In [1]:

1 import pandas as pd
2 import numpy as np
3 import matplotlib.pyplot as plt
4 import seaborn as sns
5 %matplotlib inline

In [34]: 1 print(cancer.DESCR)

Optimization Methods and Software 1, 1992, 23-34].

This database is also available through the UW CS ftp server:

ftp ftp.cs.wisc.edu
cd math-prog/cpo-dataset/machine-learn/WDBC/

.. topic:: References

- W.N. Street, W.H. Wolberg and O.L. Mangasarian. Nuclear feature extraction for breast tumor diagnosis. IS&T/SPIE 1993 International Symposium on Electronic Imaging: Science and Technology, volume 1905, pages 861-870, San Jose, CA, 1993.
- O.L. Mangasarian, W.N. Street and W.H. Wolberg. Breast cancer diagnosis and prognosis via linear programming. Operations Research, 43(4), pages 570-577, July-August 1995.
- W.H. Wolberg, W.N. Street, and O.L. Mangasarian. Machine learning techniques to diagnose breast cancer from fine-needle aspirates. Cancer Letters 77 (1994) 163-171.

Out[2]:

| mean ncavity | mean concave points | mean symmetry | mean fractal dimension | worst radius | worst texture | worst perimeter | worst area | worst smoothness | worst compactness | worst concavity | worst concave points | worst symmetry | worst fractal dimension |
|-----------------|---------------------------|------------------|------------------------------|---------------------|------------------|--------------------|---------------|---------------------|----------------------|--------------------|----------------------------|-------------------|-------------------------------|
| 0.3001 | 0.14710 | 0.2419 | 0.07871 | 25.38 | 17.33 | 184.60 | 2019.0 | 0.1622 | 0.6656 | 0.7119 | 0.2654 | 0.4601 | 0.11890 |
| 0.0869 | 0.07017 | 0.1812 | 0.05667 | 24.99 | 23.41 | 158.80 | 1956.0 | 0.1238 | 0.1866 | 0.2416 | 0.1860 | 0.2750 | 0.08902 |
| 0.1974 | 0.12790 | 0.2069 | 0.05999 | 23.57 | 25.53 | 152.50 | 1709.0 | 0.1444 | 0.4245 | 0.4504 | 0.2430 | 0.3613 | 0.08758 |
| 0.2414 | 0.10520 | 0.2597 | 0.09744 | 14.91 | 26.50 | 98.87 | 567.7 | 0.2098 | 0.8663 | 0.6869 | 0.2575 | 0.6638 | 0.17300 |
| 0.1980 | 0.10430 | 0.1809 | 0.05883 | 22.54 | 16.67 | 152.20 | 1575.0 | 0.1374 | 0.2050 | 0.4000 | 0.1625 | 0.2364 | 0.07678 |

Testing Accuracy:

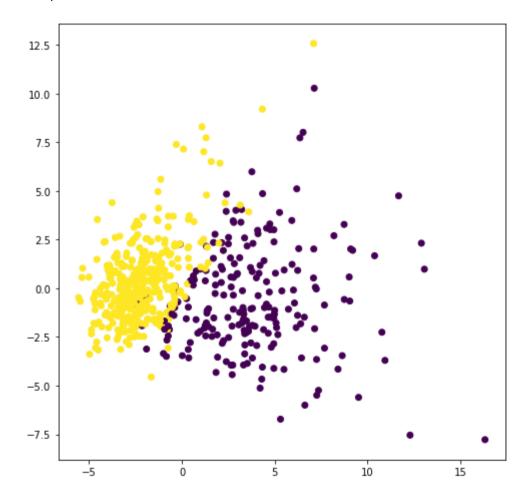
| | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| 0 | 0.96 | 0.98 | 0.97 | 54 |
| 1 | 0.99 | 0.98 | 0.98 | 89 |
| accuracy | | | 0.98 | 143 |
| macro avg | 0.98 | 0.98 | 0.98 | 143 |
| weighted avg | 0.98 | 0.98 | 0.98 | 143 |

Training Accuracy:

| | precision | recall | f1-score | support | |
|--------------|-----------|--------|----------|---------|--|
| 0 | 0.98 | 0.99 | 0.99 | 155 | |
| 1 | 1.00 | 0.99 | 0.99 | 271 | |
| accuracy | | | 0.99 | 426 | |
| macro avg | 0.99 | 0.99 | 0.99 | 426 | |
| weighted avg | 0.99 | 0.99 | 0.99 | 426 | |

```
In [17]:
           1 from sklearn.preprocessing import StandardScaler
              scaler = StandardScaler()
              scaler.fit(df)
Out[17]: StandardScaler()
In [19]:
           1 scaled data = scaler.transform(df)
           1 from sklearn.decomposition import PCA
In [20]:
In [40]:
           1 pca = PCA(n components=4)
           1 pca_df = pca.fit_transform(df)
In [41]:
           pca df = pd.DataFrame(pca df,columns = ['PC1','PC2','PC3','PC4'])
In [42]:
           1 pca df.head()
In [43]:
Out[43]:
                 PC1
                          PC2
                                   PC3
                                           PC4
          0 9.192837
                      1.948583 -1.123167 3.633731
            2.387802 -3.768172 -0.529292
                                       1.118264
          2 5.733896 -1.075174 -0.551747 0.912083
          3 7.122953 10.275589 -3.232790 0.152547
          4 3.935302 -1.948072 1.389767 2.940640
```

Out[44]: <matplotlib.collections.PathCollection at 0x20ef8dd4508>



```
1 pca.explained variance ratio .sum()
In [45]:
Out[45]: 0.7923850582446045
In [51]:
           1 from sklearn.linear model import LogisticRegression
           2 from sklearn.model selection import train test split
           3 from sklearn.metrics import classification report
             model = LogisticRegression()
             x train,x test,y train,y test = train test split(pca df,cancer['target'],random state=1)
             model.fit(x train,y train)
              print("Testing Accuracy : \n",classification report(model.predict(x test),y test),"\n")
          10
          11
          12 print("Training Accuracy : \n", classification report(model.predict(x train), y train))
         Testing Accuracy:
                        precision
                                      recall f1-score
                                                         support
                             0.93
                                                 0.94
                                                             53
                    0
                                       0.96
                                                 0.97
                                                             90
                    1
                             0.98
                                       0.96
                                                 0.96
                                                            143
             accuracy
                                                 0.96
                            0.95
            macro avg
                                       0.96
                                                            143
         weighted avg
                             0.96
                                       0.96
                                                 0.96
                                                            143
         Training Accuracy:
                        precision
                                     recall f1-score
                                                         support
                    0
                             0.96
                                       0.97
                                                 0.96
                                                            154
                            0.99
                                                 0.98
                    1
                                       0.97
                                                            272
                                                 0.97
                                                            426
             accuracy
```

426

426

0.97

0.97

0.97

0.97

0.97

0.97

macro avg

weighted avg

In []: 1