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In [ ]: # Name : PHULE PRATHAMESH DNYANDEV
```

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In [ ]: # Roll No : COTB42
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```
In [4]: import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
```

```
In [5]: iris = pd.read_csv("Iris.csv")
```

```
In [6]: iris.head()
```

Out[6]:

	Id	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm	Species
0	1	5.1	3.5	1.4	0.2	Iris-setosa
1	2	4.9	3.0	1.4	0.2	Iris-setosa
2	3	4.7	3.2	1.3	0.2	Iris-setosa
3	4	4.6	3.1	1.5	0.2	Iris-setosa
4	5	5.0	3.6	1.4	0.2	Iris-setosa

```
In [7]: iris.dtypes
```

```
Out[7]: Id                int64
SepalLengthCm          float64
SepalWidthCm           float64
PetalLengthCm          float64
PetalWidthCm           float64
Species                object
dtype: object
```

```
In [8]: from sklearn import preprocessing
```

```
In [9]: enc = preprocessing.OneHotEncoder()  
enc_iris = pd.DataFrame(enc.fit_transform(iris[['SepalLengthCm']]).toarray())  
enc_iris
```

Out[9]:

	0	1	2	3
0	0.0	1.0	0.0	0.0
1	1.0	0.0	0.0	0.0
2	1.0	0.0	0.0	0.0
3	1.0	0.0	0.0	0.0
4	0.0	1.0	0.0	0.0
5	0.0	1.0	0.0	0.0
6	1.0	0.0	0.0	0.0
7	0.0	1.0	0.0	0.0
8	1.0	0.0	0.0	0.0
9	1.0	0.0	0.0	0.0
10	0.0	1.0	0.0	0.0
11	1.0	0.0	0.0	0.0
12	1.0	0.0	0.0	0.0
13	1.0	0.0	0.0	0.0
14	0.0	1.0	0.0	0.0
15	0.0	1.0	0.0	0.0
16	0.0	1.0	0.0	0.0
17	0.0	1.0	0.0	0.0
18	0.0	1.0	0.0	0.0
19	0.0	1.0	0.0	0.0
20	0.0	1.0	0.0	0.0
21	0.0	1.0	0.0	0.0
22	1.0	0.0	0.0	0.0
23	0.0	1.0	0.0	0.0
24	1.0	0.0	0.0	0.0
25	0.0	1.0	0.0	0.0
26	0.0	1.0	0.0	0.0
27	0.0	1.0	0.0	0.0
28	0.0	1.0	0.0	0.0
29	1.0	0.0	0.0	0.0
...
120	0.0	0.0	1.0	0.0

	0	1	2	3
121	0.0	1.0	0.0	0.0
122	0.0	0.0	0.0	1.0
123	0.0	0.0	1.0	0.0
124	0.0	0.0	1.0	0.0
125	0.0	0.0	0.0	1.0
126	0.0	0.0	1.0	0.0
127	0.0	0.0	1.0	0.0
128	0.0	0.0	1.0	0.0
129	0.0	0.0	0.0	1.0
130	0.0	0.0	0.0	1.0
131	0.0	0.0	0.0	1.0
132	0.0	0.0	1.0	0.0
133	0.0	0.0	1.0	0.0
134	0.0	0.0	1.0	0.0
135	0.0	0.0	0.0	1.0
136	0.0	0.0	1.0	0.0
137	0.0	0.0	1.0	0.0
138	0.0	0.0	1.0	0.0
139	0.0	0.0	1.0	0.0
140	0.0	0.0	1.0	0.0
141	0.0	0.0	1.0	0.0
142	0.0	1.0	0.0	0.0
143	0.0	0.0	1.0	0.0
144	0.0	0.0	1.0	0.0
145	0.0	0.0	1.0	0.0
146	0.0	0.0	1.0	0.0
147	0.0	0.0	1.0	0.0
148	0.0	0.0	1.0	0.0
149	0.0	1.0	0.0	0.0

150 rows × 4 columns

```
In [10]: from sklearn import preprocessing  
le = preprocessing.LabelEncoder()  
iris['Species'] = le.fit_transform(iris['Species'])  
newiris=iris  
iris.head()
```

Out[10]:

	Id	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm	Species
0	1	5.1	3.5	1.4	0.2	0
1	2	4.9	3.0	1.4	0.2	0
2	3	4.7	3.2	1.3	0.2	0
3	4	4.6	3.1	1.5	0.2	0
4	5	5.0	3.6	1.4	0.2	0

```
In [11]: iris.isnull().head(5)
```

Out[11]:

	Id	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm	Species
0	False	False	False	False	False	False
1	False	False	False	False	False	False
2	False	False	False	False	False	False
3	False	False	False	False	False	False
4	False	False	False	False	False	False

```
In [12]: x = iris.iloc[:, :-1].values
```

In [13]: `print(x)`

[[1.00000000e+00 2.00000000e-01]	5.10000000e+00	3.50000000e+00	1.40000000e+00
[2.00000000e+00 2.00000000e-01]	4.90000000e+00	3.00000000e+00	1.40000000e+00
[3.00000000e+00 2.00000000e-01]	4.70000000e+00	3.20000000e+00	1.30000000e+00
[4.00000000e+00 2.00000000e-01]	4.60000000e+00	3.10000000e+00	1.50000000e+00
[5.00000000e+00 2.00000000e-01]	5.00000000e+00	3.60000000e+00	1.40000000e+00
[6.00000000e+00 4.00000000e-01]	5.40000000e+00	3.90000000e+00	1.70000000e+00
[7.00000000e+00 3.00000000e-01]	4.60000000e+00	3.40000000e+00	1.40000000e+00
[8.00000000e+00 2.00000000e-01]	5.00000000e+00	3.40000000e+00	1.50000000e+00
[9.00000000e+00 2.00000000e-01]	4.40000000e+00	2.90000000e+00	1.40000000e+00
[1.00000000e+01 1.00000000e-01]	4.90000000e+00	3.10000000e+00	1.50000000e+00
[1.10000000e+01 2.00000000e-01]	5.40000000e+00	3.70000000e+00	1.50000000e+00
[1.20000000e+01 2.00000000e-01]	4.80000000e+00	3.40000000e+00	1.60000000e+00
[1.30000000e+01 1.00000000e-01]	4.80000000e+00	3.00000000e+00	1.40000000e+00
[1.40000000e+01 1.00000000e-01]	4.30000000e+00	3.00000000e+00	1.10000000e+00
[1.50000000e+01 2.00000000e-01]	5.80000000e+00	4.00000000e+00	1.20000000e+00
[1.60000000e+01 4.00000000e-01]	5.70000000e+00	4.40000000e+00	1.50000000e+00
[1.70000000e+01 4.00000000e-01]	5.40000000e+00	3.90000000e+00	1.30000000e+00
[1.80000000e+01 3.00000000e-01]	5.10000000e+00	3.50000000e+00	1.40000000e+00
[1.90000000e+01 3.00000000e-01]	5.70000000e+00	3.80000000e+00	1.70000000e+00
[2.00000000e+01 3.00000000e-01]	5.10000000e+00	3.80000000e+00	1.50000000e+00
[2.10000000e+01 2.00000000e-01]	5.40000000e+00	3.40000000e+00	1.70000000e+00
[2.20000000e+01 4.00000000e-01]	5.10000000e+00	3.70000000e+00	1.50000000e+00
[2.30000000e+01 2.00000000e-01]	4.60000000e+00	3.60000000e+00	1.00000000e+00
[2.40000000e+01 5.00000000e-01]	5.10000000e+00	3.30000000e+00	1.70000000e+00
[2.50000000e+01 2.00000000e-01]	4.80000000e+00	3.40000000e+00	1.90000000e+00
[2.60000000e+01 2.00000000e-01]	5.00000000e+00	3.00000000e+00	1.60000000e+00
[2.70000000e+01 4.00000000e-01]	5.00000000e+00	3.40000000e+00	1.60000000e+00
[2.80000000e+01 2.00000000e-01]	5.20000000e+00	3.50000000e+00	1.50000000e+00
[2.90000000e+01	5.20000000e+00	3.40000000e+00	1.40000000e+00

2.00000000e-01]			
[3.00000000e+01	4.70000000e+00	3.20000000e+00	1.60000000e+00
2.00000000e-01]			
[3.10000000e+01	4.80000000e+00	3.10000000e+00	1.60000000e+00
2.00000000e-01]			
[3.20000000e+01	5.40000000e+00	3.40000000e+00	1.50000000e+00
4.00000000e-01]			
[3.30000000e+01	5.20000000e+00	4.10000000e+00	1.50000000e+00
1.00000000e-01]			
[3.40000000e+01	5.50000000e+00	4.20000000e+00	1.40000000e+00
2.00000000e-01]			
[3.50000000e+01	4.90000000e+00	3.10000000e+00	1.50000000e+00
1.00000000e-01]			
[3.60000000e+01	5.00000000e+00	3.20000000e+00	1.20000000e+00
2.00000000e-01]			
[3.70000000e+01	5.50000000e+00	3.50000000e+00	1.30000000e+00
2.00000000e-01]			
[3.80000000e+01	4.90000000e+00	3.10000000e+00	1.50000000e+00
1.00000000e-01]			
[3.90000000e+01	4.40000000e+00	3.00000000e+00	1.30000000e+00
2.00000000e-01]			
[4.00000000e+01	5.10000000e+00	3.40000000e+00	1.50000000e+00
2.00000000e-01]			
[4.10000000e+01	5.00000000e+00	3.50000000e+00	1.30000000e+00
3.00000000e-01]			
[4.20000000e+01	4.50000000e+00	2.30000000e+00	1.30000000e+00
3.00000000e-01]			
[4.30000000e+01	4.40000000e+00	3.20000000e+00	1.30000000e+00
2.00000000e-01]			
[4.40000000e+01	5.00000000e+00	3.50000000e+00	1.60000000e+00
6.00000000e-01]			
[4.50000000e+01	5.10000000e+00	3.80000000e+00	1.90000000e+00
4.00000000e-01]			
[4.60000000e+01	4.80000000e+00	3.00000000e+00	1.40000000e+00
3.00000000e-01]			
[4.70000000e+01	5.10000000e+00	3.80000000e+00	1.60000000e+00
2.00000000e-01]			
[4.80000000e+01	4.60000000e+00	3.20000000e+00	1.40000000e+00
2.00000000e-01]			
[4.90000000e+01	5.30000000e+00	3.70000000e+00	1.50000000e+00
2.00000000e-01]			
[5.00000000e+01	5.00000000e+00	3.30000000e+00	1.40000000e+00
2.00000000e-01]			
[5.10000000e+01	7.00000000e+00	3.20000000e+00	4.70000000e+00
1.40000000e+00]			
[5.20000000e+01	6.40000000e+00	3.20000000e+00	4.50000000e+00
1.50000000e+00]			
[5.30000000e+01	6.90000000e+00	3.10000000e+00	4.90000000e+00
1.50000000e+00]			
[5.40000000e+01	5.50000000e+00	2.30000000e+00	4.00000000e+00
1.30000000e+00]			
[5.50000000e+01	6.50000000e+00	2.80000000e+00	4.60000000e+00
1.50000000e+00]			
[5.60000000e+01	5.70000000e+00	2.80000000e+00	4.50000000e+00
1.30000000e+00]			
[5.70000000e+01	6.30000000e+00	3.30000000e+00	4.70000000e+00
1.60000000e+00]			

[5.80000000e+01 1.00000000e+00]	4.90000000e+00	2.40000000e+00	3.30000000e+00
[5.90000000e+01 1.30000000e+00]	6.60000000e+00	2.90000000e+00	4.60000000e+00
[6.00000000e+01 1.40000000e+00]	5.20000000e+00	2.70000000e+00	3.90000000e+00
[6.10000000e+01 1.00000000e+00]	5.00000000e+00	2.00000000e+00	3.50000000e+00
[6.20000000e+01 1.50000000e+00]	5.90000000e+00	3.00000000e+00	4.20000000e+00
[6.30000000e+01 1.00000000e+00]	6.00000000e+00	2.20000000e+00	4.00000000e+00
[6.40000000e+01 1.40000000e+00]	6.10000000e+00	2.90000000e+00	4.70000000e+00
[6.50000000e+01 1.30000000e+00]	5.60000000e+00	2.90000000e+00	3.60000000e+00
[6.60000000e+01 1.40000000e+00]	6.70000000e+00	3.10000000e+00	4.40000000e+00
[6.70000000e+01 1.50000000e+00]	5.60000000e+00	3.00000000e+00	4.50000000e+00
[6.80000000e+01 1.00000000e+00]	5.80000000e+00	2.70000000e+00	4.10000000e+00
[6.90000000e+01 1.50000000e+00]	6.20000000e+00	2.20000000e+00	4.50000000e+00
[7.00000000e+01 1.10000000e+00]	5.60000000e+00	2.50000000e+00	3.90000000e+00
[7.10000000e+01 1.80000000e+00]	5.90000000e+00	3.20000000e+00	4.80000000e+00
[7.20000000e+01 1.30000000e+00]	6.10000000e+00	2.80000000e+00	4.00000000e+00
[7.30000000e+01 1.50000000e+00]	6.30000000e+00	2.50000000e+00	4.90000000e+00
[7.40000000e+01 1.20000000e+00]	6.10000000e+00	2.80000000e+00	4.70000000e+00
[7.50000000e+01 1.30000000e+00]	6.40000000e+00	2.90000000e+00	4.30000000e+00
[7.60000000e+01 1.40000000e+00]	6.60000000e+00	3.00000000e+00	4.40000000e+00
[7.70000000e+01 1.40000000e+00]	6.80000000e+00	2.80000000e+00	4.80000000e+00
[7.80000000e+01 1.70000000e+00]	6.70000000e+00	3.00000000e+00	5.00000000e+00
[7.90000000e+01 1.50000000e+00]	6.00000000e+00	2.90000000e+00	4.50000000e+00
[8.00000000e+01 1.00000000e+00]	5.70000000e+00	2.60000000e+00	3.50000000e+00
[8.10000000e+01 1.10000000e+00]	5.50000000e+00	2.40000000e+00	3.80000000e+00
[8.20000000e+01 1.00000000e+00]	5.50000000e+00	2.40000000e+00	3.70000000e+00
[8.30000000e+01 1.20000000e+00]	5.80000000e+00	2.70000000e+00	3.90000000e+00
[8.40000000e+01 1.60000000e+00]	6.00000000e+00	2.70000000e+00	5.10000000e+00
[8.50000000e+01 1.50000000e+00]	5.40000000e+00	3.00000000e+00	4.50000000e+00
[8.60000000e+01	6.00000000e+00	3.40000000e+00	4.50000000e+00

[1.60000000e+00]			
[8.70000000e+01	6.70000000e+00	3.10000000e+00	4.70000000e+00
1.50000000e+00]			
[8.80000000e+01	6.30000000e+00	2.30000000e+00	4.40000000e+00
1.30000000e+00]			
[8.90000000e+01	5.60000000e+00	3.00000000e+00	4.10000000e+00
1.30000000e+00]			
[9.00000000e+01	5.50000000e+00	2.50000000e+00	4.00000000e+00
1.30000000e+00]			
[9.10000000e+01	5.50000000e+00	2.60000000e+00	4.40000000e+00
1.20000000e+00]			
[9.20000000e+01	6.10000000e+00	3.00000000e+00	4.60000000e+00
1.40000000e+00]			
[9.30000000e+01	5.80000000e+00	2.60000000e+00	4.00000000e+00
1.20000000e+00]			
[9.40000000e+01	5.00000000e+00	2.30000000e+00	3.30000000e+00
1.00000000e+00]			
[9.50000000e+01	5.60000000e+00	2.70000000e+00	4.20000000e+00
1.30000000e+00]			
[9.60000000e+01	5.70000000e+00	3.00000000e+00	4.20000000e+00
1.20000000e+00]			
[9.70000000e+01	5.70000000e+00	2.90000000e+00	4.20000000e+00
1.30000000e+00]			
[9.80000000e+01	6.20000000e+00	2.90000000e+00	4.30000000e+00
1.30000000e+00]			
[9.90000000e+01	5.10000000e+00	2.50000000e+00	3.00000000e+00
1.10000000e+00]			
[1.00000000e+02	5.70000000e+00	2.80000000e+00	4.10000000e+00
1.30000000e+00]			
[1.01000000e+02	6.30000000e+00	3.30000000e+00	6.00000000e+00
2.50000000e+00]			
[1.02000000e+02	5.80000000e+00	2.70000000e+00	5.10000000e+00
1.90000000e+00]			
[1.03000000e+02	7.10000000e+00	3.00000000e+00	5.90000000e+00
2.10000000e+00]			
[1.04000000e+02	6.30000000e+00	2.90000000e+00	5.60000000e+00
1.80000000e+00]			
[1.05000000e+02	6.50000000e+00	3.00000000e+00	5.80000000e+00
2.20000000e+00]			
[1.06000000e+02	7.60000000e+00	3.00000000e+00	6.60000000e+00
2.10000000e+00]			
[1.07000000e+02	4.90000000e+00	2.50000000e+00	4.50000000e+00
1.70000000e+00]			
[1.08000000e+02	7.30000000e+00	2.90000000e+00	6.30000000e+00
1.80000000e+00]			
[1.09000000e+02	6.70000000e+00	2.50000000e+00	5.80000000e+00
1.80000000e+00]			
[1.10000000e+02	7.20000000e+00	3.60000000e+00	6.10000000e+00
2.50000000e+00]			
[1.11000000e+02	6.50000000e+00	3.20000000e+00	5.10000000e+00
2.00000000e+00]			
[1.12000000e+02	6.40000000e+00	2.70000000e+00	5.30000000e+00
1.90000000e+00]			
[1.13000000e+02	6.80000000e+00	3.00000000e+00	5.50000000e+00
2.10000000e+00]			
[1.14000000e+02	5.70000000e+00	2.50000000e+00	5.00000000e+00
2.00000000e+00]			

[1.15000000e+02 2.40000000e+00]	5.80000000e+00	2.80000000e+00	5.10000000e+00
[1.16000000e+02 2.30000000e+00]	6.40000000e+00	3.20000000e+00	5.30000000e+00
[1.17000000e+02 1.80000000e+00]	6.50000000e+00	3.00000000e+00	5.50000000e+00
[1.18000000e+02 2.20000000e+00]	7.70000000e+00	3.80000000e+00	6.70000000e+00
[1.19000000e+02 2.30000000e+00]	7.70000000e+00	2.60000000e+00	6.90000000e+00
[1.20000000e+02 1.50000000e+00]	6.00000000e+00	2.20000000e+00	5.00000000e+00
[1.21000000e+02 2.30000000e+00]	6.90000000e+00	3.20000000e+00	5.70000000e+00
[1.22000000e+02 2.00000000e+00]	5.60000000e+00	2.80000000e+00	4.90000000e+00
[1.23000000e+02 2.00000000e+00]	7.70000000e+00	2.80000000e+00	6.70000000e+00
[1.24000000e+02 1.80000000e+00]	6.30000000e+00	2.70000000e+00	4.90000000e+00
[1.25000000e+02 2.10000000e+00]	6.70000000e+00	3.30000000e+00	5.70000000e+00
[1.26000000e+02 1.80000000e+00]	7.20000000e+00	3.20000000e+00	6.00000000e+00
[1.27000000e+02 1.80000000e+00]	6.20000000e+00	2.80000000e+00	4.80000000e+00
[1.28000000e+02 1.80000000e+00]	6.10000000e+00	3.00000000e+00	4.90000000e+00
[1.29000000e+02 2.10000000e+00]	6.40000000e+00	2.80000000e+00	5.60000000e+00
[1.30000000e+02 1.60000000e+00]	7.20000000e+00	3.00000000e+00	5.80000000e+00
[1.31000000e+02 1.90000000e+00]	7.40000000e+00	2.80000000e+00	6.10000000e+00
[1.32000000e+02 2.00000000e+00]	7.90000000e+00	3.80000000e+00	6.40000000e+00
[1.33000000e+02 2.20000000e+00]	6.40000000e+00	2.80000000e+00	5.60000000e+00
[1.34000000e+02 1.50000000e+00]	6.30000000e+00	2.80000000e+00	5.10000000e+00
[1.35000000e+02 1.40000000e+00]	6.10000000e+00	2.60000000e+00	5.60000000e+00
[1.36000000e+02 2.30000000e+00]	7.70000000e+00	3.00000000e+00	6.10000000e+00
[1.37000000e+02 2.40000000e+00]	6.30000000e+00	3.40000000e+00	5.60000000e+00
[1.38000000e+02 1.80000000e+00]	6.40000000e+00	3.10000000e+00	5.50000000e+00
[1.39000000e+02 1.80000000e+00]	6.00000000e+00	3.00000000e+00	4.80000000e+00
[1.40000000e+02 2.10000000e+00]	6.90000000e+00	3.10000000e+00	5.40000000e+00
[1.41000000e+02 2.40000000e+00]	6.70000000e+00	3.10000000e+00	5.60000000e+00
[1.42000000e+02 2.30000000e+00]	6.90000000e+00	3.10000000e+00	5.10000000e+00
[1.43000000e+02	5.80000000e+00	2.70000000e+00	5.10000000e+00

```

1.90000000e+00]
[ 1.44000000e+02  6.80000000e+00  3.20000000e+00  5.90000000e+00
 2.30000000e+00]
[ 1.45000000e+02  6.70000000e+00  3.30000000e+00  5.70000000e+00
 2.50000000e+00]
[ 1.46000000e+02  6.70000000e+00  3.00000000e+00  5.20000000e+00
 2.30000000e+00]
[ 1.47000000e+02  6.30000000e+00  2.50000000e+00  5.00000000e+00
 1.90000000e+00]
[ 1.48000000e+02  6.50000000e+00  3.00000000e+00  5.20000000e+00
 2.00000000e+00]
[ 1.49000000e+02  6.20000000e+00  3.40000000e+00  5.40000000e+00
 2.30000000e+00]
[ 1.50000000e+02  5.90000000e+00  3.00000000e+00  5.10000000e+00
 1.80000000e+00]]

```

```
In [14]: y = iris.iloc[:, -1].values
print(y)
```

```

[0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
 0 0 0 0 0 0 0 0 0 0 0 0 0 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1
 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 2 2 2 2 2 2 2 2 2
 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2
 2 2]

```

```
In [15]: x = iris.drop(['SepalLengthCm'], axis = 1)
y = iris['SepalLengthCm']
```

```
In [16]: from sklearn.model_selection import train_test_split
from sklearn import metrics
from sklearn.metrics import accuracy_score
```

```
In [26]: def train_test_rmse(x,y):
    x = iris_data[x]
    y = iris_data[y]
    x_train,X_test,y_train,y_test= train_test_split(x,y,test_size=0.2,random_s
tate=4)
    linreg = LinearRegression()
    linreg.fit(x_train,y_train)
    y_pred = linreg.predict(x_test)
    print(accuracy_score(y_test,y_pred))
    return np.sqrt(metrics.mean_squared_error(y_test,y_pred))
```

```
In [30]: import numpy as np
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_s
tate=4)
```

```
In [31]: X = np.array([[ -1, -1], [-2, -1], [-3, -2], [1, 1], [2, 1], [3, 2]])
Y = np.array([1, 1, 1, 2, 2, 2])
```

```
In [46]: from sklearn.naive_bayes import GaussianNB  
gaussian = GaussianNB()  
gaussian.fit(X,Y)
```

```
Out[46]: GaussianNB(priors=None)
```

```
In [47]: Y_pred = gaussian.predict(X_test)
```

```
In [48]: from sklearn.metrics import confusion_matrix,accuracy_score,precision_score,recall_score  
y_true = [2,0,2,2,0,1]  
y_pred = [0,0,2,2,0,2]  
confusion_matrix(y_true,y_pred)
```

```
Out[48]: array([[2, 0, 0],  
               [0, 0, 1],  
               [1, 0, 2]])
```

```
In [49]: ac = accuracy_score(y_test,Y_pred)  
print(ac)
```

```
1.0
```

```
In [50]: precision = precision_score(y_test,Y_pred,average='micro')  
print(precision)
```

```
1.0
```

```
In [45]: recall = recall_score(y_test, Y_pred, average='micro')  
print(recall)
```

```
1.0
```