Sales Prediction System

June 20, 2021

[1]:
```
import numpy as np
import pandas as pd
```

1 Reading the data set

[2]:
```
data=pd.read_csv("/home/sumon/Data Science Note/Advertising.csv",index_col=0)
data
```

[2]:

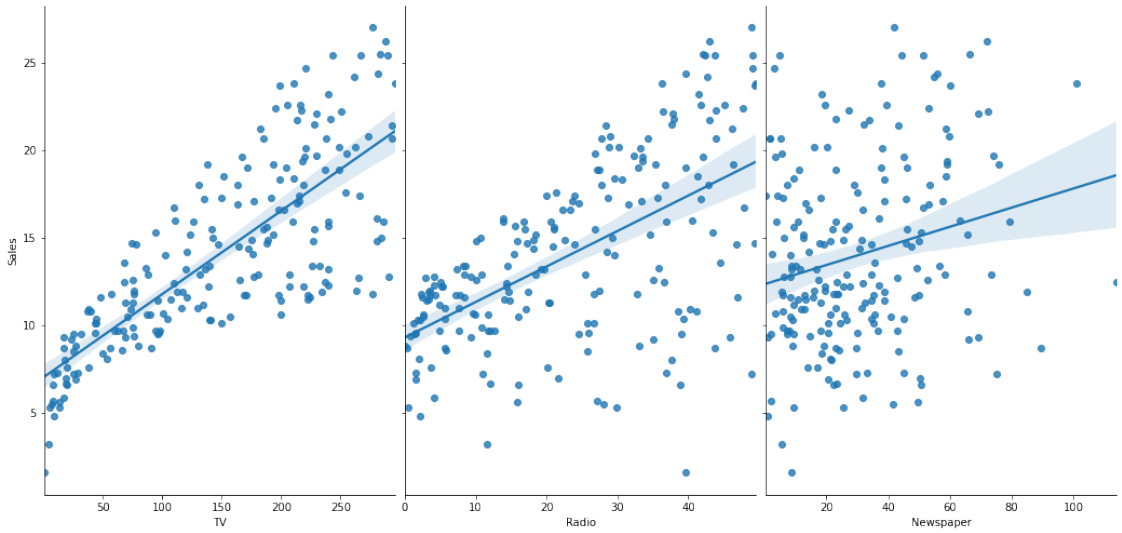
	TV	Radio	Newspaper	Sales
1	230.1	37.8	69.2	22.1
2	44.5	39.3	45.1	10.4
3	17.2	45.9	69.3	9.3
4	151.5	41.3	58.5	18.5
5	180.8	10.8	58.4	12.9
..	…	…	…	…
196	38.2	3.7	13.8	7.6
197	94.2	4.9	8.1	9.7
198	177.0	9.3	6.4	12.8
199	283.6	42.0	66.2	25.5
200	232.1	8.6	8.7	13.4

[200 rows x 4 columns]

2 Visualizing the different columns of the dataset

[6]:
```
import seaborn as sb
sb.pairplot(data,x_vars=['TV','Radio','Newspaper'],y_vars=['Sales'],aspect=0.
↪7,height=7,kind='reg')
```

[6]:
```
<seaborn.axisgrid.PairGrid at 0x7fcf386f7eb0>
```



3 Correlation matrix

[7]:
```
data.corr()
```

[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

[8]:
```
x=data[['TV','Radio','Newspaper']]
y=data['Sales']
```

[9]:
```
x
```

[9]:

	TV	Radio	Newspaper
1	230.1	37.8	69.2
2	44.5	39.3	45.1
3	17.2	45.9	69.3
4	151.5	41.3	58.5
5	180.8	10.8	58.4
..	…	…	…
196	38.2	3.7	13.8
197	94.2	4.9	8.1
198	177.0	9.3	6.4
199	283.6	42.0	66.2
200	232.1	8.6	8.7

```
[200 rows x 3 columns]
```

[10]:
```
y
```

[10]:
```
1      22.1
2      10.4
3       9.3
4      18.5
5      12.9
       …
196     7.6
197     9.7
198    12.8
199    25.5
200    13.4
Name: Sales, Length: 200, dtype: float64
```

4 Split the data set into training data and testing data

[11]:
```
from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.3)
```

[12]:
```
x_train
```

[12]:
```
        TV  Radio  Newspaper
28   240.1   16.7       22.9
26   262.9    3.5       19.5
88   110.7   40.6       63.2
158  149.8    1.3       24.3
191   39.5   41.1        5.8
..     …    …         …
132  265.2    2.9       43.0
170  284.3   10.6        6.4
45    25.1   25.7       43.3
59   210.8   49.6       37.7
131    0.7   39.6        8.7

[140 rows x 3 columns]
```

[13]:
```
y_train
```

[13]:
```
28     15.9
26     12.0
88     16.0
158    10.1
191    10.8
       …
```

```
132    12.7
170    15.0
45      8.5
59     23.8
131     1.6
Name: Sales, Length: 140, dtype: float64
```

[14]:
```
x_test
```

[14]:

	TV	Radio	Newspaper
69	237.4	27.5	11.0
151	280.7	13.9	37.0
46	175.1	22.5	31.5
15	204.1	32.9	46.0
70	216.8	43.9	27.2
135	36.9	38.6	65.6
169	215.4	23.6	57.6
54	182.6	46.2	58.7
9	8.6	2.1	1.0
34	265.6	20.0	0.3
130	59.6	12.0	43.1
35	95.7	1.4	7.4
93	217.7	33.5	59.0
126	87.2	11.8	25.9
55	262.7	28.8	15.9
199	283.6	42.0	66.2
182	218.5	5.4	27.4
71	199.1	30.6	38.7
129	220.3	49.0	3.2
31	292.9	28.3	43.2
13	23.8	35.1	65.9
110	255.4	26.9	5.5
78	120.5	28.5	14.2
143	220.5	33.2	37.9
118	76.4	0.8	14.8
187	139.5	2.1	26.6
119	125.7	36.9	79.2
192	75.5	10.8	6.0
27	142.9	29.3	12.6
7	57.5	32.8	23.5
134	219.8	33.5	45.1
128	80.2	0.0	9.2
41	202.5	22.3	31.6
44	206.9	8.4	26.4
61	53.5	2.0	21.4
63	239.3	15.5	27.3
89	88.3	25.5	73.4

```
163  188.4   18.1       25.6
165  117.2   14.7        5.4
198  177.0    9.3        6.4
77    27.5    1.6       20.7
53   216.4   41.7       39.6
22   237.4    5.1       23.5
48   239.9   41.5       18.5
121  141.3   26.8       46.2
177  248.4   30.2       20.3
73    26.8   33.0       19.3
145   96.2   14.8       38.9
52   100.4    9.6        3.6
174  168.4    7.1       12.8
65   131.1   42.8       28.9
120   19.4   16.0       22.3
85   213.5   43.0       33.8
4    151.5   41.3       58.5
162   85.7   35.8       49.3
82   239.8    4.1       36.9
107   25.0   11.0       29.7
116   75.1   35.0       52.7
146  140.3    1.9        9.0
124  123.1   34.6       12.4
```

[15]:
```
y_test
```

[15]:
```
69     18.9
151    16.1
46     14.9
15     19.0
70     22.3
135    10.8
169    17.1
54     21.2
9       4.8
34     17.4
130     9.7
35      9.5
93     19.4
126    10.6
55     20.2
199    25.5
182    12.2
71     18.3
129    24.7
31     21.4
13      9.2
```

```
110    19.8
78     14.2
143    20.1
118     9.4
187    10.3
119    15.9
192     9.9
27     15.0
7      11.8
134    19.6
128     8.8
41     16.6
44     12.9
61      8.1
63     15.7
89     12.9
163    14.9
165    11.9
198    12.8
77      6.9
53     22.6
22     12.5
48     23.2
121    15.5
177    20.2
73      8.8
145    11.4
52     10.7
174    11.7
65     18.0
120     6.6
85     21.7
4      18.5
162    13.3
82     12.3
107     7.2
116    12.6
146    10.3
124    15.2
Name: Sales, dtype: float64
```

5 Training our model

[17]:
```
from sklearn.linear_model import LinearRegression
linreg=LinearRegression()
linreg.fit(x_train,y_train)
```

[17]:
```
LinearRegression()
```

6 Interpreting model coefficients

[18]:
```
print(linreg.coef_)
print(linreg.intercept_)
```

```
[0.04703762 0.1873887  0.00137635]
2.534126914516472
```

7 Making Predictions

[19]:
```
y_pred=linreg.predict(x_test)
y_pred
```

[19]:
```
array([18.86918738, 18.39321512, 15.03001512, 18.36290562, 20.99568379,
       11.59330714, 17.16768147, 19.861346  ,  3.33354307, 18.7755062 ,
        7.64555405,  7.30815649, 19.13294299,  8.88264154, 20.30958862,
       23.8354359 , 13.86145818, 17.68667618, 22.08296553, 21.67400469,
       10.32166675, 19.59586146, 13.56228235, 19.17939083,  6.29808212,
        9.52600226, 15.4704055 ,  8.11752339, 14.7636339 , 11.41748357,
       19.21259079,  6.31920659, 16.28150586, 13.87661151,  5.45487089,
       16.73232893, 11.56698448, 14.82298479, 10.80898234, 12.61130953,
        4.15597379, 20.58168029, 14.68888488, 21.62054579, 14.26614716,
       19.90535072, 10.00512566,  9.88603872,  9.06059051, 11.80333945,
       16.7607718 ,  6.47556844, 20.68089369, 17.47999607, 13.34162032,
       14.63282951,  5.8122206 , 12.69779014,  9.50193093, 14.8251738 ])
```

[20]:
```
y_test
```

[20]:
```
69     18.9
151    16.1
46     14.9
15     19.0
70     22.3
135    10.8
169    17.1
54     21.2
9       4.8
34     17.4
130     9.7
35      9.5
93     19.4
126    10.6
55     20.2
199    25.5
```

```
182    12.2
71     18.3
129    24.7
31     21.4
13      9.2
110    19.8
78     14.2
143    20.1
118     9.4
187    10.3
119    15.9
192     9.9
27     15.0
7      11.8
134    19.6
128     8.8
41     16.6
44     12.9
61      8.1
63     15.7
89     12.9
163    14.9
165    11.9
198    12.8
77      6.9
53     22.6
22     12.5
48     23.2
121    15.5
177    20.2
73      8.8
145    11.4
52     10.7
174    11.7
65     18.0
120     6.6
85     21.7
4      18.5
162    13.3
82     12.3
107     7.2
116    12.6
146    10.3
124    15.2
Name: Sales, dtype: float64
```

[24]:
```
comparison_data=pd.DataFrame()
comparison_data['y_test']=y_test
comparison_data['y_predict']=y_pred
comparison_data
```

[24]:

	y_test	y_predict
69	18.9	18.869187
151	16.1	18.393215
46	14.9	15.030015
15	19.0	18.362906
70	22.3	20.995684
135	10.8	11.593307
169	17.1	17.167681
54	21.2	19.861346
9	4.8	3.333543
34	17.4	18.775506
130	9.7	7.645554
35	9.5	7.308156
93	19.4	19.132943
126	10.6	8.882642
55	20.2	20.309589
199	25.5	23.835436
182	12.2	13.861458
71	18.3	17.686676
129	24.7	22.082966
31	21.4	21.674005
13	9.2	10.321667
110	19.8	19.595861
78	14.2	13.562282
143	20.1	19.179391
118	9.4	6.298082
187	10.3	9.526002
119	15.9	15.470406
192	9.9	8.117523
27	15.0	14.763634
7	11.8	11.417484
134	19.6	19.212591
128	8.8	6.319207
41	16.6	16.281506
44	12.9	13.876612
61	8.1	5.454871
63	15.7	16.732329
89	12.9	11.566984
163	14.9	14.822985
165	11.9	10.808982
198	12.8	12.611310
77	6.9	4.155974

```
53     22.6  20.581680
22     12.5  14.688885
48     23.2  21.620546
121    15.5  14.266147
177    20.2  19.905351
73      8.8  10.005126
145    11.4   9.886039
52     10.7   9.060591
174    11.7  11.803339
65     18.0  16.760772
120     6.6   6.475568
85     21.7  20.680894
4      18.5  17.479996
162    13.3  13.341620
82     12.3  14.632830
107     7.2   5.812221
116    12.6  12.697790
146    10.3   9.501931
124    15.2  14.825174
```

8 Evaluation metrics of model

[25]:
```
from sklearn import metrics
```

[26]:
```
#mean absolute error
m1=metrics.mean_absolute_error(y_test,y_pred)
m1
```

[26]: 1.094765870204584

[28]:
```
#mean squared error
m2=metrics.mean_squared_error(y_test,y_pred)
m2
```

[28]: 1.8706491944519692

[29]:
```
#root mean squared error
m3=np.sqrt(metrics.mean_squared_error(y_test,y_pred))
m3
```

[29]: 1.367716781520198

[ ]:
```

```