

## TASK 4

### ✓ SALES PREDICTION USING PYTHON

Importing libraries

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

IMPORT DATASET

```
df=pd.read_csv('advertising.csv')
```

```
df.head(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




Next steps: [Generate code with df](#) [View recommended plots](#) [New interactive sheet](#)


```
df.shape
```

```
(200, 4)
```

```
df.describe()
```



	TV	Radio	Newspaper	Sales
<b>count</b>	200.000000	200.000000	200.000000	200.000000
<b>mean</b>	147.042500	23.264000	30.554000	15.130500
<b>std</b>	85.854236	14.846809	21.778621	5.283892
<b>min</b>	0.700000	0.000000	0.300000	1.600000
<b>25%</b>	74.375000	9.975000	12.750000	11.000000
<b>50%</b>	149.750000	22.900000	25.750000	16.000000
<b>75%</b>	218.825000	36.525000	45.100000	19.050000
<b>max</b>	296.400000	49.600000	114.000000	27.000000



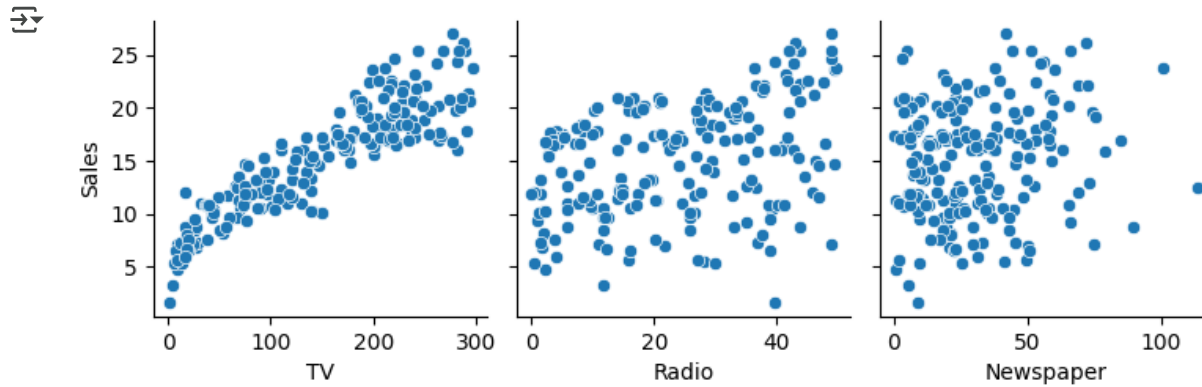
Basic observation

Avg expense spend is highest on TV.

Avg expense spend is lowest on Radio

Max sale is 27 & min is 1.6

```
sns.pairplot(df,x_vars=['TV','Radio','Newspaper'],y_vars='Sales',kind='scatter')
plt.show()
```



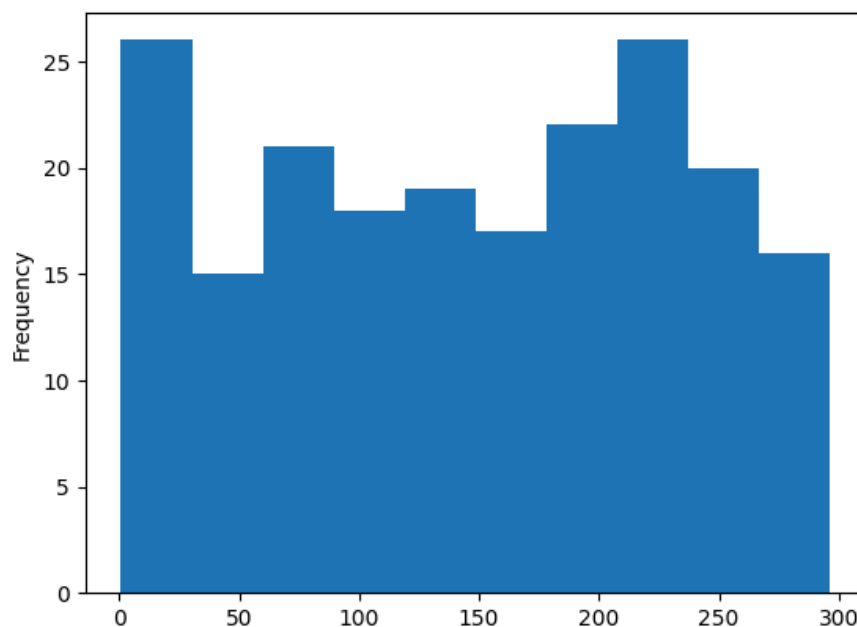
## ✓ Pair Plot Observation

When advertising cost increases in TV ads the sales will increase as well, while for newspaper and Radio it is unpredictable.

If we cannot determine the correlation using a scatter plot, we can use the seaborn heatmap to visualize the data.

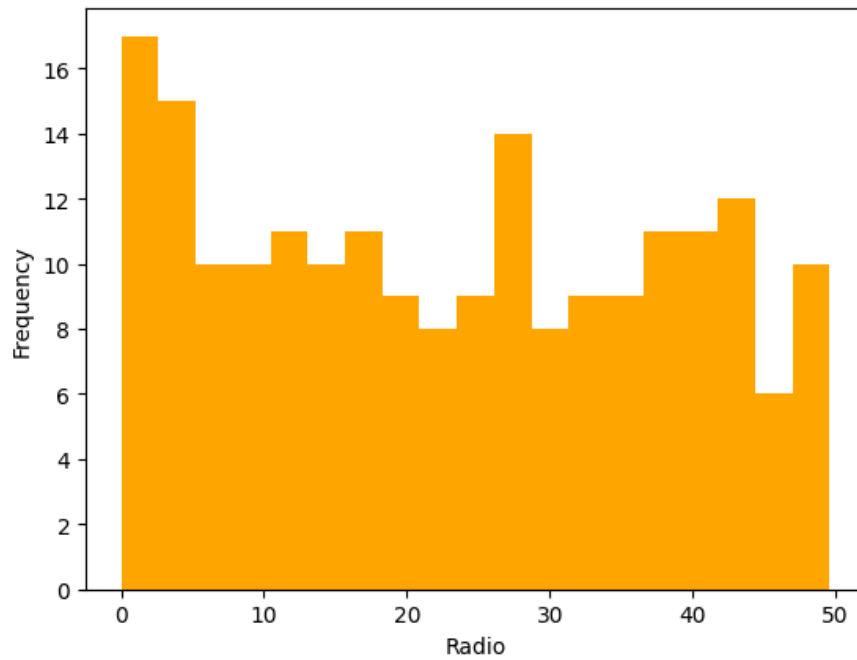
```
df['TV'].plot.hist(bins=10)
```

<Axes: ylabel='Frequency'>



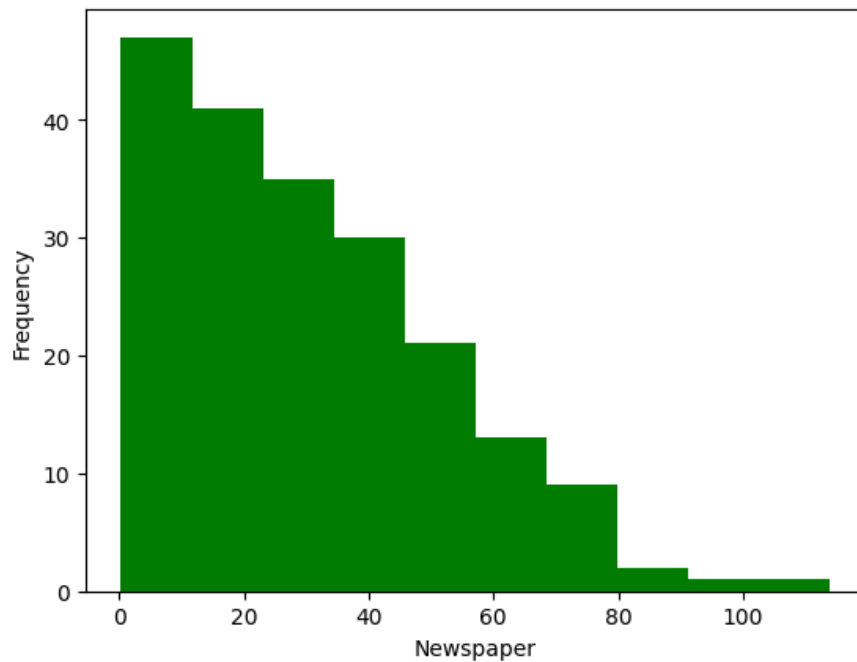
```
df["Radio"].plot.hist(bins=19,color="orange",xlabel="Radio")
```

```
<Axes: xlabel='Radio', ylabel='Frequency'>
```



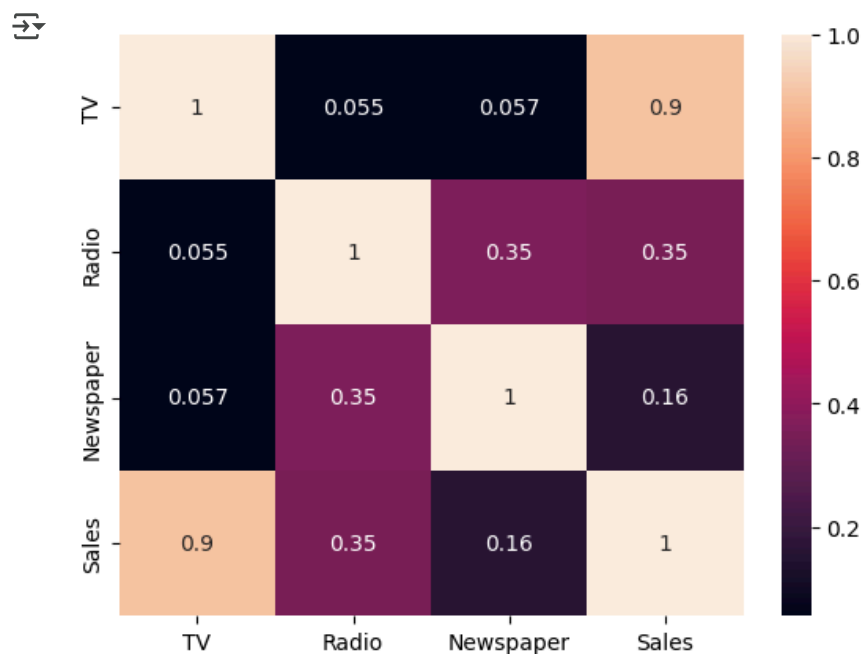
```
df["Newspaper"].plot.hist(bins=10,color="Green",xlabel="Newspaper")
```

```
<Axes: xlabel='Newspaper', ylabel='Frequency'>
```



The Majority Sales is the result of Now advertising cost in Newspaper

```
sns.heatmap(df.corr(),annot=True)
plt.show()
```



Sales is highly correlated with the TV

Lets train our model using linear regression as it is correlated with only

To generate independent variable represent X and Y represents the largest variable in a simple linear regression model

```
x=df['TV']
y=df['Sales']
```

```
from sklearn.model_selection import train_test_split
xTrain,xTest,yTrain,yTest=train_test_split(df[['TV']],df[['Sales']],test_size=0.3,random_state=0)
```

```
print(xTrain)
```

```

TV
131 265.2
96 197.6
181 218.5
19 147.3
153 171.3
.. ...
67 139.3
192 17.2
117 76.4
47 239.9
172 19.6

[140 rows x 1 columns]
```

```
print(xTest)
```

```

TV
107 90.4
98 289.7
177 170.2
182 56.2
5 8.7
```


```
61 201.5
125 87.2
180 156.6
154 187.8
80 76.4
7 120.2
33 265.6
130 0.7
37 74.7
74 213.4
183 287.6
145 140.3
45 175.1
159 131.7
60 53.5
123 123.1
179 165.6
185 205.0
122 224.0
44 25.1
16 67.8
55 198.9
150 280.7
111 241.7
22 13.2
189 18.7
129 59.6
4 180.8
83 68.4
106 25.0
134 36.9
66 31.5
26 142.9
113 209.6
168 215.4
63 102.7
8 8.6
75 16.9
118 125.7
143 104.6
71 109.8
124 229.5
184 253.8
97 184.9
149 44.7
24 62.3
30 292.9
160 172.5
40 202.5
56 7.3
```

```
print(yTrain)
```

```
⇒ Sales
131 17.7
96 16.7
181 17.2
19 14.6
153 16.0
.. ...
67 13.4
192 5.9
117 9.4
47 23.2
172 7.6
```

```
[140 rows x 1 columns]
```

```
print(yTest)
```



107	12.0
98	25.4
177	16.7
182	8.7
5	7.2
146	18.2
12	9.2
152	16.6
61	24.2
125	10.6
180	15.5
154	20.6
80	11.8
7	13.2
33	17.4
130	1.6
37	14.7
74	17.0
183	26.2
145	10.3
45	16.1
159	12.9
60	8.1
123	15.2
179	17.6
185	22.6
122	16.6
44	8.5
16	12.5
55	23.7
150	16.1
111	21.8
22	5.6
189	6.7
129	9.7
4	17.9
83	13.6
106	7.2
134	10.8
66	11.0
26	15.0
113	20.9
168	17.1
63	14.0
8	4.8
75	8.7
118	15.9
143	10.4
71	12.4
124	19.7
184	17.6
97	20.5
149	10.1
24	9.7
30	21.4
160	16.4
40	16.6
56	5.5

```
from sklearn.linear_model import LinearRegression
model=LinearRegression()
s=model.fit(xTrain,yTrain)
s
```



▼ LinearRegression ⓘ ?

```
LinearRegression()
```

```
res=model.predict(xTest)
print(res)
```

```
[ 12.09159447]
[ 22.99968079]
[ 16.45920756]
[ 10.21976029]
[  7.6199906 ]
[ 20.28497391]
[  8.4464437 ]
[ 17.95886418]
[ 21.44529217]
[ 11.91645209]
[ 15.71485245]
[ 17.42249065]
[ 11.32534656]
[ 13.72260788]
[ 21.68063975]
[  7.18213465]
[ 11.23230217]
[ 18.82362968]
[ 22.88474361]
[ 14.82272095]
[ 16.72739433]
[ 14.35202581]
[ 10.07198391]
[ 13.88133066]
[ 16.20744039]
[ 18.36388094]
[ 19.40378881]
[  8.51759529]
[ 10.85465142]
[ 18.03001578]
[ 22.50709285]
[ 20.3725451 ]
[  7.86628457]
[  8.16731053]
[ 10.40584907]
[ 17.03936669]
[ 10.88749061]
[  8.51212209]
[  9.16343282]
[  8.86788005]
[ 14.96502414]
[ 18.61564811]
[ 18.93309367]
[ 12.76479799]
[  7.6145174 ]
[  8.06879294]
[ 14.02363385]
[ 12.86878878]
[ 13.15339515]
[ 19.70481478]
[ 21.03480222]
[ 17.26376787]
[  9.59034237]
[ 10.55362545]
[ 23.17482317]
[ 16.58509115]
[ 18.22705095]
[  7.54336581]]
```

```
model.coef_
```

```
[ 0.05473199]]
```

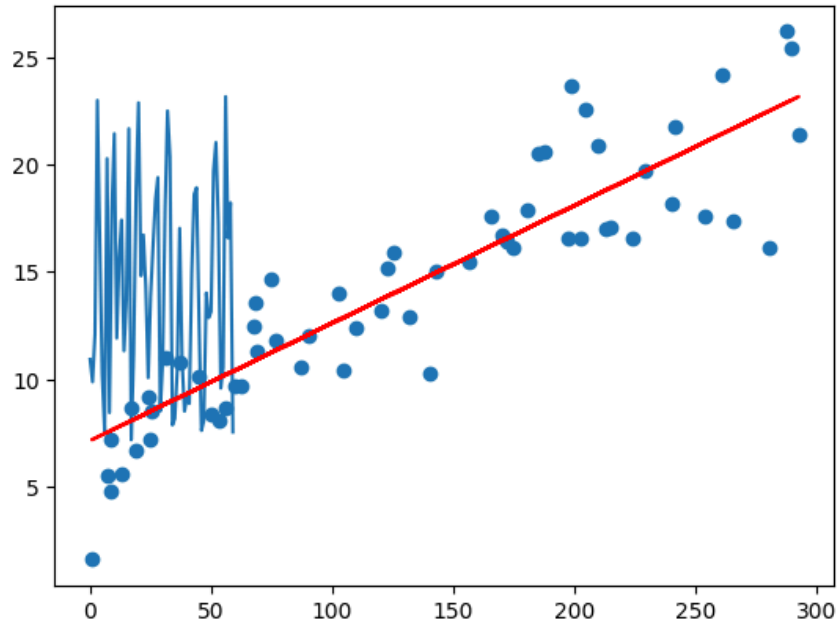
```
model.intercept_
```

```
[ 7.14382225]]
```

## visulization the linear plot

```
plt.plot(res)  
plt.scatter(xTest,yTest)  
plt.plot(xTest,7.143822+0.054731*xTest,'r')
```

[<matplotlib.lines.Line2D at 0x7814e76d1790>]



```
yTrain_pred=model.predict(xTrain)  
yTrain_pred
```





```
[15.49273351],
[11.25419497],
[ 9.22363801],
[16.27311878],
[ 8.61063968],
[13.73902748],
[21.53286336],
[19.97847475],
[20.26855431],
[22.79717242],
[19.31621762],
[ 9.305736 ],
[19.62271679],
[18.6813265 ],
[11.16115058],
[11.97665729],
[18.04096217],
[13.20265394],
[21.75179134],
[20.13719753],
[18.07927457],
[17.42796385],
[14.76798896],
[ 8.08521254],
[11.32534656],
[20.27402751],
[ 8.2165693311]
```

```
res=(yTrain-yTrain_pred)
print(res)
```

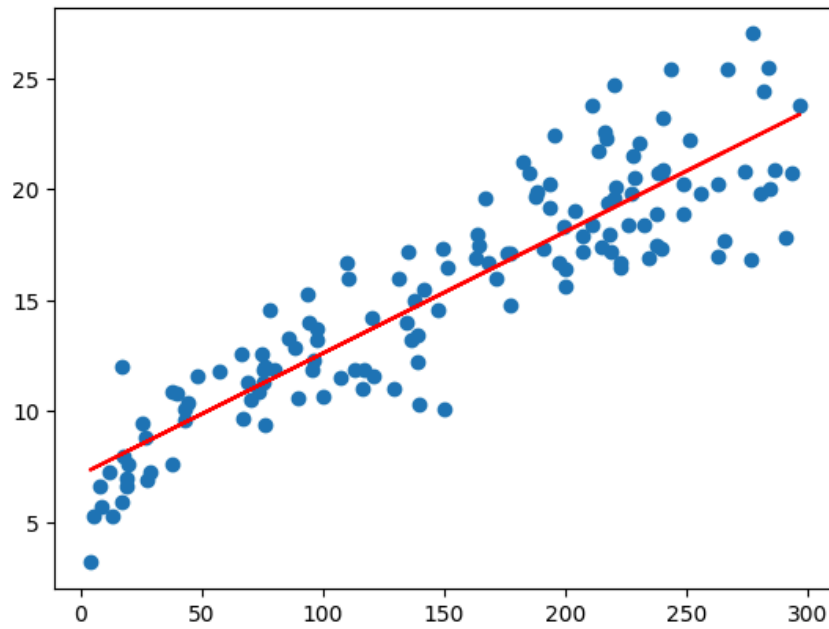
```
→ Sales
131 -3.958747
96 -1.258864
181 -1.902763
19 -0.605845
153 -0.519413
.. ...
67 -1.367989
192 -2.185213
117 -1.925347
47 2.925972
172 -0.616569
```

```
[140 rows x 1 columns]
```

## ✓ visualization the regression line


```
plt.scatter(xTrain,yTrain)
plt.plot(xTrain,7.143822+0.054731*xTrain,'r')
```

[<matplotlib.lines.Line2D at 0x7814e773a2d0>]




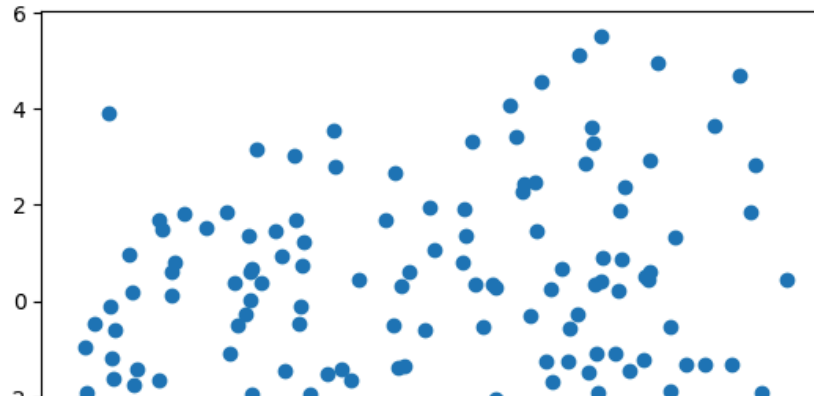
✓ plotting the histogram using the residual values

```
fig=plt.figure()
sns.distplot(res,bins=15)
plt.xlabel('res')
plt.show()
```

 <ipython-input-30-64ebfee4856b>:2: UserWarning:

```
plt.scatter(xTrain,res)  
plt.show()
```

 similar flexibility) or `histplot` (an axes-level function for histograms).



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