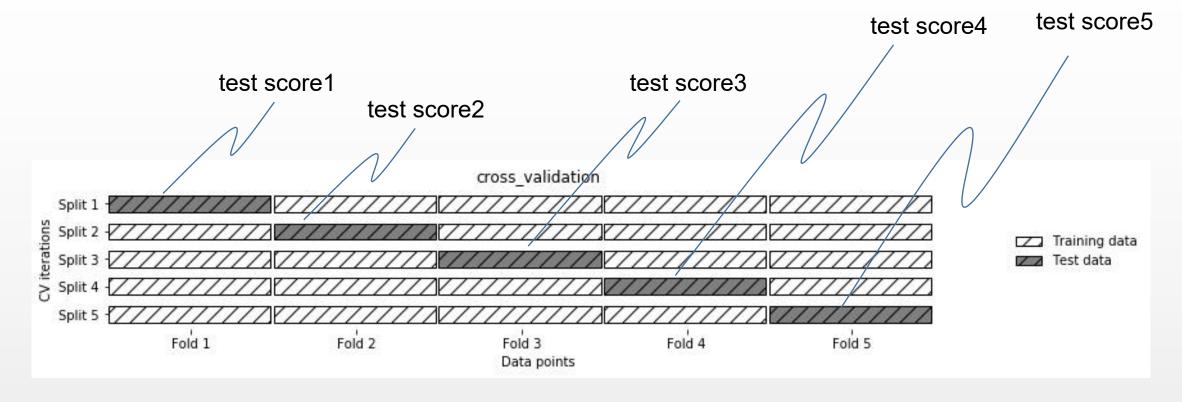
模型改进

交叉验证



使用交叉验证

scores = cross_val_score(<model>, X, y, cv = 5)

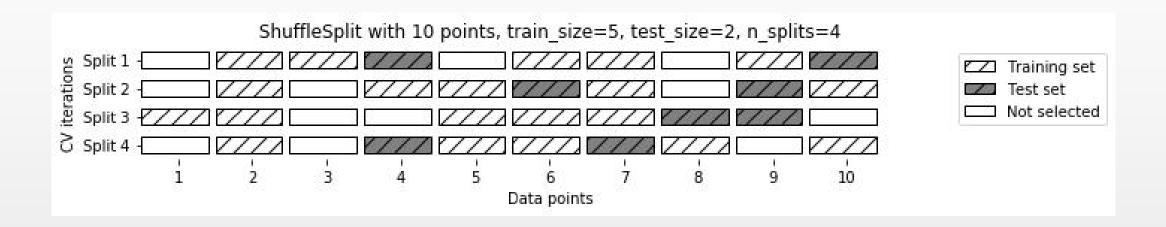
X, y 使用原数据,不需要划分 training 和 test

简单交叉验证与分层交叉验证

- 对于回归问题: cross_val_score 默认使用简单交叉验证
- 对于分类问题: cross_val_score 默认使用分层交叉验证

打乱划分交叉验证

- ShuffleSplit
- StratifiedShuffleSplit



网格搜索

训练集 training set 验证集 validation set

测试集

test set

网格搜索 + 交叉验证

test_score = model.score(X_test, y_test)

```
trainval set 
 test set
```

GridSearchCV

```
model = <model>
grid_search = GridSearchCV(model, params)
grid_search.fit(X_train, y_train)

grid_search.best_params_
grid_search.best_score_
grid_search.best_estimator_

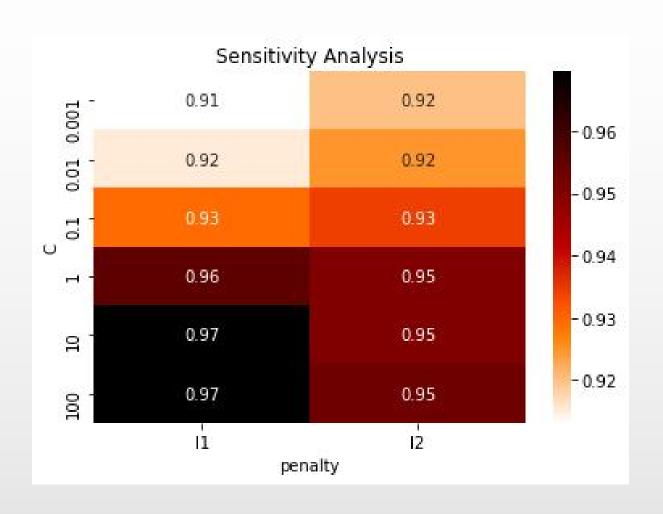
grid_search.score(X_test, y_test) = grid_search.best_estimator_.score(X_test, y_test)
y_pred = grid_search.predict(X_test) = grid_search.best_estimator_.pred(X_test)
```

cv_results_

params	mean_test_score
('C': 10, 'penalty': '11'}	0.969484
('C': 100, 'penalty': '11'}	0.969484
('C': 1, 'penalty': '11'}	0.955399
('C': 100, 'penalty': '12'}	0.953052
('C': 1, 'penalty': '12'}	0.950704
('C': 10, 'penalty': '12'}	0.950704
('C': 0.1, 'penalty': '12'}	0.934272
{'C': 0.1, 'penalty': '11'}	0.929577
('C': 0.01, 'penalty': '12'}	0.924883
('C': 0.001, 'penalty': '12'}	0.920188
('C': 0.01, 'penalty': '11'}	0.915493
{'C': 0.001, 'penalty': '11'}	0.913146

```
['mean_fit_time',
 'std_fit_time',
'mean_score_time',
 'std_score_time',
 'param_C',
 'param_penalty',
 'params',
 'split0_test_score',
 'split1_test_score',
 'split2_test_score',
 'split3_test_score',
 'split4_test_score',
 'mean_test_score',
 'std test score',
 'rank_test_score']
```

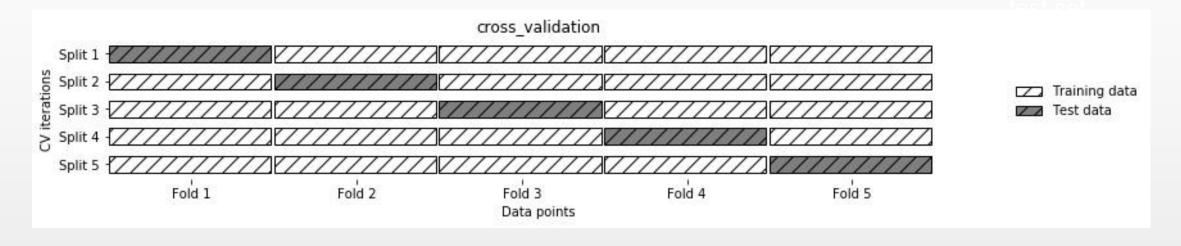
参数敏感性



在非网格空间中搜索

嵌套交叉验证

```
grid_search = GridSearchCV(<model>, params) 将 grid_search 看成一个模型 scores = cross_val_score(grid_search, X, y)
```



在模型选择中使用评估指标

使用 GridSearchCV, cross_val_score 的 scoring 参数

分类问题	回归问题
accuracy (默认) roc auc	r2 (默认值) mean_squared_error
average_precision	mean_absolute_error
f1 f1_macro	
f1_micro	
f1_weighted	

测试集 test set