```
In [1]: 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
```

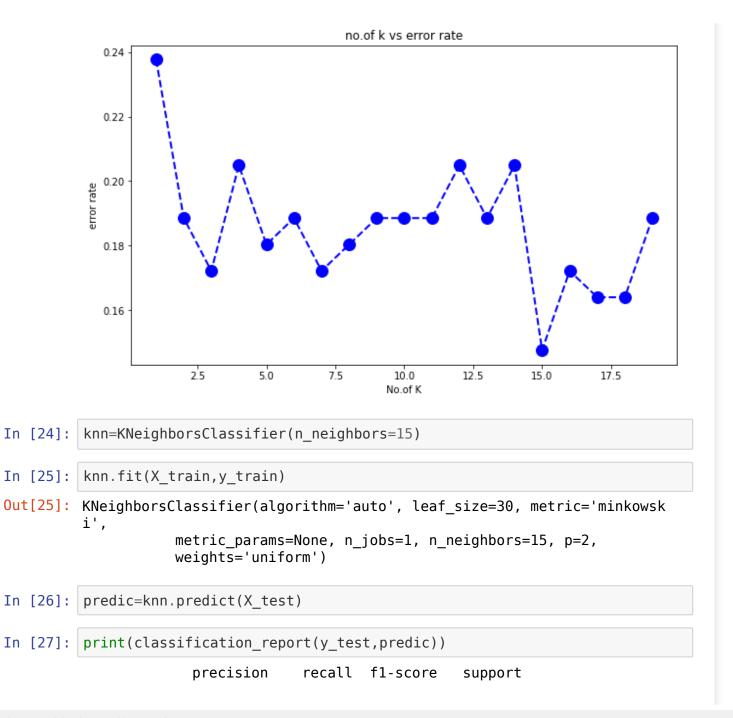
```
303 non-null float64
          oldpeak
         slope
                        303 non-null int64
                        303 non-null int64
          ca
         thal
                        303 non-null int64
                        303 non-null int64
         target
         dtypes: float64(1), 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
                                        0.000000
                                                           126.000000
                                                                        0.000000
            min
                  47.500000
                             0.000000
                                                 120.000000 211.000000
                                                                                   0.000000 133.
            25%
                                        0.000000
                                                                        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 [6]:
         std=StandardScaler()
In [7]:
In [8]: std.fit(df.drop('target',axis=1))
Out[8]: StandardScaler(copy=True, with mean=True, with std=True)
In [9]: a=std.transform(df.drop('target',axis=1))
          а
```

```
Out[9]: 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 [10]: df h=pd.DataFrame(a,columns=df.columns[:-1])
In [11]: df h.head()
Out[11]:
                                        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
         from sklearn.neighbors import KNeighborsClassifier
         knn=KNeighborsClassifier(n neighbors=2)
In [14]: 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
```

```
and functions are moved. Also note that the interface of the new CV ite
         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 [15]: df.columns
Out[15]: Index(['age', 'sex', 'cp', 'trestbps', 'chol', 'fbs', 'restecg', 'thala
         ch',
                'exang', 'oldpeak', 'slope', 'ca', 'thal', 'target'],
               dtype='object')
In [16]: X=df h[['age', 'sex', 'cp', 'trestbps', 'chol', 'fbs', 'restecg', 'thal
         ach',
                 'exang', 'oldpeak', 'slope', 'ca', 'thal']]
         v=df['target']
         X train, X test, y train, y test = train test split(X, y, test size=0.
In [17]:
         4, random state=101)
In [18]: X=df h
         v=df['target']
In [19]: knn.fit(X train,y train)
Out[19]: KNeighborsClassifier(algorithm='auto', leaf size=30, metric='minkowsk
         i',
                    metric params=None, n jobs=1, n neighbors=2, p=2,
                    weights='uniform')
In [20]: predict=knn.predict(X test)
In [21]: from sklearn.metrics import classification report
In [22]: print(classification report(y test, predict))
```

vor of the model selection module into which all the refactored classes

```
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 [23]: 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 [28]: print(confusion matrix(y test,predict))
         [[53 7]
          [16 46]]
In [29]: \# what if i take k=7
In [30]: knn=KNeighborsClassifier(n neighbors=7)
In [31]: knn.fit(X train,y train)
Out[31]: KNeighborsClassifier(algorithm='auto', leaf size=30, metric='minkowsk
         i',
                    metric params=None, n jobs=1, n neighbors=7, p=2,
                    weights='uniform')
In [32]: predictions=knn.predict(X_test)
In [34]: print(classification report(y test,predictions))
                      precision
                                   recall f1-score
                                                      support
                   0
                           0.84
                                     0.80
                                               0.82
                                                           60
                           0.82
                                     0.85
                                               0.83
                                                           62
                           0.83
                                     0.83
                                               0.83
                                                           122
         avg / total
In [37]: print(confusion_matrix(y_test,predictions))
         [[48 12]
```

	[9 53]]
In [36]:	# result is better with k=7
In []:	