K近邻算法

使用sklearn库来进行K近邻算法的运算

```
In [8]: import pandas as pd import warnings warnings.filterwarnings("ignore")

df = pd.read_excel('葡萄酒.xlsx')
# 特征变量和目标变量的切分
X_train = df[['酒精含量(%)','苹果酸含量(%)']]
y_train = df['分类']

df
```

Out[8]:

	原始样本	酒精含量(%)	苹果酸含量(%)	分类
0	样本1	5	2	0
1	样本2	6	1	0
2	样本3	4	1	0
3	样本4	8	3	1
4	样本5	10	2	1

```
In [9]: # 模型训练
```

```
# 侯望训练
from sklearn.neighbors import KNeighborsClassifier as KNN
knn = KNN(n_neighbors=3)
knn.fit(X_train, y_train)
# 模型预测: 预测多个样本
X_test = [[7, 1], [8, 3], [7, 3]] # 这里能帮助理解为什么要写成二维数组的样式
answer = knn.predict(X_test)
print(answer)
```

 $[0 \ 1 \ 0]$

编写代码实现K近邻算法。其中距离计算使用了欧氏距离,同时提供了计算马氏距离的实现代码(第7行到第9行):

```
In [10]: from sklearn.metrics import accuracy_score
         from scipy. spatial import distance
          import numpy as np
         import operator
         def classify(inX, dataSet, labels, k):
             #S=np.cov(dataSet.T) #协方差矩阵,为计算马氏距离
             #SI = np. linalg. inv(S) #协方差矩阵的逆矩阵
             #distances = np. array(distance.cdist(dataSet, [inX], 'mahalanobis', VI=SI)).reshape(-1)
             distances = np. array(distance.cdist(dataSet, [inX], 'euclidean').reshape(-1))
             sortedDistIndicies = distances.argsort() # 取排序的索引,用于label排序
             classCount={}
             for i in range(k): # 访问距离最近的k个实例
                 voteILabel = labels[sortedDistIndicies[i]]
                 classCount[voteILabel]=classCount.get(voteILabel, 0)+1
             sortedClassCount = sorted(classCount.items(),
                      key=operator.itemgetter(1), reverse=True)
             return sortedClassCount[0][0] # 取最多的分类
         ret = [classify(X_test[i], X_train, y_train, 3) for i in range(len(X_test))]
         print(ret)
```

[0, 1, 0]

```
SVM算法
In [11]: from sklearn import svm
          from sklearn. model selection import train test split
          from sklearn.datasets import load_iris
          import matplotlib.pyplot as plt
          %matplotlib inline
          iris=load iris()
          X = iris.data # 获取自变量
          y = iris.target # 获取因变量
          X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=0)
          clf = svm.SVC(C=0.8, kernel='rbf', gamma=1) # 高斯核, 松弛度0.8
          #clf = svm. SVC(C=0.5, kernel='linear') # 线性核, 松弛度0.5
          clf.fit(X_train, y_train.ravel())
          print('trian pred:%.3f' %(clf.score(X_train, y_train))) # 对训练集打分
          print('test pred:%.3f' %(clf.score(X_test, y_test))) # 对测试集打分
          print(clf. support_vectors_) #支持向量列表,从中看到切分边界
          print(clf.n_support_) # 每类别持向量个数
          plt.plot(X_train[:,0], X_train[:,1],'o', color = '#bbbbbb')
          plt.plot(clf.support_vectors_[:, 0], clf.support_vectors_[:, 1], 'o')
          trian pred:0.967
          test pred:1.000
          [[5. 3.5 1.6 0.6]
           [4.33. 1.10.1]
           [5. 4 3. 4 1. 7 0. 2]
           [5. 7 4. 4 1. 5 0. 4]
           [4. 5 2. 3 1. 3 0. 3]
           [5. 1 3. 3 1. 7 0. 5]
           [5. 2 4. 1 1. 5 0. 1]
           [5.84. 1.20.2]
           [5. 4 3. 4. 5 1. 5]
           [5. 2 2. 7 3. 9 1. 4]
           [6. 2.7 5.1 1.6]
           [5. 2. 3.5 1.]
           [6. 3 3. 3 4. 7 1. 6]
           [7. \quad 3.2 \quad 4.7 \quad 1.4]
           [6. 3.4 4.5 1.6]
           [5.6 2.9 3.6 1.3]
           [6.1 3. 4.6 1.4]
           [6.9 \ 3.1 \ 4.9 \ 1.5]
           [5. 1 2. 5 3. 1. 1]
           [6. 2 2. 2 4. 5 1. 5]
           [5. 7 2. 8 4. 5 1. 3]
```

Out[11]: [<matplotlib.lines.Line2D at 0xlec4ecbea30>]

[8 18 18]

[5. 7 2. 6 3. 5 1.] $[6.3 \ 2.5 \ 4.9 \ 1.5]$ [6. 7 3. 5. 1. 7] [5.9 3.2 4.8 1.8] [6. 3 2. 3 4. 4 1. 3] [6. 1 3. 4. 9 1. 8] [6. 9 3. 1 5. 1 2. 3] [6. 2.2 5. 1.5] $[6.7 \ 2.5 \ 5.8 \ 1.8]$ [6. 3 2. 7 4. 9 1. 8] [6. 3 2. 8 5. 1 1. 5] [4.9 2.5 4.5 1.7] [6. 3 2. 5 5. 1. 9] [5.93. 5.11.8] [7. 2 3. 6 6. 1 2. 5] [6. 5 3. 2 5. 1 2.] [6. 2 3. 4 5. 4 2. 3] [7. 2 3. 5.8 1.6] [7.7 2.6 6.9 2.3] $[7.9 \ 3.8 \ 6.4 \ 2.]$ [6. 3. 4.8 1.8] $[6.3 \ 3.4 \ 5.6 \ 2.4]$ [7.7 3.8 6.7 2.2]

