In [30]: import pandas as pd
import numpy as np

Out[2]:

	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 [3]: df.head()

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

In [4]: df.tail()

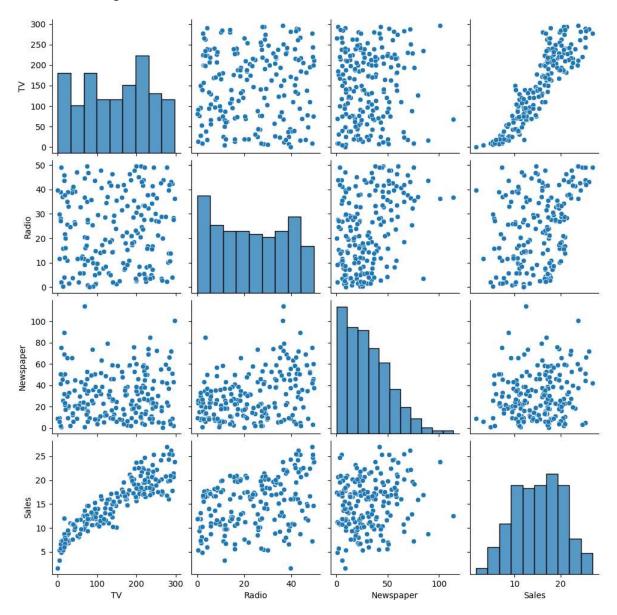
Out[4]:

	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 [5]: df.shape
Out[5]: (200, 4)
In [6]: df.describe()
Out[6]:
                       TV
                               Radio
                                     Newspaper
                                                     Sales
          count 200,000000
                           200.000000
                                     200.000000
                                                200.000000
                                                  15.130500
          mean 147.042500
                            23.264000
                                       30.554000
            std
                 85.854236
                            14.846809
                                       21.778621
                                                  5.283892
                  0.700000
                            0.000000
                                       0.300000
                                                  1.600000
           min
           25%
                 74.375000
                             9.975000
                                       12.750000
                                                  11.000000
           50% 149.750000
                            22.900000
                                       25.750000
                                                  16.000000
           75% 218.825000
                            36.525000
                                       45.100000
                                                  19.050000
           max 296.400000
                            49.600000
                                      114.000000
                                                 27.000000
In [7]: df.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 200 entries, 0 to 199
         Data columns (total 4 columns):
          #
              Column
                          Non-Null Count
                                            Dtype
              _____
                           _____
                                            ____
          0
              TV
                          200 non-null
                                            float64
          1
              Radio
                          200 non-null
                                            float64
          2
              Newspaper
                          200 non-null
                                            float64
          3
              Sales
                          200 non-null
                                            float64
         dtypes: float64(4)
         memory usage: 6.4 KB
In [8]:
         import seaborn as sns
         import matplotlib.pyplot as plt
```

In [9]: sns.pairplot(df)

Out[9]: <seaborn.axisgrid.PairGrid at 0x22d9f352170>



In [100]: features =df.columns[0:3]

In [101]: target = df.columns[-1]

```
Advertising using Ridge and Lasso - Jupyter Notebook
In [102]: sns.pairplot(df,x_vars=['TV','Radio','Newspaper'],y_vars='Sales',height=7,aspe
Out[102]: <seaborn.axisgrid.PairGrid at 0x22dc48368c0>
            Sales 15
In [103]: from sklearn.model selection import train test split
           from sklearn.linear_model import LinearRegression
```

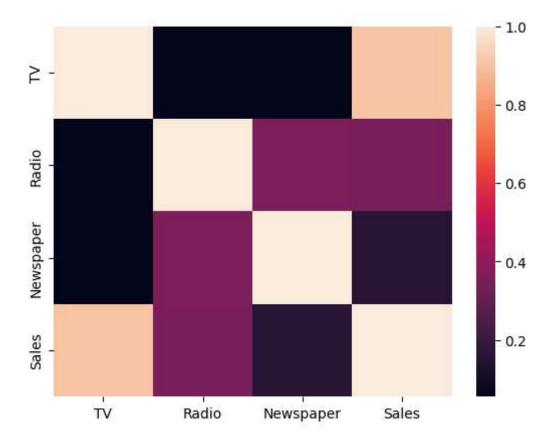
```
In [104]:
          x=np.array(df[features])
          y=np.array(df[target])
```

```
In [105]:
          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(regr.score(x_test,y_test))
```

0.8631814433474745

```
In [106]: sns.heatmap(df.corr())
```

Out[106]: <Axes: >



```
In [107]: from sklearn.linear_model import LinearRegression
lm=LinearRegression()
lm.fit(x_train,y_train)
```

Out[107]: LinearRegression()

In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.

On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.

```
In [111]: | x=Addf[['TV', 'Radio', 'Newspaper']]
           y=df['Sales']
In [112]: | from sklearn.linear_model import LinearRegression
           lm=LinearRegression()
           lm.fit(x_train,y_train)
Out[112]: LinearRegression()
           In a Jupyter environment, please rerun this cell to show the HTML representation or
           trust the notebook.
           On GitHub, the HTML representation is unable to render, please try loading this page
           with nbviewer.org.
In [113]: |print(lm.intercept_)
           4.675527241458383
In [114]:
          coeff_df = pd.DataFrame(lm.coef_)
           coeff df
Out[114]:
                     0
              0.056040
              0.099342
           2 -0.000878
          predictions=lm.predict(x test)
In [115]:
In [116]: from sklearn import metrics
           print('MAE:',metrics.mean_absolute_error(y_test,predictions))
           print('MSE:',metrics.mean_squared_error(y_test,predictions))
           print('RMSE:',np.sqrt(metrics.mean squared error(y test,predictions)))
           MAE: 1.2450972910731728
           MSE: 2.9342791752483883
           RMSE: 1.712973781249552
```

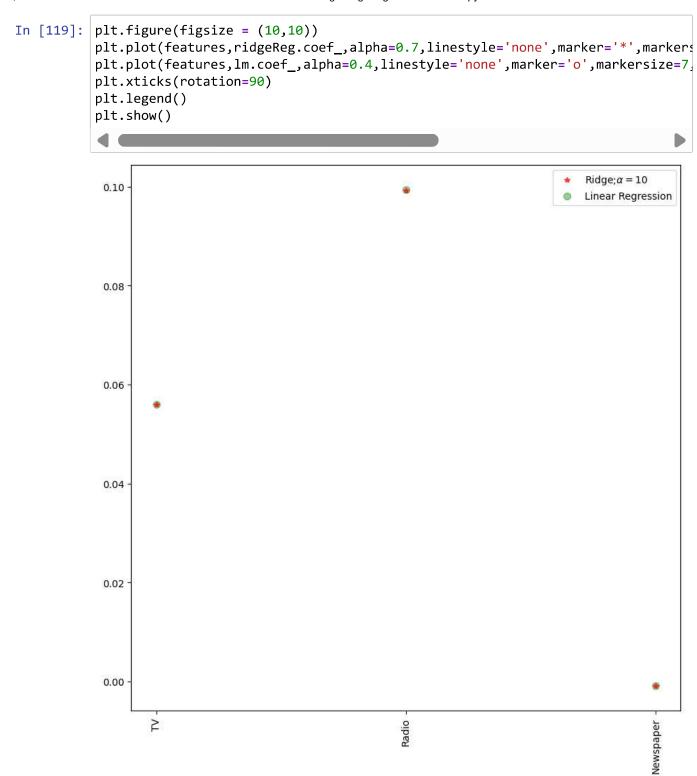
Ridge Regression Model

```
In [117]: | from sklearn.linear_model import LinearRegression
          from sklearn.linear_model import Ridge,RidgeCV,Lasso
          from sklearn.preprocessing import StandardScaler
          from sklearn.model_selection import train_test_split
```

```
In [118]: ridgeReg = Ridge(alpha=10)
    ridgeReg.fit(x_train,y_train)
    train_score_ridge = ridgeReg.score(x_train,y_train)
    test_score_ridge = ridgeReg.score(x_test,y_test)
    print('\nRidge Model\n')
    print('Train score for ridge model is {}' .format(train_score_ridge))
    print('Test score for ridge model is {}' .format(test_score_ridge))
```

Ridge Model

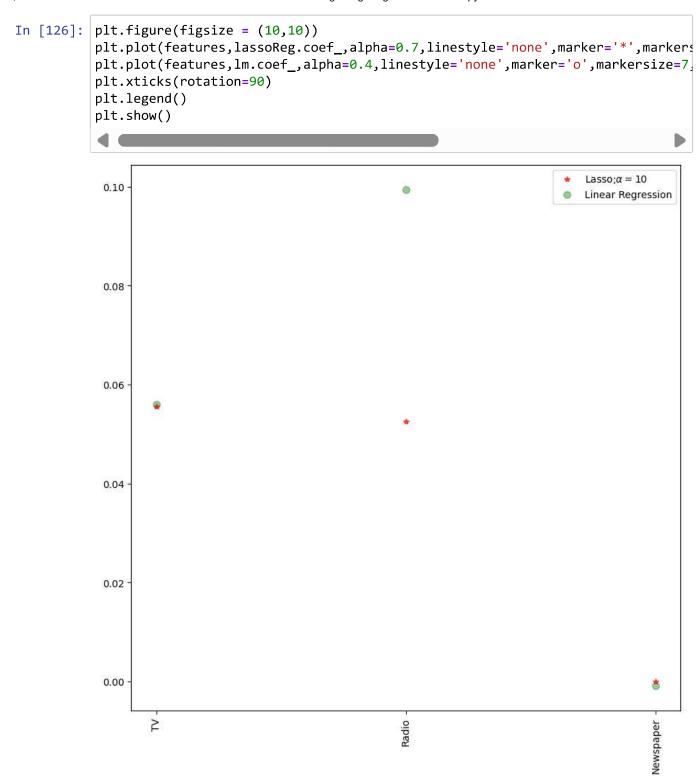
Train score for ridge model is 0.9103297738689969 Test score for ridge model is 0.8631625911226438



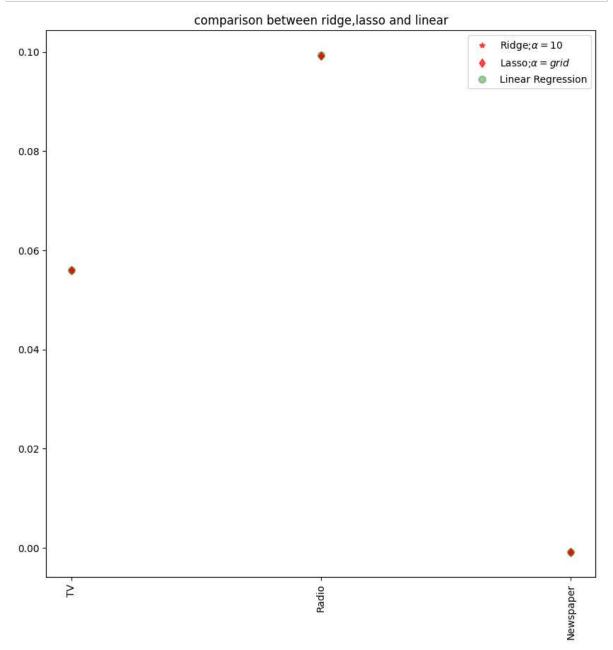
```
In [125]: lassoReg = Lasso(alpha=10)
lassoReg.fit(x_train,y_train)
train_score_lasso = lassoReg.score(x_train,y_train)
test_score_lasso = lassoReg.score(x_test,y_test)
print('\nRidge Model\n')
print('Train score for lasso model is {}'.format(train_score_lasso))
print('Test score for lasso model is{}'.format(test_score_lasso))
```

Ridge Model

Train score for lasso model is 0.8946969173105311 Test score for lasso model is 0.8123990929003627



```
In [121]: # Comparison between Ridge, Lasso and RidgeCV
plt.figure(figsize=(10,10))
plt.plot(features, ridgeReg.coef_, alpha=0.7, linestyle='none', marker='*', markers
plt.plot(features, lm.coef_, alpha=0.7, linestyle='none', marker='d', markers
plt.plot(features, lm.coef_, alpha=0.4, linestyle='none', marker='o', markersize=7,
plt.xticks(rotation=90)
plt.title('comparison between ridge, lasso and linear')
plt.legend()
plt.show()
```



In [123]: # Linear CV model using Ridge from sklearn.linear_model import RidgeCV ridge_CV=RidgeCV(alphas=[0.0001,0.01,0.001,0.1]).fit(x_train,y_train) print("The train score for ridge model is {}".format(ridge_CV.score(x_train,y_print("The test score for ridge model is {}".format(ridge_CV.score(x_test,y_text)).

The train score for ridge model is 0.9103297817545735 The test score for ridge model is 0.863181256614648

In [124]: # Linear CV model using Losso from sklearn.linear_model import LassoCV lasso_CV=L=LassoCV(alphas=[0.0001,0.01,0.001,0.1]).fit(x_train,y_train) print("The train score for lasso model is {}".format(lasso_CV.score(x_train,y_print("The test score for lasso model is {}".format(lasso_CV.score(x_test,y_text))

The train score for lasso model is 0.9103264044782 The test score for lasso model is 0.8629561862241895

In []:		