# 机器学习第一次作业

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#### 实验要求

题目:基于KNN的手写数字识别 实验条件:给定semeion手写数字数据集,给定kNN分类算

法 实验要求:

1.基本要求:编程实现kNN算法;给出在不同k值 (1,3,5)情况下,kNN算法对手写数字

的识别精度 (要求采用留一法)

2.中级要求:与weka机器学习包中的kNN分类结果进行对比

3.提高要求:将实验过程结果等图示展出

### 导入需要的包

```
In [12]: import numpy as np
from collections import Counter
import math
import csv
import matplotlib.pyplot as plt
```

#### 导入数据集 semesion

```
In [13]: # 导入数据
          def Img2Mat(fileName):
             f = open(fileName)
             ss = f. readlines()
              1 = 1en(ss)
              f. close()
              returnMat = np. zeros((1, 256))
              returnClassVector = np. zeros((1, 1))
              for i in range(1):
                  s1 = ss[i]. split()
                  for j in range (256):
                     returnMat[i][j] = np. float(s1[j])
                  c1Count = 0
                  for j in range (256, 266):
                      if s1[j] != '1':
                          c1Count += 1
                      else:
                          break
                  returnClassVector[i] = clCount
              return returnMat, returnClassVector
```

```
In [14]: X, y = Img2Mat('D:\Desktop\ML\Lab\lab1\semeion.data')
np. shape(X), np. shape(y)
```

```
C:\Users\PengYuting\AppData\Local\Temp\ipykernel_4472\2757158993.py:12: DeprecationW arning: `np.float` is a deprecated alias for the builtin `float`. To silence this wa rning, use `float` by itself. Doing this will not modify any behavior and is safe. I f you specifically wanted the numpy scalar type, use `np.float64` here.

Deprecated in NumPy 1.20; for more details and guidance: https://numpy.org/devdocs/release/1.20.0-notes.html#deprecations
   returnMat[i][j] = np.float(s1[j])
((1593, 256), (1593, 1))
```

Out[14]:

#### 基本要求

编程实现kNN算法;给出不同k值(1,3,5)情况下,kNN算法对手写数字的识别精度,采用留一法。

```
# KNN算法手动实现
In [15]:
         def MyKnnClassifier(data_X, data_y, neighbors):
            testround = 0
            acc = 0 # 用于计算ACC
            data_X = X. tolist()
            data_y = y. tolist()
            # 自己实现留一法
            for test_index in range(1593):
                X train = data X[:]
                X train. pop(test index)
                X_test = data_X[test_index]
                y_train = data_y[:]
                y_train. pop(test_index)
                y_test = data_y[test_index]
                X train = np. array (X train)
                X \text{ test} = \text{np. array}(X \text{ test})
                X_{test} = 1ist(X_{test})
                y_train = np. array(y_train)
                y_test = np. array(y_test)
                testround += 1
                trainShape = X train. shape[0] # 获得训练集的大小
                testShape = len(X test) # 获得测试集的大小
                # 差异矩阵 = 该样本与训练集中所有样本之差构成的矩阵
                testDiffMat = np. tile(X_test, (trainShape , 1)) - X_train
                sqTestDiffMat = testDiffMat ** 2 # 将差异矩阵平方
                # 方差距离为方差矩阵的整行求和,是一个一位列向量
                sqTestDiffDis = sqTestDiffMat. sum(axis=1)
                testDiffDis = sqTestDiffDis ** 0.5 # 开方生成标准差距离
                sortIndex = np. argsort (testDiffDis) # 将标准差距离按照下标排序
                labelCount = []
                for j in range (neighbors): #考察k近邻属于哪些类
                    labelCount.append(y train[sortIndex[j]][0])
                classifyRes = Counter(labelCount) # 把k近邻中最多的那个标签作为分类结果
                classifyRes = classifyRes.most common(2)[0][0]
                if classifyRes == y test[0]: # 分类正确则将accRate+1
                    acc += 1
```

```
accRate = acc / testround print('k={0}时, 测试个数为{1} 正确个数为: {2} 准确率为: {3}'.format(neighbors, return accRate
```

#### 实验结果:

```
In [16]: MyKnnClassifier(X, y, 1)
MyKnnClassifier(X, y, 3)
MyKnnClassifier(X, y, 5)

k=1时,测试个数为1593 正确个数为: 1459 准确率为: 0.9158819836785939
k=3时,测试个数为1593 正确个数为: 1464 准确率为: 0.9190207156308852
k=5时,测试个数为1593 正确个数为: 1458 准确率为: 0.9152542372881356

Out[16]:
```

## 中级要求

与weka机器学习包中的kNN分类结果进行对比。

下载weka软件,配置所需环境,在k分别取1,3,5的情况下使用weka的knn预测。选择lazy-IBK为weka的KNN Classifier,选择"Cross-Vlidation"并设置Folds=1593,点击"Start",得到结果如图所示:

```
=== Run information ===
            weka.classifiers.lazy.IBk -K 1 -W 0 -A "weka.core.neighboursearch.LinearNNSear
            almost4-weka.filters.unsupervised.attribute.NumericToNominal-Rfirst-last
Instances: 1593
Attributes: 257
            [list of attributes omitted]
Test mode: 1593-fold cross-validation
=== Classifier model (full training set) ===
IB1 instance-based classifier
using 1 nearest neighbour(s) for classification
Time taken to build model: 0.01 seconds
=== Stratified cross-validation ===
=== Summary ===
                                  1459
                                                    91.5882 %
Correctly Classified Instances
Incorrectly Classified Instances
                                   134
                                                     8.4118 %
                                     0.9065
Kappa statistic
Mean absolute error
                                      0.0175
                                      0.126
Root mean squared error
Relative absolute error
                                      9.6972 %
Root relative squared error
                                    41.9716 %
Total Number of Instances
                                    1593
```

```
=== Run information ===
            weka.classifiers.lazy.IBk -K 3 -W 0 -A "weka.core.neighboursearch.LinearNNSear
Relation:
           almost4-weka.filters.unsupervised.attribute.NumericToNominal-Rfirst-last
Instances: 1593
Attributes: 257
            [list of attributes omitted]
Test mode: 1593-fold cross-validation
=== Classifier model (full training set) ===
IB1 instance-based classifier
using 3 nearest neighbour(s) for classification
Time taken to build model: 0 seconds
=== Stratified cross-validation ===
=== Summary ===
Correctly Classified Instances 1436
Incorrectly Classified Instances
                                   157
                                                      9.8556 %
Kappa statistic
                                     0.8905
Mean absolute error
                                      0.0235
Root mean squared error
                                      0.1157
                                   13.0707 %
38.5308 %
Relative absolute error
Relative apportunity Root relative squared error
Total Number of Instances
=== Detailed Accuracy By Class ===
```

```
=== Run information ===
             weka.classifiers.lazy.IBk -K 5 -W 0 -A "weka.core.neighboursearch.LinearNNSea:
             almost4-weka.filters.unsupervised.attribute.NumericToNominal-Rfirst-last
Instances:
             1593
Attributes: 257
             [list of attributes omitted]
Test mode: 1593-fold cross-validation
=== Classifier model (full training set) ===
IB1 instance-based classifier
using 5 nearest neighbour(s) for classification
Time taken to build model: 0 seconds
=== Stratified cross-validation ===
=== Summary ===
Correctly Classified Instances 1445
                                                     90.7094 %
                                   148
Incorrectly Classified Instances
                                       0.8968
Kappa statistic
Mean absolute error
                                       0.0279
                                       0.116
Root mean squared error
                                     15.482 %
Relative absolute error
Root relative squared error
Total Number of Instances
                                    38.6282 %
                                   1593
```

可以看到,K=1时weka结果的acc=91.5882%(自己实现的也为91.5882%),K=3时weka结果的acc=90.1444%(自己实现的为91.9021%),K=5时weka结果的acc=90.7094%(自己实现的为91.5254%)。经比对可看出,weka模型在1593折时效果不如自己实现的KNN,但二者差别十分细微。

### 高级要求

#### 将实验过程结果等图示展出

```
scores1 = []
In [17]:
        scores2 = []
        for k in range (1, 31):
           score1 = MyKnnClassifier(X, y, k)
           scores1. append (score1)
        k=1时,测试个数为1593
                           正确个数为: 1459
                                          准确率为: 0.9158819836785939
        k=2时,测试个数为1593
                           正确个数为: 1459
                                          准确率为: 0.9158819836785939
        k=3时,测试个数为1593
                           正确个数为: 1464
                                          准确率为: 0.9190207156308852
        k=4时,测试个数为1593
                           正确个数为: 1465
                                          准确率为: 0.9196484620213434
        k=5时,测试个数为1593
                           正确个数为: 1458
                                          准确率为: 0.9152542372881356
        k=6时,测试个数为1593
                           正确个数为: 1464
                                          准确率为: 0.9190207156308852
        k=7时,测试个数为1593
                           正确个数为: 1469
                                          准确率为: 0.9221594475831764
                           正确个数为: 1472
        k=8时,测试个数为1593
                                          准确率为: 0.9240426867545511
        k=9时,测试个数为1593
                           正确个数为: 1471
                                          准确率为: 0.9234149403640929
        k=10时,测试个数为1593
                            正确个数为: 1465
                                           准确率为: 0.9196484620213434
        k=11时,测试个数为1593
                            正确个数为: 1456
                                           准确率为: 0.9139987445072191
        k=12时,测试个数为1593
                            正确个数为: 1460
                                           准确率为: 0.9165097300690521
        k=13时,测试个数为1593
                            正确个数为: 1461
                                           准确率为: 0.9171374764595104
        k=14时,测试个数为1593
                            正确个数为: 1453
                                           准确率为: 0.9121155053358443
        k=15时,测试个数为1593
                            正确个数为: 1446
                                           准确率为: 0.9077212806026366
                            正确个数为: 1446
                                           准确率为: 0.9077212806026366
        k=16时,测试个数为1593
        k=17时,测试个数为1593
                            正确个数为: 1442
                                           准确率为: 0.9052102950408035
        k=18时,测试个数为1593
                            正确个数为: 1438
                                           准确率为: 0.9026993094789705
        k=19时,测试个数为1593
                            正确个数为: 1436
                                           准确率为: 0.901443816698054
                            正确个数为: 1435
        k=20时,测试个数为1593
                                           准确率为: 0.9008160703075957
        k=21时,测试个数为1593
                            正确个数为: 1436
                                           准确率为: 0.901443816698054
        k=22时,测试个数为1593
                            正确个数为: 1440
                                           准确率为: 0.903954802259887
        k=23时,测试个数为1593
                            正确个数为: 1435
                                           准确率为: 0.9008160703075957
                            正确个数为: 1427
        k=24时,测试个数为1593
                                           准确率为: 0.8957940991839297
        k=25时,测试个数为1593
                            正确个数为: 1426
                                           准确率为: 0.8951663527934715
        k=26时,测试个数为1593
                            正确个数为: 1424
                                           准确率为: 0.8939108600125549
                            正确个数为: 1424
        k=27时,测试个数为1593
                                           准确率为: 0.8939108600125549
        k=28时,测试个数为1593
                            正确个数为: 1417
                                           准确率为: 0.8895166352793471
        k=29时,测试个数为1593
                            正确个数为: 1423
                                           准确率为: 0.8932831136220967
        k=30时,测试个数为1593
                            正确个数为: 1422
                                           准确率为: 0.8926553672316384
In [18]:
        path = r'./result.csv'
        with open(path) as f:
           f csv = csv. reader(f)
           headers = next(f csv)
           for row in f csv:
               scores2. append(float(row[2]))
        plt. xlabel ('k value:', fontsize=18)
In [22]:
        plt. ylabel ('accuracy', fontsize=18)
        x major locator = plt. MultipleLocator(1)
        ax = plt. gca()
        ax. xaxis. set_major_locator(x_major_locator)
        plt. xlim(0, 31)
        plt. ylim (0.75, 1)
        # 普通kNN分类精度
        plt. plot (range (1, 31), scores1, 'r')
        plt. plot (range (1, 31), scores2, 'b')
        plt. show()
```

