

```
In [1]: import pandas as pd
import warnings
warnings.filterwarnings('ignore')
```

```
In [2]: data=pd.read_csv("/home/placement/Downloads/Advertising.csv")
```

```
In [3]: data.describe()
```

Out[3]:

	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 [4]: data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 200 entries, 0 to 199
Data columns (total 5 columns):
#   Column      Non-Null Count  Dtype
---  -
0   Unnamed: 0   200 non-null    int64
1   TV           200 non-null    float64
2   radio        200 non-null    float64
3   newspaper    200 non-null    float64
4   sales        200 non-null    float64
dtypes: float64(4), int64(1)
memory usage: 7.9 KB
```

```
In [5]: data1=data.drop(["Unnamed: 0"],axis=1)
```

```
In [6]: data1
```

```
Out[6]:
```

	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	9.3
3	151.5	41.3	58.5	18.5
4	180.8	10.8	58.4	12.9
...	...	...	...	...
195	38.2	3.7	13.8	7.6
196	94.2	4.9	8.1	9.7
197	177.0	9.3	6.4	12.8
198	283.6	42.0	66.2	25.5
199	232.1	8.6	8.7	13.4

200 rows × 4 columns

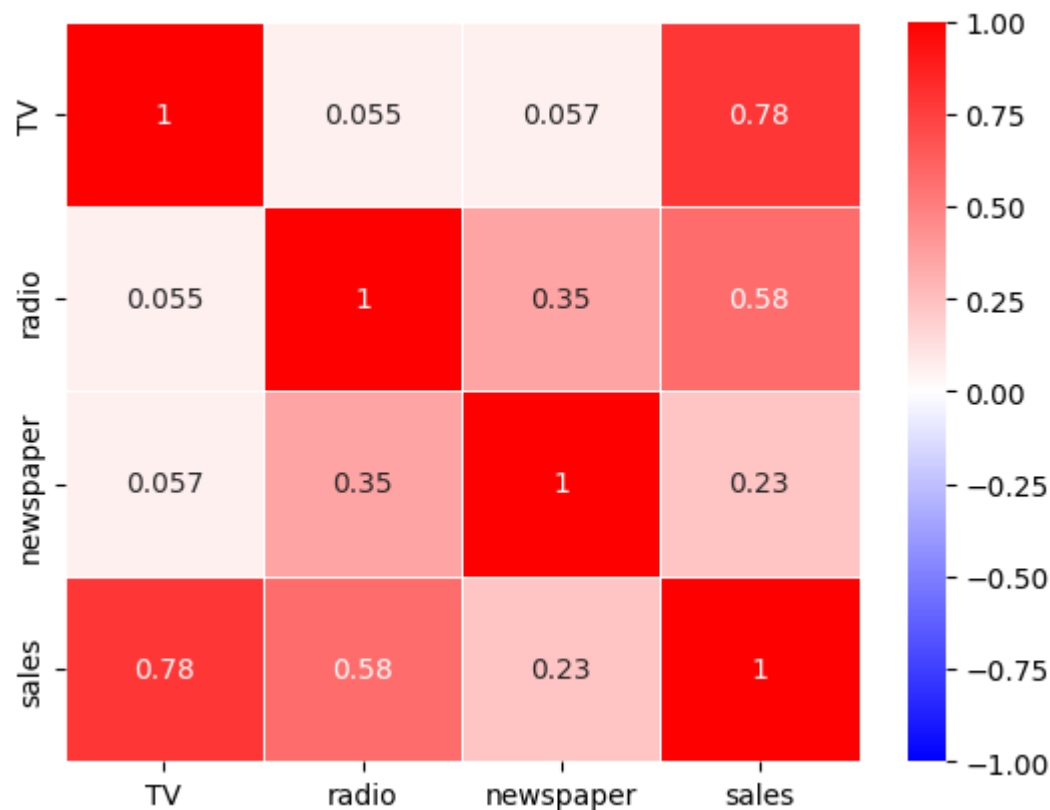
```
In [7]: cor_mat=data1.corr()  
cor_mat#correlation of data
```

```
Out[7]:
```

	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 [8]: import seaborn as sns#data of correlation in a graoh
sns.heatmap(cor_mat,vmax=1,vmin=-1,annot=True,linewidth=.5,cmap='bwr')#plotting of graph using seaborn
```

Out[8]: <Axes: >



```
In [9]: y=data1['sales']
x=data1.drop(['sales'],axis=1)
```

In [10]:

y

Out[10]:

```
0      22.1
1      10.4
2       9.3
3      18.5
4      12.9
...
195     7.6
196     9.7
197    12.8
198    25.5
199    13.4
Name: sales, Length: 200, dtype: float64
```

In [11]:

x

Out[11]:

	TV	radio	newspaper
0	230.1	37.8	69.2
1	44.5	39.3	45.1
2	17.2	45.9	69.3
3	151.5	41.3	58.5
4	180.8	10.8	58.4
...	...	...	...
195	38.2	3.7	13.8
196	94.2	4.9	8.1
197	177.0	9.3	6.4
198	283.6	42.0	66.2
199	232.1	8.6	8.7

200 rows × 3 columns

```
In [12]: 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)#dividing training data a
```

```
In [13]: x_test.head(5)#display top 5 data in testing data
```

Out[13]:

	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

```
In [14]: y_test.head(5)#display top 5 data in testing data price dataframe
```

Out[14]:

95	16.9
15	22.4
30	21.4
158	7.3
128	24.7

Name: sales, dtype: float64

```
In [15]: x_train.head(5)#display top 5 data in training data
```

Out[15]:

	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

```
In [16]: y_train.head(5)#display top 5 data in training data price dataframe
```

```
Out[16]: 42      20.7  
        189      6.7  
        90      11.2  
        136      9.5  
        51      10.7  
        Name: sales, dtype: float64
```

```
In [17]: from sklearn.linear_model import LinearRegression  
        reg=LinearRegression()#creating the object of linear Regression  
        reg.fit(x_train,y_train)#training and fitting linear Regression using training data
```

```
Out[17]: 

▼ LinearRegression



LinearRegression()


```

```
In [18]: ypred=reg.predict(x_test)#calculating predicting value
```

```
In [19]: ypred
```

```
Out[19]: 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,  
                6.90611478])
```

```
In [20]: from sklearn.metrics import r2_score  
r2_score(y_test,ypred)#finding the efficieny
```

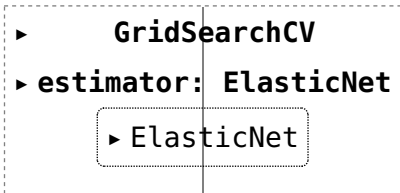
Out[20]: 0.8555568430680086

```
In [21]: from sklearn.metrics import mean_squared_error#mean_squared error  
a=mean_squared_error(y_test,ypred)  
a
```

Out[21]: 3.7279283306815105

```
In [22]: from sklearn.linear_model import ElasticNet  
from sklearn.model_selection import GridSearchCV  
elastic = ElasticNet()  
  
parameters = {'alpha': [1e-15, 1e-10, 1e-8, 1e-4, 1e-3,1e-2, 1, 5, 10, 20]}  
  
elastic_regressor = GridSearchCV(elastic, parameters)  
  
elastic_regressor.fit(x_train, y_train)
```

Out[22]:



```
└─ GridSearchCV  
  └─ estimator: ElasticNet  
    └─ ElasticNet
```

```
In [23]: elastic_regressor.best_params_
```

Out[23]: {'alpha': 1}

```
In [24]: elastic=ElasticNet(alpha=0.01)  
elastic.fit(x_train,y_train)  
y_pred_elastic=elastic.predict(x_test)
```

```
In [25]: from sklearn.metrics import r2_score  
r2_score(y_test,y_pred_elastic)
```

```
Out[25]: 0.855576715693211
```

```
In [26]: x_test
```

```
Out[26]:
```

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

66 rows × 3 columns

```
In [27]: test=[[110,32,26]]  
y_pred_elastic=elastic.predict(test)
```

```
In [28]: y_pred_elastic
```

```
Out[28]: array([14.12116625])
```

```
In [ ]:
```



In [ ]: