Machine Learning with Python

Life is too short, You need Python



실습 내용

• 다양한 알고리즘으로 모델을 만들고 성능을 비교합니다.

1.환경 준비

• 기본 라이브러리와 대상 데이터를 가져와 이후 과정을 준비합니다.

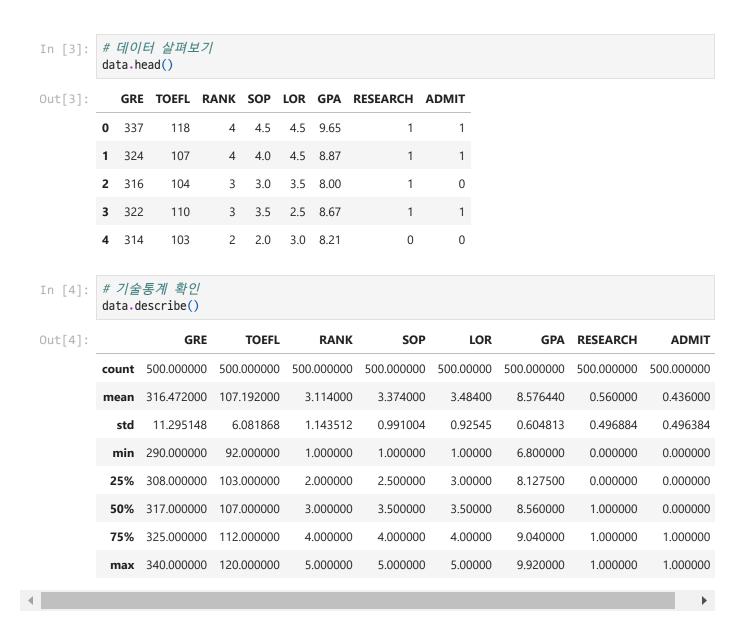
```
In [1]: # 라이브러리 불러오기
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import warnings

warnings.filterwarnings(action='ignore')
%config InlineBackend.figure_format = 'retina'
```

```
In [2]: #데이터 불러오기
path = 'https://raw.githubusercontent.com/jangrae/csv/master/admission_simple.csv'
data = pd.read_csv(path)
```

2.데이터 이해

• 분석할 데이터를 충분히 이해할 수 있도록 다양한 탐색 과정을 수행합니다.



3.데이터 준비

• 전처리 과정을 통해 머신러닝 알고리즘에 사용할 수 있는 형태의 데이터를 준비합니다.

1) x, y 분리

```
In [5]: # target 확인
target = 'ADMIT'

# 데이터 분리
x = data.drop(target, axis=1)
y = data[target]
```

2) 학습용, 평가용 데이터 분리

In [6]: # 모듈 불러오기
from sklearn.model_selection import train_test_split

7:3으로 분리
x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.3, random_state=1)

3) 정규화

In [7]: # 모듈 불러오기
from sklearn.preprocessing import MinMaxScaler

정규화
scaler = MinMaxScaler()
scaler.fit(x_train)
x_train_s = scaler.transform(x_train)
x_test_s = scaler.transform(x_test)

4.모델링

- 본격적으로 모델을 선언하고 학습하고 평가하는 과정을 진행합니다.
- 다양한 알고리즘으로 모델을 만들고 성능을 확입니다.
- 성능 평가는 confusion_matrix, classification_report를 사용합니다.
- 사용할 알고리즘을 모두 불러옵니다.

In [8]: # xgboost 설치 #!pip install xgboost Collecting xgboost

Obtaining dependency information for xgboost from https://files.pythonhosted.org/packages/2 4/ec/ad387100fa3cc2b9b81af0829b5ecfe75ec5bb19dd7c19d4fea06fb81802/xgboost-2.0.3-py3-none-win_a md64.whl.metadata

Downloading xgboost-2.0.3-py3-none-win_amd64.whl.metadata (2.0 kB)

Requirement already satisfied: numpy in c:\users\user\anaconda3\lib\site-packages (from xgboos t) (1.24.3)

Requirement already satisfied: scipy in c:\users\user\anaconda3\lib\site-packages (from xgboos t) (1.11.1)

Downloading xgboost-2.0.3-py3-none-win amd64.whl (99.8 MB)

ownloading xgboost-2.0.3-py3-none-win_amd6						
	2.5/99.8	MB	4.0	MB/s	eta	0:00:25
	3.1/99.8	MB	4.6	MB/s	eta	0:00:22
	4.6/99.8	MB	5.4	MB/s	eta	0:00:18
	4.7/99.8	MB	5.5	MB/s	eta	0:00:18
	4.7/99.8	MB	5.5	MB/s	eta	0:00:18
	4.7/99.8	MB	5.5	MB/s	eta	0:00:18
	4.7/99.8	MB	5.5	MB/s	eta	0:00:18
	4.7/99.8	MB	5.5	MB/s	eta	0:00:18
	4.7/99.8	MB	5.5	MB/s	eta	0:00:18
	4.9/99.8	MB	4.1	MB/s	eta	0:00:23
	4.9/99.8	MB	4.1	MB/s	eta	0:00:23
	4.9/99.8	MB	4.1	MB/s	eta	0:00:23
	4.9/99.8	MB	4.1	MB/s	eta	0:00:23
	5.0/99.8	MB	3.4	MB/s	eta	0:00:28
	5.4/99.8	MB	3.6	MB/s	eta	0:00:27
	5.4/99.8	MB	3.6	MB/s	eta	0:00:27
	5.4/99.8	MB	3.6	MB/s	eta	0:00:27
	5.4/99.8	MB	3.6	MB/s	eta	0:00:27
	5.4/99.8	MB	3.6	MB/s	eta	0:00:27
	5.7/99.8	MB	3.1	MB/s	eta	0:00:31
	5.7/99.8	MB	3.1	MB/s	eta	0:00:31
	5.7/99.8	MB	3.1	MB/s	eta	0:00:31
	6.0/99.8	MB	2.8	MB/s	eta	0:00:34
				•		-

ML04_04_앙상블(Admission)
 6.9/99.8 MB 3.0 MB/s eta 0:00:31
 7.5/99.8 MB 3.2 MB/s eta 0:00:29
 7.5/99.8 MB 3.2 MB/s eta 0:00:29
 7.7/99.8 MB 2.9 MB/s eta 0:00:32
 8.2/99.8 MB 3.0 MB/s eta 0:00:31
 8 8/00 8 MR 3 2 MR/s eta 0.00.20
0.0/99.0 MD 3.2 MD/3 Cta 0.00.29
 8.9/99.8 MB 3.2 MB/S eta 0:00:29
 9.2/99.8 MB 3.2 MB/s eta 0:00:28
 9.3/99.8 MB 3.3 MB/s eta 0:00:28
 9.3/99.8 MB 3.3 MB/s eta 0:00:28
 9.5/99.8 MB 3.0 MB/s eta 0:00:31
 9 9/99 8 MR 3 1 MR/s eta 0.00.30
 10 F/00 9 MP 2 2 MP/s ota 0:00:30
 11.5/99.8 MB 3.2 MB/s eta 0:00:28
 12.0/99.8 MB 3.2 MB/s eta 0:00:28
 13.6/99.8 MB 3.6 MB/S eta 0:00:25
 14.2/99.8 MB 3.6 MB/s eta 0:00:24
 14.7/99.8 MB 3.6 MB/s eta 0:00:24
 15.2/99.8 MB 4.5 MB/s eta 0:00:19
 15.4/33.6 NB 4.3 NB/S Ctd 0.00.13
13.0/99.0 MD 3.1 MD/S eta 0.00.1/
 16.1/99.8 MB 5.9 MB/s eta 0:00:15
 16.6/99.8 MB 6.1 MB/s eta 0:00:14
 16.6/99.8 MB 6.1 MB/s eta 0:00:14
16.6/99.8 MB 6.1 MB/s eta 0:00:14
16.6/99.8 MB 6.1 MB/s eta 0:00:14
 16.6/99.8 MB 6.1 MB/s eta 0:00:14
 16.9/99.8 MB 5.2 MB/s eta 0:00:16
 17.4/99.8 MB 5.2 MB/s eta 0:00:16
 17.7/99.8 MB 5.2 MB/s eta 0:00:16
 17.7/99.8 MB 5.2 MB/s eta 0:00:16
17.7/99.8 MB 5.2 MB/s eta 0:00:16
 17.7/99.8 MB 5.2 MB/s eta 0:00:16
 17.7/99.8 MB 5.2 MB/s eta 0:00:16
17.7/99.8 MB 4.5 MB/s eta 0:00:19
•
 •
 17.8/99.8 MB 5.2 MB/s eta 0:00:16
 17.8/99.8 MB 5.2 MB/s eta 0:00:16
18.2/99.8 MB 4.5 MB/s eta 0:00:18
 18.2/99.8 MB 4.5 MB/s eta 0:00:18

IV	ILU4_U4_성성글	(Aar	nissic	on)		
	- 18.2/99.8	MB	4.5	MB/s	eta	0:00:18
		MB	4.0	MB/s	eta	0:00:21
	- 18.6/99.8	MB	4.0	MB/s	eta	0:00:21
	- 18.6/99.8	MB	4.0	MB/s	eta	0:00:21
		MB	3.6	MB/s	eta	0:00:23
		MB	3.6	MB/s	eta	0:00:23
	- 21.3/99.8	MB	3.6	MB/s	eta	0:00:22
	- 21.8/99.8	MB	3.7	MB/s	eta	0:00:22
	- 22.0/99.8	MB	3.7	MB/s	eta	0:00:22
	- 22.4/99.8	MB	3.6	MB/s	eta	0:00:22
	-			-		
	- 25.9/99.8	MB	3.3	MB/s	eta	0:00:23
	- 26.2/99.8	MB	3.3	MB/s	eta	0:00:23
	- 26.2/99.8	MB	3.3	MB/s	eta	0:00:23
	-					
	- 31.0/99.8	MB	6.3	MB/s	eta	0:00:11
	- 31.6/99.8	MB	6.3	MB/s	eta	0:00:11

ML	04_04_성성글	(Aar	nissic	n)		
	32.7/99.8	MB	6.5	MB/s	eta	0:00:11
	33.2/99.8	MB	6.5	MB/s	eta	0:00:11
	33.9/99.8	MB	6.7	MB/s	eta	0:00:10
	34.0/99.8	MB	5.8	MB/s	eta	0:00:12
	34.1/99.8	MB	5.7	MB/s	eta	0:00:12
	34.6/99.8	MB	5.7	MB/s	eta	0:00:12
	36.5/99.8	MB	6.6	MB/s	eta	0:00:10
	36.5/99.8	MB	6.6	MB/s	eta	0:00:10
	36.5/99.8	MB	6.6	MB/s	eta	0:00:10
	36.5/99.8	MB	6.6	MB/s	eta	0:00:10
	36.6/99.8	MB	5.6	MB/s	eta	0:00:12
	38 2/99 8	MR	5 8	MR/s	eta	0.00.11
	39.7/99.8	MB	5.1	MB/s	eta	0:00:12
	40.0/99.8	MB	4.5	MB/s	eta	0:00:14
	40.6/99.8	MB	4.5	MB/s	eta	0:00:14
	41.1/99.8	MB	4.5	MB/s	eta	0:00:14
	41.6/99.8	MB	4.5	MB/s	eta	0:00:13
	10 0 100 0		J (1	MK/C	eta	и • ии • 15
	42.6/99.8	MB	3.9	MB/s	eta	0:00:15
	42.6/99.8 42.6/99.8	MB MB	3.9 3.9	MB/s MB/s	eta eta	0:00:15 0:00:15

ML	04_04_앙상블	(Adn	nissio	n)		
	42.6/99.8	MB	3.9	MB/s	eta	0:00:15
				-		
	43.6/99.8	MB	3.5	MB/s	eta	0:00:16
	43.6/99.8	MB	3.5	MB/s	eta	0:00:16
	43.7/99.8	MB	3.3	MB/s	eta	0:00:18
	44.2/99.8	MB	3.6	MB/s	eta	0:00:16
	-					
	45.0/99.8	MB	3.0	MB/s	eta	0:00:19
	45.0/99.8	MB	3.0	MB/s	eta	0:00:19
	45.0/99.8	MB	3.0	MB/s	eta	0:00:19
	45.0/99.8	MB	3.0	MB/s	eta	0:00:19
	-					
	-			-		
	-			-		
	49.1/99.8	MB	3.2	MB/s	eta	0:00:16
	49.7/99.8	MB	3.2	MB/s	eta	0:00:16
	49.7/99.8	MB	3.2	MB/s	eta	0:00:16
	49.7/99.8	MB	3.2	MB/s	eta	0:00:16
	49.7/99.8	MB	3.2	MB/s	eta	0:00:16
	49.7/99.8	MB	3.2	MB/s	eta	0:00:16
	20.0/99.8	ыR	∠.9	MD/2	eta	וו:טט:ט/

ML	.04_04_성성글	(Aan	nissio	n)		
	51.1/99.8	MB	2.9	MB/s	eta	0:00:17
	51.7/99.8	MB	2.9	MB/s	eta	0:00:17
	53.3/99.0	MD	2.0	MD/S	cta	0.00.14
	54.4/99.8	WR	3.9	MR/S	ета	0:00:12
	55.8/99.8	MB	5.7	MB/s	eta	0:00:08
	57.5/99.8	MB	4.4	MB/s	eta	0:00:10
	57.5/99.8	MB	4.4	MB/s	eta	0:00:10
	57.8/99.8	MB	4.0	MB/s	eta	0:00:11
	58.3/99.8	MB	4.0	MB/s	eta	0:00:11
	58.6/99.8	MB	4.5	MB/s	eta	0:00:10
	-			-		
	60.4/99.8	MB	4.6	MB/s	eta	0:00:09
	60.4/99.8	MB	4.6	MB/s	eta	0:00:09
	60.4/99.8	MB	4.6	MB/s	eta	0:00:09
	60.4/99.8	MB	4.6	MB/s	eta	0:00:09
	03.1/33.0	מו״ו	٠.۵	רווי) 2	Cla	0.00.10

ML	04_04_앙상블	(Adn	nissio	n)		
	63.4/99.8	MB	3.9	MB/s	eta	0:00:10
	64.2/99.8	MB	3.9	MB/s	eta	0:00:10
	64.2/99.8	MB	3.9	MB/s	eta	0:00:10
	64.2/99.8	MB	3.9	MB/s	eta	0:00:10
	64.8/99.8	MB	3.2	MB/s	eta	0:00:11
	65.2/99.8	MB	3.2	MB/s	eta	0:00:11
	66.5/99.8	MB	3.2	MB/s	eta	0:00:11
	67.1/99.8	MB	3.2	MB/s	eta	0:00:11
	68.1/99.8	MB	3.8	MB/s	eta	0:00:09
	71.1/99.8	MB	4.9	MB/s	eta	0:00:06
	71.7/99.8	MB	4.9	MB/s	eta	0:00:06
	71.7/99.8	MB	4.9	MB/s	eta	0:00:06
	-			-		
	72.2/99.8	MB	4.3	MB/s	eta	0:00:07
	72.2/99.8	MB	4.3	MB/s	eta	0:00:07
	72.2/99.8	MB	4.3	MB/s	eta	0:00:07
	72.4/99.8	MB	3.9	MB/s	eta	0:00:08
	12.4/99.8	ΜĎ	5.9	rid/S	CLA	מש:שש.ש

ML	04_04_8성글	(Aar	nissio	n)		
	72.7/99.8	MB	3.5	MB/s	eta	0:00:08
	72.7/99.8	MB	3.5	MB/s	eta	0:00:08
	72.7/99.8	MB	3.5	MB/s	eta	0:00:08
	72.7/99.8	MB	3.5	MB/s	eta	0:00:08
	73.1/99.8	MB	3.2	MB/s	eta	0:00:09
	73.1/99.8	MR	3.2	MR/s	eta	0:00:09
	74.2/99.8	MB	2.7	MB/s	eta	0:00:10
	74.8/99.8	MB	2.9	MB/s	eta	0:00:09
	75.3/99.8	MB	3.2	MB/s	eta	0:00:08
	/8.3/99.8	MR	3.5	MR/S	eta	0:00:07
	79.4/99.8	MB	3.2	MB/s	eta	0:00:07
	79.9/99.8	MB	3.5	MB/s	eta	0:00:06
	80.5/99.8	MB	3.5	MB/s	eta	0:00:06
	81.0/99.8	MB	3.5	MB/s	eta	0:00:06
	81.6/99.8	MB	3.5	MB/s	eta	0:00:06
	82.6/99.8	MB	4.4	MB/s	eta	0:00:04
				-		
	84.0/99.8	MB	6.8	MB/s	eta	0:00:03

MLC)4_04_앙싱글	(Aar	nissio	n)		
 	84.2/99.8	MB	5.7	MB/s	eta	0:00:03
 	86.3/99.8	MB	5.5	MB/s	eta	0:00:03
 	86.8/99.8	MB	5.5	MB/s	eta	0:00:03
 	87.4/99.8	MB	5.5	MB/s	eta	0:00:03
 	87.8/99.8	MB	4.8	MB/s	eta	0:00:03
 	90.1/99.8	MB	5.4	MB/s	eta	0:00:02
 	90.1/99.8	MB	5.4	MB/s	eta	0:00:02
 	90.1/99.8	MB	5.4	MB/s	eta	0:00:02
 	90.9/99.8	MB	4.5	MB/s	eta	0:00:02
 	90.9/99.8	MB	4.0	MB/s	eta	0:00:03
 	91.4/99.8	MB	4.0	MB/s	eta	0:00:03
 	92.4/99.8	MB	3.9	MB/s	eta	0:00:02
 	92.4/99.8	MB	3.9	MB/s	eta	0:00:02
 	92.4/99.8	MB	3.9	MB/s	eta	0:00:02
 	92.8/99.8	MB	3.5	MB/s	eta	0:00:02
 	93.3/99.8	MB	3.5	MB/s	eta	0:00:02
 -	90.5/99.8	MR	4.2	11D/ 2	Cla	0.00.01
 -	96.5/99.8	MB	4.1	MB/s	eta	0:00:01

```
----- 97.5/99.8 MB 4.1 MB/s eta 0:00:01
----- 97.6/99.8 MB 4.2 MB/s eta 0:00:01
    ----- 97.6/99.8 MB 4.2 MB/s eta 0:00:01
----- 97.6/99.8 MB 4.2 MB/s eta 0:00:01
  ----- 97.6/99.8 MB 4.2 MB/s eta 0:00:01
----- 97.6/99.8 MB 4.2 MB/s eta 0:00:01
----- 97.6/99.8 MB 4.2 MB/s eta 0:00:01
----- 97.8/99.8 MB 4.1 MB/s eta 0:00:01
 ----- 97.8/99.8 MB 4.1 MB/s eta 0:00:01
  ----- 97.8/99.8 MB 4.1 MB/s eta 0:00:01
  ----- 97.8/99.8 MB 4.1 MB/s eta 0:00:01
 ----- 97.8/99.8 MB 4.1 MB/s eta 0:00:01
  ----- 97.8/99.8 MB 4.1 MB/s eta 0:00:01
----- 97.8/99.8 MB 4.1 MB/s eta 0:00:01
----- 98.1/99.8 MB 3.7 MB/s eta 0:00:01
  ----- 98.2/99.8 MB 3.7 MB/s eta 0:00:01
----- 98.2/99.8 MB 3.7 MB/s eta 0:00:01
  ------ 98.2/99.8 MB 3.7 MB/s eta 0:00:01
----- 98.2/99.8 MB 3.7 MB/s eta 0:00:01
----- 98.2/99.8 MB 3.7 MB/s eta 0:00:01
----- 98.2/99.8 MB 3.7 MB/s eta 0:00:01
 ----- 98.4/99.8 MB 3.3 MB/s eta 0:00:01
----- 98.9/99.8 MB 3.3 MB/s eta 0:00:01
 ----- 99.4/99.8 MB 3.3 MB/s eta 0:00:01
----- 99.7/99.8 MB 3.3 MB/s eta 0:00:01
----- 99.8/99.8 MB 3.0 MB/s eta 0:00:00
```

Installing collected packages: xgboost Successfully installed xgboost-2.0.3

```
In [9]: # lightgbm 설치
         # !pip install lightgbm
```

Collecting lightgbm

Obtaining dependency information for lightgbm from https://files.pythonhosted.org/packages/e 1/4c/4685ccfae9806f561de716e32549190c1f533dde5bcadaf83bdf23972cf0/lightgbm-4.3.0-py3-none-win amd64.whl.metadata

Downloading lightgbm-4.3.0-py3-none-win amd64.whl.metadata (19 kB)

Requirement already satisfied: numpy in c:\users\user\anaconda3\lib\site-packages (from lightg bm) (1.24.3)

Requirement already satisfied: scipy in c:\users\user\anaconda3\lib\site-packages (from lightg bm) (1.11.1)

Downloading lightgbm-4.3.0-py3-none-win amd64.whl (1.3 MB)

```
----- 0.0/1.3 MB ? eta -:--:-
--- 0.1/1.3 MB 3.6 MB/s eta 0:00:01
----- 0.6/1.3 MB 8.1 MB/s eta 0:00:01
----- 1.0/1.3 MB 10.0 MB/s eta 0:00:01
------ 1.0/1.3 MB 10.0 MB/s eta 0:00:01
----- 1.0/1.3 MB 10.0 MB/s eta 0:00:01
------ 1.0/1.3 MB 10.0 MB/s eta 0:00:01
------ 1.0/1.3 MB 10.0 MB/s eta 0:00:01
----- 1.0/1.3 MB 10.0 MB/s eta 0:00:01
----- 1.0/1.3 MB 2.3 MB/s eta 0:00:01
----- 1.3/1.3 MB 3.0 MB/s eta 0:00:01
----- 1.3/1.3 MB 2.9 MB/s eta 0:00:00
```

Installing collected packages: lightgbm Successfully installed lightgbm-4.3.0

```
In [10]: # 라이브러리 불러오기
from sklearn.neighbors import KNeighborsClassifier
from sklearn.tree import DecisionTreeClassifier
from sklearn.linear_model import LogisticRegression
from sklearn.ensemble import RandomForestClassifier
from xgboost import XGBClassifier
from lightgbm import LGBMClassifier
from sklearn.metrics import *
```

1) KNN

• KNN 알고리즘으로 모델링하고 성능을 확인합니다.

```
In [11]:
          # 선언하기
          model = KNeighborsClassifier(n neighbors=5)
          # 학습하기
In [12]:
          model.fit(x_train_s, y_train)
Out[12]: v KNeighborsClassifier
          KNeighborsClassifier()
In [13]: # 예측하기
          y_pred = model.predict(x_test_s)
In [14]: # 평가하기
          print(confusion_matrix(y_test, y_pred))
          print(classification_report(y_test, y_pred))
          [[79 5]
           [15 51]]
                                    recall f1-score
                       precision
                                                      support
                            0.84
                                      0.94
                                               0.89
                                                           84
                            0.91
                                                           66
                    1
                                      0.77
                                               0.84
                                               0.87
                                                          150
             accuracy
                                               0.86
                            0.88
                                                          150
             macro avg
                                      0.86
          weighted avg
                            0.87
                                      0.87
                                               0.86
                                                          150
```

2) Decision Tree

• Decision Tree 알고리즘으로 모델링하고 성능을 확인합니다.

```
In [15]: # 선언하기
model = DecisionTreeClassifier(max_depth=5, random_state=1)

In [16]: # 학습하기
model.fit(x_train, y_train)
```

```
Out[16]: 

DecisionTreeClassifier

DecisionTreeClassifier(max_depth=5, random_state=1)
```

```
# 예측하기
In [17]:
          y_pred = model.predict(x_test)
         # 5단계: 평가하기
In [18]:
          print(confusion_matrix(y_test, y_pred))
          print(classification_report(y_test, y_pred))
          [[77 7]
           [13 53]]
                       precision
                                    recall f1-score
                                                       support
                            0.86
                    0
                                      0.92
                                                0.89
                                                            84
                            0.88
                                      0.80
                                                0.84
                                                            66
              accuracy
                                                0.87
                                                           150
                            0.87
                                      0.86
                                                0.86
                                                           150
             macro avg
```

0.87

150

3) Logistic Regression

0.87

weighted avg

• Logistic Regression 알고리즘으로 모델링하고 성능을 확인합니다.

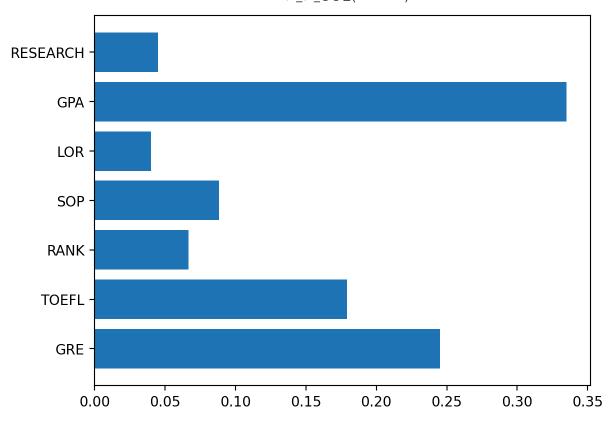
0.87

```
# 선언하기
In [35]:
          model = LogisticRegression()
          # 학습하기
In [36]:
          model.fit(x_train, y_train)
Out[36]:
          ▼ LogisticRegression
          LogisticRegression()
          # 예측하기
In [37]:
          y_pred = model.predict(x_test)
          # 5단계: 평가하기
In [38]:
          print(confusion_matrix(y_test, y_pred))
          print(classification_report(y_test, y_pred))
          [[75 9]
           [14 52]]
                       precision
                                    recall f1-score
                                                      support
                    0
                            0.84
                                      0.89
                                                0.87
                                                           84
                                                           66
                    1
                            0.85
                                      0.79
                                                0.82
                                                0.85
                                                          150
              accuracy
                                                0.84
             macro avg
                            0.85
                                      0.84
                                                          150
          weighted avg
                            0.85
                                      0.85
                                                0.85
                                                          150
```

4) Random Forest

• Random Forest 알고리즘으로 모델링하고 성능을 확인합니다.

```
# 선언하기
In [40]:
          model = RandomForestClassifier(max_depth=5, n_estimators=100) # 깊이를 5로 통일
          # 학습하기
In [41]:
          model.fit(x_train, y_train)
Out[41]:
                   RandomForestClassifier
         RandomForestClassifier(max_depth=5)
In [42]: # 예측하기
          y_pred = model.predict(x_test)
In [43]: # 5단계: 평가하기
          print(confusion_matrix(y_test, y_pred))
          print(classification report(y test, y pred))
          [[78 6]
          [12 54]]
                       precision
                                   recall f1-score
                                                     support
                            0.87
                                     0.93
                                               0.90
                                                          84
                    1
                            0.90
                                     0.82
                                               0.86
                                                          66
                                               0.88
                                                         150
             accuracy
                                               0.88
                                                         150
                            0.88
                                     0.87
             macro avg
         weighted avg
                            0.88
                                     0.88
                                               0.88
                                                         150
         # Feature 중요도 확인
In [44]:
          plt.barh(y=list(x), width=model.feature_importances_)
          plt.show()
```



5) XGBoost

In [32]:

In [33]:

예측하기

평가하기

• XGBoost 알고리즘으로 모델링하고 성능을 확인합니다.

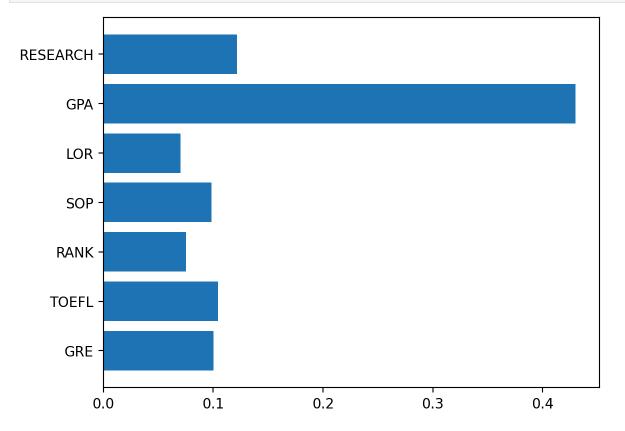
```
# 선언하기
In [30]:
         model = XGBClassifier(max depth=5, n estimators=100)
         # 학습하기
In [31]:
         model.fit(x_train, y_train)
Out[31]:
                                           XGBClassifier
         XGBClassifier(base_score=None, booster=None, callbacks=None,
                       colsample_bylevel=None, colsample_bynode=None,
                       colsample_bytree=None, device=None, early_stopping_rounds=No
         ne,
                       enable_categorical=False, eval_metric=None, feature_types=No
         ne,
                       gamma=None, grow_policy=None, importance_type=None,
                       interaction_constraints=None, learning_rate=None, max_bin=No
         ne,
                       max_cat_threshold=None, max_cat_to_onehot=None,
                       max delta step=None, max depth=5, max leaves=None,
```

print(confusion_matrix(y_test, y_pred))
print(classification_report(y_test, y_pred))

y_pred = model.predict(x_test)

```
[[77 7]
 [15 51]]
              precision
                         recall f1-score
                                              support
           0
                   0.84
                             0.92
                                       0.88
                                                   84
           1
                   0.88
                             0.77
                                       0.82
                                                   66
                                       0.85
                                                  150
    accuracy
   macro avg
                   0.86
                             0.84
                                       0.85
                                                  150
weighted avg
                   0.86
                             0.85
                                       0.85
                                                  150
```





6) LightGBM

• LightGBM 알고리즘으로 모델링하고 성능을 확인합니다.

```
In [63]: # 선언하기
model = LGBMClassifier(max_depth=5, importance_type='gain', verbose=-1)

In [64]: # 학습하기
model.fit(x_train, y_train)

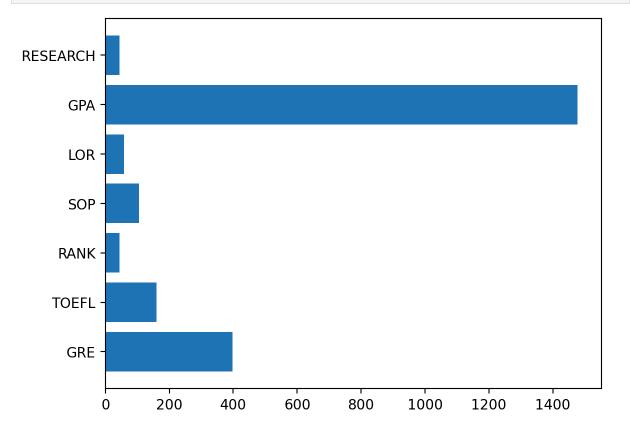
Out[64]:

LGBMClassifier

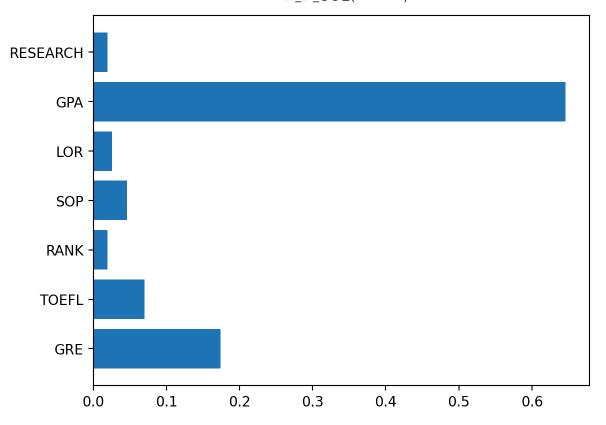
LGBMClassifier(importance_type='gain', max_depth=5, verbose=-1)
```

```
# 예측하기
In [65]:
          y_pred = model.predict(x_test)
          # 평가하기
In [66]:
          print(confusion_matrix(y_test, y_pred))
          print(classification_report(y_test, y_pred))
          [[77 7]
           [14 52]]
                        precision
                                     recall f1-score
                                                        support
                     0
                             0.85
                                       0.92
                                                 0.88
                                                             84
                     1
                             0.88
                                       0.79
                                                 0.83
                                                             66
              accuracy
                                                 0.86
                                                            150
                             0.86
                                       0.85
                                                 0.86
                                                            150
             macro avg
          weighted avg
                             0.86
                                       0.86
                                                 0.86
                                                            150
```

In [67]: # Feature 중요도 확인 plt.barh(y=list(x), width=model.feature_importances_) plt.show()



In [68]: # 0 ~ 1 사이 값으로 변환 feature_importance norm = model.feature_importances_/ np.sum(model.feature_importances_) plt.barh(y=list(x), width=feature_importance_norm) plt.show()



Τη []·