## # 交叉验证案例

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# 手动 k 折和分层 k 折交叉验证
from sklearn.datasets import load iris
from sklearn.cross validation import KFold
from sklearn.cross_validation import StratifiedKFold
iris = load_iris()
X = iris['data']
y= iris['target']
# k折
kfolds = KFold(n = y.shape[0],n_folds=3)
for train, test in kfolds:
   print "Fold %d: X train shape" % (fc), X[train].shape
   print "Fold %d: X_test shape" % (fc), X[test].shape
   fc += 1
Fold 1: X train shape (100, 4)
Fold 1: X_test shape (50, 4)
Fold 2: X train shape (100, 4)
Fold 2: X test shape (50, 4)
Fold 3: X train shape (100, 4)
Fold 3: X_test shape (50, 4)
# 分层 k 折
skfolds = StratifiedKFold(y,n folds=3)
fc = 1
for train, test in skfolds:
   print "Fold %d: X_train shape" % (fc), X[train].shape
   print "Fold %d: X_test shape" % (fc), X[test].shape
   y train = y[train]
   y_test = y[test]
   print "Train Class Distribution"
   class dist = {}
   total = 0
   for entry in y train:
      try:
          class_dist[entry] += 1
      except KeyError:
          class_dist[entry] = 1
      total += 1
   for k,v in class_dist.items():
      print "class %d percentage = %0.2f" % (k,1.0*v/total)
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print "Test Class Distribution"
   class_dist = {}
   total = 0
   for entry in y test:
      try:
          class dist[entry] += 1
      except KeyError:
          class dist[entry] = 1
      total += 1
   for k,v in class dist.items():
      print "class %d percentage = %0.2f" % (k,1.0*v/total)
   fc += 1
Fold 1: X train shape (99, 4)
Fold 1: X test shape (51, 4)
Train Class Distribution
class 0 percentage = 0.33
class 1 percentage = 0.33
class 2 percentage = 0.33
Test Class Distribution
class 0 percentage = 0.33
class 1 percentage = 0.33
class 2 percentage = 0.33
Fold 2: X_train shape (99, 4)
Fold 2: X test shape (51, 4)
Train Class Distribution
class 0 percentage = 0.33
class 1 percentage = 0.33
class 2 percentage = 0.33
Test Class Distribution
class 0 percentage = 0.33
class 1 percentage = 0.33
class 2 percentage = 0.33
Fold 3: X train shape (102, 4)
Fold 3: X_test shape (48, 4)
Train Class Distribution
class 0 percentage = 0.33
class 1 percentage = 0.33
class 2 percentage = 0.33
Test Class Distribution
```

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class 0 percentage = 0.33
class 1 percentage = 0.33
class 2 percentage = 0.33
# 案例:对 Boston 数据集的网格搜索交叉验证
from sklearn.datasets import load boston
from sklearn.cross validation import train test split
from sklearn.linear_model import Ridge
from sklearn.grid search import GridSearchCV
from sklearn.metrics import mean squared error
from sklearn.preprocessing import PolynomialFeatures
import numpy as np
bos = load boston()
X = bos['data']
y = bos['target']
# 为训练集测试集添加多项式特征
X_train,X_test,y_train,y_test =
   train_test_split(X,y,test_size=0.3,random_state=9)
poly f = PolynomialFeatures(interaction only=True)
poly f.fit(X train)
X train poly = poly f.transform(X train)
X_test_poly = poly_f.transform(X_test)
model = Ridge(normalize=True)
alpha range = np.linspace(0.0015,0.0017,30)
gparam = {'alpha':alpha range}
grid = GridSearchCV(estimator=model,param grid = gparam,
                    cv=5,scoring='mean squared error')
grid.fit(X_train_poly,y_train)
f=1
for s in grid.grid_scores_:
   print "Fold %d MSE %0.2f" % (f,abs(s[1])),s[0]
   f += 1
Fold 1 MSE 14.24 { 'alpha': 0.0015}
Fold 2 MSE 14.24 {'alpha': 0.001506896551724138}
Fold 3 MSE 14.24 {'alpha': 0.0015137931034482758}
Fold 19 MSE 14.24 {'alpha': 0.0016241379310344827}
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Fold 20 MSE 14.25 {'alpha': 0.0016310344827586206}
Fold 21 MSE 14.25 {'alpha': 0.0016379310344827587}
Fold 22 MSE 14.25 {'alpha': 0.0016448275862068966}
Fold 29 MSE 14.25 {'alpha': 0.001693103448275862}
Fold 30 MSE 14.25 {'alpha': 0.0017}
print grid.best_params_
{'alpha': 0.0015}
# 交叉验证为我们找到的最佳模型
chosen_model = grid.best_estimator_
y_pred = chosen_model.predict(X_train_poly)
print "MSE = %0.2f" % (mean_squared_error(y_train,y_pred))
MSE = 7.57
for i,c in enumerate(chosen model.coef):
   print "Coefficient %d %0.3f" % (i+1,c)
print "Intercept %0.3f" % (choosen_model.intercept_)
Coefficient 1 0.000
Coefficient 2 0.145
Coefficient 3 -0.126
Coefficient 4 0.260
Coefficient 92 -0.001
Intercept -78.963
y_pred = chosen_model.predict(X_test_poly)
print "MSE = %0.2f" % (mean squared error(y test,y pred))
MSE = 11.72
```