Ridge and Lasso Regression-Advertising

```
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\sneha\Downloads\Advertising.csv")
data

Out[3]:

	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
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

200 rows × 4 columns

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

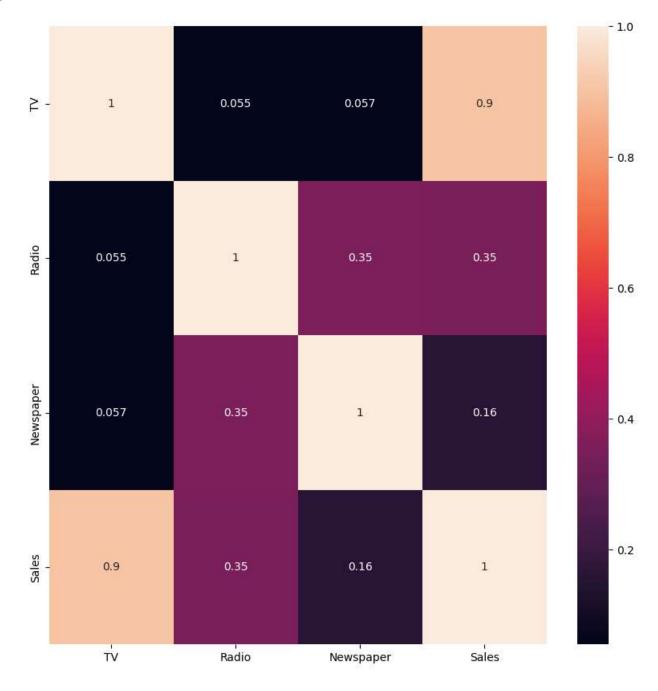
In [5]: data.tail()

Out[5]:

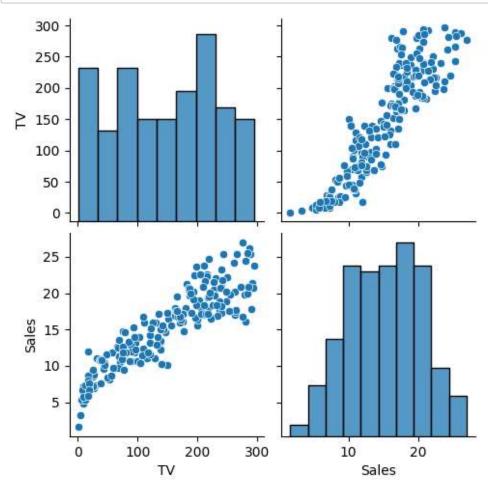
	TV	Radio	Newspaper	Sales
195	38.2	3.7	13.8	7.6
196	94.2	4.9	8.1	14.0
197	177.0	9.3	6.4	14.8
198	283.6	42.0	66.2	25.5
199	232.1	8.6	8.7	18.4

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

Out[6]: <Axes: >



```
In [7]: data.drop(columns = ["Radio", "Newspaper"], inplace = True)
#pairplot
sns.pairplot(data)
data.Sales = np.log(data.Sales)
```



The dimension of X_train is (140, 2) The dimension of X test is (60, 2)

```
In [19]: #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))
```

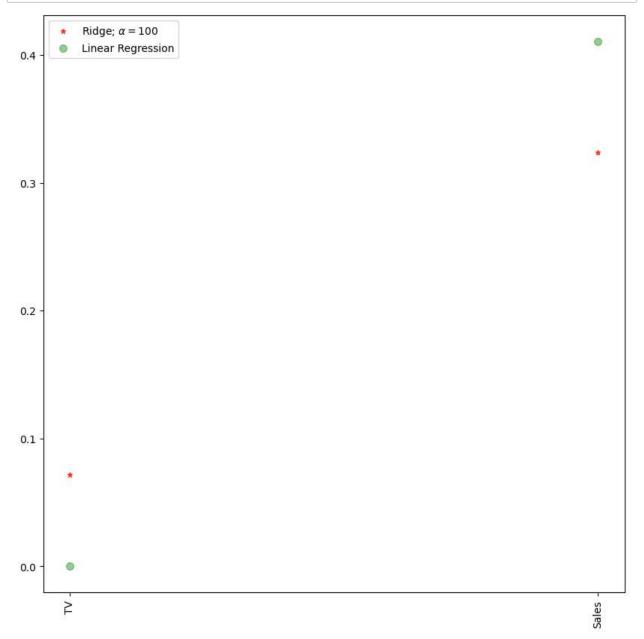
Linear Regression Model:

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

```
In [20]: #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 0.9902871391941609 The test score for ridge model is 0.984426628514122



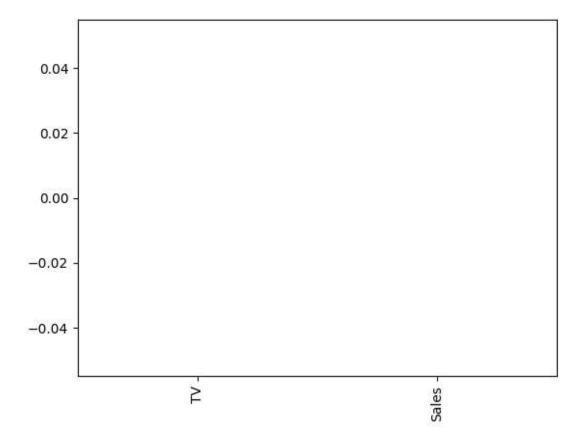
```
In [22]: #Lasso regression model
    print("\nLasso Model: \n")
    lasso = Lasso(alpha = 10)
    lasso.fit(X_train,y_train)
    train_score_ls =lasso.score(X_train,y_train)
    test_score_ls =lasso.score(X_test,y_test)
    print("The train score for ls model is {}".format(train_score_ls))
    print("The test score for ls model is {}".format(test_score_ls))
```

Lasso Model:

The train score for ls model is 0.0
The test score for ls model is -0.0042092253233847465

```
In [23]: pd.Series(lasso.coef_,features).sort_values(ascending=True).plot(kind="bar")
```

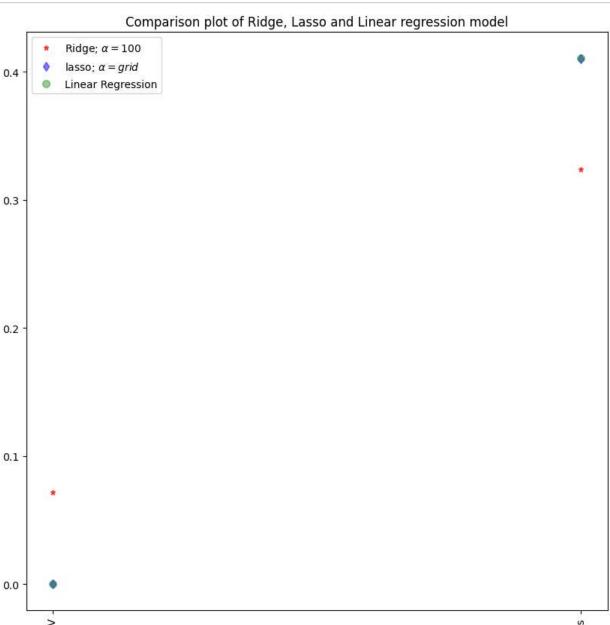
Out[23]: <Axes: >



0.9999999343798134

0.9999999152638072

```
#plot size
In [25]:
         plt.figure(figsize = (10, 10))
         #add plot for ridge regression
         plt.plot(features,ridgeReg.coef ,alpha=0.7,linestyle='none',marker='*',
                  markersize=5,color='red',label=r'Ridge; $\alpha=100$')
         #add plot for lasso regression
         plt.plot(lasso cv.coef ,alpha=0.5,linestyle='none',marker='d',
                  markersize=6,color='blue',label=r'lasso; $\alpha = grid$')
         #add plot for linear model
         plt.plot(features, lr.coef , alpha=0.4, linestyle='none', marker='o',
                  markersize=7,color='green',label='Linear Regression')
         #rotate axis
         plt.xticks(rotation = 90)
         plt.legend()
         plt.title("Comparison plot of Ridge, Lasso and Linear regression model")
         plt.show()
```



```
#Using the linear CV model
In [32]:
         from sklearn.linear model import RidgeCV
         #Ridge Cross validation
         ridge cv=RidgeCV(alphas=[0.0001,0.001,0.01,0.1,1,10]).fit(X train,y train)
         #score
         print("The train score for ridge model is {}"
               .format(ridge cv.score(X train,y train)))
         print("The train score for ridge model is {}"
               .format(ridge_cv.score(X_test,y_test)))
         The train score for ridge model is 0.999999999997627
         The train score for ridge model is 0.9999999999962466
         ELASTIC NET
In [28]: from sklearn.linear_model import ElasticNet
         regr = ElasticNet()
         regr.fit(X,y)
         print(regr.coef )
         print(regr.intercept_)
         [0.00417976 0.
         2.026383919311004
```

```
In [30]: y_pred_elastic = regr.predict(X_train)
```

```
In [31]: mean_squared_error = np.mean((y_pred_elastic-y_train)**2)
    print("Mean squared Error on test set", mean_squared_error)
```

Mean squared Error on test set 0.5538818050142158

Vehicle selection

```
In [33]: 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
```

Out[34]:

	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	рор	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	рор	51	2223	60457	1	45.481541	9.413480	7500
1536	1537	lounge	51	2557	80750	1	45.000702	7.682270	5990
1537	1538	рор	51	1766	54276	1	40.323410	17.568270	7900

1538 rows × 9 columns

```
In [39]: data = data[['engine_power','price']]
   data.columns=['Eng','pri']
```

In [35]: data.head()

Out[35]:

	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	рор	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

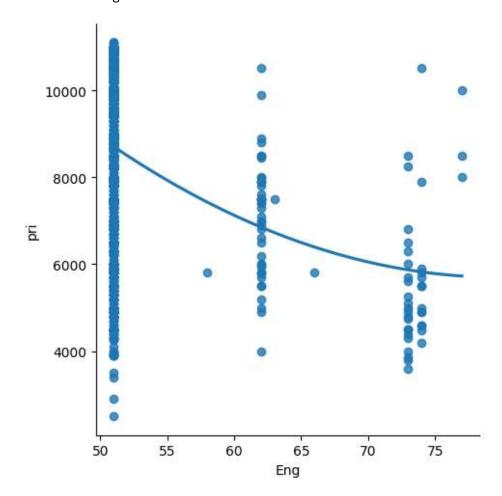
In [36]: data.tail()

Out[36]:

	ID	model	engine_power	age_in_days	km	previous_owners	lat	lon	price
1533	1534	sport	51	3712	115280	1	45.069679	7.70492	5200
1534	1535	lounge	74	3835	112000	1	45.845692	8.66687	4600
1535	1536	рор	51	2223	60457	1	45.481541	9.41348	7500
1536	1537	lounge	51	2557	80750	1	45.000702	7.68227	5990
1537	1538	рор	51	1766	54276	1	40.323410	17.56827	7900

In [40]: sns.lmplot(x='Eng',y='pri',data=data,order=2,ci=None)

Out[40]: <seaborn.axisgrid.FacetGrid at 0x167df2b3880>



```
data.info()
In [41]:
          data.describe()
          <class 'pandas.core.frame.DataFrame'>
          RangeIndex: 1538 entries, 0 to 1537
          Data columns (total 2 columns):
                Column Non-Null Count Dtype
           0
                Eng
                        1538 non-null
                                          int64
           1
                pri
                        1538 non-null
                                          int64
          dtypes: int64(2)
          memory usage: 24.2 KB
Out[41]:
                                      pri
                        Eng
                              1538.000000
           count 1538.000000
           mean
                   51.904421
                              8576.003901
             std
                    3.988023
                              1939.958641
            min
                   51.000000
                              2500.000000
            25%
                   51.000000
                              7122.500000
            50%
                   51.000000
                              9000.000000
            75%
                   51.000000 10000.000000
                   77.000000
                             11100.000000
            max
In [42]: data.fillna(method='ffill')
Out[42]:
                Eng
                       pri
              0
                  51
                     8900
              1
                     8800
              2
                  74
                     4200
              3
                  51
                     6000
                  73 5700
              4
           1533
                  51 5200
           1534
                  74 4600
                  51 7500
           1535
           1536
                     5990
           1537
                  51 7900
          1538 rows × 2 columns
In [43]:
          x=np.array(data['Eng']).reshape(-1,1)
          y=np.array(data['pri']).reshape(-1,1)
```

In [44]: | data.dropna(inplace=True)

C:\Users\sneha\AppData\Local\Temp\ipykernel_15004\1368182302.py:1: SettingWithCopyW
arning:

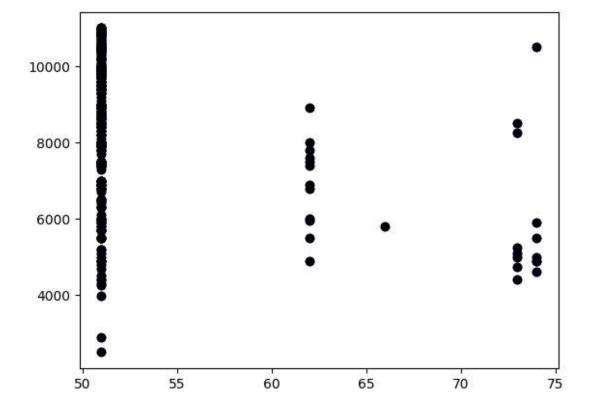
A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_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 [45]: 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))
```

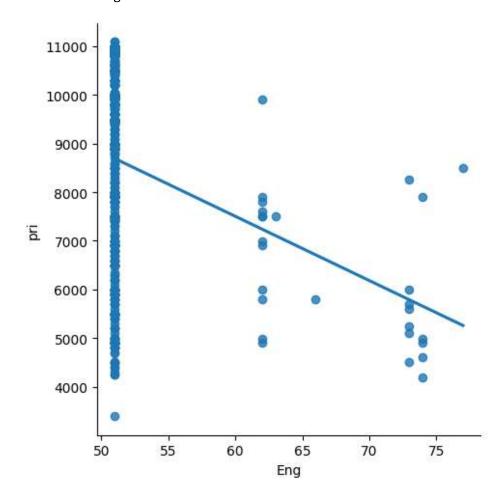
0.10416218443333636

```
In [46]: y_pred = regr.predict(X_test)
plt.scatter(X_test, y_test, color = 'b')
plt.scatter(X_test, y_test, color = 'k')
plt.show()
```



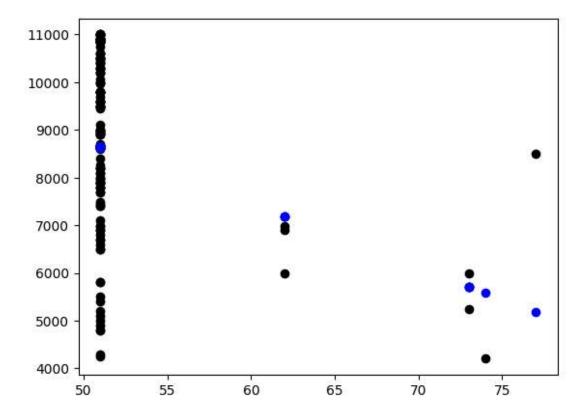
```
In [47]: df500 = data[:][:500]
# Selecting the 1st 500 rows of teh data
sns.lmplot(x = "Eng", y = "pri", data = df500, order = 1, ci = None)
```

Out[47]: <seaborn.axisgrid.FacetGrid at 0x167dfeb52a0>



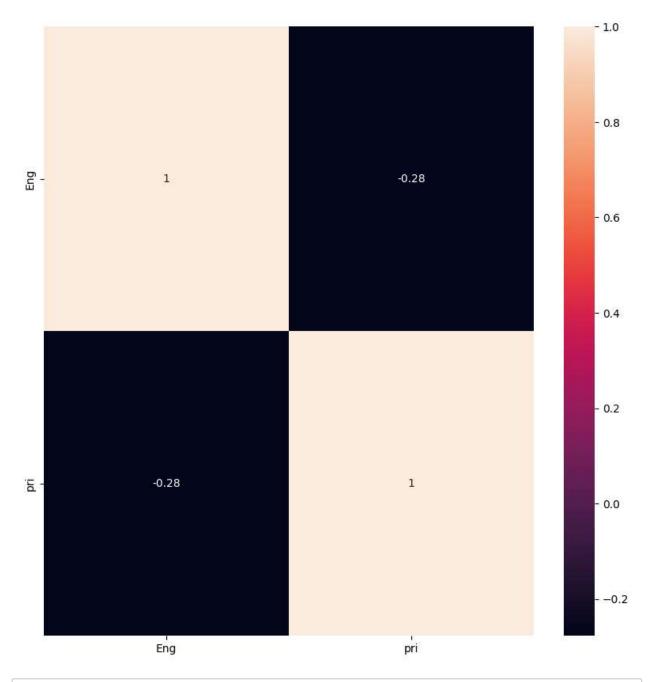
```
In [48]: df500.fillna(method = 'ffill', inplace = True)
    x = np.array(df500['Eng']).reshape(-1,1)
    y = np.array(df500['pri']).reshape(-1,1)
    df500.dropna(inplace = True)
    X_train, X_test, y_train, y_test = train_test_split(x,y,test_size = 0.25)
    regr = LinearRegression()
    regr.fit(X_train, y_train)
    print("Regression:",regr.score(X_test, y_test))
    y_pred = regr.predict(X_test)
    plt.scatter(X_test, y_test, color = 'k')
    plt.scatter(X_test, y_pred, color = 'b')
    plt.show()
```

Regression: 0.09996326546147283



```
In [49]: plt.figure(figsize = (10, 10))
sns.heatmap(data.corr(), annot = True)
```

Out[49]: <Axes: >



```
In [50]: from sklearn.linear_model import LinearRegression
    from sklearn.metrics import r2_score
    #Train the model
    model = LinearRegression()
    model.fit(X_train, y_train)
    #Evaluating the model on the test set
    y_pred = model.predict(X_test)
    r2 = r2_score(y_test, y_pred)
    print("R2 score:",r2)
```

R2 score: 0.09996326546147283

Linear Regression is not fit in this model

Linear Regression Model:

The train score for lr model is 0.05927530779021295 The test score for lr model is 0.09996326546147283

Linear Regression Model:

The train score for lr model is 0.05927530779021295 The test score for lr model is 0.09996326546147283

```
In [53]: #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 ls model is 0.059273621960479805 The test score for ls model is 0.1000081328294059

- 0.05927530779021284
- 0.0999632659422679

C:\Users\sneha\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\li
near_model_coordinate_descent.py:1568: DataConversionWarning: A column-vector y wa
s passed when a 1d array was expected. Please change the shape of y to (n_samples,
), for example using ravel().
 y = column_or_1d(y, warn=True)

In []: