```
kNN 案例
```

6.0

0.87

0.90

0.88

89

MNIST 训练集由来自 250 个人手写的数字构成, 其中 50%是高中学生,50%来自人口普查局的工作人员. 测试集也是同样比例的手写数字数据

```
from sklearn.neighbors import KNeighborsClassifier
from sklearn.metrics import classification report
from sklearn.utils import shuffle
from sklearn.datasets import fetch mldata
from sklearn.cross_validation import train_test_split
# 数据集 MNIST
mnist = fetch mldata("MNIST original")
mnist.data,mnist.target = shuffle(mnist.data,mnist.target)
mnist.data = mnist.data[:1000]
mnist.target = mnist.target[:1000]
X_train,X_test,y_train,y_test =
     train test split(mnist.data,mnist.target,
                     test_size=0.8,random_state=0)
# 训练模型并测试
model = KNeighborsClassifier(3)
model.fit(X_train,y_train)
y_predicted = model.predict(X_test)
print classification_report(y_test,y_predicted)
          precision recall f1-score support
      0.0
              0.75 0.89
                                0.81
                                          81
             0.60
                      0.99
                                0.75
      1.0
                                          83
      2.0
              0.94
                      0.64
                               0.76
                                          92
      3.0
             0.76
                     0.70
                               0.73
                                          87
              0.71
                      0.57
                               0.63
      4.0
                                          74
      5.0
             0.76
                      0.78
                               0.77
                                          54
```

```
7.0
            0.86 0.79
                            0.83
                                      87
     8.0
            0.85 0.56
                            0.67
                                       72
      9.0
            0.65
                    0.73
                            0.69
                                      81
avg / total 0.78 0.76 0.76 800
%timeit clf.fit(X_train,y_train)
100 loops, best of 3: 5.51 ms per loop
%timeit clf.predict(X test)
1 loop, best of 3: 535 ms per loop
算法在训练阶段仅仅拷贝数据。预测速度与训练阶段使用的样本数目和构成数据集的特征
数目有关。其他算法中,预测速度都独立于所使用的数据集的训练样本数。
总之,kNN 对小数据集非常好,但处理大数据集时不宜使用。
# 再看一个例子,介绍人工生成分类数据集及分层划分训练/测试集
from sklearn.datasets import make classification
from sklearn.cross validation import StratifiedShuffleSplit
# 生成一个数据集 (100x4) 及类标签 (0/1)
X,y = make classification(n features=4)
ds = np.column_stack([X,y])
strat_split = StratifiedShuffleSplit(ds[:,-1],test_size=0.2,
                                 n iter=1)
for train idx, test idx in strat split:
  X_train = ds[train_idx,:-1]
  y_train = ds[train_idx,-1]
  X test = ds[test idx,:-1]
  y_test = ds[test_idx,-1]
model = KNeighborsClassifier(n neighbors=3)
model.fit(X_train,y_train)
# 对训练集预测评估
y predicted = model.predict(X train)
```

```
print classification_report(y_train,y_predicted)
         precision recall f1-score support
              1.00
      0.0
                     0.95
                              0.97
                                        40
              0.95
                      1.00
                              0.98
      1.0
                                        40
              0.98 0.97 0.97
avg / total
                                        80
# 对测试集预测
y_predicted = model.predict(X_test)
accuracy_score(y_test,y_predicted)
0.95
confusion_matrix(y_test,y_predicted)
array([[ 9, 1],
      [ 0, 10]])
print classification_report(y_test,y_predicted)
         precision recall f1-score support
      0.0
              1.00
                      0.90
                              0.95
                                        10
      1.0
              0.91
                      1.00
                              0.95
                                        10
```

avg / total 0.95 0.95 0.95

20