curb_weight', 'Fuel_capacity', 'Fuel_efficiency', 'Power_perf_factor'],
              dtype='object')
       Index(['Sales_in_thousands', 'Engine_size', 'Horsepower', 'Wheelbase', 'Width',
               'Length', 'Curb_weight', 'Fuel_capacity', 'Fuel_efficiency',
               'Power_perf_factor'],
              dtype='object')
In [15]: def plot():
             for p in df_num.columns:
                 plt.figure(figsize=(5, 5))
                 sns.histplot(df_num[p], bins=10,kde=True)
                 plt.title(p)
                 plt.show()
         plot()
```

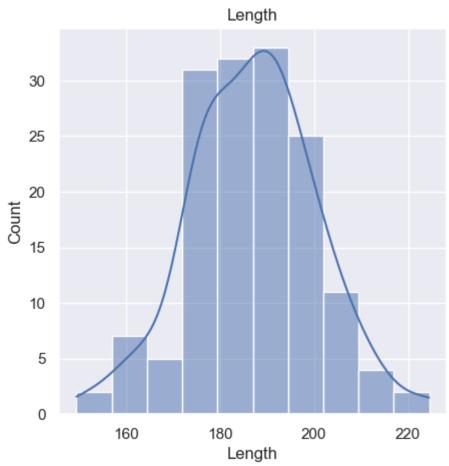


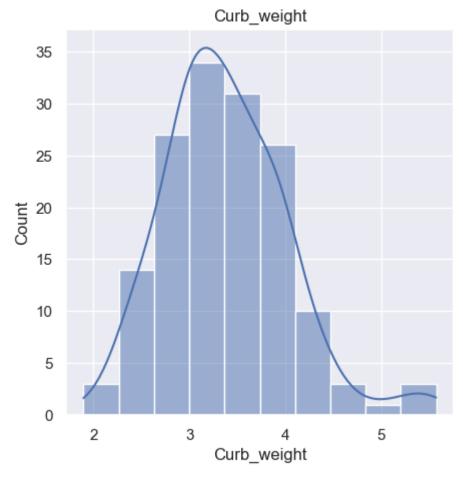


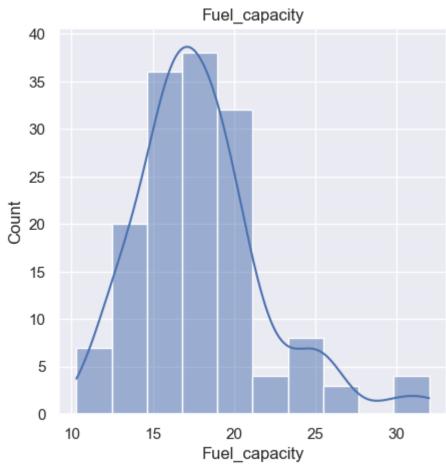


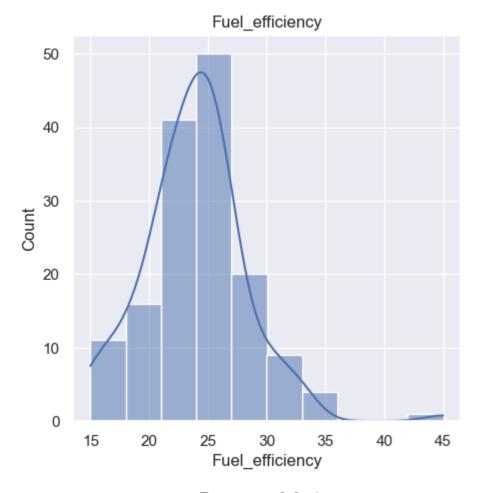


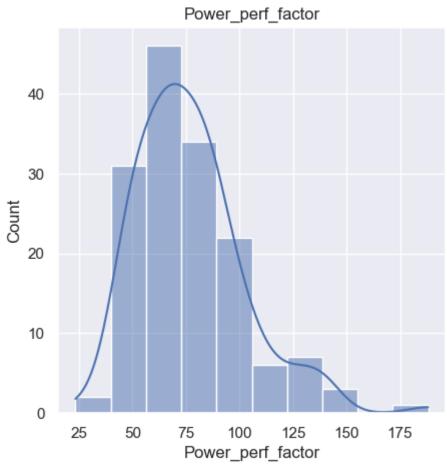






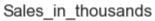


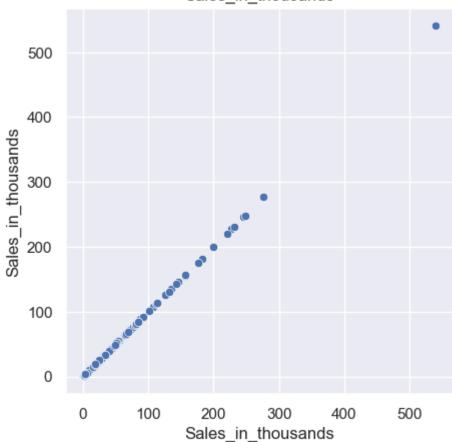


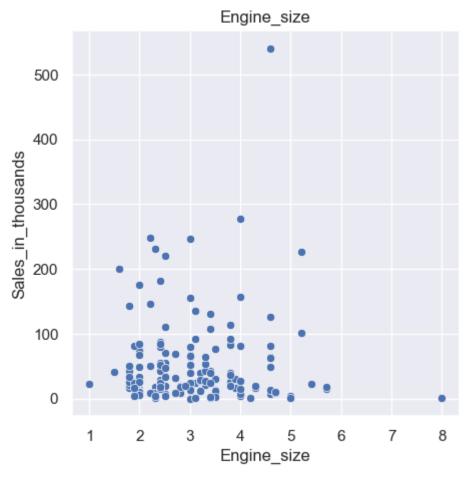


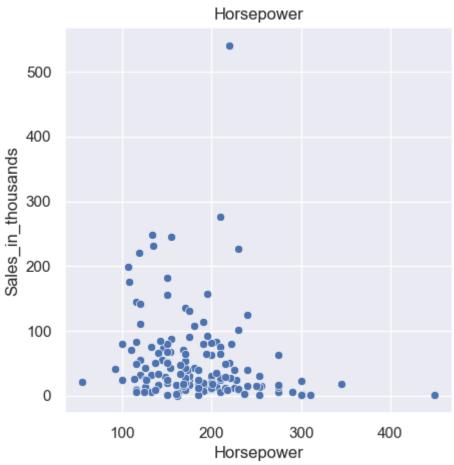
## **Bivariate Analysis**

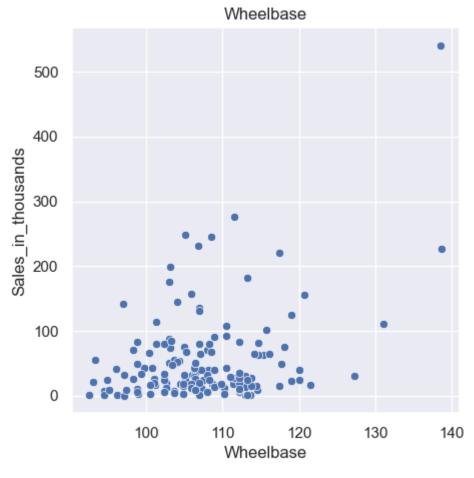
```
In [16]:
    def Biplot():
        for u in df_num.columns:
            plt.figure(figsize=(5, 5))
            sns.scatterplot(x = u , y="Sales_in_thousands" , data = df_num )
            plt.title(u)
            plt.show()
Biplot()
```

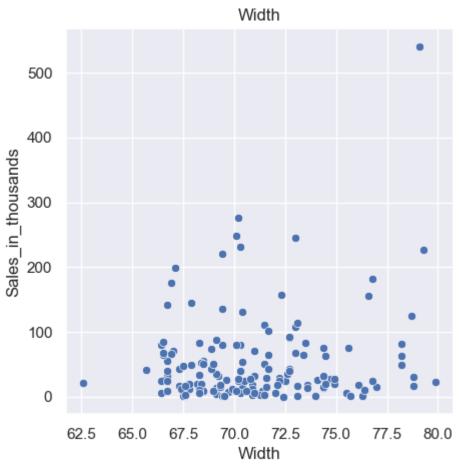


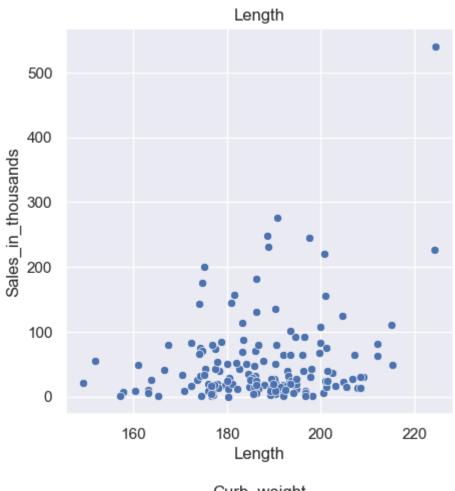


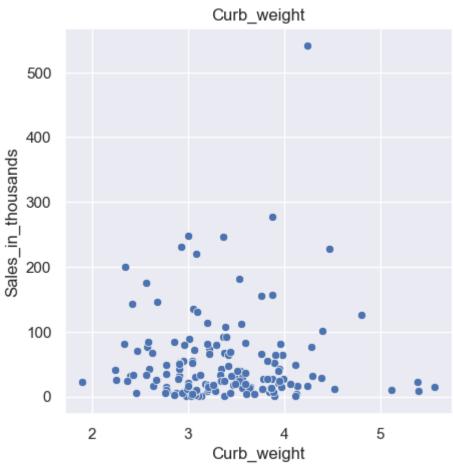


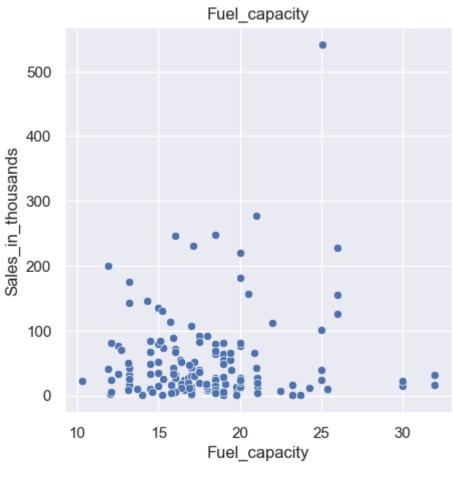


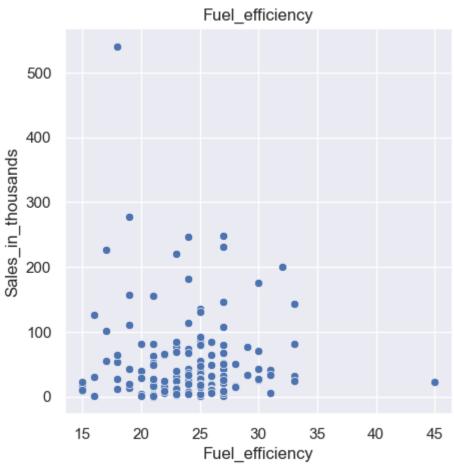


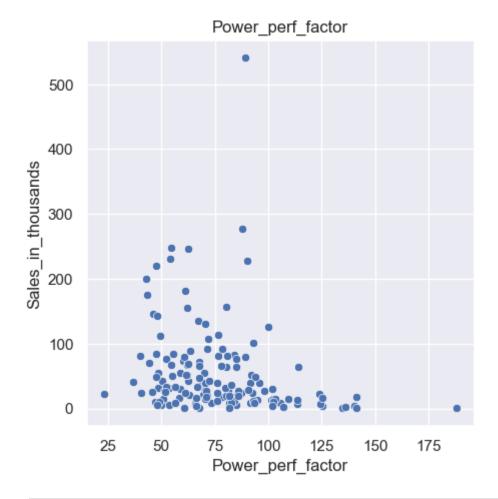












In [ ]: