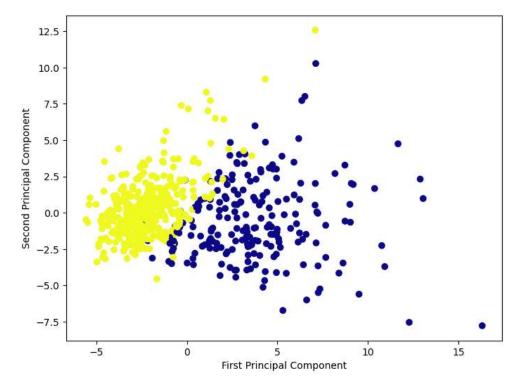
```
In [1]:
               import pandas as pd
               import numpy as np
               import matplotlib.pyplot as plt
               import seaborn as sns
              %matplotlib inline
 In [4]:
               from sklearn.datasets import load_breast_cancer
            3
               cancer=load_breast_cancer()
               df=pd.DataFrame(cancer['data'],columns=cancer['feature_names'])
            5
               df.head()
 Out[4]:
                                                                                 mean
                                                                                                     mean
              mean
                      mean
                               mean
                                                                        mean
                                                                                           mean
                                                                                                               worst
                                                                                                                      worst
                                                                                                                                worst
                                                                                                                                       worst
                                      mean
                                                  mean
                                                               mean
                                                                              concave
                                                                                                    fractal
                                       area smoothness
              radius texture
                                                                                                              radius texture perimeter
                           perimeter
                                                        compactness
                                                                     concavity
                                                                                       symmetry
                                                                                                 dimension
                                                                                points
              17.99
                      10.38
                              122.80 1001.0
                                                0.11840
                                                             0.27760
                                                                               0.14710
                                                                                          0.2419
                                                                                                               25.38
                                                                                                                       17.33
                                                                                                                               184.60 2019.0
                                                                       0.3001
                                                                                                   0.07871
              20.57
                      17.77
                               132.90 1326.0
                                                0.08474
                                                             0.07864
                                                                       0.0869
                                                                               0.07017
                                                                                          0.1812
                                                                                                   0.05667
                                                                                                               24.99
                                                                                                                       23.41
                                                                                                                               158.80
                                                                                                                                      1956.0
               19.69
                      21.25
                               130.00 1203.0
                                                0.10960
                                                             0.15990
                                                                        0.1974
                                                                               0.12790
                                                                                          0.2069
                                                                                                    0.05999
                                                                                                               23.57
                                                                                                                       25.53
                                                                                                                               152.50
                                                                                                                                     1709.0
               11.42
                      20.38
                               77.58
                                      386.1
                                                0.14250
                                                             0.28390
                                                                       0.2414
                                                                               0.10520
                                                                                          0.2597
                                                                                                   0.09744 ...
                                                                                                               14.91
                                                                                                                       26.50
                                                                                                                                98.87
                                                                                                                                       567.7
              20.29
                      14.34
                               135.10 1297.0
                                                0.10030
                                                             0.13280
                                                                       0.1980 0.10430
                                                                                          0.1809
                                                                                                   0.05883 ... 22.54
                                                                                                                       16.67
                                                                                                                               152.20 1575.0
          5 rows × 30 columns
 In [5]:
               from sklearn.preprocessing import StandardScaler
               scaler=StandardScaler()
            3
            4
            5
               scaler.fit(df)
              scaled_data=scaler.transform(df)
 In [6]: 1 scaled_data
 Out[6]: array([[ 1.09706398, -2.07333501, 1.26993369, ..., 2.29607613,
                    2.75062224, 1.93701461],
                                                1.68595471, ..., 1.0870843,
                  [\ 1.82982061,\ -0.35363241,
                   -0.24388967, 0.28118999],
                  [1.57988811, 0.45618695, 1.56650313, ..., 1.95500035,
                    1.152255 , 0.20139121],
                  [ 0.70228425, 2.0455738
                                               0.67267578, ..., 0.41406869,
                   -1.10454895, -0.31840916],
                  [\ 1.83834103,\ 2.33645719,\ 1.98252415,\ \ldots,\ 2.28998549,
                    1.91908301, 2.21963528],
                  [-1.80840125, 1.22179204, -1.81438851, ..., -1.74506282,
                   -0.04813821, -0.75120669]])
In [18]:
            1 from sklearn.decomposition import PCA
            3
               pca=PCA(n_components=2)
            4
               pca.fit(scaled_data)
            5
              x_pca=pca.transform(scaled_data)
            6 x_pca.shape
Out[18]: (569, 2)
In [19]: 1 x pca
Out[19]: array([[ 9.19283683, 1.94858307],
                  [ 2.3878018 , -3.76817174], [ 5.73389628, -1.0751738 ],
                  [ 1.25617928, -1.90229671],
                  [10.37479406, 1.67201011],
[-5.4752433 , -0.67063679]])
```

Out[20]: Text(0, 0.5, 'Second Principal Component')



In [22]: 1 dataset

Out[22]:		sepal-length	sepal-width	petal-length	petal-width	Class
·	0	5.1	3.5	1.4	0.2	Iris-setosa
	1	4.9	3.0	1.4	0.2	Iris-setosa
	2	4.7	3.2	1.3	0.2	Iris-setosa
	3	4.6	3.1	1.5	0.2	Iris-setosa
	4	5.0	3.6	1.4	0.2	Iris-setosa
	145	6.7	3.0	5.2	2.3	Iris-virginica
	146	6.3	2.5	5.0	1.9	Iris-virginica
	147	6.5	3.0	5.2	2.0	Iris-virginica

3.4

3.0

5 X_test = lda.transform(X_test)

5.4

5.1

150 rows × 5 columns

6.2

5.9

148

149

```
In [23]:
            1
               X=dataset.iloc[:,0:4].values
               y=dataset.iloc[:,4].values
In [24]:
            1 from sklearn.model_selection import train_test_split
               X_train,X_test,y_train,y_test=train_test_split(X,y,test_size=0.2,random_state=1)
In [25]:
               sc=StandardScaler()
               X_train=sc.fit_transform(X_train)
            3 X_test=sc.transform(X_test)
In [29]:
               {\bf from} \  \  {\bf sklearn.discriminant\_analysis} \  \  {\bf import} \  \  {\bf Linear DiscriminantAnalysis} \  \  {\bf as} \  \  {\bf LDA}
            1
            3
               lda = LDA(n_components=1)
            4 X_train = lda.fit_transform(X_train, y_train)
```

2.3 Iris-virginica

1.8 Iris-virginica

```
In [30]:
          1 lda
Out[30]: LinearDiscriminantAnalysis(n_components=1)
In [31]: 1 X_train
Out[31]: array([[-2.1560854],
                [-6.11662239],
               [-0.90912129],
                [-6.26379402],
                [-6.44115272],
                [ 8.56164613],
                [-2.01994399],
               [ 8.31277554],
               [-2.13442116],
               [-6.07270723],
               [-5.24089883],
                [ 8.07646049],
               [-4.56862701],
               [-5.558475],
               [-1.40925096],
               [-8.69494915],
                [ 8.44562193],
                [ 8.31096247],
               [ 7.0599544 ],
In [32]: 1 X_test
Out[32]: array([[10.35675286],
                 0.78309218],
                [-1.00007779],
               [ 8.98976393],
               [-4.91813822],
               [-2.25040201],
               [-4.42164596],
               [ 6.92657595],
               [ 6.90332131],
               [-5.84849778],
               [-1.90327928],
                [ 7.95404718],
                [-5.89372605],
               [-1.49622857],
                [-2.38691789],
               [ 7.29721934],
                [-1.021476],
                [-2.77071396],
               [ 7.838289 ],
In [33]:
          1 from sklearn.ensemble import RandomForestClassifier
          3 classifier=RandomForestClassifier(max_depth=2,random_state=0)
          4 classifier.fit(X_train,y_train)
          5 y_pred=classifier.predict(X_test)
In [34]: 1 y_pred
In [35]:
          1 from sklearn.metrics import accuracy_score
          3 accuracy_score(y_test,y_pred)
Out[35]: 1.0
 In [ ]: 1
```