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In [1]:
        import pandas as pd
         import numpy as np
         import matplotlib.pyplot as plt
         import seaborn as sns
         %matplotlib inline
In [2]: | from sklearn.datasets import load_breast_cancer
In [3]: | cancer = load_breast_cancer()
In [5]: | cancer.keys()
Out[5]: dict_keys(['data', 'target', 'target_names', 'DESCR', 'feature_names'])
In [6]: print(cancer['DESCR'])
        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)
                 - symmetry
                 - fractal dimension ("coastline approximation" - 1)
In [7]: | df_feat = pd.DataFrame(cancer['data'],columns=cancer['feature_names'])
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In [8]: df feat.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 569 entries, 0 to 568
         Data columns (total 30 columns):
                                     569 non-null float64
         mean radius
         mean texture
                                     569 non-null float64
                                     569 non-null float64
         mean perimeter
                                     569 non-null float64
         mean area
                                     569 non-null float64
         mean smoothness
         mean compactness
                                     569 non-null float64
                                     569 non-null float64
         mean concavity
         mean concave points
                                     569 non-null float64
         mean symmetry
                                     569 non-null float64
         mean fractal dimension
                                     569 non-null float64
                                     569 non-null float64
         radius error
         texture error
                                     569 non-null float64
                                     569 non-null float64
         perimeter error
         area error
                                     569 non-null float64
         smoothness error
                                     569 non-null float64
                                     569 non-null float64
         compactness error
                                     569 non-null float64
         concavity error
         concave points error
                                     569 non-null float64
                                     569 non-null float64
         symmetry error
         fractal dimension error
                                     569 non-null float64
         worst radius
                                     569 non-null float64
         worst texture
                                     569 non-null float64
                                     569 non-null float64
         worst perimeter
         worst area
                                     569 non-null float64
         worst smoothness
                                     569 non-null float64
         worst compactness
                                     569 non-null float64
                                     569 non-null float64
         worst concavity
                                     569 non-null float64
         worst concave points
                                     569 non-null float64
         worst symmetry
                                     569 non-null float64
         worst fractal dimension
         dtypes: float64(30)
         memory usage: 133.4 KB
In [10]: | cancer['target_names']
Out[10]: array(['malignant', 'benign'], dtype='<U9')</pre>
         from sklearn.cross_validation import train_test_split
In [11]:
In [14]: | X =df feat
         y= cancer['target']
         X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.33, random_
In [15]:
         from sklearn.svm import SVC
         model = SVC()
In [16]:
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In [17]: | model.fit(X_train,y_train)
Out[17]: SVC(C=1.0, cache size=200, class weight=None, coef0=0.0,
           decision_function_shape='ovr', degree=3, gamma='auto', kernel='rbf',
           max_iter=-1, probability=False, random_state=None, shrinking=True,
           tol=0.001, verbose=False)
In [18]:
         predictions = model.predict(X_test)
         from sklearn.metrics import classification report , confusion matrix
In [19]:
         print(confusion_matrix(y_test,predictions))
In [20]:
         print('\n')
         print(classification_report(y_test,predictions))
             0 71]
         ] ]
             0 117]]
                      precision
                                    recall f1-score
                                                       support
                                      0.00
                   0
                           0.00
                                                0.00
                                                            71
                   1
                                      1.00
                                                           117
                           0.62
                                                0.77
         avg / total
                           0.39
                                      0.62
                                                           188
                                                0.48
         C:\Users\q21\Anaconda3\lib\site-packages\sklearn\metrics\classification.py:113
         5: UndefinedMetricWarning: Precision and F-score are ill-defined and being set
         to 0.0 in labels with no predicted samples.
            'precision', 'predicted', average, warn for)
In [21]: # grid allow find right parameters
         from sklearn.grid_search import GridSearchCV
In [22]:
In [23]:
         param_grid = {'C':[0.1,1,10,100], 'gamma':[1,0.1,0.01,0.001]}
In [24]: | grid = GridSearchCV(SVC(),param_grid,verbose =3)
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In [25]: grid.fit(X train,y train)
        [CV] ...... C=100, gamma=0.01, score=0.629921 -
        [CV] C=100, gamma=0.01 ......
        [CV] ...... C=100, gamma=0.01, score=0.629921 -
        [CV] C=100, gamma=0.001 .....
        [CV] ...... C=100, gamma=0.001, score=0.905512 -
        [CV] C=100, gamma=0.001 ......
        [CV] ..... C=100, gamma=0.001, score=0.921260 -
        [CV] C=100, gamma=0.001 ......
        [CV] ...... C=100, gamma=0.001, score=0.937008 - 0.0s
        [Parallel(n jobs=1)]: Done 48 out of 48 | elapsed:
                                                        0.5s finished
Out[25]: GridSearchCV(cv=None, error_score='raise',
              estimator=SVC(C=1.0, cache_size=200, class_weight=None, coef0=0.0,
         decision_function_shape='ovr', degree=3, gamma='auto', kernel='rbf',
         max iter=-1, probability=False, random state=None, shrinking=True,
         tol=0.001, verbose=False),
              fit params={}, iid=True, n jobs=1,
              param_grid={'C': [0.1, 1, 10, 100], 'gamma': [1, 0.1, 0.01, 0.001]},
              pre_dispatch='2*n_jobs', refit=True, scoring=None, verbose=3)
In [26]: grid.best_params_
Out[26]: {'C': 1, 'gamma': 0.001}
In [27]: grid.best estimator
Out[27]: SVC(C=1, cache size=200, class weight=None, coef0=0.0,
         decision_function_shape='ovr', degree=3, gamma=0.001, kernel='rbf',
         max iter=-1, probability=False, random state=None, shrinking=True,
         tol=0.001, verbose=False)
In [28]: | grid_predictions = grid.predict(X_test)
In [29]:
        print(confusion_matrix(y_test,grid_predictions))
        print('\n')
        print(classification_report(y_test,grid_predictions))
        [[ 63
               8]
         [ 7 110]]
                   precision
                              recall f1-score
                                               support
                0
                       0.90
                                0.89
                                         0.89
                                                   71
                       0.93
                                0.94
                                         0.94
                                                   117
        avg / total
                       0.92
                                0.92
                                         0.92
                                                   188
In [ ]:
```