

Group Practical 1

Shiyu Yi,2016141231175

Chuang Du,2016141462277

Jiali Shang,2016141462137

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1 Problem overview

1. Using iris data to assess the classification performance by tuning the KNN classifiers:

- Splitting the data using different percentage
- Change cross validation folds
- Changing the value of K
- Normalise the data

- 1) Summarise the above classification performances of the above settings using tables/figures
- 2) Discuss the results.

2 Problem one

This data is from the software weka:

To compare we use the sklearn to train the data:

The code:https://github.com/chanchann/Bio_Machine_Learning

Table 1: Splitting the data using different percentage(weka)

Training proportion	Test Proportion	Evaluate Score
0.9	0.1	0.933333
0.8	0.2	0.966667
0.7	0.3	0.955556
0.6	0.4	0.950000
0.5	0.5	0.96000
0.4	0.6	0.966667
0.3	0.7	0.942857
0.2	0.8	0.958333
0.1	0.9	0.903704

Table 2: Splitting the data using different percentage(sklearn)

Training proportion	Test Proportion	Evaluate Score
0.9	0.1	0.9333333333333333
0.8	0.2	0.9333333333333333
0.7	0.3	0.9555555555555556
0.6	0.4	0.9666666666666667
0.5	0.5	0.9733333333333334
0.4	0.6	0.9666666666666667
0.3	0.7	0.9333333333333333
0.2	0.8	0.9166666666666666
0.1	0.9	0.837037037037037

Splitting the data using different percentage

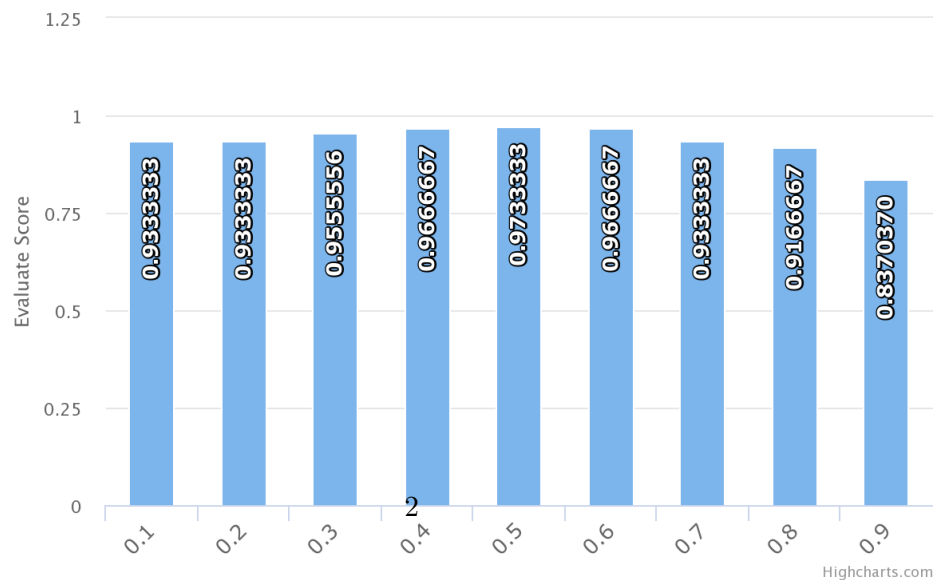


fig.1

3 Problem two/three

Table 3: cross validation folds

cv's folder	Accuracy
2	0.94(+/-0.04)
3	0.99(+/-0.02)
4	0.97(+/-0.04)
5	0.97(+/-0.05)
6	0.97(+/-0.07)
7	0.97(+/-0.07)
8	0.97(+/-0.08)

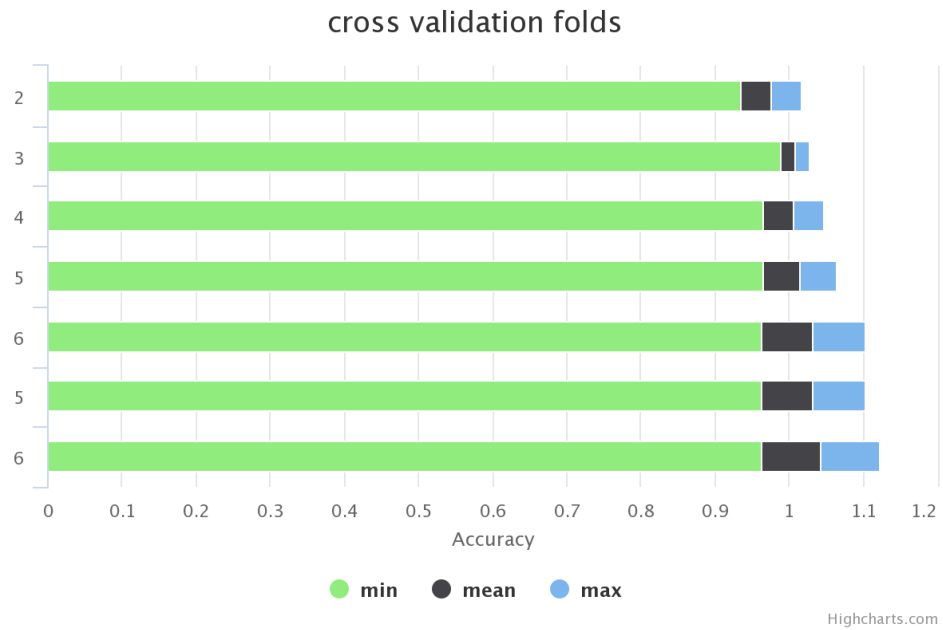


fig.2

Table 4: Normalise the data	
Method	Accuracy
Min-Max scaling	0.97(+/-0.05)
Standardization	0.97(+/-0.02)
Normalizer	0.97(+/-0.04)

4 Problem Four

5 Classification

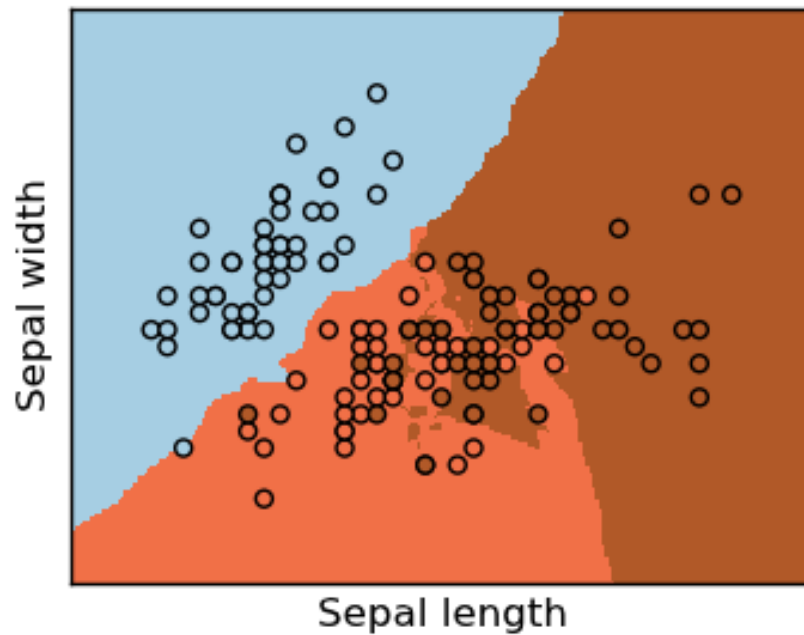


fig.2

split.py

To implement Splitting the data using different percentage

```
import numpy as np
import pandas as pd
from sklearn import datasets
from sklearn.neighbors import KNeighborsClassifier

iris=datasets.load_iris()
```

```

iris_x=iris.data
iris_y=iris.target

# Split iris data in train and test data
# A random permutation, to split the data randomly

np.random.seed(0)
indices=np.random.permutation(len(iris_x))
for i in np.arange(0.1,1,0.1):
    test_size=-1*i*len(indices)
    test_size=int(test_size)
    iris_x_train=iris_x[indices[:test_size]]
    iris_y_train=iris_y[indices[:test_size]]
    iris_x_test=iris_x[indices[test_size:]]
    iris_y_test=iris_y[indices[test_size:]]
    #Create a nearest-neighbor classifier
    knn=KNeighborsClassifier()
    knn.fit(iris_x_train,iris_y_train)
    #iris_x_predict=knn.predict(iris_x_test)
    #print(iris_x_predict)
    #print(iris_y_test)
    score=knn.score(iris_x_test,iris_y_test)
    #print(score)
    print('test propotion: {0},\nevaluate score:{1}'.format(i,score))
    print('-----')

```

crossVal.py

To use cross validation folds.

```

import numpy as np
import pandas as pd
from sklearn import datasets
from sklearn.neighbors import KNeighborsClassifier
from sklearn.model_selection import train_test_split
from sklearn.model_selection import cross_val_score

#Create the KNNClassifier
knn=KNeighborsClassifier()
# knn.fit(x_train,y_train)
for i in range(2,9):
    scores=cross_val_score(knn,iris.data,iris.target,cv=i)
    print(scores)
    print("Accuracy:%0.2f(+/-%0.2f)"%(scores.mean(),scores.std()*2))
    print("-----")

```

normalize.py

To normalize the data to train.

```

import numpy as np
import pandas as pd
from sklearn import datasets
from sklearn.neighbors import KNeighborsClassifier
from sklearn.model_selection import train_test_split
from sklearn.model_selection import cross_val_score
from sklearn.preprocessing import MinMaxScaler
from sklearn.preprocessing import StandardScaler
from sklearn.preprocessing import Normalizer

iris=datasets.load_iris()
#Min-Max scaling
#MinMaxScaler().fit_transform(iris.data)

#Standardization
# StandardScaler().fit_transform(iris.data)

# Normalizer
Normalizer().fit_transform(iris.data)

# Here we split the data 0.6:0.4
x_train,x_test,y_train,y_test=train_test_split(
iris.data,iris.target,test_size=0.4,random_state=0)

#Create the KNNClassifier
knn=KNeighborsClassifier()
scores=cross_val_score(knn,iris.data,iris.target,cv=5)
print("Accuracy:%0.2f(+/-%0.2f)"%(scores.mean(),scores.std()*2))

```

plot.py

To plot the classification.

```

import numpy as np
import pylab as pl
from sklearn import neighbors, datasets

# import some data to play with
iris = datasets.load_iris()
X = iris.data[:, :2] # we only take the first two features.
Y = iris.target

h = .02 # step size in the mesh

knn=neighbors.KNeighborsClassifier()

# we create an instance of Neighbours Classifier and fit the data.

```

```

knn.fit(X, Y)

# Plot the decision boundary. For that, we will assign a color to
# each
# point in the mesh [x_min, m_max]x[y_min, y_max].
x_min, x_max = X[:,0].min() - .5, X[:,0].max() + .5
y_min, y_max = X[:,1].min() - .5, X[:,1].max() + .5
xx, yy = np.meshgrid(np.arange(x_min, x_max, h), np.arange(y_min,
y_max, h))
Z = knn.predict(np.c_[xx.ravel(), yy.ravel()])

# Put the result into a color plot
Z = Z.reshape(xx.shape)
pl.figure(1, figsize=(4, 3))
pl.set_cmap(pl.cm.Paired)
pl.pcolormesh(xx, yy, Z)

# Plot also the training points
pl.scatter(X[:,0], X[:,1],c=Y )
pl.xlabel('Sepal length')
pl.ylabel('Sepal width')

pl.xlim(xx.min(), xx.max())
pl.ylim(yy.min(), yy.max())
pl.xticks(())
pl.yticks(())

pl.show()

```