

devl4a

November 13, 2024

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

```
[3]: df=pd.read_csv("/content/Iris.csv")
df
```

```
[3]:      Id  SepalLengthCm  SepalWidthCm  PetalLengthCm  PetalWidthCm  \
0      1           5.1           3.5           1.4           0.2
1      2           4.9           3.0           1.4           0.2
2      3           4.7           3.2           1.3           0.2
3      4           4.6           3.1           1.5           0.2
4      5           5.0           3.6           1.4           0.2
..    ...           ...           ...           ...           ...
145   146           6.7           3.0           5.2           2.3
146   147           6.3           2.5           5.0           1.9
147   148           6.5           3.0           5.2           2.0
148   149           6.2           3.4           5.4           2.3
149   150           5.9           3.0           5.1           1.8
```

```
      Species
0      Iris-setosa
1      Iris-setosa
2      Iris-setosa
3      Iris-setosa
4      Iris-setosa
..    ...
145  Iris-virginica
146  Iris-virginica
147  Iris-virginica
148  Iris-virginica
149  Iris-virginica
```

[150 rows x 6 columns]

```
[4]: df.info()
df.drop("Id",axis=1,inplace=True)
```

<class 'pandas.core.frame.DataFrame'>

```

RangeIndex: 150 entries, 0 to 149
Data columns (total 6 columns):
#   Column          Non-Null Count  Dtype
---  -
0   Id               150 non-null   int64
1   SepalLengthCm    150 non-null   float64
2   SepalWidthCm     150 non-null   float64
3   PetalLengthCm    150 non-null   float64
4   PetalWidthCm     150 non-null   float64
5   Species          150 non-null   object
dtypes: float64(4), int64(1), object(1)
memory usage: 7.2+ KB

```

```
[5]: df.describe()
```

```

[5]:      SepalLengthCm  SepalWidthCm  PetalLengthCm  PetalWidthCm
count      150.000000      150.000000      150.000000      150.000000
mean         5.843333         3.054000         3.758667         1.198667
std          0.828066         0.433594         1.764420         0.763161
min          4.300000         2.000000         1.000000         0.100000
25%          5.100000         2.800000         1.600000         0.300000
50%          5.800000         3.000000         4.350000         1.300000
75%          6.400000         3.300000         5.100000         1.800000
max          7.900000         4.400000         6.900000         2.500000

```

```
[6]: df.dtypes
```

```

[6]: SepalLengthCm    float64
     SepalWidthCm    float64
     PetalLengthCm    float64
     PetalWidthCm    float64
     Species         object
     dtype: object

```

```
[7]: df.isnull().sum()
```

```

[7]: SepalLengthCm    0
     SepalWidthCm    0
     PetalLengthCm    0
     PetalWidthCm    0
     Species         0
     dtype: int64

```

```

[8]: # Determine X and Y
     x=df.iloc[:,0:4].values
     y=df.iloc[:,4].values

```

```
[9]: #Label Encoding
from sklearn.preprocessing import LabelEncoder
le=LabelEncoder()
y=le.fit_transform(y)
```

```
[11]: #Accuracy Prediction
from sklearn.metrics import
    classification_report,confusion_matrix,accuracy_score,precision_score,recall_score,f1_score
#Model Selection
from sklearn.model_selection import train_test_split
from sklearn.naive_bayes import GaussianNB
```

```
[14]: #Training and Testing the model
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.3,random_state=0)
gaussian=GaussianNB()
gaussian.fit(x_train,y_train)
y_pred=gaussian.predict(x_test)
accuracyscore=round(accuracy_score(y_test,y_pred)*100,2)
print(accuracyscore)
precisionscore=precision_score(y_test,y_pred,average='weighted')*100
print(precisionscore)
recallscore=recall_score(y_test,y_pred,average='weighted')*100
print(recallscore)
f1score=f1_score(y_test,y_pred,average='weighted')*100
print(f1score)
conf_mat=confusion_matrix(y_test,y_pred)
print(conf_mat)
class_report=classification_report(y_test,y_pred)
print(class_report)
```

```
100.0
100.0
100.0
100.0
```

```
[[16  0  0]
 [ 0 18  0]
 [ 0  0 11]]
```

	precision	recall	f1-score	support
0	1.00	1.00	1.00	16
1	1.00	1.00	1.00	18
2	1.00	1.00	1.00	11
accuracy			1.00	45
macro avg	1.00	1.00	1.00	45
weighted avg	1.00	1.00	1.00	45

[]: