SVM in Binary Calssification

By using SVM to predict whether a tumor is malignant or benign. From the built in breast cancer dataset from Scikit Learn. Challenge: all the data are trained into one class (due to using deafult parameters in svc), using grid search to test combination of parameters

```
In [2]: # data library
import pandas as pd
import numpy as np
# visualization libaray
import matplotlib.pyplot as plt
import seaborn as sns
# since using jupyter notebook
%matplotlib inline
```

Load the Data

Create a data frame of X, y

```
In [19]:
           X = pd.DataFrame(cancer['data'], columns=cancer['feature_names'])
           X.head(5)
Out[19]:
                                                                                          mean
                mean
                        mean
                                   mean
                                           mean
                                                        mean
                                                                      mean
                                                                                 mean
                                                                                                     mean
                                                                                        concave
               radius
                       texture perimeter
                                            area
                                                 smoothness
                                                               compactness
                                                                             concavity
                                                                                                 symmetry
                                                                                          points
            0
                17.99
                         10.38
                                  122.80
                                          1001.0
                                                      0.11840
                                                                    0.27760
                                                                                0.3001
                                                                                                     0.2419
                                                                                        0.14710
                20.57
                         17.77
                                  132.90
                                          1326.0
                                                      0.08474
                                                                    0.07864
                                                                                0.0869
                                                                                        0.07017
                                                                                                    0.1812
            2
                19.69
                         21.25
                                  130.00
                                          1203.0
                                                      0.10960
                                                                    0.15990
                                                                                0.1974
                                                                                        0.12790
                                                                                                    0.2069
                         20.38
                                   77.58
                                           386.1
                                                                                        0.10520
                                                                                                    0.2597
            3
                11.42
                                                      0.14250
                                                                    0.28390
                                                                                0.2414
                20.29
                         14.34
                                  135.10 1297.0
                                                      0.10030
                                                                    0.13280
                                                                                0.1980
                                                                                        0.10430
                                                                                                    0.1809
           5 rows × 30 columns
In [27]:
           y = cancer['target']
           len(y)
Out[27]: 569
In [29]:
           ## statistic of X
           X.describe()
Out[29]:
                        mean
                                    mean
                                                mean
                                                                           mean
                                                                                         mean
                                                                                                     mean
                                                         mean area
                        radius
                                   texture
                                             perimeter
                                                                    smoothness
                                                                                  compactness
                                                                                                 concavity
                   569.000000
                               569.000000
                                           569.000000
                                                         569.000000
                                                                      569.000000
                                                                                    569.000000
                                                                                                569.000000
            count
                                             91.969033
                    14.127292
                                19.289649
                                                         654.889104
                                                                        0.096360
                                                                                      0.104341
                                                                                                  0.088799
            mean
                     3.524049
                                 4.301036
                                             24.298981
                                                         351.914129
                                                                        0.014064
                                                                                      0.052813
                                                                                                  0.079720
              std
                     6.981000
              min
                                 9.710000
                                            43.790000
                                                         143.500000
                                                                        0.052630
                                                                                      0.019380
                                                                                                  0.000000
             25%
                     11.700000
                                16.170000
                                            75.170000
                                                         420.300000
                                                                        0.086370
                                                                                      0.064920
                                                                                                  0.029560
             50%
                    13.370000
                                18.840000
                                            86.240000
                                                         551.100000
                                                                        0.095870
                                                                                      0.092630
                                                                                                  0.061540
             75%
                                                                                                  0.130700
                    15.780000
                                21.800000
                                           104.100000
                                                         782.700000
                                                                        0.105300
                                                                                      0.130400
             max
                    28.110000
                                39.280000
                                           188.500000
                                                       2501.000000
                                                                        0.163400
                                                                                      0.345400
                                                                                                  0.426800
           8 rows × 30 columns
In [31]: | # is there any object type column
            (X.dtypes == 'object').any()
```

Out[31]: False

```
In [33]: from sklearn.model_selection import train_test_split
In [34]: X_train, X_test, y_train, y_test = train_test_split(X,y, test_size=0.30, rando m_state=101)
```

build the model

Using gridsearch for parameters

In [57]: grid.fit(X_train,y_train)

```
Fitting 3 folds for each of 25 candidates, totalling 75 fits
[CV] C=0.1, gamma=1, kernel=rbf ......
[CV] C=0.1, gamma=1, kernel=rbf, score=0.631578947368421, total= 0.0s
[CV] C=0.1, gamma=1, kernel=rbf ......
[CV] C=0.1, gamma=1, kernel=rbf, score=0.631578947368421, total= 0.0s
[CV] C=0.1, gamma=1, kernel=rbf ......
[CV] C=0.1, gamma=1, kernel=rbf, score=0.63636363636364, total= 0.0s
[CV] C=0.1, gamma=0.1, kernel=rbf, score=0.631578947368421, total=
[CV] C=0.1, gamma=0.1, kernel=rbf, score=0.631578947368421, total=
[CV] C=0.1, gamma=0.1, kernel=rbf, score=0.6363636363636364, total=
                                                 0.0s
[CV] C=0.1, gamma=0.01, kernel=rbf, score=0.631578947368421, total=
[CV] C=0.1, gamma=0.01, kernel=rbf ...............................
[CV] C=0.1, gamma=0.01, kernel=rbf, score=0.631578947368421, total=
                                                 0.0s
[CV] C=0.1, gamma=0.01, kernel=rbf, score=0.6363636363636364, total=
                                                  0.0s
[Parallel(n jobs=1)]: Using backend SequentialBackend with 1 concurrent worke
[Parallel(n_jobs=1)]: Done
                    1 out of
                           1 | elapsed:
                                       0.0s remaining:
                                                    0.
[Parallel(n jobs=1)]: Done
                   2 out of
                           2 | elapsed:
                                      0.0s remaining:
                                                    0.
0s
```

```
[CV] C=0.1, gamma=0.001, kernel=rbf ......
[CV] C=0.1, gamma=0.001, kernel=rbf, score=0.631578947368421, total=
                                                      0.0s
[CV] C=0.1, gamma=0.001, kernel=rbf ..............................
[CV] C=0.1, gamma=0.001, kernel=rbf, score=0.631578947368421, total=
                                                      0.0s
[CV] C=0.1, gamma=0.001, kernel=rbf ......
[CV] C=0.1, gamma=0.001, kernel=rbf, score=0.6363636363636364, total=
                                                       0.0s
[CV] C=0.1, gamma=0.0001, kernel=rbf, score=0.9022556390977443, total=
                                                        0.0
[CV] C=0.1, gamma=0.0001, kernel=rbf ......
[CV] C=0.1, gamma=0.0001, kernel=rbf, score=0.9624060150375939, total=
                                                        0.0
[CV] C=0.1, gamma=0.0001, kernel=rbf, score=0.91666666666666666, total=
                                                        0.0
[CV] C=1, gamma=1, kernel=rbf .....
[CV] C=1, gamma=1, kernel=rbf, score=0.631578947368421, total= 0.0s
[CV] C=1, gamma=1, kernel=rbf .....
[CV] C=1, gamma=1, kernel=rbf, score=0.631578947368421, total= 0.0s
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[CV] C=1, gamma=1, kernel=rbf, score=0.6363636363636364, total= 0.0s
[CV] C=1, gamma=0.1, kernel=rbf ......
[CV] C=1, gamma=0.1, kernel=rbf, score=0.631578947368421, total= 0.0s
[CV] C=1, gamma=0.1, kernel=rbf ......
[CV] C=1, gamma=0.1, kernel=rbf, score=0.631578947368421, total= 0.0s
[CV] C=1, gamma=0.1, kernel=rbf ......
[CV] C=1, gamma=0.1, kernel=rbf, score=0.63636363636364, total= 0.0s
[CV] C=1, gamma=0.01, kernel=rbf ......
[CV] C=1, gamma=0.01, kernel=rbf, score=0.631578947368421, total= 0.0s
[CV] C=1, gamma=0.01, kernel=rbf ......
[CV] C=1, gamma=0.01, kernel=rbf, score=0.631578947368421, total= 0.0s
[CV] C=1, gamma=0.01, kernel=rbf ......
[CV] C=1, gamma=0.01, kernel=rbf, score=0.6363636363636364, total=
[CV] C=1, gamma=0.001, kernel=rbf ................................
[CV] C=1, gamma=0.001, kernel=rbf, score=0.9022556390977443, total=
[CV] C=1, gamma=0.001, kernel=rbf, score=0.9398496240601504, total=
                                                      0.0s
[CV] C=1, gamma=0.001, kernel=rbf ................................
[CV] C=1, gamma=0.001, kernel=rbf, score=0.9545454545454546, total=
                                                      0.0s
[CV] C=1, gamma=0.0001, kernel=rbf ...............................
[CV] C=1, gamma=0.0001, kernel=rbf, score=0.9398496240601504, total=
                                                      0.0s
[CV] C=1, gamma=0.0001, kernel=rbf, score=0.9699248120300752, total=
                                                      0.0s
[CV] C=1, gamma=0.0001, kernel=rbf ......
[CV] C=1, gamma=0.0001, kernel=rbf, score=0.946969696969697, total=
[CV] C=10, gamma=1, kernel=rbf .....
[CV] C=10, gamma=1, kernel=rbf, score=0.631578947368421, total= 0.0s
[CV] C=10, gamma=1, kernel=rbf .....
[CV] C=10, gamma=1, kernel=rbf, score=0.631578947368421, total= 0.0s
[CV] C=10, gamma=1, kernel=rbf .....
[CV] C=10, gamma=1, kernel=rbf, score=0.6363636363636364, total= 0.0s
[CV] C=10, gamma=0.1, kernel=rbf ......
[CV] C=10, gamma=0.1, kernel=rbf, score=0.631578947368421, total= 0.0s
[CV] C=10, gamma=0.1, kernel=rbf ......
[CV] C=10, gamma=0.1, kernel=rbf, score=0.631578947368421, total= 0.0s
[CV] C=10, gamma=0.1, kernel=rbf ......
[CV] C=10, gamma=0.1, kernel=rbf, score=0.6363636363636364, total=
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[CV] C=10, gamma=0.01, kernel=rbf, score=0.631578947368421, total=
[CV] C=10, gamma=0.01, kernel=rbf ......
[CV] C=10, gamma=0.01, kernel=rbf, score=0.631578947368421, total=
[CV] C=10, gamma=0.01, kernel=rbf, score=0.636363636363636364, total=
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[CV] C=10, gamma=0.001, kernel=rbf, score=0.8947368421052632, total=
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[CV] C=10, gamma=0.001, kernel=rbf, score=0.9323308270676691, total=
                                               0.0s
0.0s
[CV] C=10, gamma=0.0001, kernel=rbf ......
[CV] C=10, gamma=0.0001, kernel=rbf, score=0.9323308270676691, total=
                                                0.0s
[CV] C=10, gamma=0.0001, kernel=rbf, score=0.9699248120300752, total=
                                                0.0s
[CV] C=10, gamma=0.0001, kernel=rbf ......
[CV] C=10, gamma=0.0001, kernel=rbf, score=0.9621212121212122, total=
                                                0.0s
[CV] C=100, gamma=1, kernel=rbf ......
[CV] C=100, gamma=1, kernel=rbf, score=0.631578947368421, total= 0.0s
[CV] C=100, gamma=1, kernel=rbf ......
[CV] C=100, gamma=1, kernel=rbf, score=0.631578947368421, total= 0.0s
[CV] C=100, gamma=1, kernel=rbf ......
[CV] C=100, gamma=1, kernel=rbf, score=0.63636363636364, total= 0.0s
[CV] C=100, gamma=0.1, kernel=rbf ................................
[CV] C=100, gamma=0.1, kernel=rbf, score=0.631578947368421, total=
[CV] C=100, gamma=0.1, kernel=rbf, score=0.631578947368421, total=
[CV] C=100, gamma=0.1, kernel=rbf, score=0.6363636363636364, total=
                                              0.0s
[CV] C=100, gamma=0.01, kernel=rbf ................................
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[CV] C=100, gamma=0.01, kernel=rbf, score=0.631578947368421, total=
[CV] C=100, gamma=0.01, kernel=rbf, score=0.6363636363636364, total=
                                               0.0s
[CV] C=100, gamma=0.001, kernel=rbf ............
[CV] C=100, gamma=0.001, kernel=rbf, score=0.8947368421052632, total=
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[CV] C=100, gamma=0.001, kernel=rbf .......
[CV] C=100, gamma=0.001, kernel=rbf, score=0.9323308270676691, total=
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[CV] C=100, gamma=0.0001, kernel=rbf ..............................
[CV] C=100, gamma=0.0001, kernel=rbf, score=0.9172932330827067, total=
                                                0.0
[CV] C=100, gamma=0.0001, kernel=rbf, score=0.9774436090225563, total=
                                                0.0
[CV] C=100, gamma=0.0001, kernel=rbf .......
[CV] C=100, gamma=0.0001, kernel=rbf, score=0.9393939393939394, total=
                                                0.0
[CV] C=1000, gamma=1, kernel=rbf, score=0.631578947368421, total= 0.0s
[CV] C=1000, gamma=1, kernel=rbf ......
[CV] C=1000, gamma=1, kernel=rbf, score=0.631578947368421, total= 0.0s
[CV] C=1000, gamma=1, kernel=rbf ......
[CV] C=1000, gamma=1, kernel=rbf, score=0.6363636363636364, total=
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[CV] C=1000, gamma=0.1, kernel=rbf, score=0.631578947368421, total=
                                                                       0.0s
        [CV] C=1000, gamma=0.1, kernel=rbf ................................
        [CV] C=1000, gamma=0.1, kernel=rbf, score=0.631578947368421, total=
        [CV] C=1000, gamma=0.1, kernel=rbf, score=0.636363636363636364, total=
                                                                        0.0s
        [CV] C=1000, gamma=0.01, kernel=rbf ......
        [CV] C=1000, gamma=0.01, kernel=rbf, score=0.631578947368421, total=
                                                                        0.0s
        [CV] C=1000, gamma=0.01, kernel=rbf ......
        [CV] C=1000, gamma=0.01, kernel=rbf, score=0.631578947368421, total=
                                                                        0.0s
        [CV] C=1000, gamma=0.01, kernel=rbf ......
        [CV] C=1000, gamma=0.01, kernel=rbf, score=0.6363636363636364, total=
                                                                         0.0s
        [CV] C=1000, gamma=0.001, kernel=rbf ......
        [CV] C=1000, gamma=0.001, kernel=rbf, score=0.8947368421052632, total=
                                                                         0.0
        [CV] C=1000, gamma=0.001, kernel=rbf ......
        [CV] C=1000, gamma=0.001, kernel=rbf, score=0.9323308270676691, total=
                                                                         0.0
        [CV] C=1000, gamma=0.001, kernel=rbf ......
        [CV] C=1000, gamma=0.001, kernel=rbf, score=0.91666666666666666, total=
                                                                          0.0
        [CV] C=1000, gamma=0.0001, kernel=rbf .....
        [CV] C=1000, gamma=0.0001, kernel=rbf, score=0.9097744360902256, total=
                                                                          0.
        [CV] C=1000, gamma=0.0001, kernel=rbf ......
        [CV] C=1000, gamma=0.0001, kernel=rbf, score=0.9699248120300752, total=
                                                                          0.
        0s
        [CV] C=1000, gamma=0.0001, kernel=rbf .....
        [CV] C=1000, gamma=0.0001, kernel=rbf, score=0.9318181818181818, total=
                                                                          0.
        0s
        [Parallel(n_jobs=1)]: Done 75 out of 75 | elapsed:
                                                         1.1s finished
Out[57]: GridSearchCV(cv=3, error_score='raise-deprecating',
              estimator=SVC(C=1.0, cache size=200, class weight=None, coef0=0.0,
          decision_function_shape='ovr', degree=3, gamma='auto_deprecated',
          kernel='rbf', max iter=-1, probability=False, random state=None,
          shrinking=True, tol=0.001, verbose=False),
              fit params=None, iid='warn', n jobs=None,
              param_grid={'C': [0.1, 1, 10, 100, 1000], 'gamma': [1, 0.1, 0.01, 0.00
        1, 0.0001], 'kernel': ['rbf']},
              pre dispatch='2*n jobs', refit=True, return train score='warn',
              scoring=None, verbose=3)
In [60]: | grid.best_params_
Out[60]: {'C': 10, 'gamma': 0.0001, 'kernel': 'rbf'}
In [67]: | grid.best_estimator_
Out[67]: SVC(C=10, cache_size=200, class_weight=None, coef0=0.0,
          decision_function_shape='ovr', degree=3, gamma=0.0001, kernel='rbf',
          max iter=-1, probability=False, random state=None, shrinking=True,
          tol=0.001, verbose=False)
```

prediction

Evaluation

0.9473684210526315

```
In [71]: from sklearn.metrics import confusion_matrix, classification_report, accuracy_
In [77]: print('test data:\n')
         print('accuracy = %.2f' %(accuracy_score(y_test, pred)))
         print(confusion_matrix(y_test, pred))
         print(classification_report(y_test, pred))
         print(grid.score(X_test, y_test))
         test data:
         accuracy = 0.95
         [[ 60 6]
          [ 3 102]]
                       precision
                                    recall f1-score
                                                        support
                            0.95
                                      0.91
                                                 0.93
                    0
                                                             66
                    1
                            0.94
                                      0.97
                                                 0.96
                                                            105
                            0.95
                                      0.95
                                                 0.95
                                                            171
            micro avg
            macro avg
                            0.95
                                      0.94
                                                 0.94
                                                            171
                                      0.95
                                                 0.95
                                                            171
         weighted avg
                            0.95
```

```
In [79]: | print('train data:\n')
         print('accuracy = %.2f' %(accuracy_score(y_train, pred_train)))
         print(confusion_matrix(y_train, pred_train))
         print(classification_report(y_train, pred_train))
         print(grid.score(X_train, y_train))
         train data:
         accuracy = 0.97
         [[138 8]
          [ 4 248]]
                                    recall f1-score
                       precision
                                                       support
                                      0.95
                                                0.96
                    0
                            0.97
                                                           146
                    1
                            0.97
                                      0.98
                                                0.98
                                                           252
            micro avg
                            0.97
                                      0.97
                                                0.97
                                                           398
            macro avg
                            0.97
                                      0.96
                                                0.97
                                                           398
                                                           398
         weighted avg
                            0.97
                                      0.97
                                                0.97
```

0.9698492462311558

```
In [ ]:
```