```
In [6]: import pandas as pd
        import numpy as np
        import seaborn as sns
        import matplotlib.pyplot as plt
        %matplotlib inline
In [2]: df=pd.read csv('heart.csv')
In [3]: df.head()
Out[3]:
           age sex cp trestbps chol fbs restecg thalach exang oldpeak slope ca thal target
           63
                 1 3
                           145
                               233
                                    1
                                            0
                                                 150
                                                        0
                                                              2.3
                                                                     0
                                                                       0
                                                                            1
                                                                                  1
            37
                           130
                               250
                                     0
                                            1
                                                              3.5
                                                 187
                                                                                  1
         2
            41
                 0 1
                           130
                               204
                                     0
                                            0
                                                 172
                                                              1.4
                                                                     2 0
                                                                             2
                               236
                                                        0
                                                                     2 0
                                                                                  1
             56
                 1 1
                           120
                                     0
                                                 178
                                                              8.0
            57
                 0 0
                           120 354
                                     0
                                           1
                                                 163
                                                        1
                                                              0.6
                                                                     2 0
                                                                             2
                                                                                  1
In [4]: df.info()
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 303 entries, 0 to 302
        Data columns (total 14 columns):
                     303 non-null int64
        age
                     303 non-null int64
        sex
                     303 non-null int64
        ср
        trestbps
                     303 non-null int64
        chol
                     303 non-null int64
        fbs
                     303 non-null int64
                     303 non-null int64
        resteca
        thalach
                     303 non-null int64
        exang
                     303 non-null int64
```

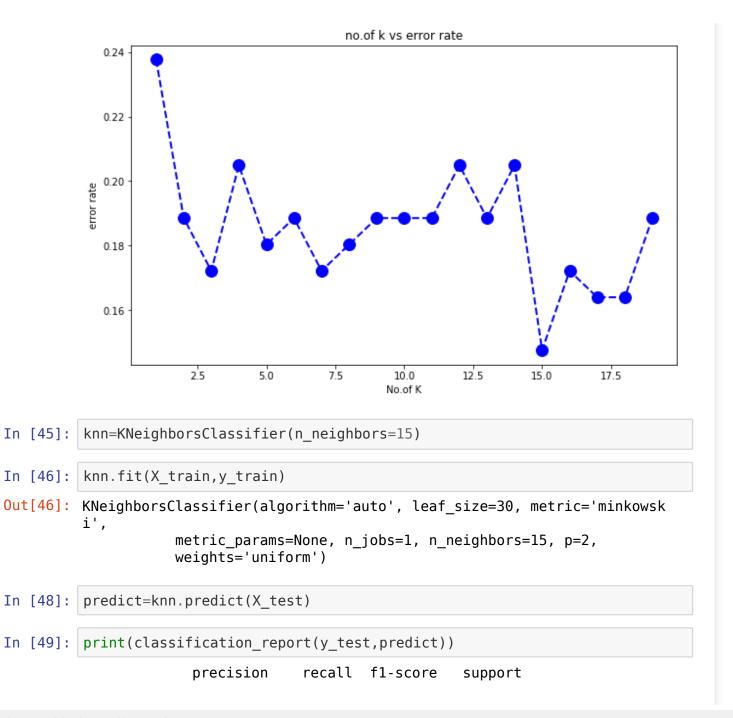
```
303 non-null float64
           oldpeak
           slope
                         303 non-null int64
                         303 non-null int64
           ca
           thal
                         303 non-null int64
                         303 non-null int64
           target
           dtypes: float64(1), int64(13)
           memory usage: 33.2 KB
          df.describe()
 In [5]:
 Out[5]:
                        age
                                   sex
                                              ср
                                                    trestbps
                                                                  chol
                                                                              fbs
                                                                                     restecg
            count 303.000000
                             303.000000
                                       303.000000
                                                  303.000000 303.000000
                                                                       303.000000
                                                                                  303.000000
                                                                                            303.0
                                                  131.623762 246.264026
            mean
                   54.366337
                              0.683168
                                         0.966997
                                                                         0.148515
                                                                                    0.528053
                                                                                            149.6
                    9.082101
              std
                               0.466011
                                         1.032052
                                                   17.538143
                                                             51.830751
                                                                         0.356198
                                                                                    0.525860
                                                                                              22.9
                   29.000000
                              0.000000
                                                   94.000000
                                                                                    0.000000
                                                                                             71.(
                                         0.000000
                                                            126.000000
                                                                         0.000000
             min
                   47.500000
                              0.000000
                                                  120.000000 211.000000
                                                                         0.000000
                                                                                    0.000000 133.
             25%
                                         0.000000
             50%
                   55.000000
                              1.000000
                                         1.000000 130.000000 240.000000
                                                                         0.000000
                                                                                    1.000000
                                                                                            153.0
             75%
                   61.000000
                              1.000000
                                         2.000000
                                                  140.000000 274.500000
                                                                         0.000000
                                                                                    1.000000
                                                                                            166.0
                                                                                    2.000000
                                                                                            202.0
                   77.000000
                              1.000000
                                         3.000000 200.000000 564.000000
                                                                         1.000000
             max
          from sklearn.preprocessing import StandardScaler
 In [7]:
 In [8]: std=StandardScaler()
 In [9]: std.fit(df.drop('target',axis=1))
 Out[9]: StandardScaler(copy=True, with mean=True, with std=True)
In [10]: a=std.transform(df.drop('target',axis=1))
           а
```

```
Out[10]: array([[ 0.9521966 ,  0.68100522,
                                                1.97312292, ..., -2.27457861,
                  -0.71442887, -2.14887271],
                 [-1.91531289, 0.68100522,
                                                1.00257707, \ldots, -2.27457861,
                  -0.71442887, -0.51292188],
                 [-1.47415758, -1.46841752,
                                                0.03203122, ..., 0.97635214,
                  -0.71442887, -0.51292188],
                 [1.50364073, 0.68100522, -0.93851463, \ldots, -0.64911323,
                    1.24459328, 1.12302895],
                 [0.29046364, 0.68100522, -0.93851463, \ldots, -0.64911323,
                    0.26508221, 1.12302895],
                 [0.29046364, -1.46841752, 0.03203122, ..., -0.64911323,
                   0.26508221, -0.51292188]
In [11]: df h=pd.DataFrame(a,columns=df.columns[:-1])
In [12]: df h.head()
Out[12]:
                                        trestbps
                                                    chol
                                                             fbs
                                                                           thalach
                 age
                          sex
                                                                   restecq
                                                                                     exanc
           0 0.952197
                      0.681005
                               1.973123
                                       0.763956
                                                -0.256334
                                                         2.394438
                                                                 -1.005832 0.015443 -0.696631
           1 -1.915313 0.681005
                               1.002577 -0.092738
                                                0.072199 -0.417635
                                                                 0.898962 1.633471
                                                                                  -0.696631
           2 -1.474158 -1.468418
                               0.032031 -0.092738 -0.816773 -0.417635 -1.005832 0.977514 -0.696631
                               0.032031 -0.663867 -0.198357 -0.417635
             0.180175
                      0.681005
                                                                 0.898962 1.239897
                                                                                  -0.696631
             0.290464 -1.468418 -0.938515 -0.663867 2.082050 -0.417635
                                                                 0.898962 0.583939
                                                                                 1.435481
In [13]: from sklearn.neighbors import KNeighborsClassifier
          knn=KNeighborsClassifier(n neighbors=2)
In [15]: from sklearn.cross validation import train test split
          C:\Users\Dell\Anaconda3\lib\site-packages\sklearn\cross validation.py:4
          1: DeprecationWarning: This module was deprecated in version 0.18 in fa
```

```
rators are different from that of this module. This module will be remo
         ved in 0.20.
           "This module will be removed in 0.20.", DeprecationWarning)
In [17]: df.columns
Out[17]: Index(['age', 'sex', 'cp', 'trestbps', 'chol', 'fbs', 'restecg', 'thala
         ch',
                'exang', 'oldpeak', 'slope', 'ca', 'thal', 'target'],
               dtype='object')
In [19]: X=df h[['age', 'sex', 'cp', 'trestbps', 'chol', 'fbs', 'restecg', 'thal
         ach',
                 'exang', 'oldpeak', 'slope', 'ca', 'thal']]
         v=df['target']
In [20]:
         X train, X test, y train, y test = train test split(X, y, test size=0.
         4, random state=101)
In [21]: X=df h
         v=df['target']
In [22]: knn.fit(X train,y train)
Out[22]: KNeighborsClassifier(algorithm='auto', leaf size=30, metric='minkowsk
         i',
                    metric params=None, n jobs=1, n neighbors=2, p=2,
                    weights='uniform')
In [23]: predict=knn.predict(X test)
In [24]: from sklearn.metrics import classification report
In [25]: print(classification report(y test, predict))
```

vor of the model\_selection module into which all the refactored classes and functions are moved. Also note that the interface of the new CV ite

```
precision
                                   recall f1-score
                                                      support
                   0
                           0.77
                                     0.88
                                               0.82
                                                           60
                           0.87
                                     0.74
                                               0.80
                                                           62
         avg / total
                           0.82
                                     0.81
                                               0.81
                                                           122
In [26]: from sklearn.metrics import confusion matrix
In [27]: print(confusion matrix(y test,predict))
         [[53 7]
          [16 46]]
In [39]: error rate=[]
         for i in range(1,20):
             knn=KNeighborsClassifier(n_neighbors=i)
             knn.fit(X train,y train)
             predict i=knn.predict(X test)
             error rate.append(np.mean(predict i != y test))
In [44]: plt.figure(figsize=(10,6))
         plt.plot(range(1,20),error_rate,color='blue', marker='o', linestyle='da
         shed',
                 linewidth=2, markersize=12)
         plt.xlabel('No.of K')
         plt.ylabel('error rate')
         plt.title('no.of k vs error rate')
Out[44]: Text(0.5,1,'no.of k vs error rate')
```



```
0
                           0.96
                                     0.73
                                               0.83
                                                            60
                           0.79
                                     0.97
                                               0.87
                                                            62
                           0.87
                                                           122
         avg / total
                                     0.85
                                               0.85
In [50]: print(confusion matrix(y test,predict))
         [[44 16]
          [ 2 60]]
In [52]: \# what if i take k=7
In [53]: knn=KNeighborsClassifier(n neighbors=7)
In [54]: knn.fit(X train,y train)
Out[54]: KNeighborsClassifier(algorithm='auto', leaf size=30, metric='minkowsk
         i',
                    metric params=None, n jobs=1, n neighbors=7, p=2,
                    weights='uniform')
In [56]: predictions=knn.predict(X_test)
In [57]: print(classification report(y test,predict))
                      precision
                                   recall f1-score
                                                       support
                   0
                           0.96
                                     0.73
                                               0.83
                                                            60
                           0.79
                                     0.97
                                               0.87
                                                            62
         avg / total
                           0.87
                                     0.85
                                               0.85
                                                           122
In [58]: # result is same as k=15
```

```
In [59]: | print(confusion_matrix(y_test,predict))
         [[44 16]
          [ 2 60]]
In [ ]: # shows same result
In [ ]:
```