```
In [1]: import pandas as pd
import warnings
warnings.filterwarnings('ignore')
data=pd.read_csv('/home/placement/Downloads/Advertising.csv')#read the titanic csv file
print(data)
```

	Unnamed: 0	TV	radio	newspaper	sales
0	1	230.1	37.8	69.2	22.1
1	2	44.5	39.3	45.1	10.4
2	3	17.2	45.9	69.3	9.3
3	4	151.5	41.3	58.5	18.5
4	5	180.8	10.8	58.4	12.9
195	196	38.2	3.7	13.8	7.6
196	197	94.2	4.9	8.1	9.7
197	198	177.0	9.3	6.4	12.8
198	199	283.6	42.0	66.2	25.5
199	200	232.1	8.6	8.7	13.4

[200 rows x 5 columns]

In [2]: data.describe()

#### Out[2]:

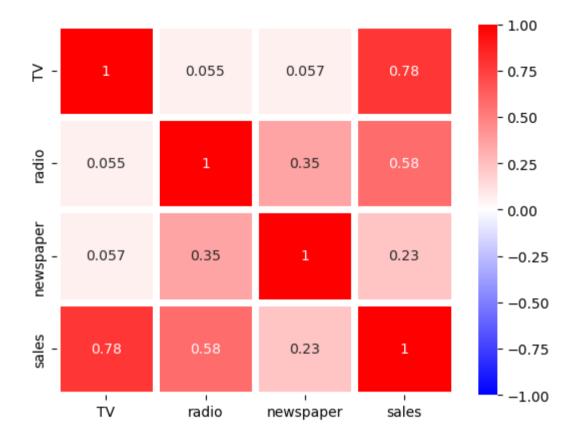
	Unnamed: 0	TV	radio	newspaper	sales
count	200.000000	200.000000	200.000000	200.000000	200.000000
mean	100.500000	147.042500	23.264000	30.554000	14.022500
std	57.879185	85.854236	14.846809	21.778621	5.217457
min	1.000000	0.700000	0.000000	0.300000	1.600000
25%	50.750000	74.375000	9.975000	12.750000	10.375000
50%	100.500000	149.750000	22.900000	25.750000	12.900000
75%	150.250000	218.825000	36.525000	45.100000	17.400000
max	200.000000	296.400000	49.600000	114.000000	27.000000

```
In [3]: data.shape
Out[3]: (200, 5)
In [4]: data.columns
Out[4]: Index(['Unnamed: 0', 'TV', 'radio', 'newspaper', 'sales'], dtype='object')
In [8]: data1=data.drop(columns='Unnamed: 0')
In [9]: data1
Out[9]:
                 TV radio newspaper sales
            0 230.1
                     37.8
                               69.2
                                     22.1
                44.5
                     39.3
                               45.1
                                     10.4
               17.2
                     45.9
                               69.3
                                      9.3
            3 151.5
                     41.3
                               58.5
                                     18.5
            4 180.8
                     10.8
                               58.4
                                     12.9
          195
                38.2
                                      7.6
                      3.7
                               13.8
                                8.1
                                      9.7
          196
               94.2
                      4.9
          197 177.0
                      9.3
                                6.4
                                     12.8
                                     25.5
          198
              283.6
                     42.0
                               66.2
                                8.7 13.4
          199 232.1
                      8.6
         200 rows × 4 columns
```

```
In [10]: data1.isna().sum()
Out[10]: TV
                         0
           radio
          newspaper
                         0
          sales
          dtype: int64
In [11]: cor=datal.corr()
In [12]: cor
Out[12]:
                          TV
                                 radio newspaper
                                                    sales
                 TV 1.000000 0.054809
                                        0.056648  0.782224
                radio 0.054809 1.000000
                                        0.354104 0.576223
           newspaper 0.056648 0.354104
                                        1.000000 0.228299
                sales 0.782224 0.576223
                                        0.228299 1.000000
```

```
In [13]: import seaborn as sns
sns.heatmap(cor,vmax=1,vmin=-1,annot=True,linewidth=5,cmap='bwr')
```

Out[13]: <Axes: >



```
In [14]: y=data1['sales']
x=data1.drop(columns='sales')
```

### In [15]: x

### Out[15]:

TV	radio	newspaper
230.1	37.8	69.2
44.5	39.3	45.1
17.2	45.9	69.3
151.5	41.3	58.5
180.8	10.8	58.4
38.2	3.7	13.8
94.2	4.9	8.1
177.0	9.3	6.4
283.6	42.0	66.2
232.1	8.6	8.7
	230.1 44.5 17.2 151.5 180.8  38.2 94.2 177.0 283.6	230.1 37.8 44.5 39.3 17.2 45.9 151.5 41.3 180.8 10.8  38.2 3.7 94.2 4.9 177.0 9.3 283.6 42.0

200 rows × 3 columns

```
In [16]: y
Out[16]: 0
                     22.1
                     10.4
                      9.3
            2
            3
                     18.5
            4
                     12.9
                     . . .
                     7.6
            195
                      9.7
            196
            197
                     12.8
            198
                     25.5
            199
                     13.4
           Name: sales, Length: 200, dtype: float64
In [52]: from sklearn.model_selection import train_test_split
    x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.33,random_state=42)
```

In [53]: x\_train

Out[53]:

	TV	radio	newspaper
42	293.6	27.7	1.8
189	18.7	12.1	23.4
90	134.3	4.9	9.3
136	25.6	39.0	9.3
51	100.4	9.6	3.6
106	25.0	11.0	29.7
14	204.1	32.9	46.0
92	217.7	33.5	59.0
179	165.6	10.0	17.6
102	280.2	10.1	21.4

134 rows × 3 columns

In [54]: x test

Out[54]:

	TV	radio	newspaper
95	163.3	31.6	52.9
15	195.4	47.7	52.9
30	292.9	28.3	43.2
158	11.7	36.9	45.2
128	220.3	49.0	3.2
97	184.9	21.0	22.0
31	112.9	17.4	38.6
12	23.8	35.1	65.9
35	290.7	4.1	8.5
119	19.4	16.0	22.3

66 rows × 3 columns

# LinearRegression

```
In [55]: from sklearn.linear_model import LinearRegression
    reg=LinearRegression()
    reg.fit(x_train,y_train)
```

Out[55]: 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 [56]: y_pred_reg=reg.predict(x_test)
```

```
In [57]: y pred reg
Out[57]: array([16.58673085, 21.18622524, 21.66752973, 10.81086512, 22.25210881,
                13.31459455, 21.23875284, 7.38400509, 13.43971113, 15.19445383,
                 9.01548612, 6.56945204, 14.4156926, 8.93560138, 9.56335776,
                12.10760805, 8.86091137, 16.25163621, 10.31036304, 18.83571624,
                19.81058732, 13.67550716, 12.45182294, 21.58072583, 7.67409148,
                 5.67090757, 20.95448184, 11.89301758, 9.13043149, 8.49435255,
                12.32217788, 9.99097553, 21.71995241, 12.64869606, 18.25348116,
                20.17390876, 14.20864218, 21.02816483, 10.91608737, 4.42671034,
                 9.59359543, 12.53133363, 10.14637196, 8.1294087, 13.32973122,
                 5.27563699, 9.30534511, 14.15272317, 8.75979349, 11.67053724,
                15.66273733, 11.75350353, 13.21744723, 11.06273296, 6.41769181,
                 9.84865789, 9.45756213, 24.32601732, 7.68903682, 12.30794356,
                17.57952015, 15.27952025, 11.45659815, 11.12311877, 16.60003773,
                 6.906114781)
In [58]: from sklearn.metrics import r2 score
         r2 score(y test, y pred reg)
Out[58]: 0.8555568430680086
In [59]: from sklearn.metrics import mean squared error
         mean squared error(y test,y pred reg)
Out[59]: 3.7279283306815105
```

## RidgeRegression

```
ridge=Ridge()
                                                                #creating an object for Ridge
          parameters={'alpha':alpha}
          ridge regressor=GridSearchCV(ridge,parameters)
          ridge regressor.fit(x train,y train)
                                                                #training and fitting
Out[60]: GridSearchCV(estimator=Ridge(),
                       param grid={'alpha': [1e-15, 1e-10, 0.0001, 0.001, 0.01, 15, 10,
                                               20. 3011)
         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 [61]: ridge regressor.best params
Out[61]: {'alpha': 30}
In [62]: ridge=Ridge(alpha=30)
          ridge.fit(x train,y train)
         y pred ridge=ridge.predict(x test)
In [63]: from sklearn.metrics import r2 score #to know the efficiency of the predicted price
         r2 score(y test,y pred ridge)
Out[63]: 0.8556430695733936
In [64]: from sklearn.metrics import mean squared error
         mean squared error(y test,y pred ridge)
Out[64]: 3.7257029138524405
In [ ]:
```

## **ElasticRegression**

In [60]: from sklearn.model\_selection import GridSearchCV
from sklearn.linear model import Ridge

alpha=[1e-15.1e-10.1e-4.1e-3.1e-2.15.10.20.30]

#creating an object for ElasticNet

```
parameters={'alpha':[1e-15,1e-10,1e-8,1e-4,1e-3,1e-2,15,10,20]}
         elastic regressor=GridSearchCV(elastic,parameters)
         elastic regressor.fit(x train,y train)
                                                     #training and fitting
Out[65]: GridSearchCV(estimator=ElasticNet(),
                       param grid={'alpha': [1e-15, 1e-10, 1e-08, 0.0001, 0.001, 0.01, 15,
                                              10. 2011)
         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 nbyiewer.org.
In [66]: elastic regressor.best params
Out[66]: {'alpha': 0.01}
In [67]: elastic=ElasticNet(alpha=0.01)
         elastic.fit(x train,y train)
         v pred enet=elastic.predict(x test)
In [68]: from sklearn.metrics import r2 score #to know the efficiency of the predicted price
         r2 score(y test,y pred enet)
Out[68]: 0.855576715693211
In [69]: from sklearn.metrics import mean squared error
         mean squared error(y test,y pred enet)
Out[69]: 3.7274154388002283
In [80]: test=[[110,33,21],[220,6,13]]
         y pred enet=elastic.predict(test)
In [81]: y pred enet #sales we get from testdata
Out[81]: array([14.28742973, 13.84367754])
```

In [65]: **from** sklearn.linear model **import** ElasticNet

elastic=ElasticNet()

from sklearn.model selection import GridSearchCV

In [ ]: