

In [ ]:

In [ ]:

# Machine Learning

## Machine Learning Procedure

Reading & Studying the Dataset

Data Cleaning & Analysis

Feature Engineering

Data Visualization or EDA (Exploratory Data Analysis)

Encoding

ip/op Creation

Train Test Split

Scaling

ML Algorithm

Prediction

Accuracy

## Advertising Dataset

```
In [1]: #Importing Packages  
import numpy as np  
import pandas as pd  
import matplotlib.pyplot as plt  
import seaborn as sns
```

```
In [2]: #Reading and Studying the Dataset  
adv = pd.read_csv(r"C:\Users\VICTUS\Downloads\advertising.csv")  
adv
```

Out[2]:

	TV	Radio	Newspaper	Sales
<b>0</b>	230.1	37.8	69.2	22.1
<b>1</b>	44.5	39.3	45.1	10.4
<b>2</b>	17.2	45.9	69.3	12.0
<b>3</b>	151.5	41.3	58.5	16.5
<b>4</b>	180.8	10.8	58.4	17.9
...	...	...	...	...
<b>195</b>	38.2	3.7	13.8	7.6
<b>196</b>	94.2	4.9	8.1	14.0
<b>197</b>	177.0	9.3	6.4	14.8
<b>198</b>	283.6	42.0	66.2	25.5
<b>199</b>	232.1	8.6	8.7	18.4

200 rows × 4 columns

In [3]: *#Data Cleaning and Analysis*  
`adv.isnull().sum()`

Out[3]: TV 0  
Radio 0  
Newspaper 0  
Sales 0  
dtype: int64

In [4]: `adv.dtypes`

Out[4]: TV float64  
Radio float64  
Newspaper float64  
Sales float64  
dtype: object

In [5]: `for i in adv.columns:`  
`print(i,':','\n',adv[i].unique(),'\n')`

TV :

[230.1 44.5 17.2 151.5 180.8 8.7 57.5 120.2 8.6 199.8 66.1 214.7  
 23.8 97.5 204.1 195.4 67.8 281.4 69.2 147.3 218.4 237.4 13.2 228.3  
 62.3 262.9 142.9 240.1 248.8 70.6 292.9 112.9 97.2 265.6 95.7 290.7  
 266.9 74.7 43.1 228. 202.5 177. 293.6 206.9 25.1 175.1 89.7 239.9  
 227.2 66.9 100.4 216.4 182.6 262.7 198.9 7.3 136.2 210.8 210.7 53.5  
 261.3 239.3 102.7 131.1 69. 31.5 139.3 216.8 199.1 109.8 26.8 129.4  
 213.4 16.9 27.5 120.5 5.4 116. 76.4 239.8 75.3 68.4 213.5 193.2  
 76.3 110.7 88.3 134.3 28.6 217.7 250.9 107.4 163.3 197.6 184.9 289.7  
 135.2 222.4 296.4 280.2 187.9 238.2 137.9 25. 90.4 13.1 255.4 225.8  
 241.7 175.7 209.6 78.2 75.1 139.2 125.7 19.4 141.3 18.8 224. 123.1  
 229.5 87.2 7.8 80.2 220.3 59.6 0.7 265.2 8.4 219.8 36.9 48.3  
 25.6 273.7 43. 73.4 193.7 220.5 104.6 96.2 140.3 243.2 38. 44.7  
 280.7 121. 171.3 187.8 4.1 93.9 149.8 11.7 131.7 172.5 85.7 188.4  
 163.5 117.2 234.5 17.9 206.8 215.4 284.3 50. 164.5 19.6 168.4 276.9  
 248.4 170.2 276.7 165.6 156.6 218.5 56.2 287.6 253.8 205. 139.5 191.1  
 286. 18.7 39.5 75.5 166.8 149.7 38.2 94.2 283.6 232.1]

Radio :

[37.8 39.3 45.9 41.3 10.8 48.9 32.8 19.6 2.1 2.6 5.8 24. 35.1 7.6  
 32.9 47.7 36.6 39.6 20.5 23.9 27.7 5.1 15.9 16.9 12.6 3.5 29.3 16.7  
 27.1 16. 28.3 17.4 1.5 20. 1.4 4.1 43.8 49.4 26.7 37.7 22.3 33.4  
 8.4 25.7 22.5 9.9 41.5 15.8 11.7 3.1 9.6 41.7 46.2 28.8 28.1 19.2  
 49.6 29.5 2. 42.7 15.5 29.6 42.8 9.3 24.6 14.5 27.5 43.9 30.6 14.3  
 33. 5.7 43.7 1.6 28.5 29.9 7.7 20.3 44.5 43. 18.4 40.6 25.5 47.8  
 4.9 33.5 36.5 14. 31.6 21. 42.3 4.3 36.3 10.1 17.2 34.3 46.4 11.  
 0.3 0.4 26.9 8.2 38. 15.4 20.6 46.8 35. 0.8 36.9 26.8 21.7 2.4  
 34.6 32.3 11.8 38.9 0. 49. 12. 2.9 27.2 38.6 47. 39. 28.9 25.9  
 17. 35.4 33.2 14.8 1.9 7.3 40.3 25.8 13.9 23.3 39.7 21.1 11.6 43.5  
 1.3 18.1 35.8 36.8 14.7 3.4 37.6 5.2 23.6 10.6 20.9 20.1 7.1 30.2  
 7.8 2.3 10. 5.4 21.3 45.1 28.7 12.1 41.1 42. 35.6 3.7 8.6]

Newspaper :

[ 69.2 45.1 69.3 58.5 58.4 75. 23.5 11.6 1. 21.2 24.2 4.  
 65.9 7.2 46. 52.9 114. 55.8 18.3 19.1 53.4 49.6 26.2 19.5  
 12.6 22.9 40.8 43.2 38.6 30. 0.3 7.4 8.5 5. 45.7 35.1  
 32. 31.6 38.7 1.8 26.4 43.3 31.5 35.7 18.5 49.9 36.8 34.6  
 3.6 39.6 58.7 15.9 60. 41.4 16.6 37.7 9.3 21.4 54.7 27.3  
 8.4 28.9 0.9 2.2 10.2 11. 27.2 31.7 19.3 31.3 13.1 89.4  
 20.7 14.2 9.4 23.1 22.3 36.9 32.5 35.6 33.8 65.7 16. 63.2  
 73.4 51.4 33. 59. 72.3 10.9 5.9 22. 51.2 45.9 49.8 100.9  
 17.9 5.3 29.7 23.2 25.6 5.5 56.5 2.4 10.7 34.5 52.7 14.8  
 79.2 46.2 50.4 15.6 12.4 74.2 25.9 50.6 9.2 3.2 43.1 8.7  
 43. 2.1 65.6 59.7 20.5 1.7 12.9 75.6 37.9 34.4 38.9 9.  
 44.3 11.9 20.6 37. 48.7 9.5 5.7 50.5 24.3 45.2 30.7 49.3  
 5.4 84.8 21.6 19.4 57.6 6.4 18.4 47.4 17. 12.8 41.8 20.3  
 35.2 23.7 17.6 8.3 27.4 71.8 19.6 26.6 18.2 3.7 23.4 5.8  
 6. 13.8 8.1 66.2]

Sales :

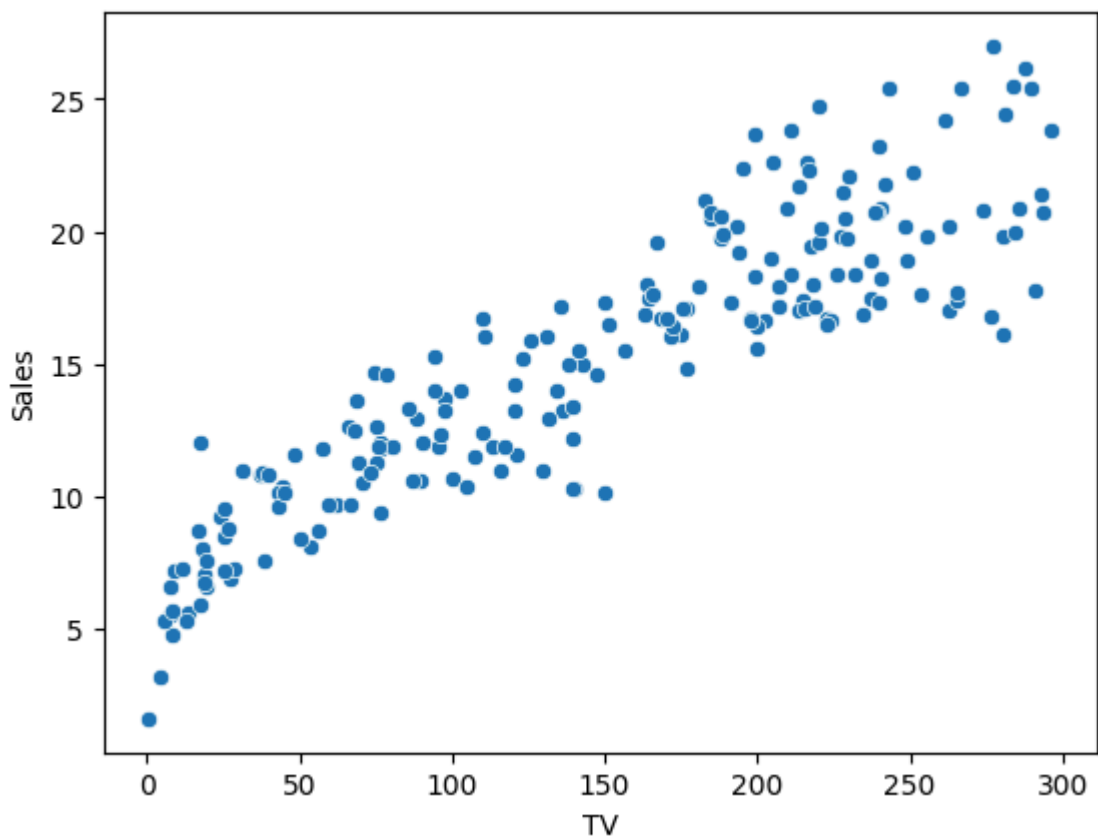
[22.1 10.4 12. 16.5 17.9 7.2 11.8 13.2 4.8 15.6 12.6 17.4 9.2 13.7  
 19. 22.4 12.5 24.4 11.3 14.6 18. 17.5 5.6 20.5 9.7 17. 15. 20.9  
 18.9 10.5 21.4 11.9 17.8 25.4 14.7 10.1 21.5 16.6 17.1 20.7 8.5 16.1  
 10.6 23.2 19.8 16.4 10.7 22.6 21.2 20.2 23.7 5.5 23.8 18.4 8.1 24.2  
 14. 16. 11. 13.4 22.3 18.3 12.4 8.8 8.7 6.9 14.2 5.3 17.3 13.6  
 21.7 12.9 16.7 7.3 19.4 22.2 11.5 16.9 17.2 19.7 21.8 12.2 9.4 15.9  
 6.6 15.5 7. 15.2 24.7 1.6 17.7 5.7 19.6 10.8 11.6 9.5 20.8 9.6  
 10.9 19.2 20.1 12.3 10.3 18.2 20.6 3.2 15.3 13.3 19.9 8. 20. 8.4  
 7.6 27. 16.8 17.6 26.2 6.7 5.9 14.8 25.5]

```
In [6]: adv.describe()
```

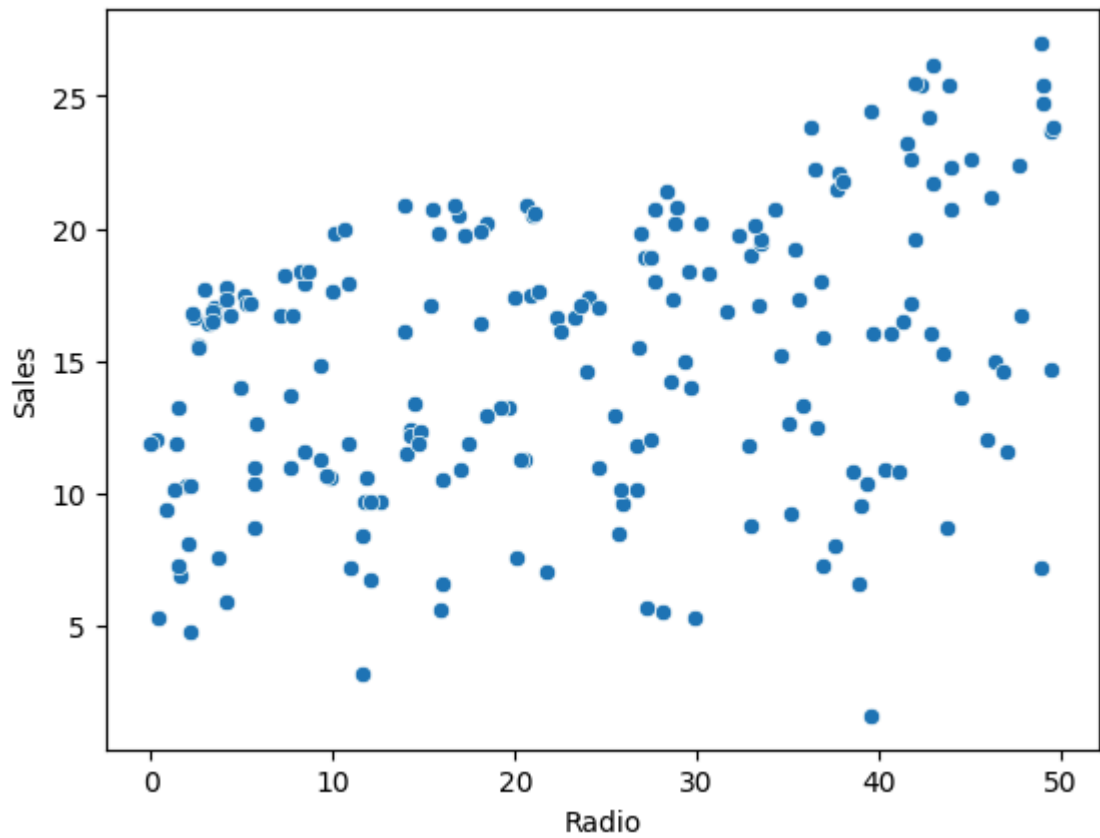
```
Out[6]:
```

	TV	Radio	Newspaper	Sales
count	200.000000	200.000000	200.000000	200.000000
mean	147.042500	23.264000	30.554000	15.130500
std	85.854236	14.846809	21.778621	5.283892
min	0.700000	0.000000	0.300000	1.600000
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

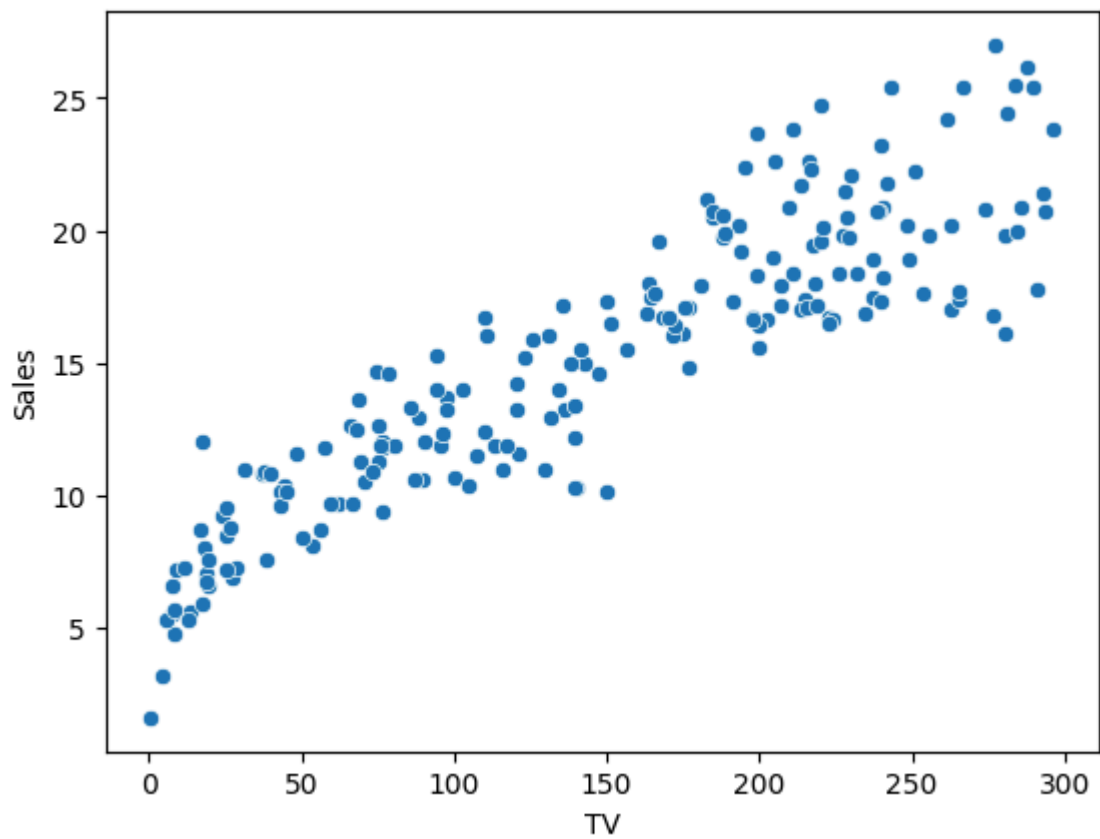
```
In [7]: #Data Visualization or EDA
sns.scatterplot(x=adv.TV,y=adv.Sales)
plt.show()
```

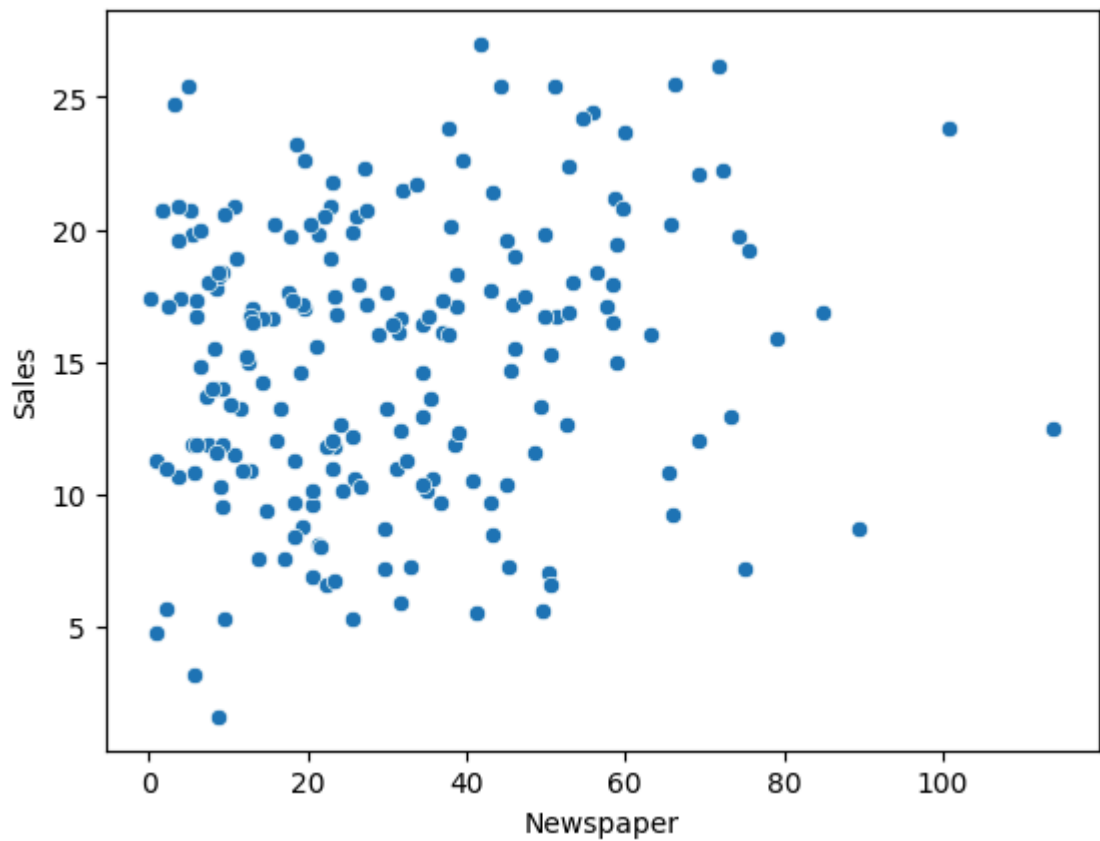
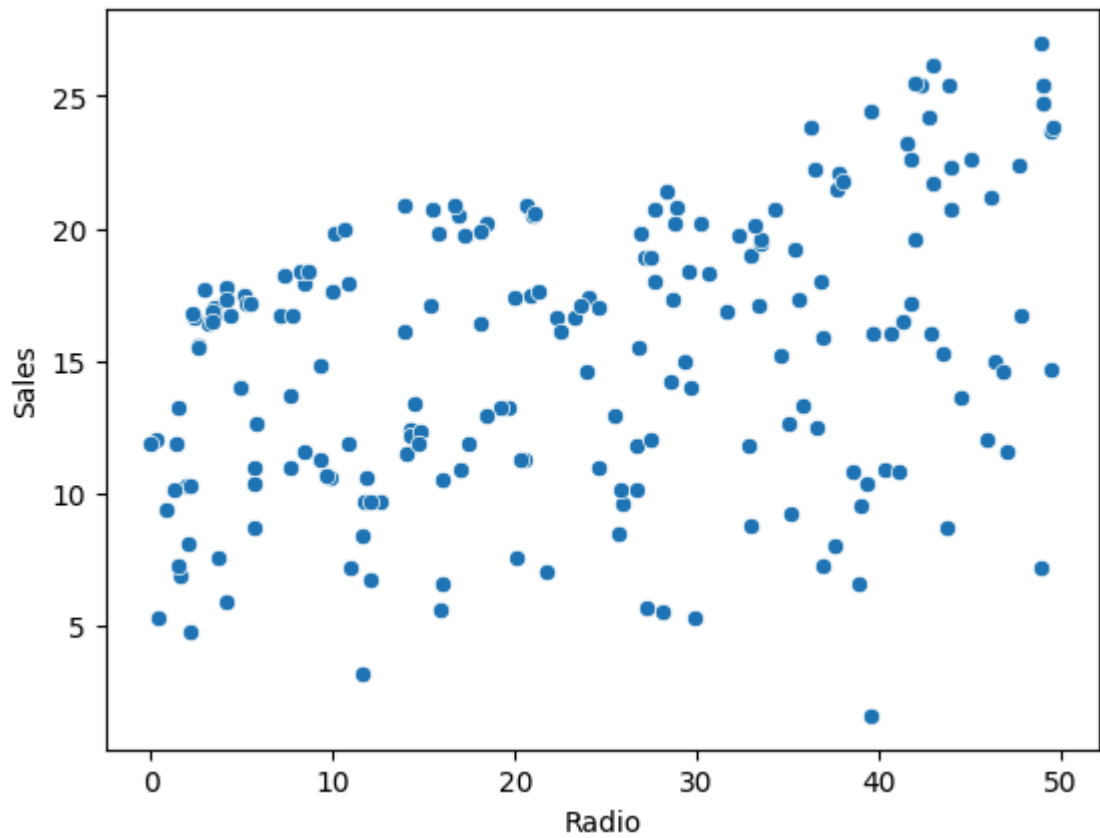


```
In [8]: sns.scatterplot(x=adv.Radio,y=adv.Sales)
plt.show()
```

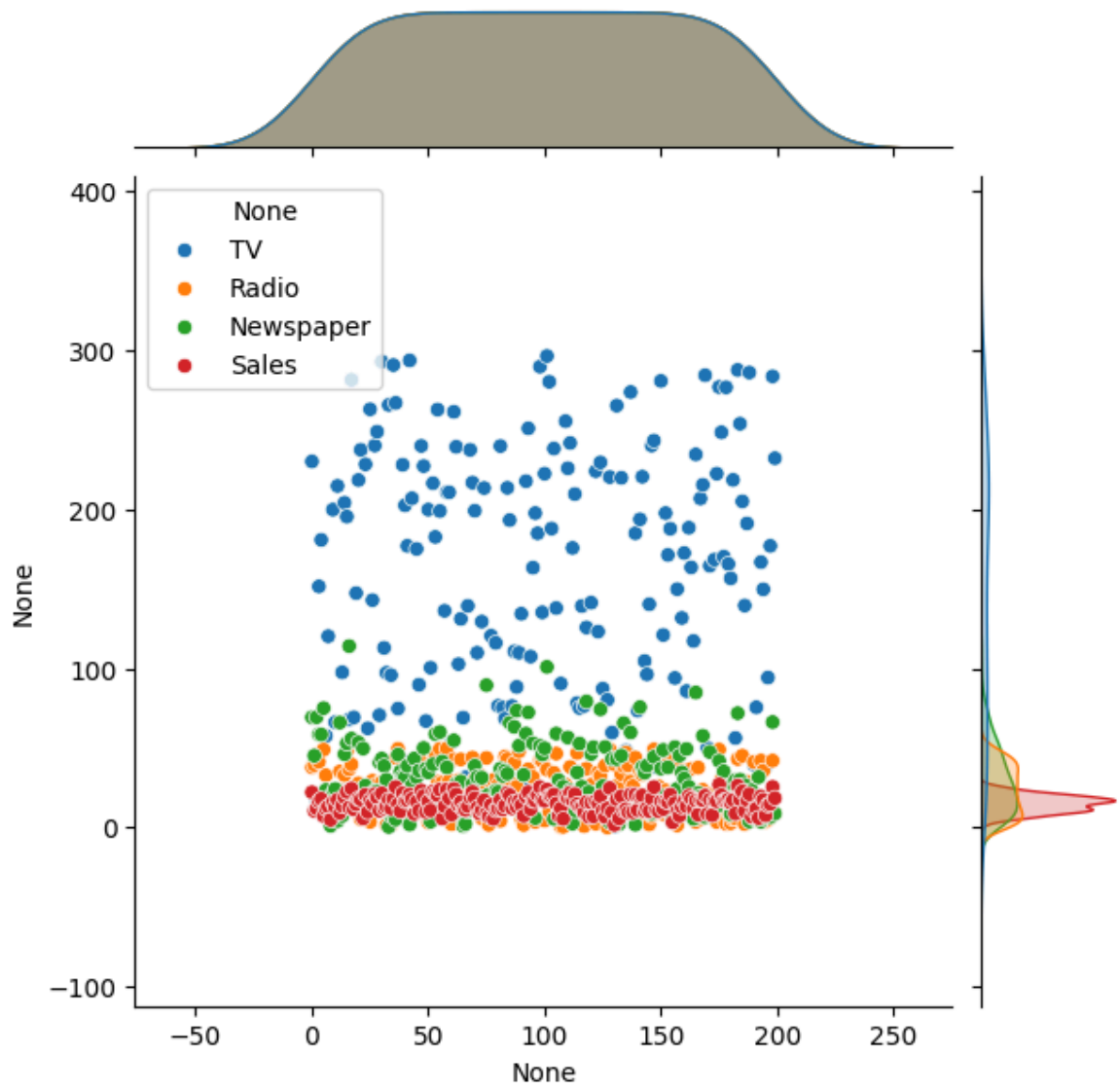


```
In [9]: for i in ['TV', 'Radio', 'Newspaper']:  
        sns.scatterplot(x=adv[i], y=adv.Sales)  
        plt.show()
```

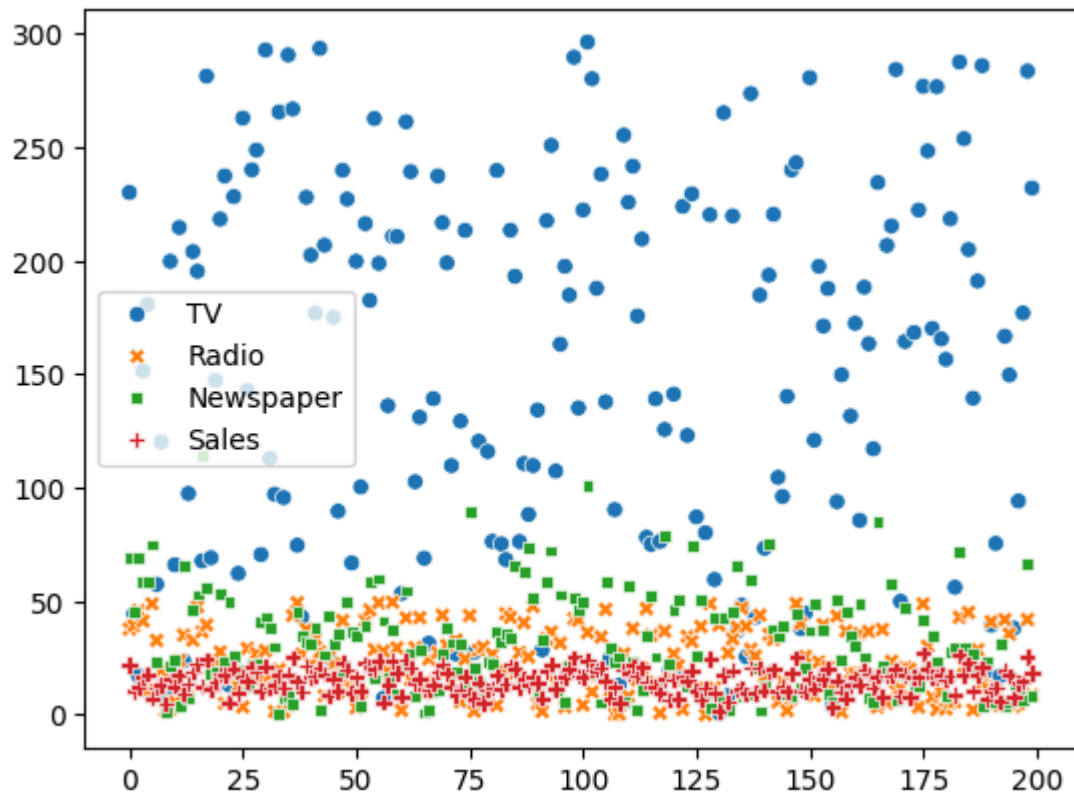




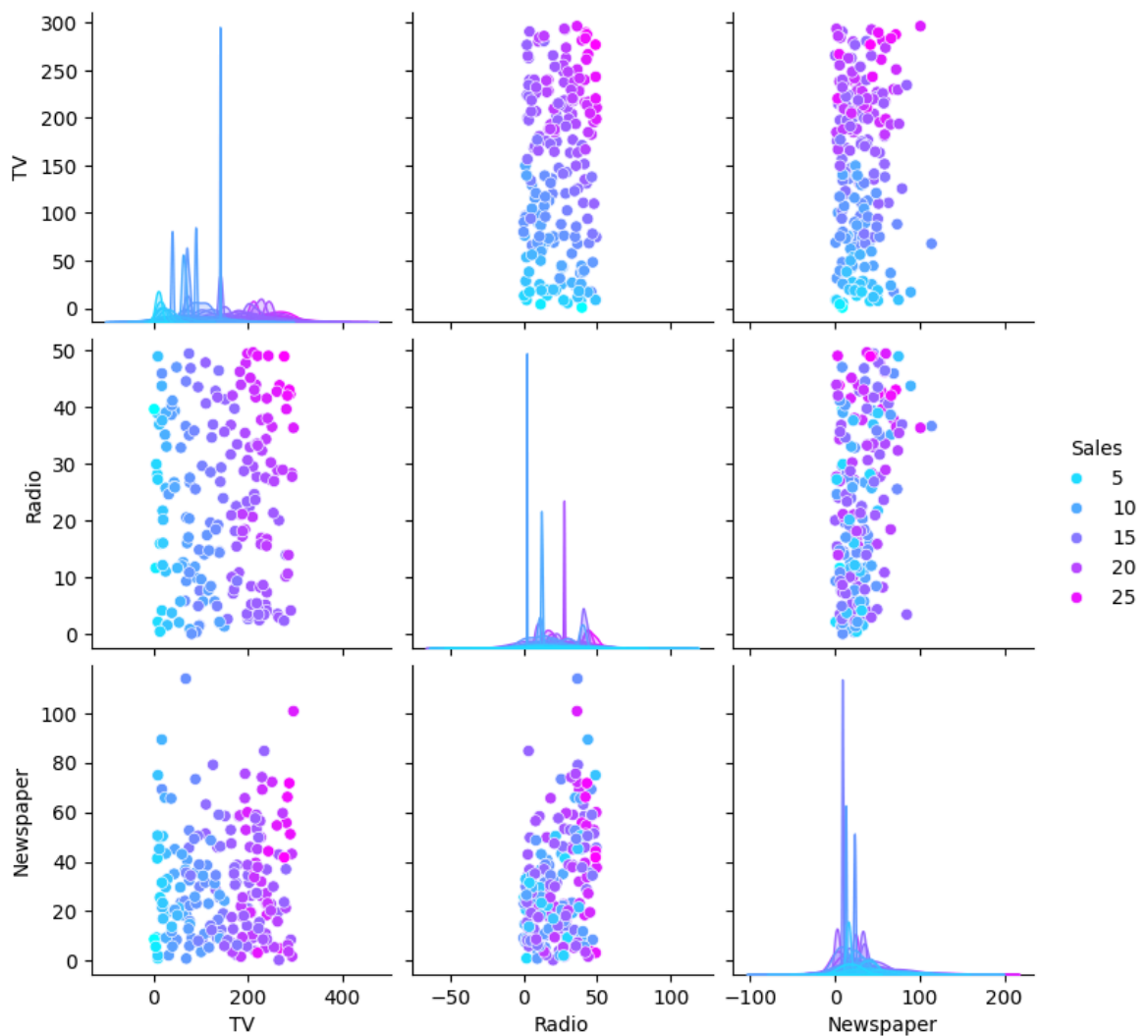
```
In [10]: sns.jointplot(data=adv)
plt.show()
```



```
In [11]: sns.scatterplot(data=adv)
plt.show()
```



```
In [12]: sns.pairplot(adv,hue='Sales',palette='cool')
plt.show()
```



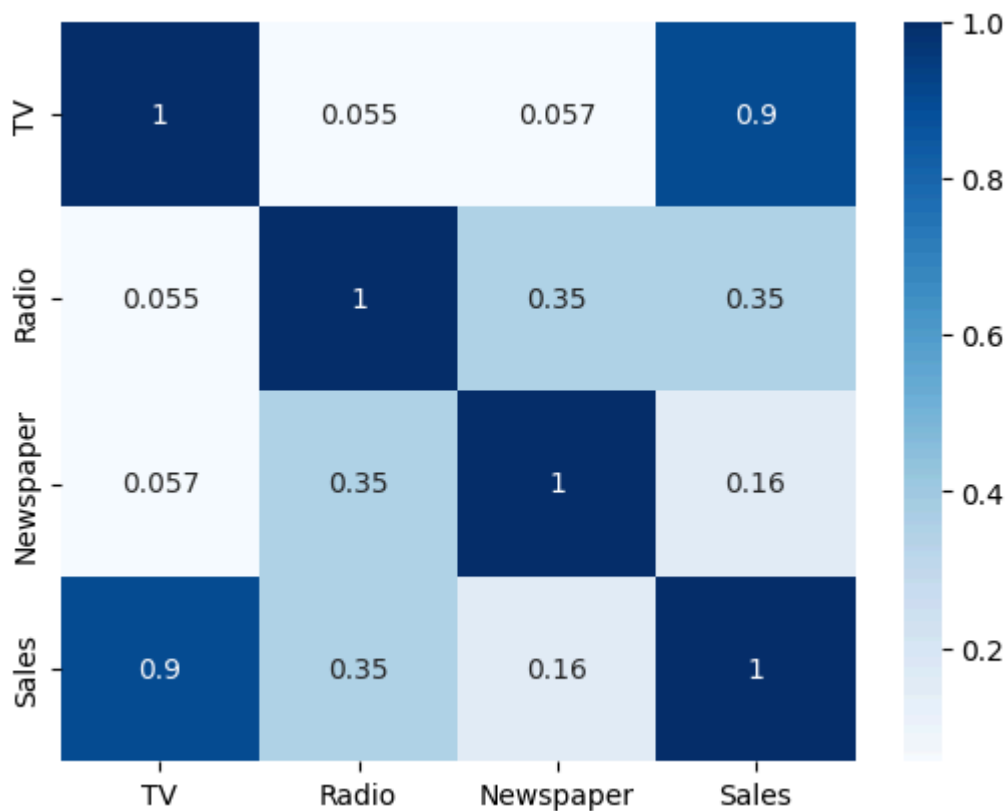


```
In [13]: c = adv.corr()
c
```

```
Out[13]:
```

	TV	Radio	Newspaper	Sales
TV	1.000000	0.054809	0.056648	0.901208
Radio	0.054809	1.000000	0.354104	0.349631
Newspaper	0.056648	0.354104	1.000000	0.157960
Sales	0.901208	0.349631	0.157960	1.000000

```
In [14]: sns.heatmap(c,annot=True,cmap='Blues')
plt.show()
```



```
In [15]: #Encoding
```

```
In [16]: #ip/op Creation
ip = adv.drop('Sales',axis=1)
op = adv.Sales
```

```
In [17]: ip.head()
```

Out[17]:

	TV	Radio	Newspaper
<b>0</b>	230.1	37.8	69.2
<b>1</b>	44.5	39.3	45.1
<b>2</b>	17.2	45.9	69.3
<b>3</b>	151.5	41.3	58.5
<b>4</b>	180.8	10.8	58.4

In [18]: `op.head()`

Out[18]:

0	22.1
1	10.4
2	12.0
3	16.5
4	17.9

Name: Sales, dtype: float64

In [19]: `#!pip install scikit-learn`

In [22]: `#Train Test Split`  
`from sklearn.model_selection import train_test_split`  
`x_train, x_test, y_train, y_test = train_test_split(ip,op,test_size=0.2,random_s`

In [23]: `x_train`

Out[23]:

	TV	Radio	Newspaper
<b>156</b>	93.9	43.5	50.5
<b>115</b>	75.1	35.0	52.7
<b>155</b>	4.1	11.6	5.7
<b>15</b>	195.4	47.7	52.9
<b>61</b>	261.3	42.7	54.7
...	...	...	...
<b>0</b>	230.1	37.8	69.2
<b>184</b>	253.8	21.3	30.0
<b>131</b>	265.2	2.9	43.0
<b>152</b>	197.6	23.3	14.2
<b>106</b>	25.0	11.0	29.7

160 rows × 3 columns

In [24]: `print(x_train.shape)`  
`print(y_train.shape)`  
`print(x_test.shape)`  
`print(y_test.shape)`

```
(160, 3)
(160,)
(40, 3)
(40,)
```

```
In [25]: #Scaling
#Standard Scaler Transform
from sklearn.preprocessing import StandardScaler
sc = StandardScaler()

x_train = sc.fit_transform(x_train)
x_test = sc.transform(x_test)
```

## Linear Regression

$$y = mx + c$$

Slope of the line  
 ↓  
 **$y$**  → Dependent Variable (op)  
 **$x$**  → Independent Variable (ip)  
 **$m$**  → Slope of the line  
 **$+$**   
 **$c$**  → y – intercept  
 (the point at which the line intersects the y – axis)

```
In [26]: #ML Algorithm
#Linear Regression
from sklearn.linear_model import LinearRegression
lr = LinearRegression()
lr.fit(x_train,y_train)
```

```
Out[26]: ▼ LinearRegression ⓘ ?
LinearRegression()
```

```
In [27]: #Prediction
pred = lr.predict(x_test)
pred
```

```
Out[27]: array([17.99747206, 11.02863798, 19.14207387, 15.16467067,  8.56496638,
                10.89550601, 24.92202604, 10.61757023, 18.6998404 , 17.29854871,
                14.66680343, 13.03094656, 19.28219899, 10.98418026, 13.7677926 ,
                14.4988402 , 16.92235556, 17.30704485, 17.78993353, 21.29009711,
                19.23977212, 10.98691684,  9.77272715, 11.18064352,  8.34136167,
                13.13093624, 21.59478295, 17.04191764, 24.83365506, 11.62133637,
                16.22765284, 21.95021111,  9.21285178,  9.94421377,  9.82032335,
                10.22691037, 15.74967394,  9.51795157, 13.67254607, 12.42186513])
```

```
In [28]: y_test
```

```
Out[28]: 40      16.6
         51      10.7
         139     20.7
         197     14.8
         170      8.4
         82      11.3
         183     26.2
         46      10.6
         70      18.3
         100     16.7
         179     17.6
         83      13.6
         25      17.0
         190     10.8
         159     12.9
         173     16.7
         95      16.9
          3      16.5
         41      17.1
         58      23.8
         14      19.0
         143     10.4
         12       9.2
          6      11.8
         182      8.7
         161     13.3
         128     24.7
         122     16.6
         101     23.8
          86     12.0
          64     16.0
          47     23.2
         158      7.3
          34     11.9
          38     10.1
         196     14.0
           4     17.9
          72      8.8
          67     13.4
         145     10.3
         Name: Sales, dtype: float64
```

```
In [29]: #Accuracy
         from sklearn.metrics import mean_squared_error, r2_score

         mse = mean_squared_error(y_test, pred)
         r2 = r2_score(y_test, pred)
```

```
print("MSE : ",mse)  
print("R2 Score : ",r2)
```

MSE : 1.9901879673862246

R2 Score : 0.9148706512256517

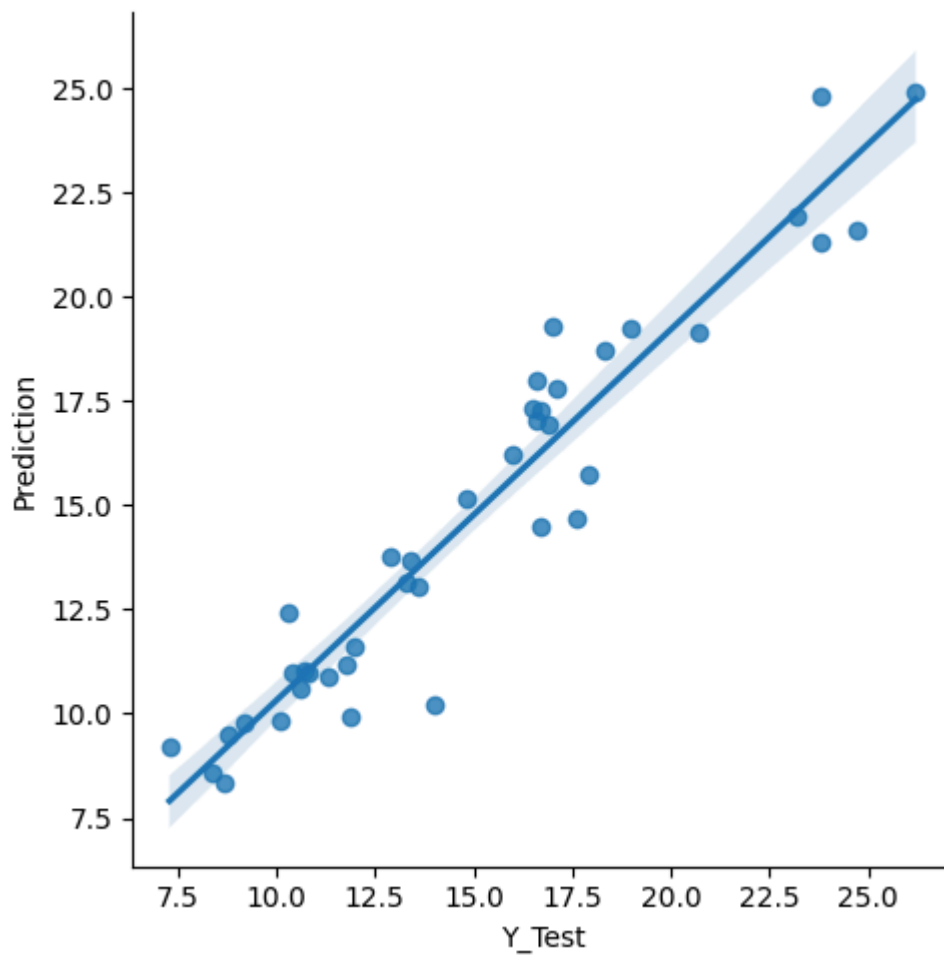
```
In [33]: df = pd.DataFrame({'Y_Test':list(y_test),'Prediction':pred})  
df
```

Out[33]:

	Y_Test	Prediction
<b>0</b>	16.6	17.997472
<b>1</b>	10.7	11.028638
<b>2</b>	20.7	19.142074
<b>3</b>	14.8	15.164671
<b>4</b>	8.4	8.564966
<b>5</b>	11.3	10.895506
<b>6</b>	26.2	24.922026
<b>7</b>	10.6	10.617570
<b>8</b>	18.3	18.699840
<b>9</b>	16.7	17.298549
<b>10</b>	17.6	14.666803
<b>11</b>	13.6	13.030947
<b>12</b>	17.0	19.282199
<b>13</b>	10.8	10.984180
<b>14</b>	12.9	13.767793
<b>15</b>	16.7	14.498840
<b>16</b>	16.9	16.922356
<b>17</b>	16.5	17.307045
<b>18</b>	17.1	17.789934
<b>19</b>	23.8	21.290097
<b>20</b>	19.0	19.239772
<b>21</b>	10.4	10.986917
<b>22</b>	9.2	9.772727
<b>23</b>	11.8	11.180644
<b>24</b>	8.7	8.341362
<b>25</b>	13.3	13.130936
<b>26</b>	24.7	21.594783
<b>27</b>	16.6	17.041918
<b>28</b>	23.8	24.833655
<b>29</b>	12.0	11.621336
<b>30</b>	16.0	16.227653
<b>31</b>	23.2	21.950211
<b>32</b>	7.3	9.212852

	Y_Test	Prediction
33	11.9	9.944214
34	10.1	9.820323
35	14.0	10.226910
36	17.9	15.749674
37	8.8	9.517952
38	13.4	13.672546
39	10.3	12.421865

```
In [34]: #lmplo - Linear Model Plot
sns.lmplot(x='Y_Test',y='Prediction',data=df)
plt.show()
```



In [ ]:

## Classification Models

### Maintenance Dataset

```
In [81]: #Importing Packages
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

```
In [82]: #Reading and Studying the Dataset
main = pd.read_csv(r"C:\Users\VICTUS\Downloads\maintenance_data.csv")
main
```

```
Out[82]:
```

	lifetime	broken	pressureInd	moistureInd	temperatureInd	team	provider
0	56	0	92.178854	104.230204	96.517159	TeamA	Provider4
1	81	1	72.075938	103.065701	87.271062	TeamC	Provider4
2	60	0	96.272254	77.801376	112.196170	TeamA	Provider1
3	86	1	94.406461	108.493608	72.025374	TeamC	Provider2
4	34	0	97.752899	99.413492	103.756271	TeamB	Provider1
...	...	...	...	...	...	...	...
995	88	1	88.589759	112.167556	99.861456	TeamB	Provider4
996	88	1	116.727075	110.871332	95.075631	TeamA	Provider4
997	22	0	104.026778	88.212873	83.221220	TeamB	Provider1
998	78	0	104.911649	104.257296	83.421491	TeamA	Provider4
999	63	0	116.901354	99.998694	47.641493	TeamB	Provider1

1000 rows × 7 columns

```
In [83]: #Data Cleaning
main.isnull().sum()
```

```
Out[83]: lifetime      0
broken      0
pressureInd  0
moistureInd  0
temperatureInd  0
team        0
provider    0
dtype: int64
```

```
In [84]: main.dtypes
```

```
Out[84]: lifetime      int64
broken      int64
pressureInd  float64
moistureInd  float64
temperatureInd  float64
team        object
provider    object
dtype: object
```



```
In [85]: for i in main.columns:  
         print(i,':','\n',main[i].unique(),'\n')
```

lifetime :

```
[56 81 60 86 34 30 68 65 23 38 29 82 80 48 92 88 74 61 35 26 63 79 53 73
13 36 31 25 58 19 84 12 15 43 1 20 16 3 18 7 47 39 57 4 24 28 49 76
52 8 40 46 5 41 93 77 62 85 55 33 17 45 9 72 50 42 44 54 64 27 22 59
66 83 14 51 71 21 78 6 69 89 2 67 87 11 10 32 37 90]
```

broken :

```
[0 1]
```

pressureInd :

```
[ 92.17885406 72.07593772 96.27225443 94.40646126 97.75289859
87.67880097 94.61417404 96.48330289 105.486158 99.17823531
97.81784409 67.81225145 86.36611059 76.14465414 103.1072633
88.41407945 84.35504868 79.66925455 86.22910861 84.17942039
100.0059233 115.6075596 97.69718903 101.4156229 118.9786971
102.1127749 129.1243378 109.0330362 107.2980695 127.2639544
138.1911205 72.55408417 95.21464918 91.24713174 81.55537698
69.13459519 132.8574784 54.73533145 84.89864628 94.15228726
84.34406647 95.75352436 123.9284399 101.5053508 120.9455502
83.06857115 113.3491219 75.64669299 113.4358321 81.77978353
68.85065142 150.665421 72.42312878 97.13286233 113.965257
57.46321326 144.3445174 80.69223153 93.91332598 73.32115289
83.95831157 56.17177135 105.9000372 115.698401 119.7459957
105.8937636 87.39284133 82.88180029 122.287991 85.07220883
119.204381 107.7859931 90.67123432 66.63788383 107.8373588
113.2726053 85.33137798 82.52821911 90.63682861 112.60533
78.66593081 70.10171881 99.12691885 129.8155876 129.3586686
86.43068027 127.6432644 95.24539906 83.49423659 96.21961865
79.1113931 106.8671887 111.2558114 96.12034618 86.10751429
108.4133588 100.2518501 117.9043996 121.6736792 60.23306632
94.23091985 74.48747426 92.453609 112.047415 125.0326918
96.00878453 109.6711377 154.9245853 105.7582145 93.35727589
99.96158153 95.92226826 111.7239334 97.9282765 128.7277348
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96.95590895	101.9202298	107.9943727	106.0739247	103.5603805
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93.2324069	82.61246511	102.4336216	100.2276629	109.0407797
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89.02241135	123.2676935	86.16872259	105.9859448	101.0983454
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88.16059474	111.7631163	99.70414132	82.84151939	90.80665259
99.04378211	97.82876698	100.5938109	94.59619158	108.2822108
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107.2141766	91.62927168	122.3492056	103.7544081	102.8531259
104.7427573	92.25017384	93.92932564	104.9092518	100.5511169
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86.51678718	88.07971633	94.06325121	97.22459227	105.454307
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97.10285708	106.5032417	88.1673482	102.8413903	103.721079
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80.05832135	87.98386127	111.4629184	92.79820374	109.7438203
129.9878939	76.08557046	100.6230519	75.19061894	110.2672376
77.21030463	93.00411663	92.92996339	77.07077762	93.34575008
124.736851	76.55626608	136.030098	128.7880801	75.9997621
84.81261253	108.8943008	107.4600703	84.04405964	73.29702688
127.4278908	124.484758	87.67851893	128.812916	112.167476
135.1693172	101.8822342	103.595199	120.4686983	73.96276076
103.067457	126.8485419	80.72306687	85.16860071	95.60684471
118.6429881	88.76889673	106.5479672	121.0385472	90.37342086
86.140575	121.6596974	104.2853153	90.687831	111.6442027
107.1263062	132.8960253	112.0732419	108.5662	80.10887306
91.46874009	98.27463019	71.80056052	123.3052926	119.9772055
117.8775732	83.42249496	137.4741755	105.3091613	125.6032844
92.37923722	105.1152794	119.8540501	102.4971134	42.27959777
105.8270944	99.98644583	141.8914187	108.3030495	81.18539743
91.39246184	102.8762405	96.76267759	92.79302585	100.4592065
104.753569	114.6001775	93.4643859	120.353069	111.5351793
99.81819568	99.63780661	67.25648332	90.11035352	75.30231632
104.1224911	90.7504789	112.6088848	127.0598802	119.861342
115.0832449	80.75742109	115.3353489	92.80935236	98.51566602
113.5163748	107.6967466	93.35842227	102.7250535	54.85463932
79.45524105	109.7863541	116.1649707	111.15933	88.37180867
97.35961631	83.03736305	132.4949068	102.7298925	91.10385148
95.52186729	134.0993917	100.0956146	111.9665528	77.19401385
89.47804982	82.35935334	82.37295228	60.61123187	85.28897211
99.32586687	110.9381936	121.4210197	105.3645153	110.1829006
127.5978295	84.35551244	57.34534891	65.21580585	77.59314255
96.76811894	87.05655228	107.4768874	76.54239502	92.02624326
106.5518926	109.8760155	123.0878047	117.8925563	114.1714076
73.99386306	93.02570296	104.9696808	111.5631001	112.4952072
101.92534	121.3259062	91.72599463	99.08231665	76.29286351
101.5804487	73.71605989	91.67263447	81.90237741	114.5670649
95.9716024	102.6007936	138.311966	81.39597051	104.7856021
86.63390737	130.5570767	65.18346688	114.9243374	119.5940377
63.18449738	70.92465014	110.7122015	93.91395318	87.24826931
82.13596405	106.7305167	128.3033183	80.55511582	46.85704695
91.61400596	64.88013098	96.95931027	128.3332476	112.4178876
80.99028054	97.91723838	97.78208935	120.3658475	93.40801931
149.924611	80.76109253	88.43730614	96.3568589	73.31904291
114.8743613	86.83447267	108.4264351	84.6399811	84.74365498
109.8976734	99.76389665	124.5358699	119.151983	101.1478423
117.1686465	95.26037259	71.32737145	124.6895599	110.7019087
77.81619344	106.867088	92.74548648	108.621829	78.4914549
53.68552772	99.18399217	127.3901657	129.6915102	115.4800775
99.01180073	86.52544178	69.38011409	72.22892252	122.1541802
95.78505053	93.85869715	119.595801	133.4597497	135.8513241
98.22838553	125.4883602	113.188661	103.6789277	93.48403916
112.9856511	100.1692002	82.56648328	114.2516554	136.1321316
105.6019068	93.4746867	133.6236876	76.443343	143.3595386
62.95617977	89.58568737	101.3501698	88.69819983	114.3900269
91.440331	96.46031059	116.5434247	93.0974028	78.25048078
87.017921	44.33108441	102.3655284	95.55979966	129.203039
69.53376531	123.3805805	110.8264747	118.0431521	85.37598469
111.2401286	118.0588863	96.96869193	103.4074918	104.4062563
109.5085553	89.18130235	66.33906123	95.73557981	108.0423921
106.6359452	114.3053581	92.50320389	82.23645826	91.12806915

```

96.27540641 98.43332131 84.73797429 84.82670487 101.7005933
106.3987993 101.4990011 104.2267675 90.21907271 92.27922542
93.81460506 123.2548138 66.868012 132.6489933 111.9414802
99.79866988 76.26973887 89.08310201 109.636018 96.71791287
96.28547323 127.8021082 112.9373963 114.6269398 112.4831253
128.2809537 110.1142502 81.1547673 60.70238138 117.2177177
93.16406456 86.24532972 90.70206847 88.00175655 124.0469975
117.2735726 107.2609915 104.0767926 122.9497116 121.1105841
143.9519908 87.93739852 99.6011896 104.29978 114.9546663
86.05930834 101.1799657 110.3341415 112.9557703 107.1117844
103.262856 99.49648475 82.25351215 102.3134457 76.71751602
70.32511371 110.0930537 122.1599585 58.30095817 110.2320489
106.6508842 114.1031517 101.2066501 118.0273944 75.63057702
99.86145565 95.07563134 83.22122036 83.42149109 47.64149344]

```

team :

```
['TeamA' 'TeamC' 'TeamB']
```

provider :

```
['Provider4' 'Provider1' 'Provider2' 'Provider3']
```

In [86]: `main.describe(include='all')`

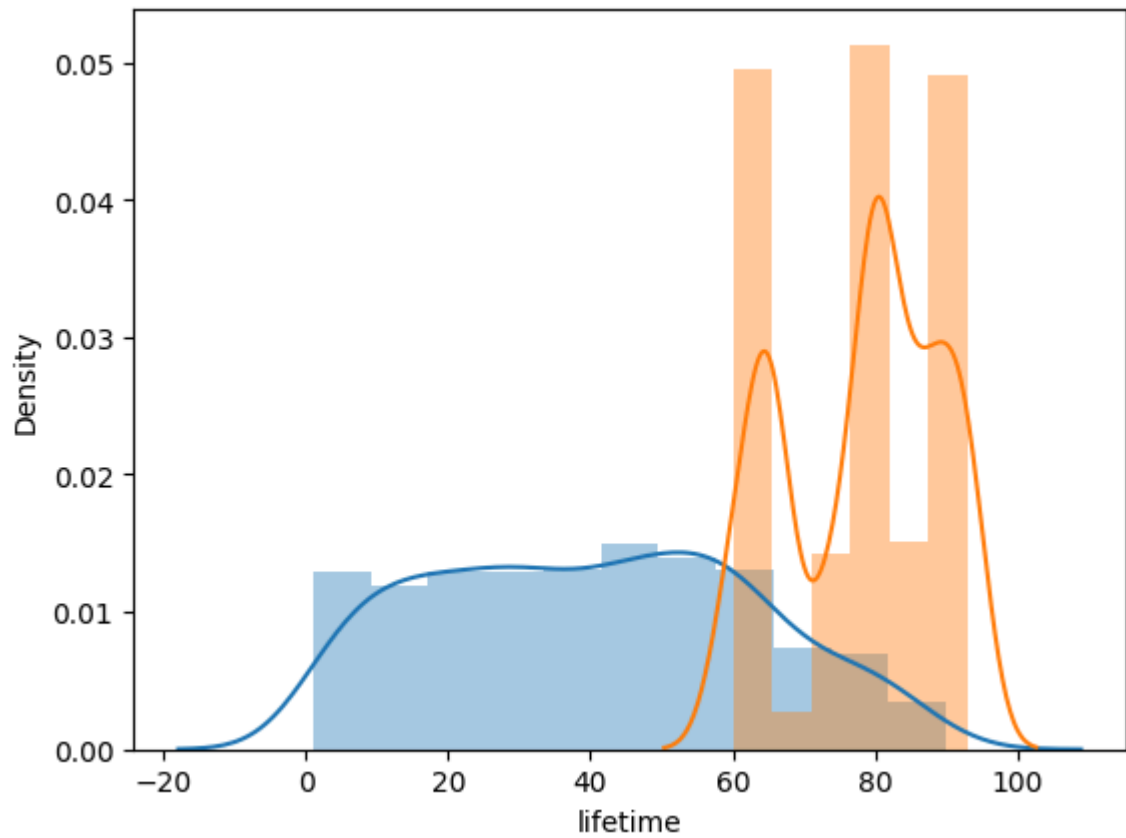
Out[86]:

	lifetime	broken	pressureInd	moistureInd	temperatureInd	team	p
<b>count</b>	1000.000000	1000.000000	1000.000000	1000.000000	1000.000000	1000	
<b>unique</b>	NaN	NaN	NaN	NaN	NaN	3	
<b>top</b>	NaN	NaN	NaN	NaN	NaN	TeamB	Pi
<b>freq</b>	NaN	NaN	NaN	NaN	NaN	356	
<b>mean</b>	55.195000	0.397000	98.599338	99.376723	100.628541	NaN	
<b>std</b>	26.472737	0.489521	19.964052	9.988726	19.633060	NaN	
<b>min</b>	1.000000	0.000000	33.481917	58.547301	42.279598	NaN	
<b>25%</b>	34.000000	0.000000	85.558076	92.771764	87.676913	NaN	
<b>50%</b>	60.000000	0.000000	97.216997	99.433959	100.592277	NaN	
<b>75%</b>	80.000000	1.000000	112.253190	106.120762	113.662885	NaN	
<b>max</b>	93.000000	1.000000	173.282541	128.595038	172.544140	NaN	

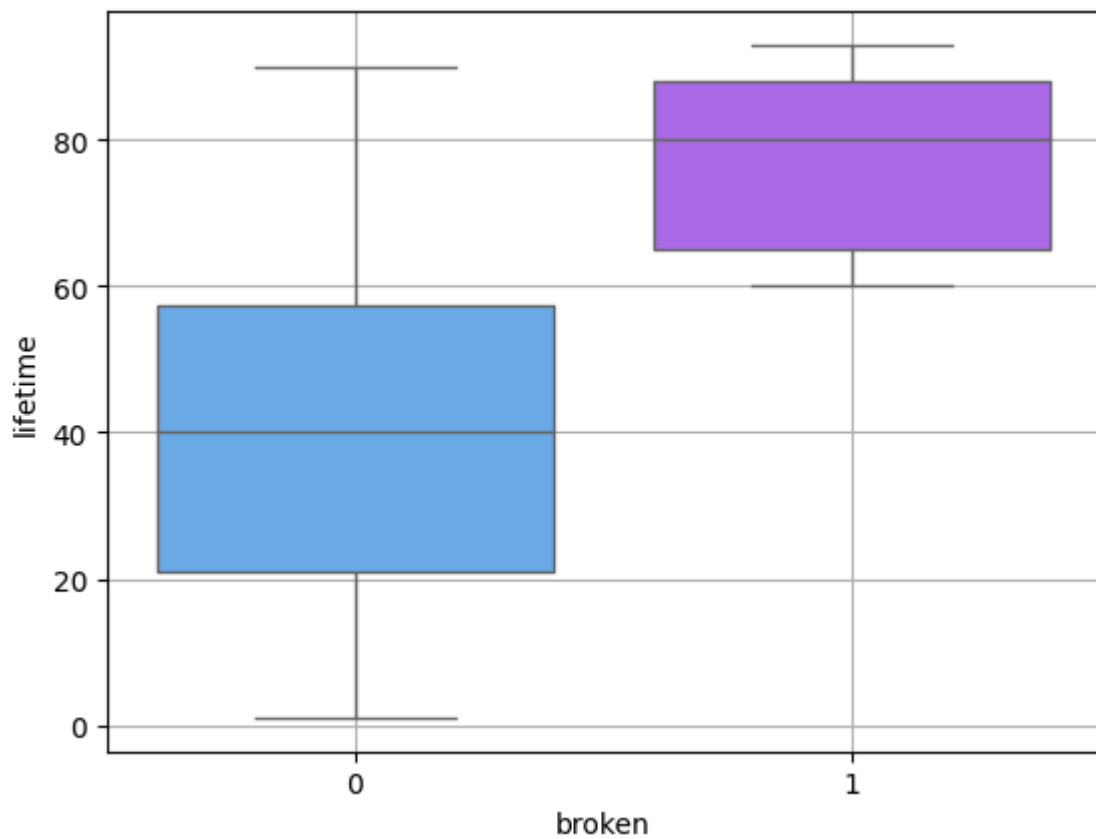


In [87]: `#Data Visualization or EDA`  
`import warnings`  
`warnings.filterwarnings('ignore')`

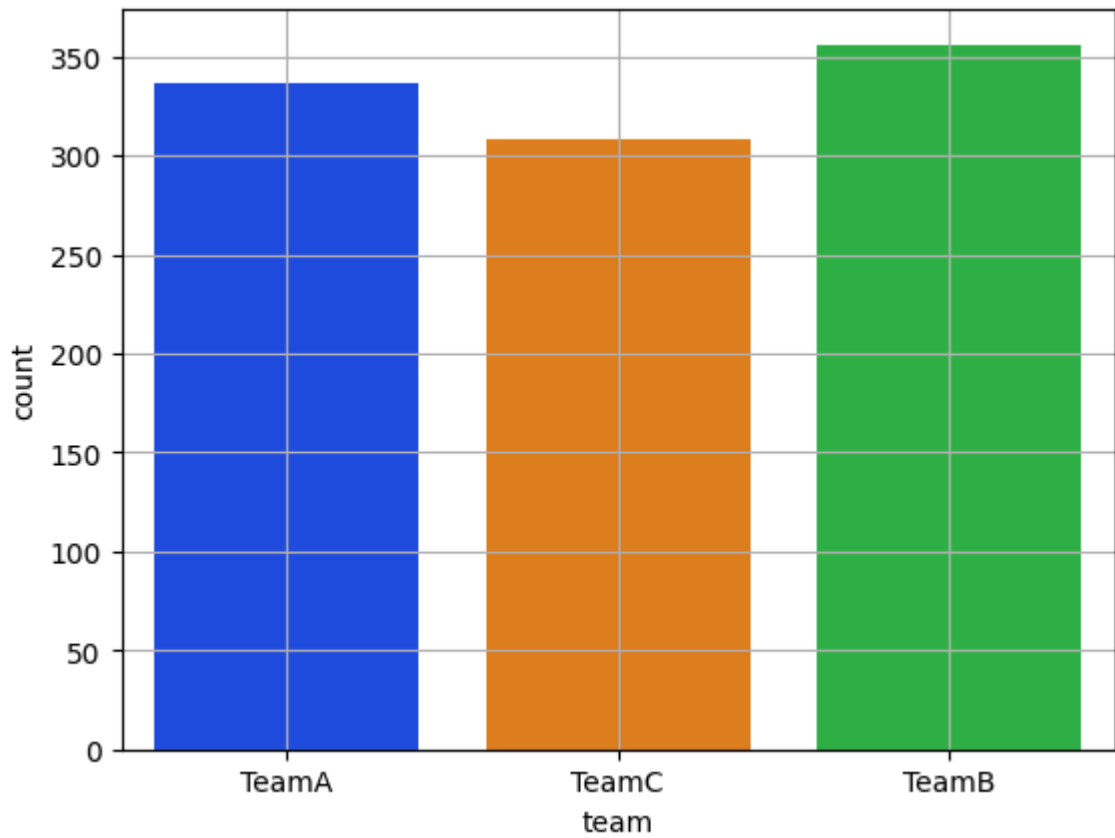
In [272...]: `sns.distplot(main.lifetime[main.broken==0])`  
`sns.distplot(main.lifetime[main.broken==1])`  
`plt.show()`



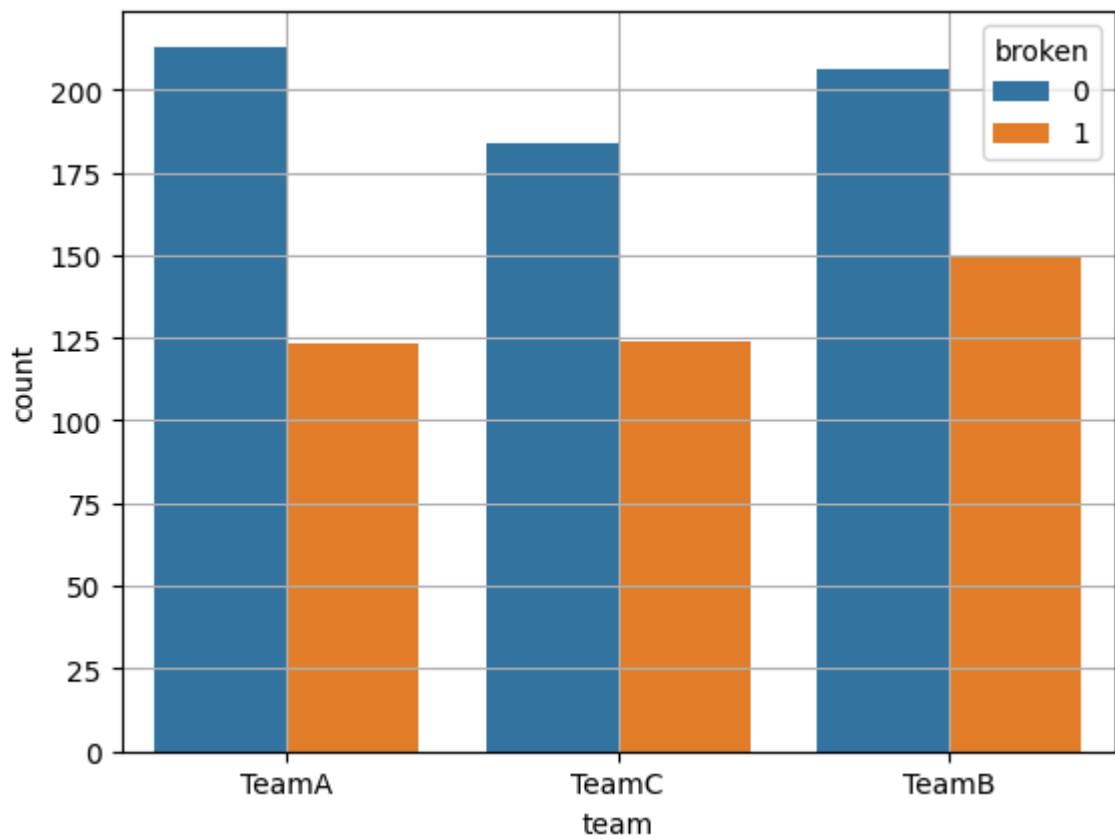
```
In [271]: sns.boxplot(x=main.broken,y=main.lifetime,palette='cool')  
plt.grid()  
plt.show()
```



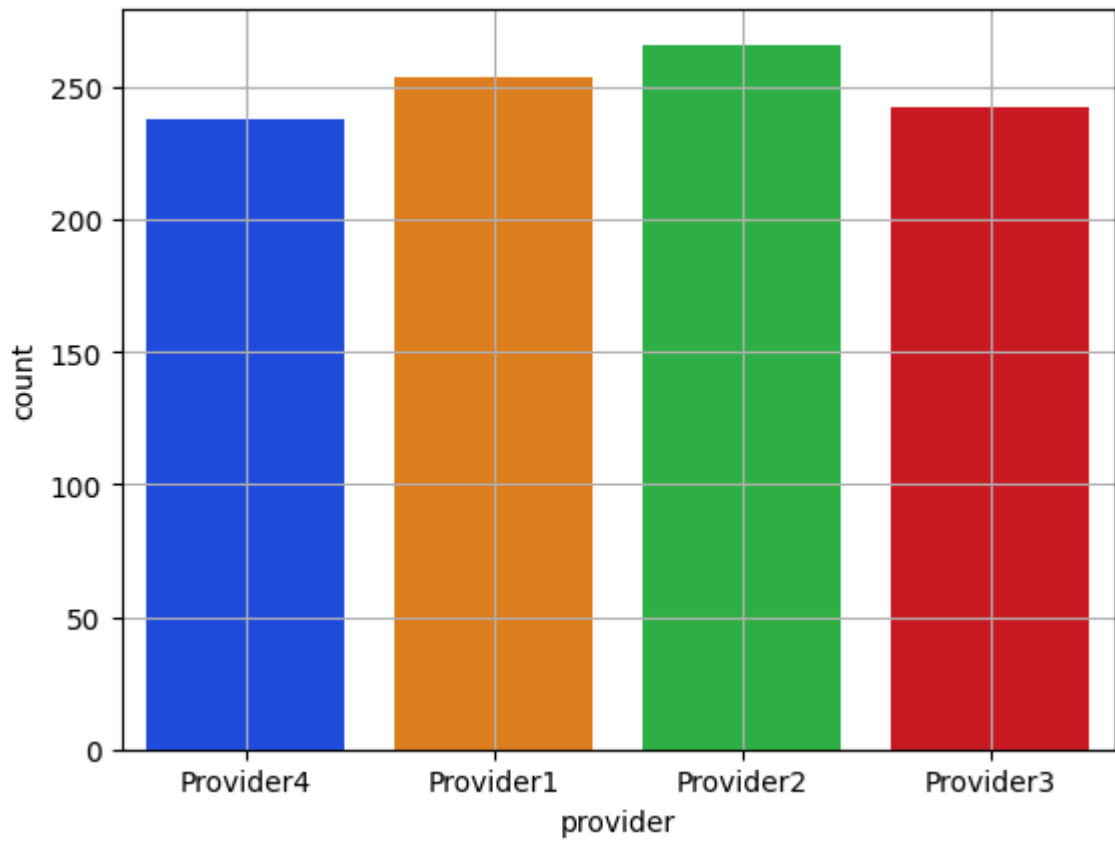
```
In [88]: sns.countplot(x=main.team,palette='bright')  
plt.grid()  
plt.show()
```



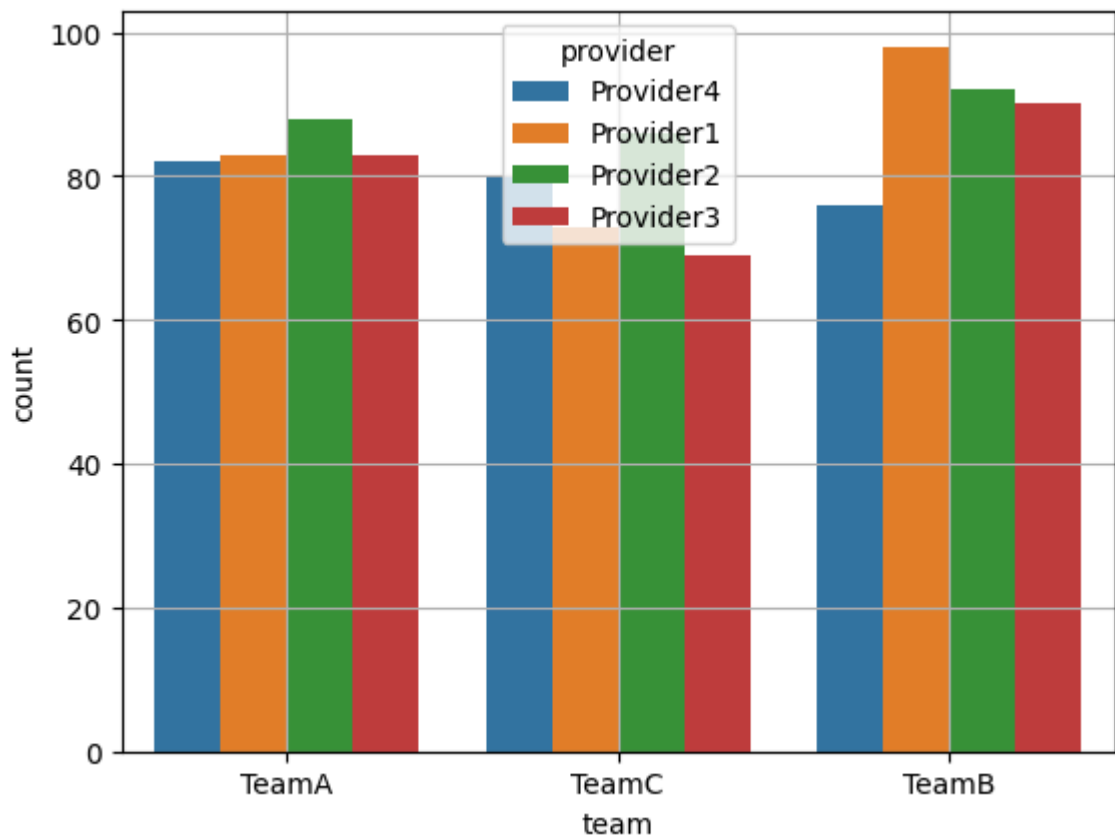
```
In [89]: sns.countplot(x=main.team, hue=main.broken)
plt.grid()
plt.show()
```



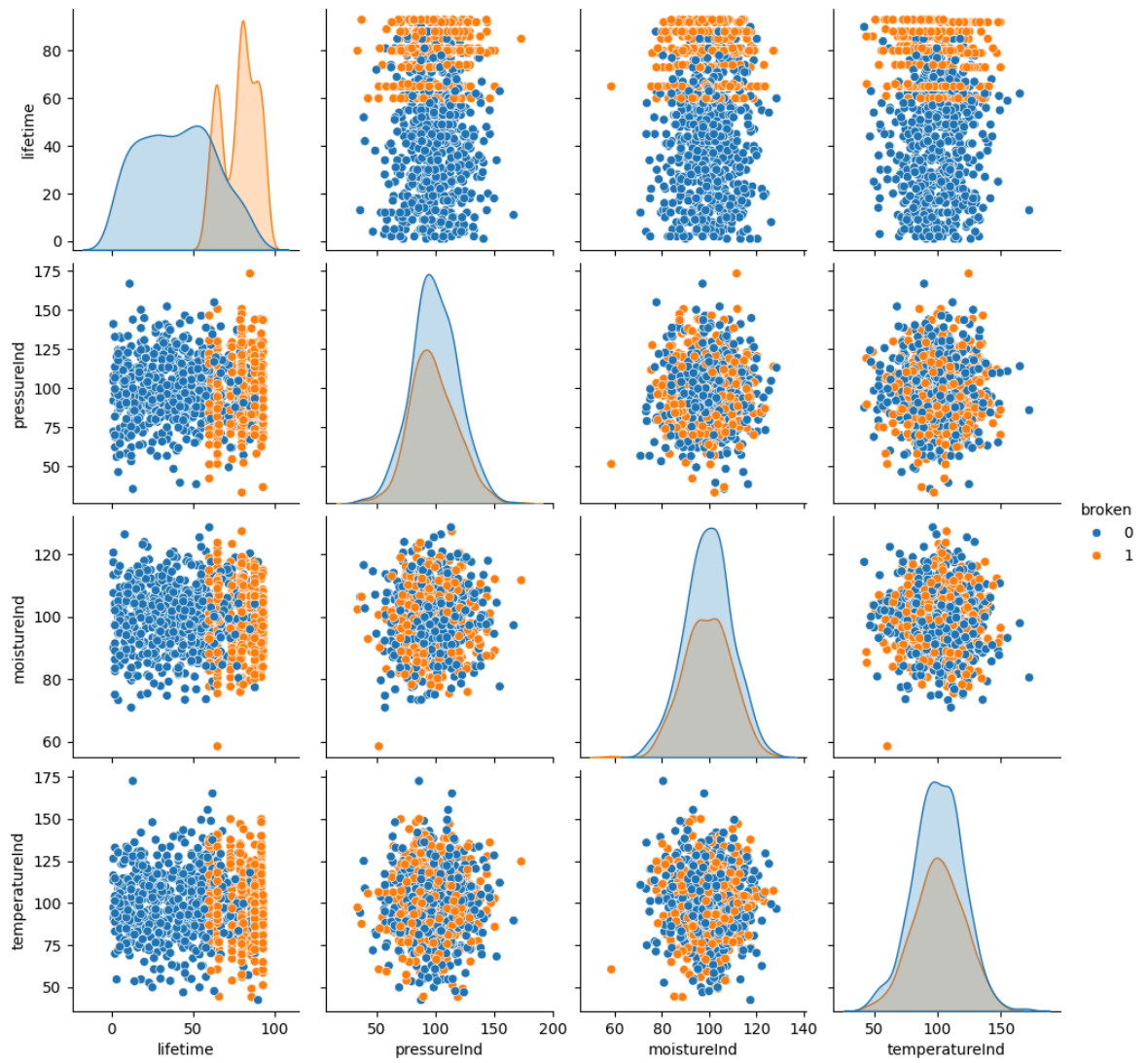
```
In [90]: sns.countplot(x=main.provider, palette='bright')
plt.grid()
plt.show()
```



```
In [91]: sns.countplot(x=main.team, hue=main.provider)
plt.grid()
plt.show()
```

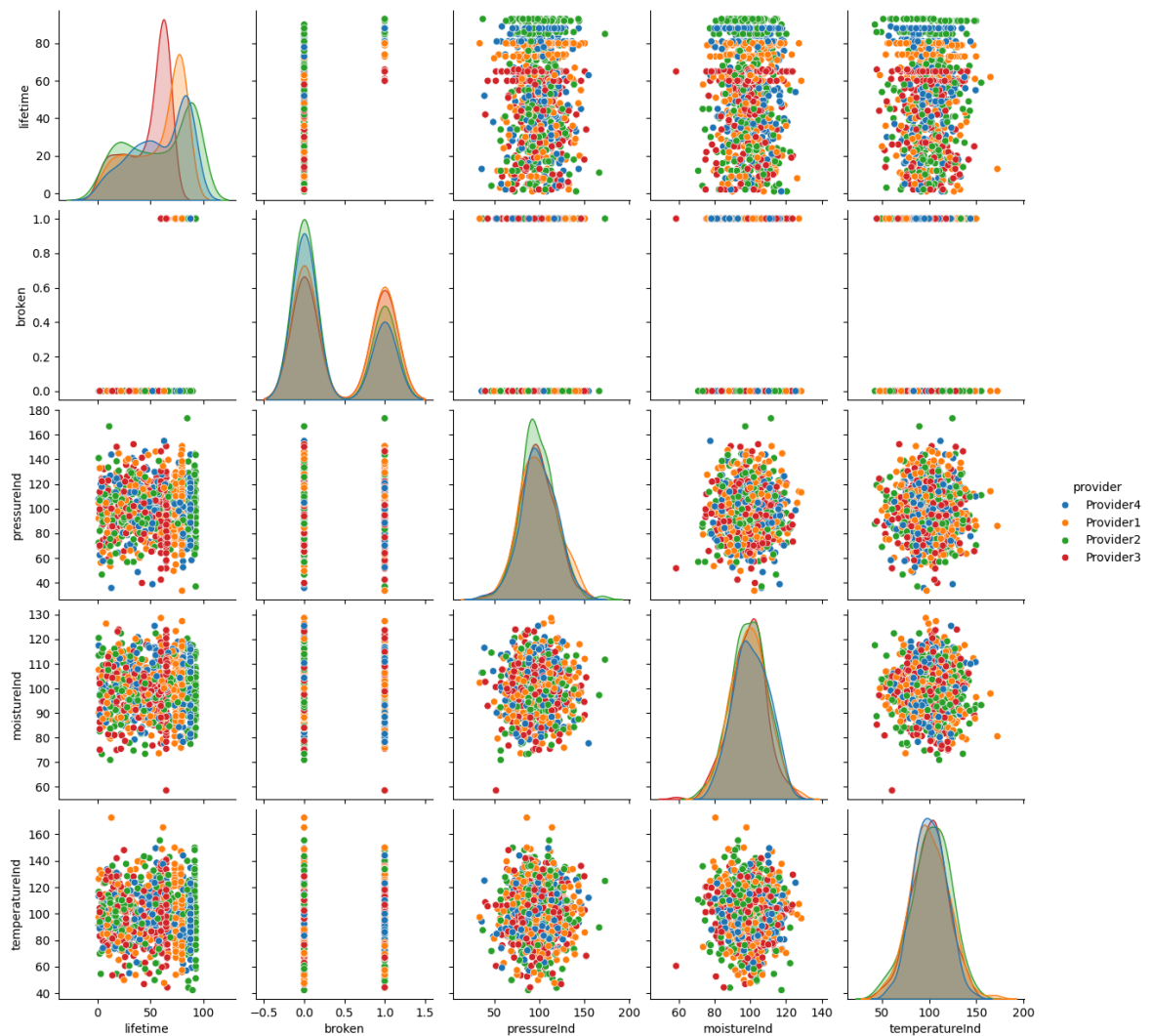


```
In [92]: sns.pairplot(main, hue='broken')
plt.show()
```



```
In [93]: sns.pairplot(main, hue='provider')  
plt.show()
```





**Encoding :** It is use to convert string to numbers.

**Types of Encoding :-**

- > Label Encoding
- > One-Hot Encoding
- > Frequency Encoding
- > Target Encoding
- > Ordinal Encoding
- > Dummy Encoding

```
In [94]: from sklearn.preprocessing import LabelEncoder
le = LabelEncoder()
```

```
In [95]: main.team = le.fit_transform(main.team)
main.team.unique()
```

```
Out[95]: array([0, 2, 1])
```

```
In [96]: le.inverse_transform([0,2,1])
```

```
Out[96]: array(['TeamA', 'TeamC', 'TeamB'], dtype=object)
```

```
In [97]: main.provider = le.fit_transform(main.provider)
main.provider.unique()
```

Out[97]: array([3, 0, 1, 2])

In [98]: le.inverse\_transform([3,0,1,2])

Out[98]: array(['Provider4', 'Provider1', 'Provider2', 'Provider3'], dtype=object)

In [99]: main

Out[99]:

	lifetime	broken	pressureInd	moistureInd	temperatureInd	team	provider
0	56	0	92.178854	104.230204	96.517159	0	3
1	81	1	72.075938	103.065701	87.271062	2	3
2	60	0	96.272254	77.801376	112.196170	0	0
3	86	1	94.406461	108.493608	72.025374	2	1
4	34	0	97.752899	99.413492	103.756271	1	0
...	...	...	...	...	...	...	...
995	88	1	88.589759	112.167556	99.861456	1	3
996	88	1	116.727075	110.871332	95.075631	0	3
997	22	0	104.026778	88.212873	83.221220	1	0
998	78	0	104.911649	104.257296	83.421491	0	3
999	63	0	116.901354	99.998694	47.641493	1	0

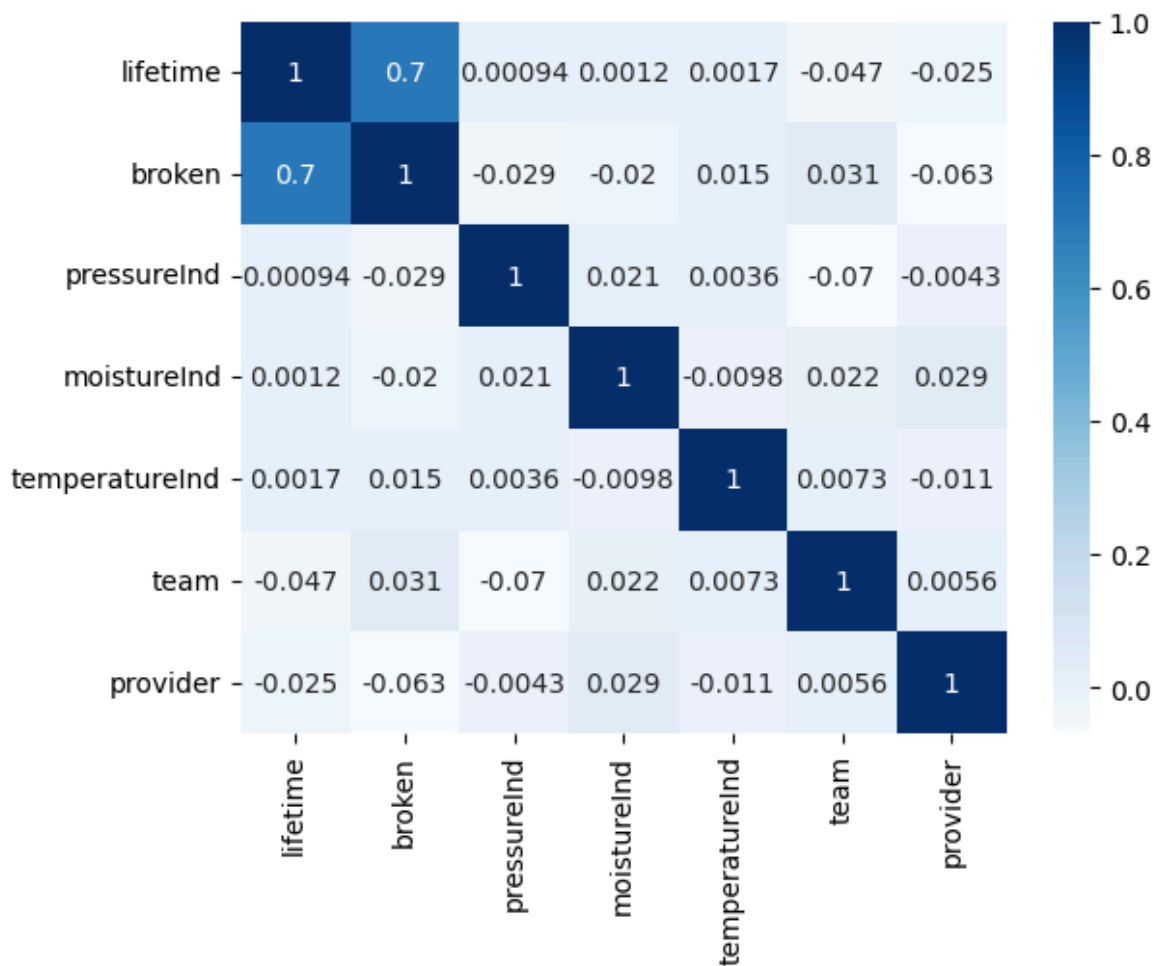
1000 rows × 7 columns

In [100... c = main.corr()  
c

Out[100...

	lifetime	broken	pressureInd	moistureInd	temperatureInd	team	provider
lifetime	1.000000	0.702656	0.000943	0.001196	0.001744	-0.046537	-0.025172
broken	0.702656	1.000000	-0.028942	-0.019520	0.015364	0.030876	-0.062972
pressureInd	0.000943	-0.028942	1.000000	0.020543	0.003641	-0.069528	-0.004337
moistureInd	0.001196	-0.019520	0.020543	1.000000	-0.009842	0.022420	0.028906
temperatureInd	0.001744	0.015364	0.003641	-0.009842	1.000000	0.007310	-0.010822
team	-0.046537	0.030876	-0.069528	0.022420	0.007310	1.000000	0.005616
provider	-0.025172	-0.062972	-0.004337	0.028906	-0.010822	0.005616	1.000000

In [101... sns.heatmap(c,annot=True,cmap='Blues')  
plt.show()



```
In [102... #ip/op Creation
ip = main.drop('broken',axis=1)
op = main.broken
```

```
In [117... #Train Test Split
from sklearn.model_selection import train_test_split
x_train, x_test, y_train, y_test = train_test_split(ip,op,test_size=0.2)
```

```
In [118... print(x_train.shape)
print(y_train.shape)
print(x_test.shape)
print(y_test.shape)
```

```
(800, 6)
```

```
(800,)
```

```
(200, 6)
```

```
(200,)
```

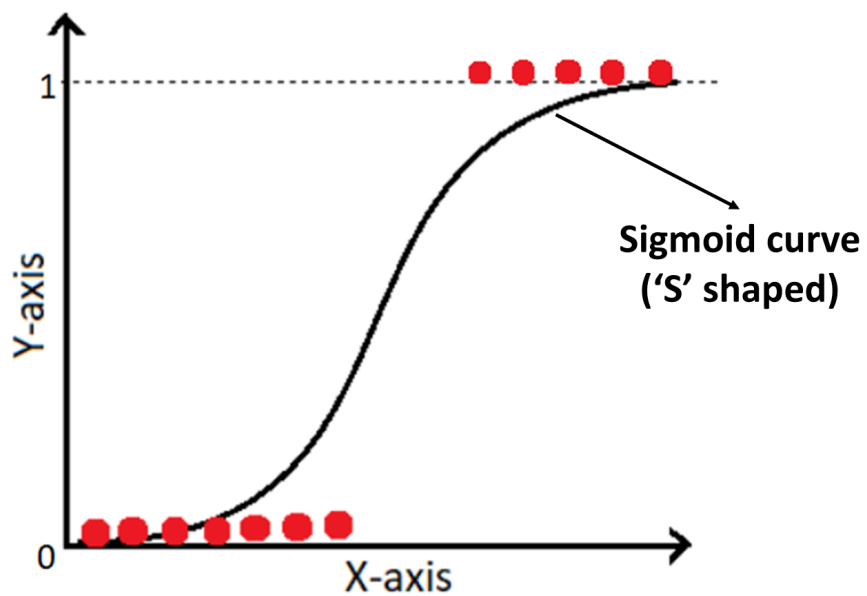
```
In [119... #Scaling
#Standard Scaler Transform
from sklearn.preprocessing import StandardScaler
sc = StandardScaler()

x_train = sc.fit_transform(x_train)
x_test = sc.transform(x_test)
```

## Logistic Regression

$$y = \frac{1}{1 + e^{-(x)}}$$

Dependent Variable (op)      Euler's constant (2.718)      Independent Variable (ip)



```
In [120... from sklearn.linear_model import LogisticRegression
lr = LogisticRegression()
lr.fit(x_train,y_train)
```

```
Out[120... LogisticRegression
LogisticRegression()
```

```
In [121... #Prediction
pred = lr.predict(x_test)
pred
```

```
Out[121...] array([0, 0, 1, 0, 1, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 1, 1, 1, 1, 0,
      1, 0, 1, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0,
      1, 1, 1, 0, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 1, 1, 1, 0, 1,
      1, 1, 0, 1, 1, 0, 1, 0, 1, 1, 0, 0, 0, 0, 1, 0, 1, 1, 0, 0, 1, 0,
      1, 0, 1, 1, 1, 1, 1, 1, 0, 1, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0, 1, 1,
      0, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 1, 0, 0, 1, 1, 1, 0, 0, 1, 1, 1,
      0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 1, 0, 0, 0, 1, 0, 0, 0, 1, 1, 1,
      0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 1, 0, 1, 1, 0, 1, 0, 1,
      0, 0, 1, 1, 0, 1, 0, 0, 0, 1, 1, 0, 0, 0, 0, 1, 1, 0, 0, 1, 0,
      0, 0], dtype=int64)
```

```
In [122...] y_test
```

```
Out[122...] 201    0
            899    0
            438    0
            211    0
            135    1
            ..
            303    0
            975    1
            988    0
            660    0
            250    0
            Name: broken, Length: 200, dtype: int64
```

```
In [123...] #Accuracy
from sklearn.metrics import accuracy_score, recall_score, precision_score, f1_score

ac = accuracy_score(y_test, pred)
re = recall_score(y_test, pred)
pr = precision_score(y_test, pred)
f1 = f1_score(y_test, pred)

print("Accuracy : ", ac)
print("Recall : ", re)
print("Precision : ", pr)
print("F1 Score : ", f1)
```

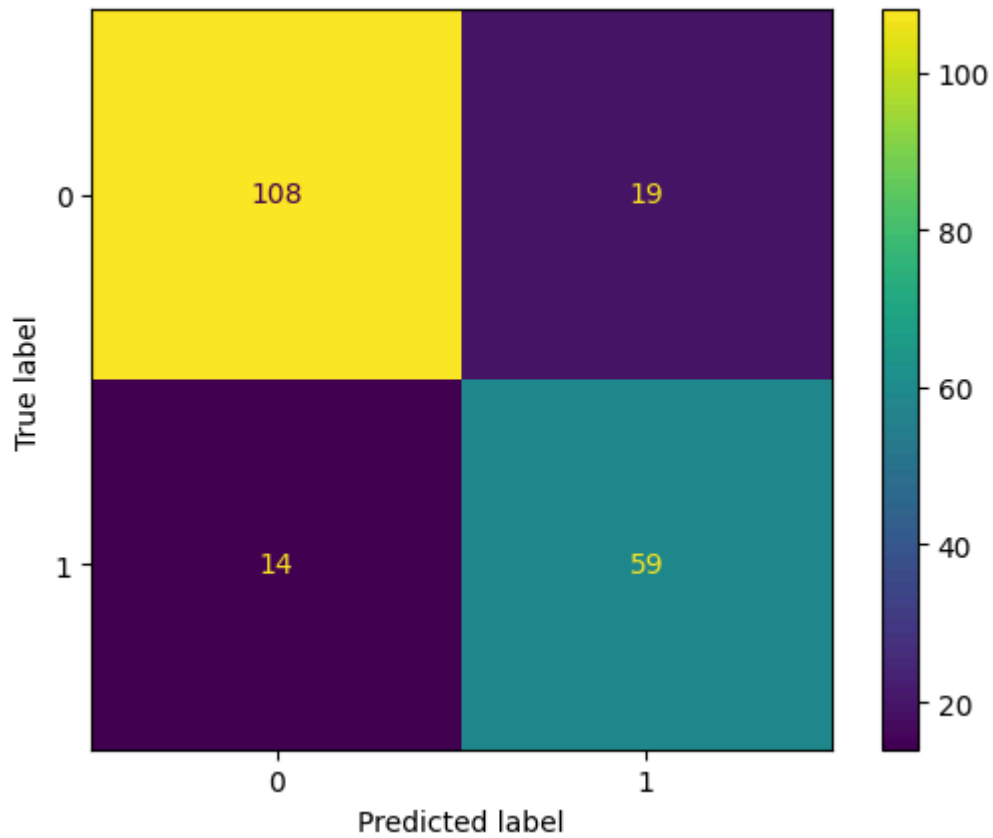
```
Accuracy : 0.835
Recall : 0.8082191780821918
Precision : 0.7564102564102564
F1 Score : 0.7814569536423841
```

```
In [124...] #Confusion Matrix
from sklearn.metrics import confusion_matrix, ConfusionMatrixDisplay

cm = confusion_matrix(y_test, pred)
cm
```

```
Out[124...] array([[108, 19],
      [ 14, 59]], dtype=int64)
```

```
In [125...] cmd = ConfusionMatrixDisplay(cm)
cmd.plot()
plt.show()
```

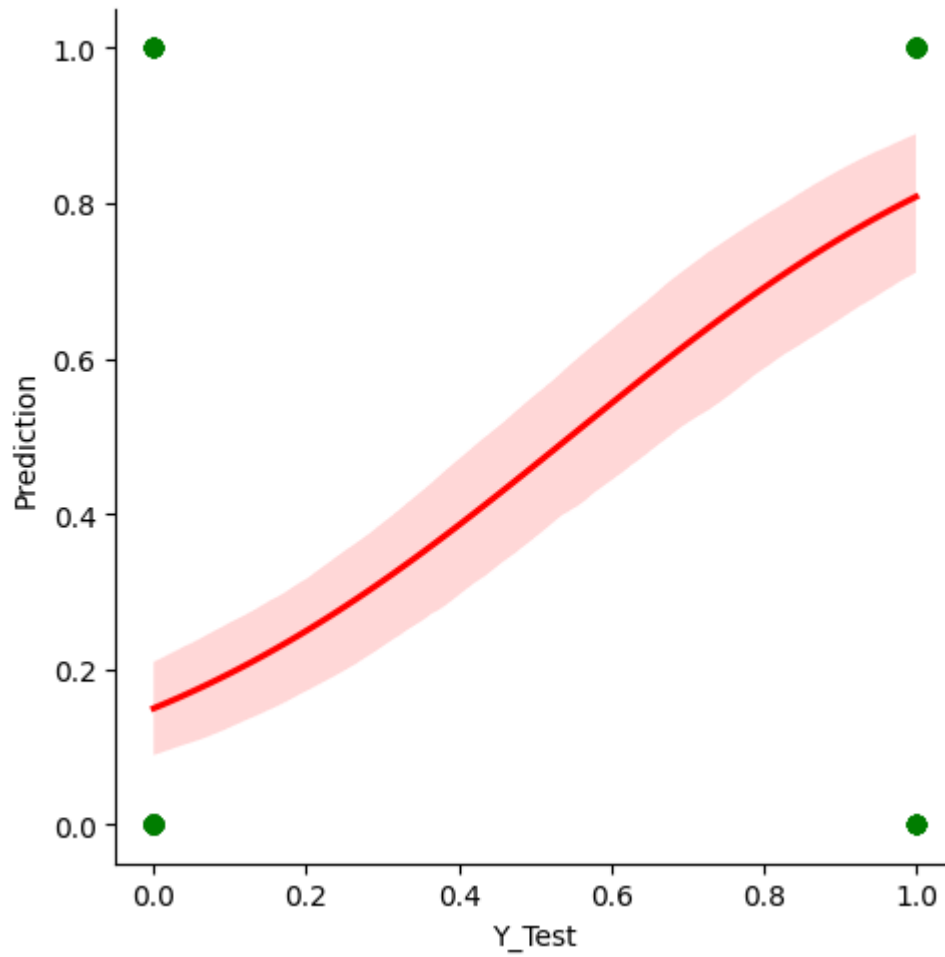


```
In [126... df = pd.DataFrame({'Y_Test':list(y_test),'Prediction':pred})
df
```

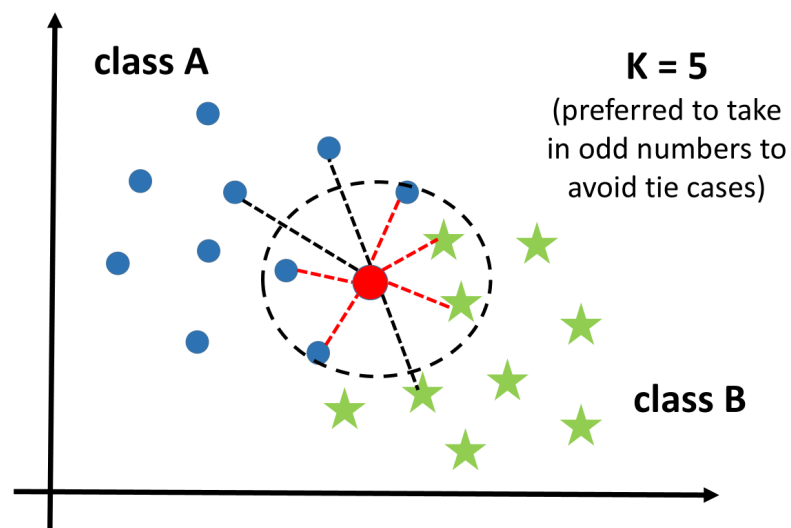
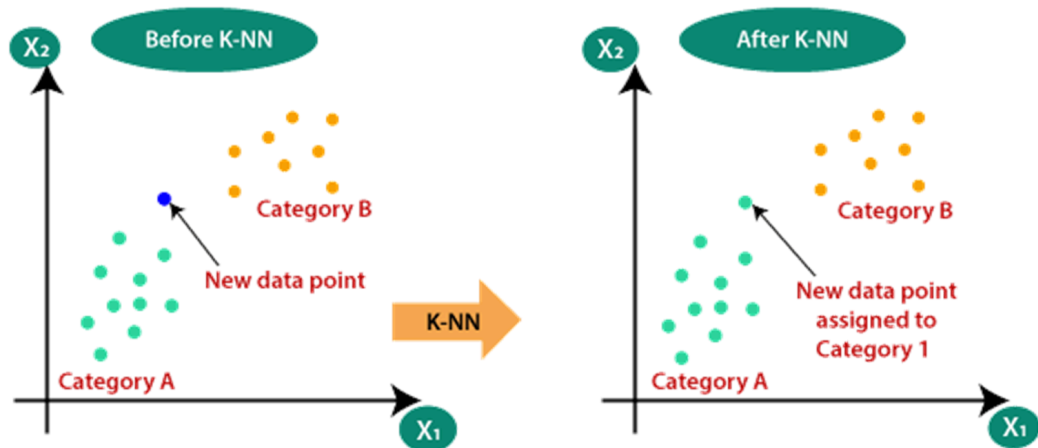
```
Out[126...
   Y_Test  Prediction
0        0           0
1        0           0
2        0           1
3        0           0
4        1           1
...      ...         ...
195       0           0
196       1           1
197       0           0
198       0           0
199       0           0
```

200 rows × 2 columns

```
In [131... #lmpplot
sns.lmplot(x='Y_Test',y='Prediction',data=df,logistic=True,
           scatter_kws={'color':'green'},line_kws={'color':'red'})
plt.show()
```



## KNN : K-Nearest Neighbors



## Euclidean Distance Formula

$$d = \sqrt{(x_2 - x_1)^2 + (y_2 - y_1)^2}$$

```
In [132... from sklearn.neighbors import KNeighborsClassifier
knn = KNeighborsClassifier(n_neighbors=7)
```



```
knn.fit(x_train,y_train)
```

Out[132...

```
▼ KNeighborsClassifier ⓘ ?  
KNeighborsClassifier(n_neighbors=7)
```

In [133...

```
pred1 = knn.predict(x_test)  
pred1
```

Out[133...

```
array([0, 0, 1, 1, 1, 1, 0, 0, 0, 1, 1, 0, 0, 0, 1, 1, 1, 1, 1, 1, 0, 1,  
       0, 0, 1, 1, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0,  
       1, 1, 1, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 1, 1, 1, 1, 0, 1,  
       1, 1, 0, 1, 1, 1, 1, 0, 1, 1, 0, 0, 0, 0, 1, 1, 1, 1, 0, 0, 1, 0,  
       0, 0, 1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 0, 0, 1, 1, 0, 0, 0, 0, 1, 0,  
       1, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 1, 0, 0, 1, 1, 1,  
       1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 1, 0, 0, 0, 1, 0, 1,  
       1, 0, 1, 0, 1, 1, 0, 0, 0, 0, 1, 0, 0, 1, 0, 1, 1, 0, 1, 0, 1,  
       0, 0, 1, 1, 0, 0, 0, 1, 1, 1, 0, 0, 0, 0, 1, 1, 1, 0, 1, 0,  
       0, 0], dtype=int64)
```

In [138...

```
#Accuracy  
from sklearn.metrics import accuracy_score, recall_score, precision_score, f1_score  
  
ac = accuracy_score(y_test,pred1)  
re = recall_score(y_test,pred1)  
pr = precision_score(y_test,pred1)  
f1 = f1_score(y_test,pred1)  
  
print("Accuracy : ",ac)  
print("Recall : ",re)  
print("Precision : ",pr)  
print("F1 Score : ",f1)
```

```
Accuracy : 0.8  
Recall : 0.8356164383561644  
Precision : 0.6853932584269663  
F1 Score : 0.7530864197530864
```

In [139...

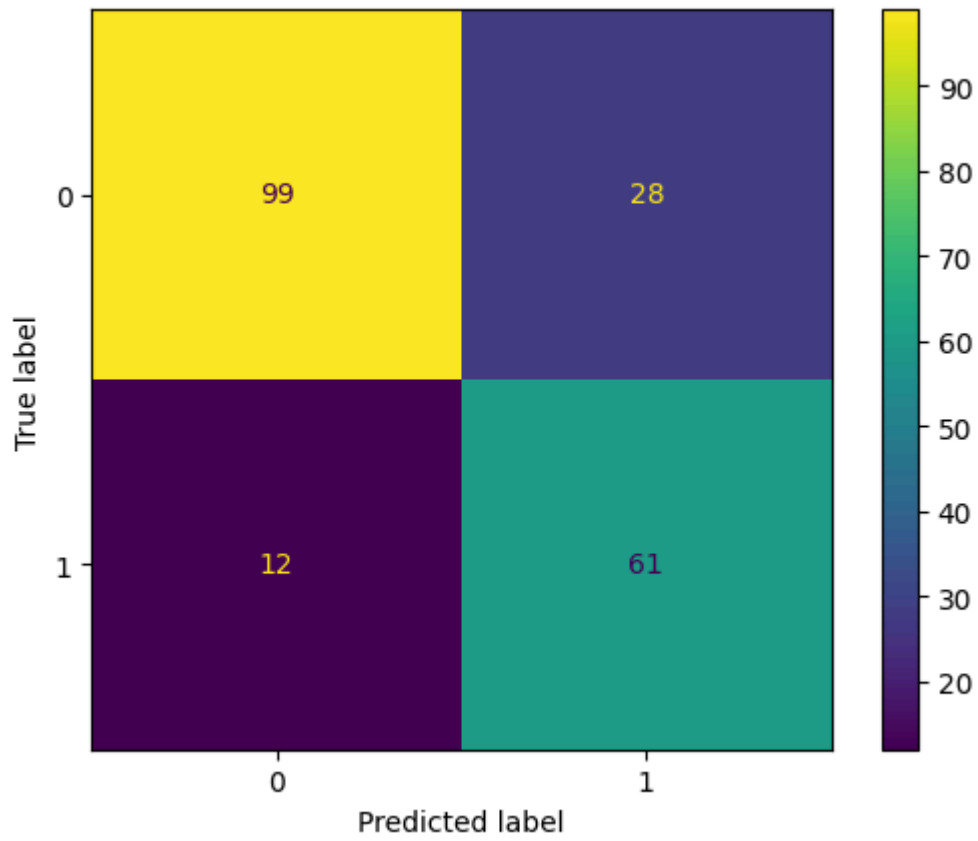
```
cm = confusion_matrix(y_test,pred1)  
cm
```

Out[139...

```
array([[99, 28],  
       [12, 61]], dtype=int64)
```

In [140...

```
cmd = ConfusionMatrixDisplay(cm)  
cmd.plot()  
plt.show()
```



In [ ]:

In [ ]: