Lab 10 06-10-2022 k-NN cancer

October 6, 2022

0.1 Breast Cancer

[1]:
```
import numpy as np
import pandas as pd
!pip install scikit-learn
```

```
Defaulting to user installation because normal site-packages is not writeable
Requirement already satisfied: scikit-learn in
/opt/anaconda3/lib/python3.9/site-packages (1.0.2)
Requirement already satisfied: numpy>=1.14.6 in
/opt/anaconda3/lib/python3.9/site-packages (from scikit-learn) (1.21.5)
Requirement already satisfied: threadpoolctl>=2.0.0 in
/opt/anaconda3/lib/python3.9/site-packages (from scikit-learn) (2.2.0)
Requirement already satisfied: joblib>=0.11 in
/opt/anaconda3/lib/python3.9/site-packages (from scikit-learn) (1.1.0)
Requirement already satisfied: scipy>=1.1.0 in
/opt/anaconda3/lib/python3.9/site-packages (from scikit-learn) (1.7.3)
```

[3]:
```
ds = pd.read_csv('BreastCancer.csv')
print(ds)
```

```
           id diagnosis  radius_mean  texture_mean  perimeter_mean  area_mean  \
0      842302         M        17.99         10.38          122.80     1001.0
1      842517         M        20.57         17.77          132.90     1326.0
2    84300903         M        19.69         21.25          130.00     1203.0
3    84348301         M        11.42         20.38           77.58      386.1
4    84358402         M        20.29         14.34          135.10     1297.0
..        …       …           …             …               …          …
564    926424         M        21.56         22.39          142.00     1479.0
565    926682         M        20.13         28.25          131.20     1261.0
566    926954         M        16.60         28.08          108.30      858.1
567    927241         M        20.60         29.33          140.10     1265.0
568     92751         B         7.76         24.54           47.92      181.0

     smoothness_mean  compactness_mean  concavity_mean  concave points_mean  \
0            0.11840           0.27760         0.30010              0.14710
1            0.08474           0.07864         0.08690              0.07017
2            0.10960           0.15990         0.19740              0.12790
3            0.14250           0.28390         0.24140              0.10520
```

```
4           0.10030           0.13280         0.19800              0.10430
..              …                 …               …                    …
564         0.11100           0.11590         0.24390              0.13890
565         0.09780           0.10340         0.14400              0.09791
566         0.08455           0.10230         0.09251              0.05302
567         0.11780           0.27700         0.35140              0.15200
568         0.05263           0.04362         0.00000              0.00000

     …  texture_worst  perimeter_worst  area_worst  smoothness_worst  \
0    …          17.33           184.60      2019.0           0.16220
1    …          23.41           158.80      1956.0           0.12380
2    …          25.53           152.50      1709.0           0.14440
3    …          26.50            98.87       567.7           0.20980
4    …          16.67           152.20      1575.0           0.13740
..   …            …                …           …                …
564  …          26.40           166.10      2027.0           0.14100
565  …          38.25           155.00      1731.0           0.11660
566  …          34.12           126.70      1124.0           0.11390
567  …          39.42           184.60      1821.0           0.16500
568  …          30.37            59.16       268.6           0.08996

     compactness_worst  concavity_worst  concave points_worst  symmetry_worst  \
0              0.66560           0.7119                0.2654          0.4601
1              0.18660           0.2416                0.1860          0.2750
2              0.42450           0.4504                0.2430          0.3613
3              0.86630           0.6869                0.2575          0.6638
4              0.20500           0.4000                0.1625          0.2364
..                 …                …                     …               …
564            0.21130           0.4107                0.2216          0.2060
565            0.19220           0.3215                0.1628          0.2572
566            0.30940           0.3403                0.1418          0.2218
567            0.86810           0.9387                0.2650          0.4087
568            0.06444           0.0000                0.0000          0.2871

     fractal_dimension_worst  Unnamed: 32
0                    0.11890          NaN
1                    0.08902          NaN
2                    0.08758          NaN
3                    0.17300          NaN
4                    0.07678          NaN
..                       …            …
564                  0.07115          NaN
565                  0.06637          NaN
566                  0.07820          NaN
567                  0.12400          NaN
568                  0.07039          NaN

[569 rows x 33 columns]
```

[5]:
```
ds.shape
cols=ds.columns
print(cols)
ds.value_counts("diagnosis")
```

```
Index(['id', 'diagnosis', 'radius_mean', 'texture_mean', 'perimeter_mean',
       'area_mean', 'smoothness_mean', 'compactness_mean', 'concavity_mean',
       'concave points_mean', 'symmetry_mean', 'fractal_dimension_mean',
       'radius_se', 'texture_se', 'perimeter_se', 'area_se', 'smoothness_se',
       'compactness_se', 'concavity_se', 'concave points_se', 'symmetry_se',
       'fractal_dimension_se', 'radius_worst', 'texture_worst',
       'perimeter_worst', 'area_worst', 'smoothness_worst',
       'compactness_worst', 'concavity_worst', 'concave points_worst',
       'symmetry_worst', 'fractal_dimension_worst', 'Unnamed: 32'],
      dtype='object')
```

[5]:
```
diagnosis
B    357
M    212
dtype: int64
```

0.1.1 Data preprocessing

[8]:
```
y = ds['diagnosis']
ds.drop('diagnosis', axis = 1,inplace = True)
ds.drop('Unnamed: 32', axis = 1,inplace=True)
ds.drop('id', axis = 1,inplace=True)
cols=ds.columns
print(cols)
x = ds
```

```
Index(['radius_mean', 'texture_mean', 'perimeter_mean', 'area_mean',
       'smoothness_mean', 'compactness_mean', 'concavity_mean',
       'concave points_mean', 'symmetry_mean', 'fractal_dimension_mean',
       'radius_se', 'texture_se', 'perimeter_se', 'area_se', 'smoothness_se',
       'compactness_se', 'concavity_se', 'concave points_se', 'symmetry_se',
       'fractal_dimension_se', 'radius_worst', 'texture_worst',
       'perimeter_worst', 'area_worst', 'smoothness_worst',
       'compactness_worst', 'concavity_worst', 'concave points_worst',
       'symmetry_worst', 'fractal_dimension_worst'],
      dtype='object')
```

[10]:
```
ds.describe()
```

[10]:

	radius_mean	texture_mean	perimeter_mean	area_mean	\
count	569.000000	569.000000	569.000000	569.000000	
mean	14.127292	19.289649	91.969033	654.889104	

std	3.524049	4.301036	24.298981	351.914129
min	6.981000	9.710000	43.790000	143.500000
25%	11.700000	16.170000	75.170000	420.300000
50%	13.370000	18.840000	86.240000	551.100000
75%	15.780000	21.800000	104.100000	782.700000
max	28.110000	39.280000	188.500000	2501.000000

	smoothness_mean	compactness_mean	concavity_mean	concave points_mean	\
count	569.000000	569.000000	569.000000	569.000000	
mean	0.096360	0.104341	0.088799	0.048919	
std	0.014064	0.052813	0.079720	0.038803	
min	0.052630	0.019380	0.000000	0.000000	
25%	0.086370	0.064920	0.029560	0.020310	
50%	0.095870	0.092630	0.061540	0.033500	
75%	0.105300	0.130400	0.130700	0.074000	
max	0.163400	0.345400	0.426800	0.201200	

	symmetry_mean	fractal_dimension_mean	…	radius_worst	\
count	569.000000	569.000000	…	569.000000	
mean	0.181162	0.062798	…	16.269190	
std	0.027414	0.007060	…	4.833242	
min	0.106000	0.049960	…	7.930000	
25%	0.161900	0.057700	…	13.010000	
50%	0.179200	0.061540	…	14.970000	
75%	0.195700	0.066120	…	18.790000	
max	0.304000	0.097440	…	36.040000	

	texture_worst	perimeter_worst	area_worst	smoothness_worst	\
count	569.000000	569.000000	569.000000	569.000000	
mean	25.677223	107.261213	880.583128	0.132369	
std	6.146258	33.602542	569.356993	0.022832	
min	12.020000	50.410000	185.200000	0.071170	
25%	21.080000	84.110000	515.300000	0.116600	
50%	25.410000	97.660000	686.500000	0.131300	
75%	29.720000	125.400000	1084.000000	0.146000	
max	49.540000	251.200000	4254.000000	0.222600	

	compactness_worst	concavity_worst	concave points_worst	\
count	569.000000	569.000000	569.000000	
mean	0.254265	0.272188	0.114606	
std	0.157336	0.208624	0.065732	
min	0.027290	0.000000	0.000000	
25%	0.147200	0.114500	0.064930	
50%	0.211900	0.226700	0.099930	
75%	0.339100	0.382900	0.161400	
max	1.058000	1.252000	0.291000	

```
       symmetry_worst  fractal_dimension_worst
count      569.000000               569.000000
mean         0.290076                 0.083946
std          0.061867                 0.018061
min          0.156500                 0.055040
25%          0.250400                 0.071460
50%          0.282200                 0.080040
75%          0.317900                 0.092080
max          0.663800                 0.207500

[8 rows x 30 columns]
```

[11]:
```
from sklearn.model_selection import train_test_split
x_train, x_test, y_train, y_test= train_test_split(x, y, test_size= 0.25,␣
 ↪random_state=0)
```

0.1.2 Feature Scaling

[13]:
```
from sklearn.preprocessing import MinMaxScaler
st_x= MinMaxScaler()
x_train= st_x.fit_transform(x_train)
x_test=st_x.fit_transform(x_test)
print(x_train)
```

```
[[0.23044157 0.32157676 0.21940433 … 0.31484671 0.30277942 0.09858323]
 [0.20062473 0.42116183 0.19452699 … 0.06965208 0.34042973 0.06677161]
 [0.62232003 0.76929461 0.60403566 … 0.56079917 0.19850187 0.07431457]
 …
 [0.11619102 0.35726141 0.11077327 … 0.17402687 0.17524147 0.17263545]
 [0.12963226 0.35311203 0.11706171 … 0.         0.06780997 0.06919848]
 [0.21434995 0.59004149 0.21235575 … 0.33251808 0.10782574 0.21172767]]
```

0.1.3 Fitiing k-NN

[15]:
```
from sklearn.neighbors import KNeighborsClassifier
classifier = KNeighborsClassifier(n_neighbors = 5)
classifier.fit(x_train, y_train)
```

[15]:
```
KNeighborsClassifier()
```

[16]:
```
y_pred= classifier.predict(x_test)
print(y_pred)
```

```
['M' 'B' 'B' 'B' 'B' 'B' 'B' 'B' 'B' 'B' 'M' 'B' 'B' 'M' 'M' 'M' 'B' 'M'
 'M' 'M' 'M' 'M' 'B' 'B' 'M' 'B' 'B' 'M' 'B' 'M' 'B' 'M' 'B' 'M' 'B' 'M'
 'B' 'M' 'B' 'M' 'M' 'B' 'M' 'B' 'B' 'M' 'B' 'B' 'B' 'M' 'M' 'M' 'M' 'B'
 'B' 'B' 'B' 'B' 'B' 'M' 'M' 'M' 'B' 'B' 'M' 'B' 'M' 'M' 'M' 'B' 'M' 'M'
```

```
 'B' 'M' 'M' 'B' 'B' 'M' 'B' 'B' 'M' 'M' 'M' 'B' 'M' 'B' 'B' 'B' 'M' 'M'
 'B' 'M' 'M' 'M' 'B' 'B' 'M' 'B' 'B' 'B' 'B' 'B' 'B' 'B' 'M' 'B' 'M' 'B'
 'M' 'M' 'B' 'M' 'M' 'M' 'B' 'B' 'B' 'B' 'B' 'B' 'B' 'B' 'M' 'B' 'M' 'M'
 'M' 'B' 'B' 'B' 'M' 'B' 'B' 'M' 'B' 'B' 'M' 'M' 'M' 'B' 'B' 'B' 'M']
```

0.1.4 Creating confusion matrix

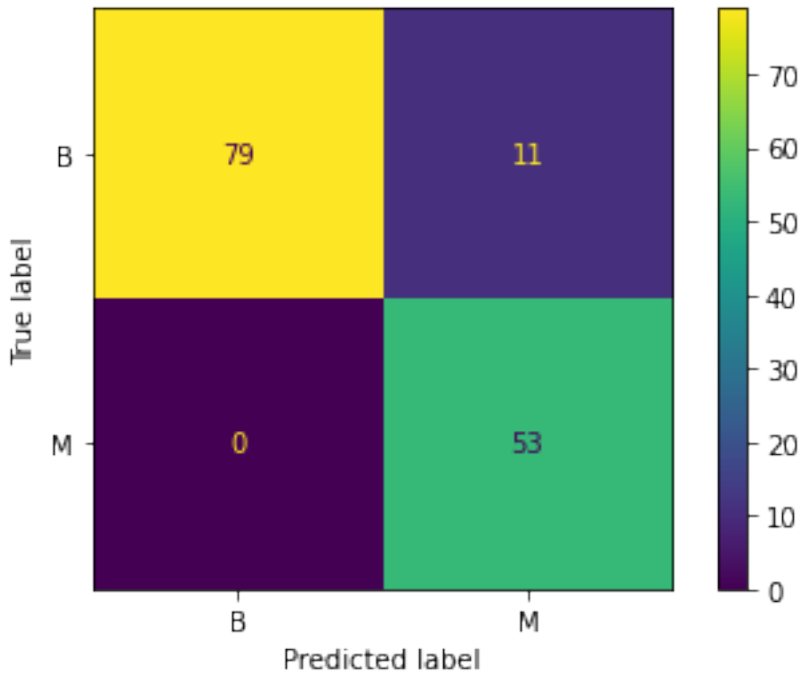
[17]:
```
from sklearn.metrics import confusion_matrix
cm= confusion_matrix (y_test, y_pred,labels=classifier.classes_)
print(cm)
```

```
[[79 11]
 [ 0 53]]
```

[20]:
```
from sklearn.metrics import ConfusionMatrixDisplay
disp = ConfusionMatrixDisplay(confusion_matrix=cm,
display_labels=classifier.classes_)

disp.plot()
```

[20]:
```
<sklearn.metrics._plot.confusion_matrix.ConfusionMatrixDisplay at
0x7eff2c581f10>
```



Accuracy

[21]:
```
training_score = classifier.score(x_train, y_train)
test_score = classifier.score(x_test, y_test)
print(training_score)
print(test_score)
```

```
0.9765258215962441
0.9230769230769231
```

New set

[25]:
```
K = []
training = []
test = []
scores = {}
for k in range(2, 22):
    clf = KNeighborsClassifier(n_neighbors = k)
    clf.fit(x_train, y_train)
    training_score = clf.score(x_train, y_train)
    test_score = clf.score(x_test, y_test)
    K.append(k)
    training.append(training_score)
    test.append(test_score)
    scores[k] = [training_score, test_score]
for keys, values in scores.items():
    print(keys, ':', values)
```

```
2 : [0.9765258215962441, 0.9230769230769231]
3 : [0.9812206572769953, 0.8951048951048951]
4 : [0.9835680751173709, 0.916083916083916]
5 : [0.9765258215962441, 0.9230769230769231]
6 : [0.9788732394366197, 0.9090909090909091]
7 : [0.9788732394366197, 0.916083916083916]
8 : [0.9812206572769953, 0.951048951048951]
9 : [0.9765258215962441, 0.9230769230769231]
10 : [0.9741784037558685, 0.9370629370629371]
11 : [0.9765258215962441, 0.9230769230769231]
12 : [0.9694835680751174, 0.9300699300699301]
13 : [0.9741784037558685, 0.9300699300699301]
14 : [0.9694835680751174, 0.9370629370629371]
15 : [0.9694835680751174, 0.9370629370629371]
16 : [0.9647887323943662, 0.9440559440559441]
17 : [0.9694835680751174, 0.9370629370629371]
18 : [0.9671361502347418, 0.9440559440559441]
19 : [0.971830985915493, 0.9440559440559441]
20 : [0.9624413145539906, 0.9440559440559441]
21 : [0.9647887323943662, 0.9440559440559441]
```

Visualisation

[27]:
```
import matplotlib.pyplot as plt
plt.scatter(K, training, color ='r')
plt.scatter(K, test, color ='g')
plt.show()
```