# In [2]: import pandas as pd import numpy as np import seaborn as sns import matplotlib.pyplot as plt from sklearn.model\_selection import train\_test\_split from sklearn.linear\_model import LinearRegression from sklearn.linear\_model import Ridge, RidgeCV, Lasso from sklearn.preprocessing import StandardScaler

# In [3]: #data data=pd.read\_csv(r"C:\Users\sudheer\Downloads\Advertising.csv")

#### Out[3]: TV Radio Newspaper Sales **0** 230.1 22.1 37.8 69.2 44.5 45.1 1 39.3 10.4 17.2 45.9 69.3 12.0 **3** 151.5 58.5 16.5 41.3 **4** 180.8 10.8 58.4 17.9 38.2 195 3.7 13.8 7.6 **196** 94.2 8.1 14.0 4.9 **197** 177.0 9.3 6.4 14.8 **198** 283.6 42.0 66.2 25.5

200 rows × 4 columns

**199** 232.1

# In [4]: data.head()

### Out[4]:

	TV	Radio	Newspaper	Sales
0	230.1	37.8	69.2	22.1
1	44.5	39.3	45.1	10.4
2	17.2	45.9	69.3	12.0
3	151.5	41.3	58.5	16.5
4	180.8	10.8	58.4	17.9

8.6

8.7

18.4

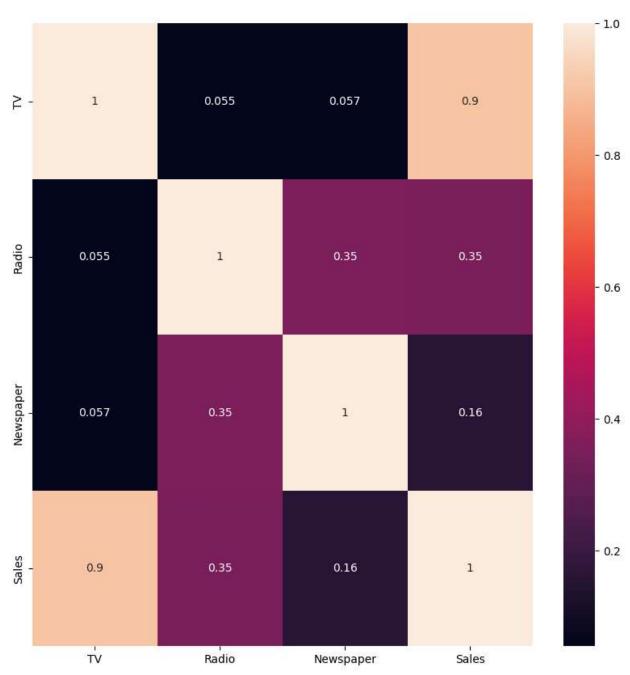
In [5]: data.tail()

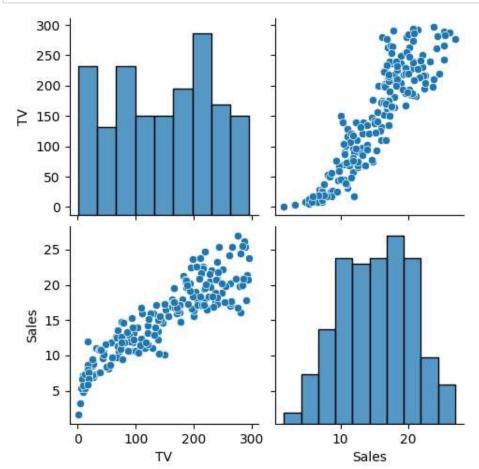
Out[5]:

	TV	Radio	Newspaper	Sales
195	38.2	3.7	13.8	7.6
196	94.2	4.9	8.1	14.0
197	177.0	9.3	6.4	14.8
198	283.6	42.0	66.2	25.5
199	232.1	8.6	8.7	18.4

In [6]: plt.figure(figsize = (10, 10))
sns.heatmap(data.corr(), annot = True)

Out[6]: <Axes: >





```
In [8]: features = data.columns[0:2]
    target = data.columns[-1]
    #X and y values
    X = data[features].values
    y = data[target].values
    #splot
    X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=1)
    print("The dimension of X_train is {}".format(X_train.shape))
    print("The dimension of X_test is {}".format(X_test.shape))
    #Scale features
    scaler = StandardScaler()
    X_train = scaler.fit_transform(X_train)
    X_test = scaler.transform(X_test)
```

The dimension of  $X_{train}$  is (140, 2) The dimension of  $X_{tst}$  is (60, 2)

#### Linear Regression Model:

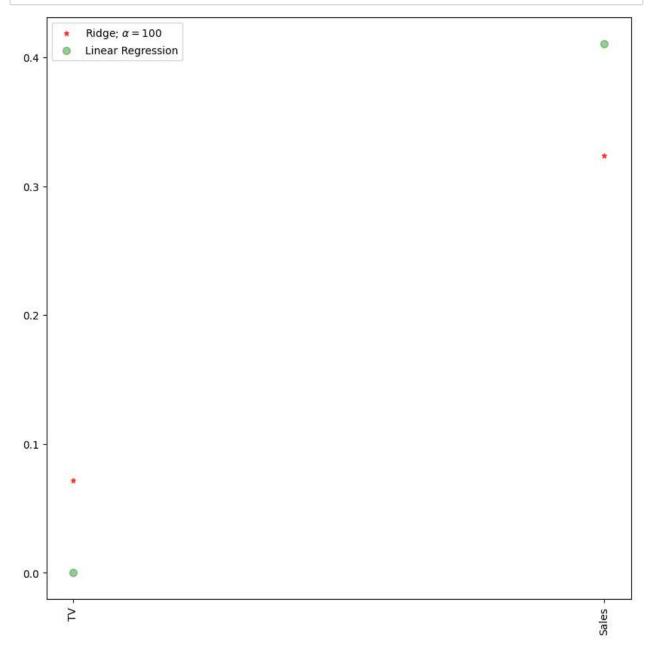
The train score for lr model is 1.0 The test score for lr model is 1.0

```
In [10]: #Ridge Regression Model
    ridgeReg = Ridge(alpha=10)
    ridgeReg.fit(X_train,y_train)
    #train and test scorefor ridge regression
    train_score_ridge = ridgeReg.score(X_train, y_train)
    test_score_ridge = ridgeReg.score(X_test, y_test)
    print("\nRidge Model:\n")
    print("The train score for ridge model is {}".format(train_score_ridge))
    print("The test score for ridge model is {}".format(test_score_ridge))
```

#### Ridge Model:

The train score for ridge model is 0.990287139194161 The test score for ridge model is 0.9844266285141221

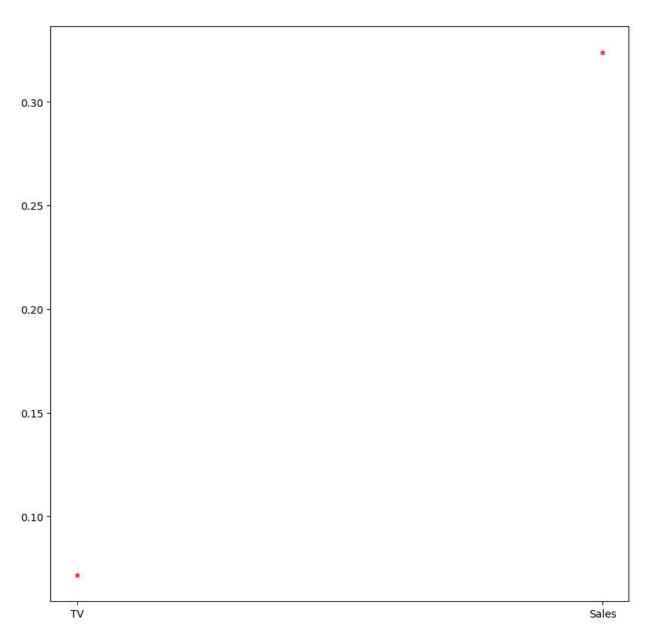
```
In [11]: plt.figure(figsize = (10, 10))
   plt.plot(features,ridgeReg.coef_,alpha=0.7,linestyle='none',marker='*',markersize=5,cole#plt.plot(rr100.coef_,alpha=0.5,linestyle='none',marker='d',markersize=6,color='blue',leplot(features,lr.coef_,alpha=0.4,linestyle='none',marker='o',markersize=7,color='greplt.xticks(rotation = 90)
   plt.legend()
   plt.show()
```



```
In [15]: #plot size
    plt.figure(figsize = (10, 10))
    #add plot for ridge regression
    plt.plot(features,ridgeReg.coef_,alpha=0.7,linestyle='none',marker='*',markersize=5,cole#add plot for lasso regression
    plt.plot(lasso_cv.coef_,alpha=0.5,linestyle='none',marker='d',markersize=6,color='blue'#add plot for linear model
    plt.plot(features,lr.coef_,alpha=0.4,linestyle='none',marker='o',markersize=7,color='gr#rotate axis
    plt.xticks(rotation = 90)
    plt.legend()
    plt.title("Comparison plot of Ridge, Lasso and Linear regression model")
    plt.show()
```

```
NameError
Cell In[15], line 6
    4 plt.plot(features,ridgeReg.coef_,alpha=0.7,linestyle='none',marker='*',markers
ize=5,color='red',label=r'Ridge; $\alpha=100$')
    5 #add plot for lasso regression
---> 6 plt.plot(lasso_cv.coef_,alpha=0.5,linestyle='none',marker='d',markersize=6,col
or='blue',label=r'lasso; $\alpha = grid$')
    7 #add plot for linear model
    8 plt.plot(features,lr.coef_,alpha=0.4,linestyle='none',marker='o',markersize=7,
color='green',label='Linear Regression')
```

NameError: name 'lasso\_cv' is not defined



```
In [16]: #Lasso regression model
print("\nLasso Model: \n")
lasso = Lasso(alpha = 10)
lasso.fit(X_train,y_train)
train_score_ls =lasso.score(X_train,y_train)
test_score_ls =lasso.score(X_test,y_test)
print("The train score for ls model is {}".format(train_score_ls))
print("The test score for ls model is {}".format(test_score_ls))
```

#### Lasso Model:

```
The train score for 1s model is 0.0
The test score for 1s model is -0.0042092253233847465
```

```
In [17]: pd.Series(lasso.coef_, features).sort_values(ascending = True).plot(kind = "bar")
Out[17]: <Axes: >

0.04 -
0.02 -
-0.02 -
-0.04 -
```

```
In [18]: #Using the linear CV model
    from sklearn.linear_model import LassoCV
    #Lasso Cross validation
    lasso_cv = LassoCV(alphas = [0.0001, 0.001, 0.1, 1, 10], random_state=0).fit(X_tra
    #score
    print(lasso_cv.score(X_train, y_train))
    print(lasso_cv.score(X_test, y_test))
```

0.9999999343798134
0.9999999152638072

# Vehicle selection

```
In [1]: import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
from sklearn.linear_model import Ridge, RidgeCV, Lasso
from sklearn.preprocessing import StandardScaler
```

In [2]: #data
 data=pd.read\_csv(r"C:\Users\sudheer\Downloads\fiat500\_VehicleSelection\_Dataset (2).csv"
 data

## Out[2]:

	ID	model	engine_power	age_in_days	km	previous_owners	lat	lon	price
0	1	lounge	51	882	25000	1	44.907242	8.611560	8900
1	2	pop	51	1186	32500	1	45.666359	12.241890	8800
2	3	sport	74	4658	142228	1	45.503300	11.417840	4200
3	4	lounge	51	2739	160000	1	40.633171	17.634609	6000
4	5	рор	73	3074	106880	1	41.903221	12.495650	5700
1533	1534	sport	51	3712	115280	1	45.069679	7.704920	5200
1534	1535	lounge	74	3835	112000	1	45.845692	8.666870	4600
1535	1536	pop	51	2223	60457	1	45.481541	9.413480	7500
1536	1537	lounge	51	2557	80750	1	45.000702	7.682270	5990
1537	1538	pop	51	1766	54276	1	40.323410	17.568270	7900

1538 rows × 9 columns

```
In [3]: data = data[['engine_power', 'price']]
    data.columns=['Eng', 'pri']
```

# In [4]: data.head()

# Out[4]:

	Eng	pri
0	51	8900
1	51	8800
2	74	4200
3	51	6000
4	73	5700

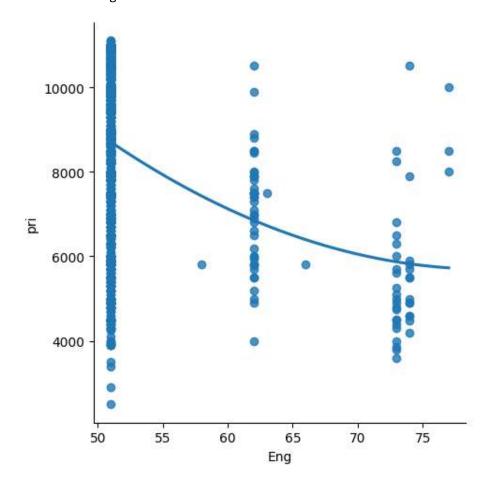
# In [5]: data.tail()

# Out[5]:

	Eng	pri
1533	51	5200
1534	74	4600
1535	51	7500
1536	51	5990
1537	51	7900

```
In [6]: sns.lmplot(x='Eng',y='pri',data=data,order=2,ci=None)
```

Out[6]: <seaborn.axisgrid.FacetGrid at 0x270a1fa5870>



# In [7]: data.info()

```
data.describe()
In [8]:
Out[8]:
                        Eng
                                       pri
           count 1538.000000
                               1538.000000
           mean
                   51.904421
                               8576.003901
             std
                    3.988023
                               1939.958641
            min
                   51.000000
                               2500.000000
            25%
                   51.000000
                               7122.500000
            50%
                               9000.000000
                   51.000000
            75%
                   51.000000
                              10000.000000
                   77.000000
                              11100.000000
            max
         data.fillna(method='ffill')
Out[9]:
                Eng
                       pri
                  51 8900
             0
             1
                  51 8800
             2
                  74 4200
                  51 6000
             3
             4
                  73 5700
```

1538 rows × 2 columns

51 5200

74 4600

51 7500

51 5990

51 7900

15331534

1535

1536

1537

```
In [10]: x=np.array(data['Eng']).reshape(-1,1)
y=np.array(data['pri']).reshape(-1,1)
```

```
In [11]: data.dropna(inplace=True)
```

 $\label{local-temp-ipy-ernel_14200} C:\Users\sudheer\AppData\Local\Temp\ipy-kernel\_14200\1580077326.py:1: SettingWithCopyWarning:$ 

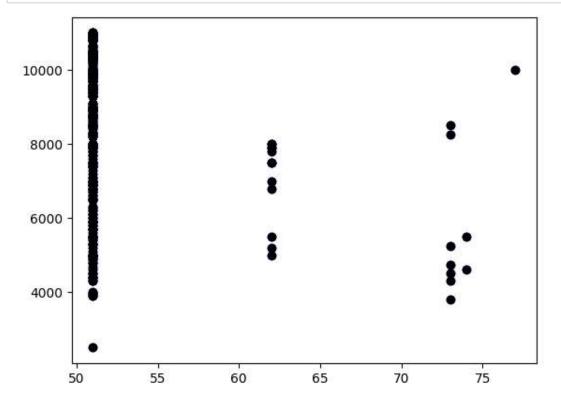
A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/use r\_guide/indexing.html#returning-a-view-versus-a-copy (https://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returning-a-view-versus-a-copy) data.dropna(inplace=True)

```
In [12]: X_train,X_test,y_train, y_test = train_test_split(x, y, test_size = 0.25)
# Splitting the data into training data and test data
regr= LinearRegression()
regr.fit(X_train, y_train)
print(regr.score(X_test, y_test))
```

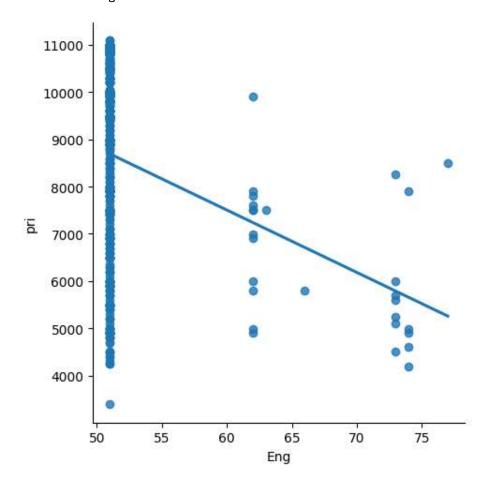
### 0.06427439865414497

```
In [13]: y_pred=regr.predict(X_test)
    plt.scatter(X_test, y_test, color = 'b')
    plt.scatter(X_test, y_test, color = 'k')
    plt.show()
```



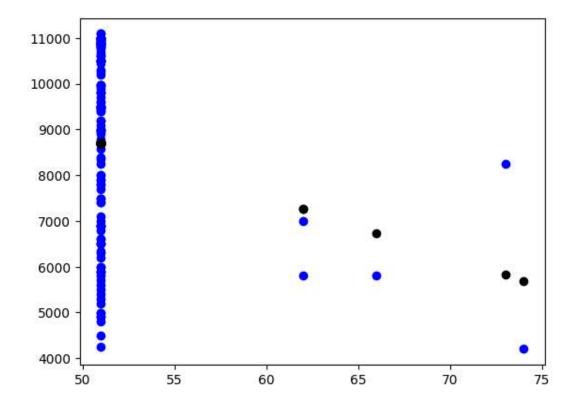
```
In [14]: df500 = data[:][:500]
# Selecting the 1st 500 rows of teh data
sns.lmplot(x = "Eng", y = "pri", data = df500, order = 1, ci = None)
```

Out[14]: <seaborn.axisgrid.FacetGrid at 0x2708fbc0550>



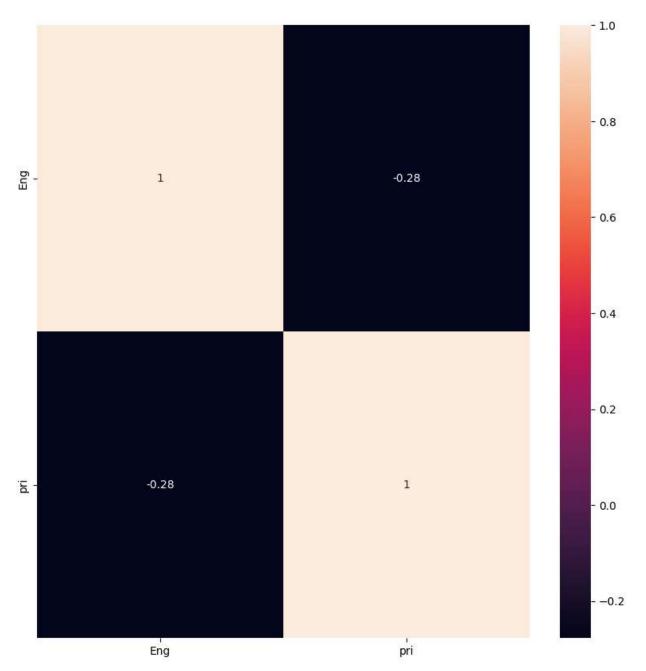
```
In [15]: df500.fillna(method='ffill',inplace=True)
    x=np.array(df500['Eng']).reshape(-1,1)
    y=np.array(df500['pri']).reshape(-1,1)
    df500.dropna(inplace=True)
    X_train,X_test,y_train, y_test = train_test_split(x, y, test_size = 0.25)
# Splitting the data into training data and test data
    regr= LinearRegression()
    regr.fit(X_train, y_train)
    print("Regression:",regr.score(X_test,y_test))
    y_pred=regr.predict(X_test)
    plt.scatter(X_test, y_test, color = 'b')
    plt.scatter(X_test, y_pred, color = 'k')
    plt.show()
```

Regression: 0.05228398076602292



```
In [16]: plt.figure(figsize = (10, 10))
sns.heatmap(data.corr(), annot = True)
```

Out[16]: <Axes: >



```
In [17]: #Model
         lr = LinearRegression()
         #Fit model
         lr.fit(X_train, y_train)
         #predict
         #prediction = Lr.predict(X test)
         #actual
         actual = y_test
         train score lr = lr.score(X train, y train)
         test_score_lr = lr.score(X_test, y_test)
         print("\nLinear Regression Model:\n")
         print("The train score for lr model is {}".format(train_score_lr))
         print("The test score for lr model is {}".format(test_score_lr))
         Linear Regression Model:
         The train score for 1r model is 0.07526429590089345
         The test score for lr model is 0.05228398076602292
In [18]: #Ridge Regression Model
         ridgeReg = Ridge(alpha=10)
         ridgeReg.fit(X_train,y_train)
         #train and test scorefor ridge regression
         train_score_ridge = ridgeReg.score(X_train, y_train)
         test score_ridge = ridgeReg.score(X_test, y_test)
         print("\nRidge Model:\n")
         print("The train score for ridge model is {}".format(train score ridge))
         print("The test score for ridge model is {}".format(test_score_ridge))
         Ridge Model:
         The train score for ridge model is 0.07526412778450209
         The test score for ridge model is 0.05227522794569095
In [19]: #Lasso regression model
         print("\nLasso Model: \n")
         lasso = Lasso(alpha = 10)
         lasso.fit(X_train,y_train)
         train_score_ls =lasso.score(X_train,y_train)
         test_score_ls =lasso.score(X_test,y_test)
         print("The train score for ls model is {}".format(train_score_ls))
         print("The test score for ls model is {}".format(test_score_ls))
         Lasso Model:
         The train score for 1s model is 0.0752629050977659
         The test score for 1s model is 0.05225824306507554
In [ ]:
```