

調參數

Learning rate

```
forest = XGBClassifier()
# Try different numbers of n_estimators - this will take a minute
estimators = [10, 1, 0.1, 0.01, 0.001, 0.0001, 0.00001]
scores = []
for n in estimators:
    forest.set_params(learning_rate=n)
    forest.fit(X_train, y_train)
    test_y_predicted = forest.predict(X_test)
    scores.append(accuracy_score(y_test, test_y_predicted))
    #scores.append(model.score(x_test, y_test))
    print(accuracy_score(y_test, test_y_predicted))
plt.title("Effect of lr")
plt.xlabel("lr")
plt.ylabel("score")
plt.plot(estimators, scores)
```

```
0.22907099602829656
0.600128194996883
0.5516189740302693
0.5432921528857045
0.5426014218522713
0.542607275505097
0.542607275505097
```

Max depth

```
from xgboost import XGBClassifier
forest = XGBClassifier()
# Try different numbers of n_estimators - this will take a minute
estimators = np.arange(12, 53, 4)
scores = []
for n in estimators:
    forest.set_params(learning_rate=1, max_depth=n)
    forest.fit(X_train, y_train)
    test_y_predicted = forest.predict(X_test)
    scores.append(accuracy_score(y_test, test_y_predicted))
    #scores.append(model.score(x_test, y_test))
    print(accuracy_score(y_test, test_y_predicted))
plt.title("Effect of max_depth")
plt.xlabel("max_depth")
plt.ylabel("score")
plt.plot(estimators, scores)
```

```
0.7002461461013209
0.736325135292551
0.738584645283273
0.7370217199788098
```

N_estimators

```
warnings.filterwarnings('ignore')

from xgboost import XGBClassifier
forest = XGBClassifier()
# Try different numbers of n_estimators - this will take a minute
estimators = np.arange(20, 201, 20)
scores = []
for n in estimators:
    forest.set_params(learning_rate=1, max_depth=20, n_estimators=n)
    forest.fit(X_train, y_train)
    test_y_predicted = forest.predict(X_test)
    scores.append(accuracy_score(y_test, test_y_predicted))
    #scores.append(model.score(x_test, y_test))
    print(accuracy_score(y_test, test_y_predicted))
plt.title("Effect of n_estimators")
plt.xlabel("n_estimator")
plt.ylabel("score")
plt.plot(estimators, scores)
```

```
0.7278519728273436
0.7356812334817234
0.7373758659747649
0.7383768406079604
0.738584645283273
0.7389914741546594
0.7389856205018337
```

最後選擇: learning_rate=1,max_depth=20,n_estimators=120

Mean Square Error - test: 0.296
 Mean Square Error - train: 0.042
 Mean Absolute Error - test: 0.273
 Mean Absolute Error - train: 0.040
 Accuracy - test: 0.739
 Accuracy - train: 0.961

Classification Report:				
	precision	recall	f1-score	support
-1	0.71	0.66	0.68	78281
0	0.75	0.80	0.78	185120
1	0.73	0.67	0.70	78266
micro avg	0.74	0.74	0.74	341667
macro avg	0.73	0.71	0.72	341667
weighted avg	0.74	0.74	0.74	341667

Accuracy: 0.7389914741546594
 Precision: 0.7307161841183701
 Recall: 0.7101680269345952
 F-1: 0.7194464825948663
 AUC: 0.7776641860748198
 (peak=-1)
 Confusion Matrix:
 [[242031 26962]
 [21355 51319]]
 True positives: 242031
 False positives: 21355
 True negatives: 51319
 False negatives: 26962
 AUC: 0.8029602079383822

(peak=0)
 Confusion Matrix:
 [[107943 36591]
 [48604 148529]]
 True positives: 107943
 False positives: 48604
 True negatives: 148529
 False negatives: 36591
 AUC: 0.7501401490591803

(peak=1)
 Confusion Matrix:
 [[107943 36591]
 [48604 148529]]
 True positives: 107943
 False positives: 48604
 True negatives: 148529
 False negatives: 36591
 AUC: 0.7501401490591803

