```
In [1]:
        import numpy as np
        import pandas as pd
        import matplotlib.pyplot as plt
        from sklearn.datasets import load breast cancer
        cancer = load breast cancer()
In [2]: cancer.keys()
Out[2]: dict_keys(['data', 'target', 'target_names', 'DESCR', 'feature_names'])
In [3]: | cancer["data"]
Out[3]: array([[1.799e+01, 1.038e+01, 1.228e+02, ..., 2.654e-01, 4.601e-01,
               1.189e-01],
              [2.057e+01, 1.777e+01, 1.329e+02, ..., 1.860e-01, 2.750e-01,
               8.902e-02],
              [1.969e+01, 2.125e+01, 1.300e+02, ..., 2.430e-01, 3.613e-01,
               8.758e-021,
              [1.660e+01, 2.808e+01, 1.083e+02, ..., 1.418e-01, 2.218e-01,
               7.820e-021,
              [2.060e+01, 2.933e+01, 1.401e+02, ..., 2.650e-01, 4.087e-01,
               1.240e-01],
              [7.760e+00, 2.454e+01, 4.792e+01, ..., 0.000e+00, 2.871e-01,
               7.039e-0211)
In [4]: | cancer["target"]
0, 0, 1, 0, 1, 1, 1, 1, 0, 0, 1, 0, 0, 1, 1, 1, 1, 0, 1, 0, 0,
              1, 1, 1, 1, 0, 1, 0, 0, 1, 0, 1, 0, 0, 1, 1, 1, 0, 0, 1, 0, 0, 0,
              1, 1, 1, 0, 1, 1, 0, 0, 1, 1, 1, 0, 0, 1, 1, 1, 1, 0, 1, 1, 0, 1,
              1, 1, 1, 1, 1, 1, 0, 0, 0, 1, 0, 0, 1, 1, 1, 0, 0, 1, 0, 1, 0,
              0, 1, 0, 0, 1, 1, 0, 1, 1, 0, 1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1, 1,
              1, 1, 0, 1, 1, 1, 1, 0,
                                    0, 1, 0, 1, 1, 0, 0, 1, 1, 0, 0, 1, 1, 1,
              1, 0, 1, 1, 0, 0, 0, 1, 0, 1, 0, 1, 1, 1, 0, 1, 1, 0, 0, 1, 0, 0,
              0, 0, 1, 0, 0, 0, 1, 0, 1, 0, 1, 1, 0, 1, 0, 0, 0, 0, 1, 1, 0, 0,
              1, 1, 1, 0, 1, 1, 1, 1, 1, 0, 0, 1, 1, 0, 1, 1, 0, 0, 1, 0, 1, 1,
              1, 1, 0, 1, 1, 1, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
              0, 0, 1, 1, 1, 1, 1, 1, 0, 1, 0, 1, 1, 0, 1, 1, 0, 1, 0, 0, 1, 1,
              1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 0, 1, 0, 1, 1, 1, 1, 1,
              1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 1, 0, 1, 0, 1, 1, 1, 1, 1, 0, 0,
              0, 1, 1, 1, 1, 0, 1, 0, 1, 0, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1, 0,
              0, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 0, 1, 0, 0, 0, 1, 0, 0,
              1, 1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 0, 1, 1, 0, 1, 1, 0, 0, 1, 1,
              1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 0, 1, 1, 0,
              1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 0, 0, 1, 0, 1, 1, 1, 1,
              1, 0, 1, 1, 0, 1, 0, 1, 1, 0, 1, 0, 1, 1, 1, 1, 1, 1, 1, 1, 0, 0,
              1, 1, 1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 1, 1, 1,
              1, 1, 1, 0, 1, 0, 1, 1, 0, 1, 1, 1, 1, 1, 0, 0, 1, 0, 1, 0, 1, 1,
              1, 1, 1, 0, 1, 1, 0, 1, 0, 1, 0, 0, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1,
              1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 0, 0, 0, 0, 0, 1
```

```
In [5]: print(cancer['DESCR']) #dataset description
        Breast Cancer Wisconsin (Diagnostic) Database
        _____
        Notes
         _ _ _ _ _
        Data Set Characteristics:
             :Number of Instances: 569
             :Number of Attributes: 30 numeric, predictive attributes and the class
             :Attribute Information:
                 - radius (mean of distances from center to points on the perimeter)

    texture (standard deviation of gray-scale values)

                 - perimeter
                 - area
                 - smoothness (local variation in radius lengths)
                 - compactness (perimeter^2 / area - 1.0)
                 - concavity (severity of concave portions of the contour)
                 - concave points (number of concave portions of the contour)
In [6]: x = cancer['data']
         y = cancer['target']
In [7]: | x_df = pd.DataFrame(x)
         x df.head()
Out[7]:
               0
                           2
                                  3
                                                 5
                                                               7
                                                                      8
                                                                                     20
                                                                                           2
         0 17.99 10.38 122.80 1001.0 0.11840 0.27760 0.3001 0.14710 0.2419 0.07871 ...
                                                                                   25.38
                                                                                        17.3
         1 20.57 17.77 132.90 1326.0 0.08474 0.07864 0.0869 0.07017 0.1812 0.05667
         2 19.69 21.25 130.00 1203.0 0.10960 0.15990 0.1974 0.12790 0.2069
                                                                                        25.5
                                                                        0.05999
                                                                                   23.57
           11.42 20.38
                        77.58
                               386.1 0.14250 0.28390 0.2414 0.10520 0.2597
                                                                        0.09744
                                                                                   14.91
                                                                                        26.5
           20.29 14.34 135.10 1297.0 0.10030 0.13280 0.1980 0.10430 0.1809 0.05883 ...
                                                                                   22.54
                                                                                        16.6
        5 rows × 30 columns
```

```
In [8]: y df = pd.DataFrame(y)
          y_df.tail(10)
 Out[8]:
               0
           559
               1
           560
               1
           561 1
           562 0
           563 0
           564
               0
           565 0
           566 0
           567 0
           568 1
 In [9]: from sklearn.model selection import train test split
          x_train,x_test,y_train,y_test=train_test_split(x,y)
In [10]: from sklearn.preprocessing import StandardScaler
          scalar = StandardScaler()
          scalar.fit(x_train)
Out[10]: StandardScaler(copy=True, with mean=True, with std=True)
In [11]: x train=scalar.transform(x train)
          x test=scalar.transform(x test)
          x train df = pd.DataFrame(x train)
In [12]:
          x train df.head()
Out[12]:
                    0
                              1
                                                                                                  8
                                        2
                                                 3
                                                                     5
                                                                                        7
             -0.269022 -1.056637
                                -0.323814 -0.317785 -1.628520 -1.021009
                                                                       -0.713341
                                                                                 -0.598087
                                                                                          -1.596108
             -0.641466
                       -1.009633
                                 -0.631010 -0.672462
                                                     1.341373
                                                              0.268735
                                                                       -0.977790
                                                                                 -0.818031
                                                                                           2.414279
              1.796348
                        0.466295
                                  1.846605
                                           1.862197
                                                     0.828232
                                                              1.075170
                                                                        1.642538
                                                                                  1.589084
                                                                                           0.130639
           3
              1.460584
                        0.997441
                                                     0.420531
                                                              0.916829
                                                                                           0.616520
                                  1.490731
                                           1.394025
                                                                        1.466660
                                                                                  0.995416
              2.197007
                        0.621408
                                  2.231111
                                           2.321858
                                                     0.673587
                                                              1.645934
                                                                                 2.559798
                                                                                           0.040938
                                                                        1.971518
          5 rows × 30 columns
In [13]: | from sklearn.neural_network import MLPClassifier #multilevel perceptron
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In [14]: | mlp = MLPClassifier(hidden layer sizes=(30,30,30,30,30,30,30,30))
In [15]: | mlp.fit(x_train,y_train) #training of neural network model
Out[15]: MLPClassifier(activation='relu', alpha=0.0001, batch_size='auto', beta_1=0.9,
                beta 2=0.999, early stopping=False, epsilon=1e-08,
                hidden_layer_sizes=(30, 30, 30, 30, 30, 30, 30, 30),
                learning_rate='constant', learning_rate_init=0.001, max_iter=200,
                momentum=0.9, nesterovs momentum=True, power t=0.5,
                random state=None, shuffle=True, solver='adam', tol=0.0001,
                validation_fraction=0.1, verbose=False, warm_start=False)
In [16]: | predictions = mlp.predict(x test)
In [17]: from sklearn.metrics import confusion matrix
         print(confusion_matrix(y_test,predictions))
         [[48 2]
          [ 1 92]]
In [18]: from sklearn.metrics import accuracy score
         accuracy nn=round(accuracy score(predictions,y test)*100,2)
         print("Accuracy of this model is ",accuracy nn,"%")
         Accuracy of this model is 97.9 %
In [19]: from sklearn.ensemble import RandomForestClassifier
         model rand=RandomForestClassifier(n estimators=100)
         model_rand.fit(x_train,y_train)
         predicted_rand=model_rand.predict(x_test)
         from sklearn.metrics import confusion matrix
         print(confusion_matrix(predicted_rand,y_test))
         from sklearn.metrics import accuracy_score
         accuracy=round(accuracy score(predicted rand,y test)*100,2)
         print("Accuracy of this model is ",accuracy,"%")
         [[45 1]
          [ 5 92]]
         Accuracy of this model is 95.8 %
In [ ]:
In [ ]:
In [ ]:
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In]:	
In]:	
In	[]:	