산업인공지능개론

XGBoost, LightGBM

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CONTENS

01. XGBoost

02. LightGBM

코드 구현

```
import numpy as np
   import pandas as pd
   from sklearn.datasets import load boston
   from sklearn.metrics import mean squared error
   from sklearn.model selection import train test split
   import xgboost as xgb
   boston = load boston()
   data = pd.DataFrame(boston.data)
   data.columns = boston.feature names
   data['PRICE'] = boston.target
11 print(data.head())
12 X, y = data.iloc[:,:-1],data.iloc[:,-1]
   X train, X test, y train, y test = train test split(X, y, test size=0.2, random state=123)
   xg reg = xgb.XGBRegressor(objective ='reg:squarederror', colsample bytree = 0.3,
                learning rate = 0.1, max depth = 5, alpha = 10, n estimators = 10)
  xg reg.fit(X train, y train)
   preds = xg reg.predict(X test)
18 rmse = np.sqrt(mean_squared_error(y test, preds))
19 print("RMSE: %f" % (rmse))
```

01. XGBoost

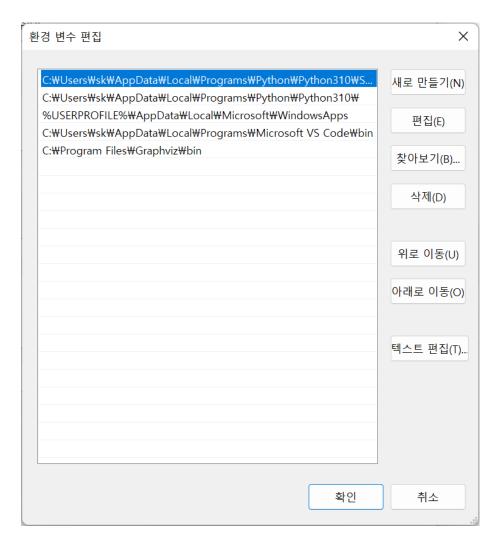
결과

	CRIM	ZN	INDUS	CHAS	NOX	RM	AGE	DIS	RAD	TAX	PTRATIO	В	LSTAT	PRICE
0	0.00632	18.0	2.31	0.0	0.538	6.575	65.2	4.0900	1.0	296.0	15.3	396.90	4.98	24.0
1	0.02731	0.0	7.07	0.0	0.469	6.421	78.9	4.9671	2.0	242.0	17.8	396.90	9.14	21.6
2	0.02729	0.0	7.07	0.0	0.469	7.185	61.1	4.9671	2.0	242.0	17.8	392.83	4.03	34.7
3	0.03237	0.0	2.18	0.0	0.458	6.998	45.8	6.0622	3.0	222.0	18.7	394.63	2.94	33.4
4	0.06905	0.0	2.18	0.0	0.458	7.147	54.2	6.0622	3.0	222.0	18.7	396.90	5.33	36.2
RMSE: 10.423243														

코드 구현

```
1 import matplotlib.pyplot as plt
 2 from lightgbm import LGBMClassifier, LGBMRegressor
 3 from lightgbm import plot_importance, plot_metric, plot_tree
 4 from sklearn.datasets import load iris
 5 from sklearn.model_selection import train_test_split
 6 from sklearn.model_selection import cross_validate
 8 iris = load_iris()
 9 X_train, X_test, y_train, y_test = train_test_split(iris.data, iris.target, test_size=0.2, random_state=123)
10 lgbmc = LGBMClassifier(n estimators=400)
11 evals = [(X_test, y_test)]
12 lgbmc.fit(X_train, y_train, early_stopping_rounds=100, eval_metric='logloss', eval_set=evals, verbose=True)
13 preds = lgbmc.predict(X test)
15 cross val = cross validate(
        estimator=lgbmc,
16
        X=iris.data, y=iris.target,
        cv=5
21 print('avg fit time: {} (+/- {})'.format(cross_val['fit_time']. mean(), cross_val['fit_time'].std()))
22 print('avg score time: {} (+/- {})'.format(cross_val['score_time']. mean(), cross_val['score_time'].std()))
23 print('avg test score: {} (+/- {})'.format(cross val['test score'].mean(), cross val['test score'].std()))
25 fig, ax = plt.subplots(figsize=(12, 8))
26 plot metric(lgbmc)
27 plot_importance(lgbmc, figsize=(10,12))
28 plot_tree(lgbmc, ax=ax, figsize=(28,24))
29 plt.show()
```

Graphviz 설치 후 환경 변수 Graphviz 경로 추가

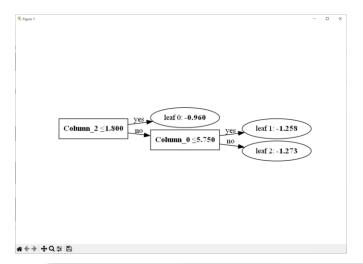


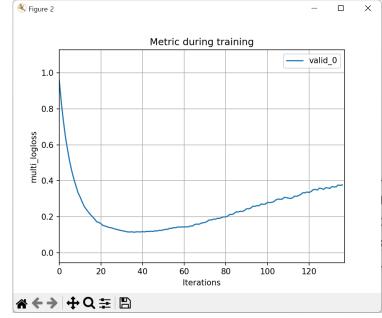
결과

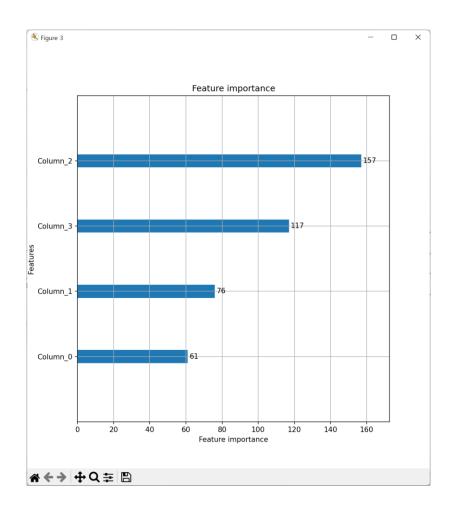
```
valid 0's multi logloss: 0.95847
        valid_0's multi_logloss: 0.832184
        valid_0's multi_logloss: 0.731164
        valid 0's multi logloss: 0.641056
        valid 0's multi logloss: 0.571726
        valid_0's multi_logloss: 0.507286
        valid_0's multi_logloss: 0.454933
        valid_0's multi_logloss: 0.410205
        valid_0's multi_logloss: 0.372194
        valid 0's multi logloss: 0.333919
        valid_0's multi_logloss: 0.310212
[12]
[13]
[14]
        valid_0's multi_logloss: 0.282326
        valid_0's multi_logloss: 0.257165
        valid_0's multi_logloss: 0.240836
        valid 0's multi logloss: 0.225383
        valid_0's multi_logloss: 0.211583
        valid_0's multi_logloss: 0.199289
[18]
[19]
[20]
[21]
[22]
[23]
[24]
[25]
[26]
        valid_0's multi_logloss: 0.186269
        valid_0's multi_logloss: 0.171556
        valid 0's multi logloss: 0.168245
        valid_0's multi_logloss: 0.161065
        valid 0's multi logloss: 0.151371
        valid_0's multi_logloss: 0.148081
        valid_0's multi_logloss: 0.143843
         valid 0's multi logloss: 0.140169
        valid_0's multi_logloss: 0.138303
[27]
[28]
[29]
[30]
[31]
[32]
[33]
[34]
[35]
        valid_0's multi_logloss: 0.134058
        valid_0's multi_logloss: 0.130884
        valid_0's multi_logloss: 0.128082
        valid_0's multi_logloss: 0.124975
        valid_0's multi_logloss: 0.122225
        valid 0's multi logloss: 0.120298
        valid_0's multi_logloss: 0.117257
        valid_0's multi_logloss: 0.115021
        valid_0's multi_logloss: 0.115037
        valid_0's multi_logloss: 0.115831
        valid_0's multi_logloss: 0.113318
        valid_0's multi_logloss: 0.115651
        valid 0's multi logloss: 0.115772
        valid_0's multi_logloss: 0.114953
        valid 0's multi logloss: 0.117056
[42]
[43]
[44]
[45]
[46]
        valid_0's multi_logloss: 0.115412
        valid_0's multi_logloss: 0.118359
        valid_0's multi_logloss: 0.117129
        valid 0's multi logloss: 0.119174
        valid 0's multi logloss: 0.117789
        valid_0's multi_logloss: 0.121333
        valid_0's multi_logloss: 0.120375
[49]
        valid_0's multi_logloss: 0.124128
        valid_0's multi_logloss: 0.123394
        valid_0's multi_logloss: 0.30682
        valid_0's multi_logloss: 0.306206
```

```
valid 0's multi logloss: 0.303895
[112]
       valid_0's multi_logloss: 0.300907
       valid 0's multi logloss: 0.304274
       valid_0's multi_logloss: 0.314218
       valid_0's multi_logloss: 0.312988
[115]
[116]
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[117]
       valid_0's multi_logloss: 0.323073
       valid_0's multi_logloss: 0.333026
[118]
[119]
       valid_0's multi_logloss: 0.332652
       valid 0's multi logloss: 0.337212
[120]
[121]
       valid 0's multi logloss: 0.334481
[122]
       valid 0's multi logloss: 0.340022
[123]
       valid_0's multi_logloss: 0.350061
       valid_0's multi_logloss: 0.351676
[124]
[125]
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[126]
       valid_0's multi_logloss: 0.358595
       valid 0's multi logloss: 0.356737
[127]
[128]
       valid_0's multi_logloss: 0.351512
[129]
       valid_0's multi_logloss: 0.361591
       valid 0's multi logloss: 0.35978
[130]
[131]
       valid 0's multi logloss: 0.357317
[132]
       valid 0's multi logloss: 0.367439
[133]
       valid 0's multi logloss: 0.365665
[134]
      valid_0's multi_logloss: 0.365745
      valid_0's multi_logloss: 0.375832
[135]
      valid_0's multi_logloss: 0.374115
       valid_0's multi_logloss: 0.376748
avg fit time: 0.07919144630432129 (+/- 0.014747029726158264)
avg score time: 0.001833200454711914 (+/- 0.0006676939103703261)
```

결과







Python 설치 패키지

