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# Bagging 演示
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from sklearn.datasets import make classification
from sklearn.neighbors import KNeighborsClassifier
from sklearn.ensemble import BaggingClassifier
from sklearn.metrics import classification report
from sklearn.cross_validation import train_test_split
                      # 特征个数
n f = 30
inf_f = int(0.6 * n_f) # 60%含信息特征
red f = int(0.1 * n f) # 10% 冗余特征(含信息特征的线性组合)
rep f = int(0.1 * n f) # 10% 重复特征(随机取自含信息特征和冗余特征)
# 创建分类数据集
X,y = make_classification(
                       # 实例个数
   n samples=500,
                      # 随机抽 3%实例改变类别,产生一些噪声
   flip y=0.03,
   n features=n f,
   n informative=inf f,
   n redundant=red f,
   n repeated=rep f,
   random state=7)
# 划分为训练集,验证集,测试集
X train,X test all,y train,y test all =
   train test split(X,y,test size=0.3,random state=9)
X_dev,X_test,y_dev,y_test =
   train_test_split(X_test_all,y_test_all,test_size=0.3,
      random_state=9)
# kNN 模型
knn = KNeighborsClassifier()
knn.fit(X_train,y_train)
# 在训练集上预测
y_pred = knn.predict(X_train)
print classification report(y train, y pred)
          precision recall f1-score support
         0
               0.88
                       0.87
                               0.88
                                         181
               0.87
                       0.88
                               0.87
                                         169
avg / total 0.87 0.87 0.87 350
```

```
# bagging 集成
bagging = BaggingClassifier(
   KNeighborsClassifier(),
                          # 构造 100 个 kNN 模型
   n estimators=100,
   bootstrap=True,
   max samples=1.0,
                           # Bootstrap 样本大小用所有实例
   bootstrap features=True,
                           # Bootstrap 特征使用 70%,各模型不一样
   max features=0.7,
   random state=9)
bagging.fit(X train,y train)
y_pred = bagging.predict(X_train)
print classification_report(y_train,y_pred)
          precision recall f1-score
                                         support
          0
                0.94
                         0.97
                                  0.95
                                            181
          1
                0.96
                         0.93
                                  0.95
                                            169
avg / total
                0.95
                         0.95
                                  0.95
                                            350
# 看几个模型各自随机抽取的特征,很不一样,所以各模型有变化.
for i,f_set in enumerate(bagging.estimators_features_[0:5]):
   print "estimator %d" % (i+1), f set
estimator 1 [20 10 6 17 18 11 17 9 14 3 10 10 23 22 18 17 11 21 20 1 16]
estimator 2 [ 3 27 28 20 20 27 25 0 21 1 12 20 21 29 1 0 28 16 4 9 10]
estimator 3 [ 5 23 19 2 16 21 4 13 27 1 15 24 5 14 1 4 25 22 26 29 15]
estimator 4 [23 10 16 7 22 11 0 14 14 17 8 17 27 12 13 23 8 7 27 0 27]
estimator 5 [ 0 26 13 23 7 27 15 18 11 26 18 26 3 22 6 11 21 6 12 19 7]
# 在验证集上预测
y pred = knn.predict(X dev)
print classification_report(y_dev,y_pred)
          precision recall f1-score support
          0
                0.83
                         0.84
                                  0.83
                                             51
          1
                0.85
                         0.83
                                  0.84
                                             54
avg / total
                0.84
                         0.84
                                  0.84
                                            105
y_pred = bagging.predict(X_dev)
print classification_report(y_dev,y_pred)
          precision recall f1-score support
                0.85
                         0.88
                                  0.87
                                             51
```

1 0.88 0.85 0.87 54

avg / total 0.87 0.87 0.87 105