# 450 samples Xgboost-bacteria

## 1. Data preprocess

```
1. scaler_bac = StandardScaler().fit(X_bac)
2. X_bac_lasso = scaler_bac.transform(X_bac)
3.
4. kfold = sk.model_selection.StratifiedKFold(n_splits=10, shuffle=True, random_state=0)
5. regr = LassoCV(cv=kfold)
6. regr.fit(X_bac_lasso, y_bac)
7. model_coef = pd.Series(regr.coef_, index = list(X_bac_all.columns[:]))
8. print("Variables Kept: ", str(sum(model_coef != 0)))  # Variables Kep t: 19
9.
10. ## get the importance features
11. mask = top_coef != 0
12. X_bac_lasso = X_bac_lasso.loc[:, mask]
```

## 2. Xgboost classifier

## 2.1 Split data

```
    X_bac_tain, X_bac_test, y_bac_train, y_bac_test = train_test_split(X_bac_lasso, y_bac, shuffle=True, test_size=0.2, random_state=31, stratify=y_bac)
```

## 2.2 Tuning parameters

#### 2.2.1 n\_estimators

```
1. kfold = StratifiedKFold(n_splits=5, shuffle=True, random_state=0)
2. param_test1 = {
```

```
'n estimators':range(1,100,1)
gsearch1 = GridSearchCV(
    estimator = XGBClassifier(
        booster = "qbtree",
        learning rate = 0.1,
        n = 17,
        max depth = 5,
        min child weight = 1,
        gamma = 0.1,
        subsample = 1.0,
        colsample by tree = 0.8,
        colsample bylevel = 1.0,
        reg alpha = 0,  # L1 regularization parameter
        reg lambda = 1.0,  # L2 regularization parameter
        objective = 'binary:logistic',
        nthread = 8,
        scale_pos_weight = 3.6,
        seed = 27
        eval metric = ["auc",],
        early stopping rounds=20,
   ),
    param grid = param test1,
   scoring = 'roc auc',
   n jobs = 8,
    cv = kfold
    refit=True,
    return train score=True,
)
gsearch1.fit( X bac train, y bac train, eval metric=["auc",] )
print(gsearch1.best params) # {'n estimators': 22}
```

### 2.2.2 max\_depth & min\_child\_weight

```
param_test2 = {
    'max_depth':range(1,10,1),
    'min_child_weight':range(1,10)
}

gsearch2 = GridSearchCV(
    estimator = XGBClassifier(
```

```
booster = "gbtree",
        learning rate = 0.1,
        n = 22,
       max depth = 5,
       min child weight = 1,
       gamma = 0.1
        subsample = 1.0,
        colsample by tree = 0.8,
       colsample bylevel = 1.0,
        reg alpha = 0,  # L1 regularization parameter
        reg lambda = 1.0,  # L2 regularization parameter
       objective = 'binary:logistic',
       nthread = 8,
        scale pos weight = 3.6,
        seed = 27
       eval metric = ["auc",],
       early stopping rounds=20,
   ),
   param grid = param test2,
   scoring = 'roc auc',
   n jobs = 8,
   cv = kfold
   refit=True,
   return train score=True,
gsearch2.fit(X_bac_tain, y_bac_tain, eval_metric="auc")
print(gsearch2.best params) # {'max depth': 3, 'min child weight':
```

### 2.2.3 gamma

```
#gsearch2.best params ["min child weight"]
        gamma = 0.1,
        subsample = 1.0,
        colsample by tree = 0.8,
        colsample bylevel = 1.0,
        reg alpha = 0,  # L1 regularization parameter
        reg lambda = 1.0,  # L2 regularization parameter
        objective = 'binary:logistic',
        nthread = 8,
        scale pos weight = 3.6,
        seed = 27
        eval metric = ["auc",],
        early stopping rounds=20,
   ),
   param grid = param test3,
   scoring = 'roc auc',
   n jobs = 8,
   cv = kfold
   refit=True,
   #verbose=2,
   return train score=True,
gsearch3.fit(X bac tain, y bac train, eval metric="auc")
print(gsearch3.best params) # {'gamma': 3.8}
```

#### 2.2.4 Subsample & Colsample\_bytree

```
## subsample and colsample bytree
 param test4 = {
 'colsample bylevel':[i/10.0 for i in range(2,11)],
  'colsample bytree':[i/10.0 for i in range(2,11)]
 }
gsearch4 = GridSearchCV(
     estimator = XGBClassifier(
         booster = "gbtree",
         learning rate = 0.1,
         n = 22,
                               # gsearch1.best params ["n estimators"]
         \max depth = 3,
                                 # gsearch2.best params ["max depth"]
         min child weight = 6,
 gsearch2.best params ["min child weight"]
         gamma = 3.8,
                                 # gsearch3.best params ["gamma"]
         subsample = 1.0,
```

```
colsample bytree = 0.8,
        colsample bylevel = 1.0,
        reg alpha = 0,  # L1 regularization parameter
        reg lambda = 1.0,  # L2 regularization parameter
        objective = 'binary:logistic',
        nthread = 8,
        scale pos weight = 3.6,
        seed = 27
        eval metric = ["auc",],
        early stopping rounds=20,
    ),
   param grid = param test4,
   scoring = 'roc auc',
   n jobs = 8,
   cv = kfold,
   refit=True,
   #verbose=2,
   return train score=True,
gsearch4.fit(X bac tain, y bac tain, eval metric="auc")
print(gsearch4.best params ) # {'colsample bylevel': 1.0,
'colsample bytree': 0.8}
```

### 2.2.5 reg\_alpha & reg\_lambda

```
param test5 = {
'reg alpha': [0, 1e-5, 1e-2, 0.05, 0.075, 0.1, 0.15, 0.2, 0.25, 0.5, 1
, 10],
'reg lambda': [0, 1e-2, 0.05, 0.075, 0.1, 0.25, 0.3, 0.4, 0.5, 0.6, 0.7
, 0.75, 0.8, 0.9, 1, 5, 10]
gsearch5 = GridSearchCV(
   estimator = XGBClassifier(
        booster = "gbtree",
        learning rate = 0.1,
        n = 22,
                              # gsearch1.best params ["n estimators"]
        max depth = 3,
                                # gsearch2.best params ["max depth"]
        min child weight = 6,
gsearch2.best params ["min child weight"]
                                # gsearch3.best params ["gamma"]
        qamma = 3.8,
        subsample = 1.0,
        colsample bytree = 0.8, #
```

```
gsearch4.best params ["colsample bytree"]
        colsample bylevel = 1.0, # gsearch4.best params ["colsample byle
vel"]
        reg alpha = 0,
                               # L1 regularization parameter
       reg lambda = 1.0,  # L2 regularization parameter
        objective = 'binary:logistic',
       nthread = 8,
        scale pos weight = 3.6,
        seed = 27,
       eval metric = ["auc",],
        early stopping rounds=20,
   ),
   param grid = param test5,
   scoring = 'roc auc',
   n jobs = 8,
   cv = kfold
   refit=True,
   #verbose=2,
   return train score=True,
gsearch5.fit(X bac tain, y bac tain, eval metric="auc")
print(gsearch5.best params ) # {'reg alpha': 0.075, 'reg lambda': 0.
9}
```

```
param test6 = {
 "learning rate": [1e-3, 1e-2, 0.05, 0.1, 0.5],
 }
gsearch6 = GridSearchCV(
    estimator = XGBClassifier(
        booster = "gbtree",
        learning rate = 0.1,
        n estimators = gsearch1.best params ["n estimators"],
        max depth = gsearch2.best params ["max depth"],
        min child weight = gsearch2.best params ["min child weight"],
        gamma = gsearch3.best params ["gamma"],
        subsample = 1.0,
        colsample bytree = gsearch4.best params ["colsample bytree"],
        colsample bylevel = 1.0,
gsearch4.best params ["colsample bylevel"],
        objective = 'binary:logistic',
        nthread = 8,
        scale pos weight = 3.6,
        reg alpha = gsearch5.best params ["reg alpha"],
```

```
reg_lambda = gsearch5.best_params_["reg_lambda"],
seed=27,
eval_metric = ["auc",],
early_stopping_rounds=20,

),
param_grid = param_test6,
scoring = 'roc_auc',
n_jobs = 8,
cv = kfold,
refit=True,
return_train_score=True,

return_train_score=True,

gsearch6.fit(X_bac_tain, y_bac_tain, eval_metric="auc")
print(gsearch6.best_params_) # {'learning_rate': 0.1}
```

```
clf para = XGBClassifier(
    \#silent = 0, \#
    nthread = 8, # CPU threads
    learning rate = gsearch6.best params ["learning rate"],
    booster = "gbtree",
    n estimators = gsearch1.best params ["n estimators"],
    min child weight = gsearch2.best params ["min child weight"],
    max depth = gsearch2.best params ["max depth"],
    gamma = gsearch3.best params ["gamma"],
    subsample = 1.0, #gsearch4.best params ["subsample"],
    colsample bytree = gsearch4.best params ["colsample bytree"],
    colsample bylevel = 1.0,
#gsearch4.best params ["colsample bylevel"],
    reg lambda = gsearch5.best params ["reg lambda"],
    reg alpha = gsearch5.best params ["reg alpha"], # L1
regularization parameter
    scale pos weight = 3.6,
                                                     # L2 regularization
parameter
    objective = 'binary:logistic',
    seed = 27
    val metric = ["auc", ],
)
clf para.fit(X bac tain, y bac tain)
```