Ensemble 퀴즈

grid_search.fit(X_train, y_train)

주어진 데이터를 활용하여 모델을 적합하고, 최적의 모델을 찾으세요.

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• Ensemble 실습 퀴즈 (bagging / boosting / BGM)
 • 모델은, RandomForestClassifier, GradientBoostingClassifier, lightgbm 을 사용하세요.
 • 각 모델별로 최적의 파라미터를 찾으세요. (평가는 f1 score)
 • 모델간의 평가는 ROC 값으로 하고, ROC 커브로 비교하세요.
                                                                                                         In [23]:
# import library
import sklearn
import pandas as pd
import numpy as np
from sklearn.model_selection import train_test_split, GridSearchCV
from sklearn.metrics import confusion_matrix, roc_curve
import warnings
warnings.simplefilter("ignore", UserWarning)
                                                                                                          In [5]:
filename = './data/pima-indians-diabetes.csv'
dataframe = pd.read csv(filename, header=None)
dataframe.columns = ['Pregnancies', 'Glucose', 'BloodPressure', 'SkinThickness', 'Insulin', 'BMI', 'DiabetesPe
array = dataframe.values
X = array[:,0:8].astype(float) # 0 - 7 column은 독립변수
Y = array[:,8].astype(int) # 마지막 column은 종속변수
X train, X test, y train, y test = train test split(X, Y, test size=0.3, random state=0)
RandomForestClassifier
                                                                                                          In [6]:
hyperparameter를 정의하세요
param_grid = {'n_estimators': [100, 200],
              'oob score': [True], # compute out of bag error
              'n jobs':[-1],
              'max_depth': [3, 5]
                                                                                                          In [7]:
. . .
모델을 정의하세요
from sklearn.ensemble import RandomForestClassifier
rf model = RandomForestClassifier()
                                                                                                          In [8]:
hyperparameter를 search 하세요
grid search = GridSearchCV(rf model, param grid=param grid, cv=5, scoring='f1')
```

```
Out[8]:
            {\tt GridSearchCV}
 ▶ estimator: RandomForestClassifier
       RandomForestClassifier
In [9]:
. . .
최적의 파라미터로 최적의 모델을 결정하세요
rf_opt = grid_search.best_estimator_
GradientBoostingClassifier
                                                                                                    In [10]:
hyperparameter를 정의하세요
param_grid = {'n_estimators': [100, 200],
             'learning_rate': [0.01, 0.001, 0.0001], 'max_depth': [1, 3, 5]
                                                                                                    In [11]:
. . .
모델을 정의하세요
from sklearn.ensemble import GradientBoostingClassifier
gb_model = GradientBoostingClassifier()
                                                                                                    In [12]:
. . .
hyperparameter를 search 하세요
grid search = GridSearchCV(gb model, param grid=param grid, cv=5, scoring='f1')
grid search.fit(X train, y train)
                                                                                                   Out[12]:
              GridSearchCV
▶ estimator: GradientBoostingClassifier
      GradientBoostingClassifier
In [13]:
. . .
최적의 파라미터로 최적의 모델을 결정하세요
gb_opt = grid_search.best_estimator_
lightgbm
                                                                                                    In [15]:
. . .
hyperparameter를 정의하세요
param dict = { "objective":['binary'], # multiclass, regression
             "max depth": [25,50, 75],
             "learning_rate" : [0.01,0.05,0.1],
             "num leaves": [300,900,1200],
             "n_estimators": [200]
             }
                                                                                                    In [16]:
```

모델을 정의하세요

```
import lightgbm as lgb
lg = lgb.LGBMClassifier(silent=True)
                                                                                                     In [17]:
hyperparameter를 search 하세요
grid_search = GridSearchCV(lg, n_jobs=-1, param_grid=param_dict, cv = 3, scoring="accuracy")
grid_search.fit(X_train,y_train)
                                                                                                    Out[17]:
        GridSearchCV
 ▶ estimator: LGBMClassifier
       LGBMClassifier
   _____
                                                                                                     In [18]:
. . .
최적의 파라미터로 최적의 모델을 결정하세요
lg opt = grid search.best estimator
ROC 커브로 모델 비교
                                                                                                     In [22]:
. . .
ROC 커브로 최적의 모델을 산출하세요
from sklearn.metrics import plot roc curve
import matplotlib.pyplot as plt
fig = plt.figure()
ax = fig.gca()
plot_roc_curve(rf_opt, X_test, y_test, ax=ax)
plot roc curve (gb opt, X test, y test, ax=ax)
plot_roc_curve(lg_opt, X_test, y_test, ax=ax)
plt.show()
```

/home/restful3/anaconda3/envs/trading/lib/python3.8/site-packages/sklearn/utils/deprecation.py:87: FutureWarni ng: Function plot_roc_curve is deprecated; Function :func:`plot_roc_curve` is deprecated in 1.0 and will be re moved in 1.2. Use one of the class methods: :meth:`sklearn.metric.RocCurveDisplay.from_predictions` or :meth:`sklearn.metric.RocCurveDisplay.from_estimator`.

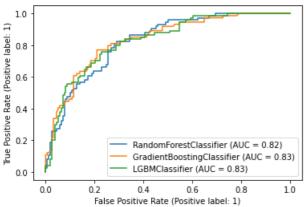
warnings.warn(msg, category=FutureWarning)

/home/restful3/anaconda3/envs/trading/lib/python3.8/site-packages/sklearn/utils/deprecation.py:87: FutureWarning: Function plot_roc_curve is deprecated; Function: func:`plot_roc_curve` is deprecated in 1.0 and will be removed in 1.2. Use one of the class methods: :meth:`sklearn.metric.RocCurveDisplay.from_predictions` or :meth:`sklearn.metric.RocCurveDisplay.from_estimator`.

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