

Alchemist 3기 아이디어 톤

df_pig 조

1. 데이터 확인

pd.read_csv로 데이터를 불러오고 head, info, describe()를 통해 데이터의 형태를 확인

| | Gender | Age | Height | Weight | family_history_with_overweight | FAVC | FCVC | NCP | CAEC |
|------|--------|-----------|----------|------------|--------------------------------|------|------|-----|-----------|
| 0 | Female | 21.000000 | 1.620000 | 64.000000 | yes | no | 2.0 | 3.0 | Sometimes |
| 1 | Female | 21.000000 | 1.520000 | 56.000000 | yes | no | 3.0 | 3.0 | Sometimes |
| 2 | Male | 23.000000 | 1.800000 | 77.000000 | yes | no | 2.0 | 3.0 | Sometimes |
| 3 | Male | 27.000000 | 1.800000 | 87.000000 | no | no | 3.0 | 3.0 | Sometimes |
| 4 | Male | 22.000000 | 1.780000 | 89.800000 | no | no | 2.0 | 1.0 | Sometimes |
| ... | ... | ... | ... | ... | ... | ... | ... | ... | ... |
| 2106 | Female | 20.976842 | 1.710730 | 131.408528 | yes | yes | 3.0 | 3.0 | Sometimes |
| 2107 | Female | 21.982942 | 1.748584 | 133.742943 | yes | yes | 3.0 | 3.0 | Sometimes |
| 2108 | Female | 22.524036 | 1.752206 | 133.689352 | yes | yes | 3.0 | 3.0 | Sometimes |
| 2109 | Female | 24.361936 | 1.739450 | 133.346641 | yes | yes | 3.0 | 3.0 | Sometimes |
| 2110 | Female | 23.664709 | 1.738836 | 133.472641 | yes | yes | 3.0 | 3.0 | Sometimes |

2111 rows × 17 columns

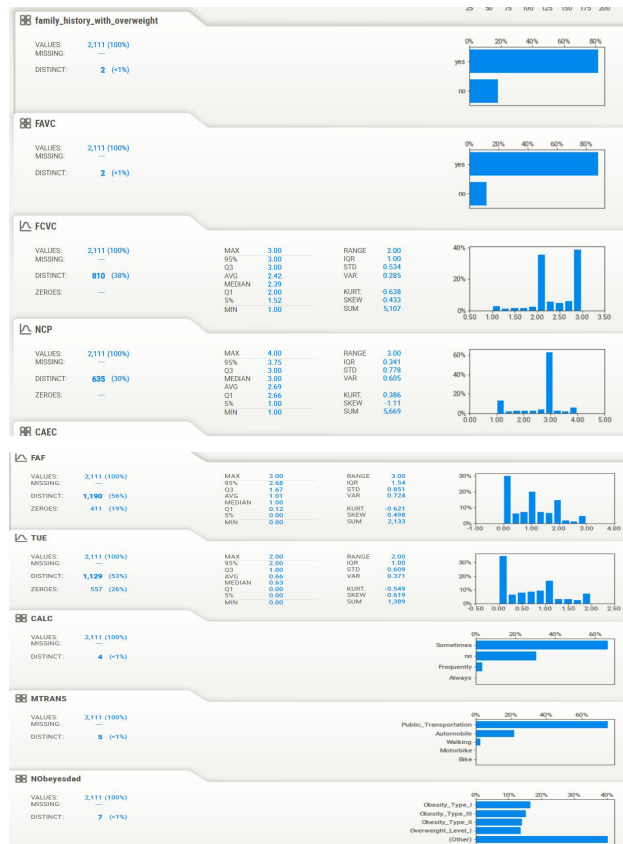
```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2111 entries, 0 to 2110
Data columns (total 17 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   Gender                                2111 non-null   object
1   Age                                  2111 non-null   float64
2   Height                              2111 non-null   float64
3   Weight                              2111 non-null   float64
4   family_history_with_overweight       2111 non-null   object
5   FAVC                                 2111 non-null   object
6   FCVC                                 2111 non-null   float64
7   NCP                                  2111 non-null   float64
8   CAEC                                 2111 non-null   object
9   SMOKE                                2111 non-null   object
10  CH2O                                  2111 non-null   float64
11  SCC                                  2111 non-null   object
12  FAF                                  2111 non-null   float64
13  TUE                                  2111 non-null   float64
14  CALC                                  2111 non-null   object
15  MTRANS                               2111 non-null   object
16  NObeyesdad                           2111 non-null   object
dtypes: float64(8), object(9)
memory usage: 388.5+ KB
```

```
pig.describe()
```

| | Age | Height | Weight | FCVC | NCP | CH2O | FAF | TUE |
|--------------|-------------|-------------|-------------|-------------|-------------|-------------|-------------|-------------|
| count | 2111.000000 | 2111.000000 | 2111.000000 | 2111.000000 | 2111.000000 | 2111.000000 | 2111.000000 | 2111.000000 |
| mean | 24.312600 | 1.701677 | 86.586058 | 2.419043 | 2.685628 | 2.008011 | 1.010298 | 0.657866 |
| std | 6.345968 | 0.093305 | 26.191172 | 0.533927 | 0.778039 | 0.612953 | 0.850592 | 0.608927 |
| min | 14.000000 | 1.450000 | 39.000000 | 1.000000 | 1.000000 | 1.000000 | 0.000000 | 0.000000 |
| 25% | 19.947192 | 1.630000 | 65.473343 | 2.000000 | 2.658738 | 1.584812 | 0.124505 | 0.000000 |
| 50% | 22.777890 | 1.700499 | 83.000000 | 2.385502 | 3.000000 | 2.000000 | 1.000000 | 0.625350 |
| 75% | 26.000000 | 1.768464 | 107.430682 | 3.000000 | 3.000000 | 2.477420 | 1.666678 | 1.000000 |
| max | 61.000000 | 1.980000 | 173.000000 | 3.000000 | 4.000000 | 3.000000 | 3.000000 | 2.000000 |

데이터에서 null값은 존재하지 않고 일부 칼럼이 비정형 데이터임을 확인

sweetviz를 통한 데이터 시각화



smote로 인한 문제 확인

```
pig[(['Age', 'FCVC', 'NCP', 'TUE'])].value_counts()
```

| Age | FCVC | NCP | TUE | |
|-----------|------|----------|----------|----|
| 21.000000 | 2.0 | 1.000000 | 0.000000 | 18 |
| 18.000000 | 2.0 | 3.000000 | 0.000000 | 16 |
| 23.000000 | 2.0 | 3.000000 | 1.000000 | 15 |
| 21.000000 | 2.0 | 3.000000 | 0.000000 | 13 |
| | | | 1.000000 | 12 |
| | | | | .. |
| 21.282238 | 3.0 | 3.000000 | 0.849236 | 1 |
| 21.274628 | 3.0 | 3.489918 | 0.128394 | 1 |
| 21.243142 | 3.0 | 1.726260 | 0.000000 | 1 |
| 21.238416 | 3.0 | 3.000000 | 0.890527 | 1 |
| 61.000000 | 3.0 | 3.000000 | 1.000000 | 1 |

Name: count, Length: 1784, dtype: int64

2. 데이터 전처리

1. 실수형 데이터 스케일링 (StandardScaler)

```
#실수형 데이터의 피쳐 세개 정규화  
from sklearn.preprocessing import StandardScaler  
scaler = StandardScaler()  
float_list=['Height','Weight','CH20']  
#타겟값 인코딩  
from sklearn.preprocessing import LabelEncoder  
label_encoder = LabelEncoder()  
pig['NObeyesdad'] = label_encoder.fit_transform(pig['NObeyesdad'])  
#실수형 데이터만 선택한 데이터프레임  
pig_float=pig[float_list]
```

데이터 전처리

2. smote로 인해 변환된 함수 round() 적용

#smote로 인해 변환된 함수 반올림 적용.

```
pig.Age = pig.Age.round(); pig.FCVC = pig.FCVC.round(); pig.NCP = pig.NCP.round(); pig.TUE = pig.TUE.round()
```

3. 피처와 타겟 구분

#피처와 타겟 분리

```
X_features = pig.drop(['NObeyesdad'], axis=1, inplace=False)
```

```
y_target = pig['NObeyesdad']
```

4. 비정형 데이터에 원핫인코딩 적용

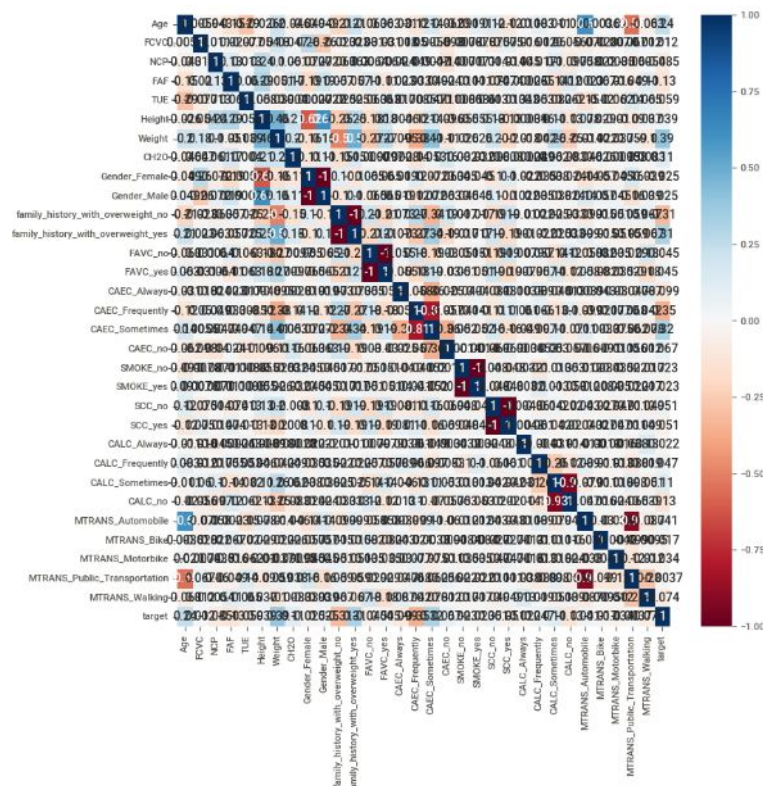
```
X_features_oh = pd.get_dummies(X_features, columns=['Gender', 'family_history_with_overweight', 'FAVC', 'CAEC', 'SMOKE', 'SCC', 'CALC', 'MTRANS'], dtype=int)  
X_features_oh
```

데이터 전처리

| | Age | FCVC | NCP | FAF | TUE | Height | Weight | CH2O | Gender_Female | Gender_Male | ... | SCC_yes | CALC_Always | CALC_Frequently |
|---|------|------|-----|----------|-----|-----------|-----------|-----------|---------------|-------------|-----|---------|-------------|-----------------|
| 0 | 21.0 | 2.0 | 3.0 | 0.000000 | 1.0 | -0.875589 | -0.862558 | -0.013073 | 1 | 0 | ... | 0 | 0 | 0 |
| 1 | 21.0 | 3.0 | 3.0 | 3.000000 | 0.0 | -1.947599 | -1.168077 | 1.618759 | 1 | 0 | ... | 1 | 0 | 0 |
| 2 | 23.0 | 2.0 | 3.0 | 2.000000 | 1.0 | 1.054029 | -0.366090 | -0.013073 | 0 | 1 | ... | 0 | 0 | 1 |
| 3 | 27.0 | 3.0 | 3.0 | 2.000000 | 0.0 | 1.054029 | 0.015808 | -0.013073 | 0 | 1 | ... | 0 | 0 | 1 |
| 4 | 22.0 | 2.0 | 1.0 | 0.000000 | 0.0 | 0.839627 | 0.122740 | -0.013073 | 0 | 1 | ... | 0 | 0 | 0 |

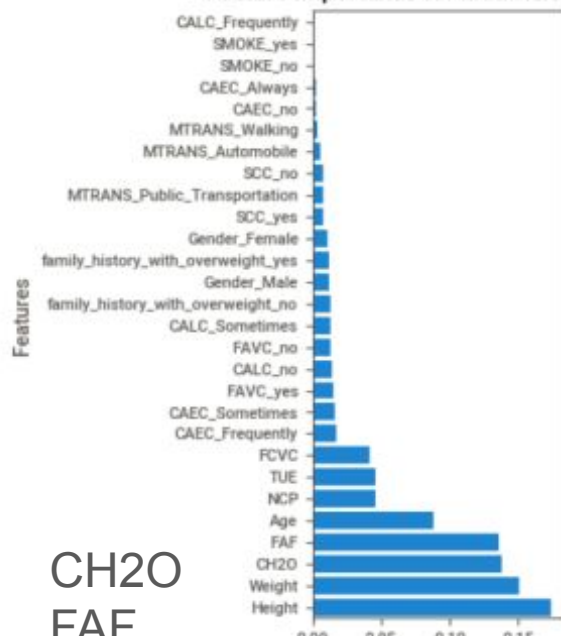
2111 rows × 31 columns

1.000000
0.387643
0.316962
0.313667
0.235660
0.114108
0.108868
0.066715
0.050679
0.047318
0.044582
0.041170
0.038986
0.024908
0.023256
0.012068
-0.003748
-0.017351
-0.022484
-0.023256
-0.024908
-0.034229
-0.044582
-0.050679
-0.059050
-0.073823
-0.085367
-0.099028
-0.129564
-0.134716
-0.313667
-0.351822



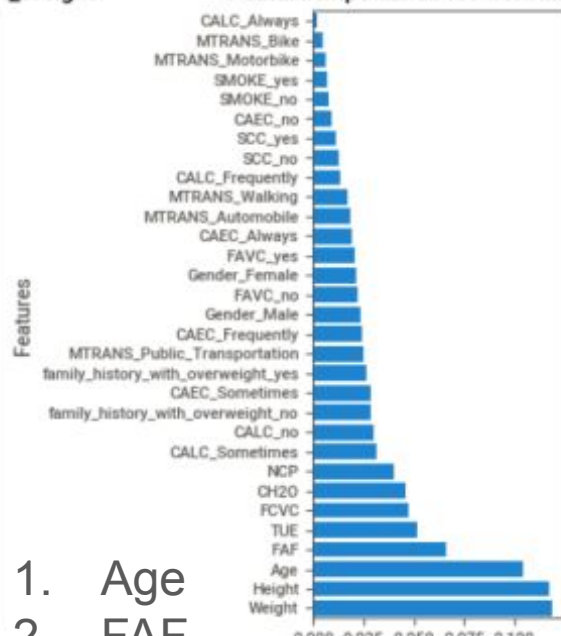
저체중 그룹, 정상체중 그룹, 과체중 1 그룹

Feature Importance for Insufficient_weight



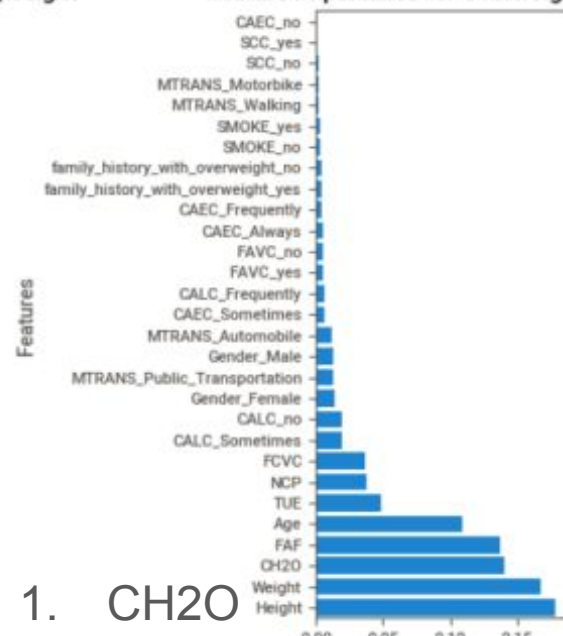
1. CH2O
2. FAF
3. Age
4. NCP

Feature Importance for Normal_weight



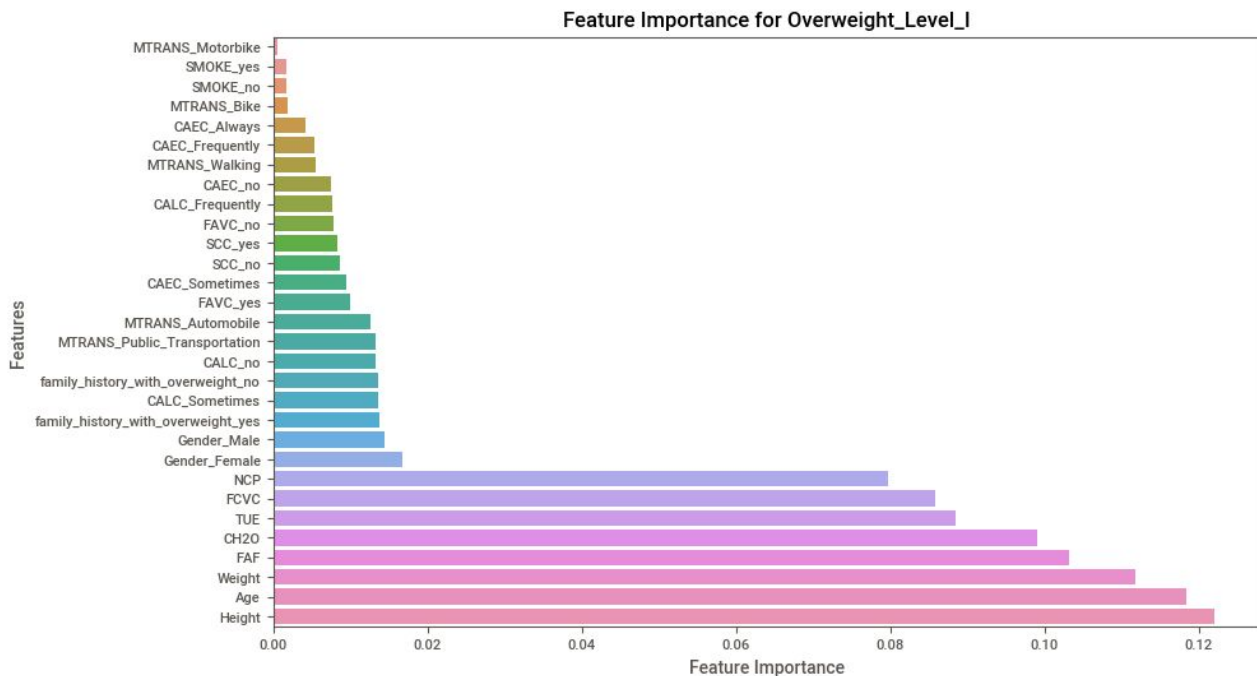
1. Age
2. FAF
3. TUE
4. FCVC

Feature Importance for Overweight



1. CH2O
2. FAF
3. Age
4. TUE

과체중 그룹-Overweight_Level_I

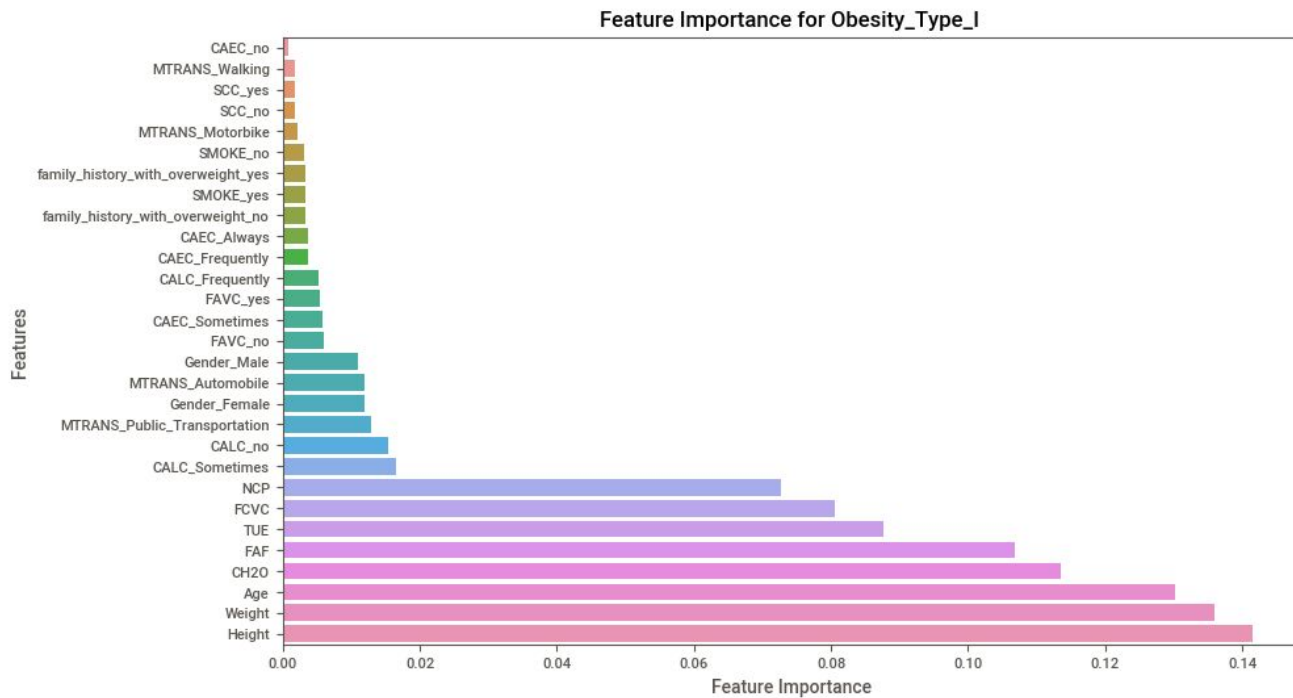


1. FAF
2. CH2O
3. TUE
4. FCVC
5. NCP

중요 인자

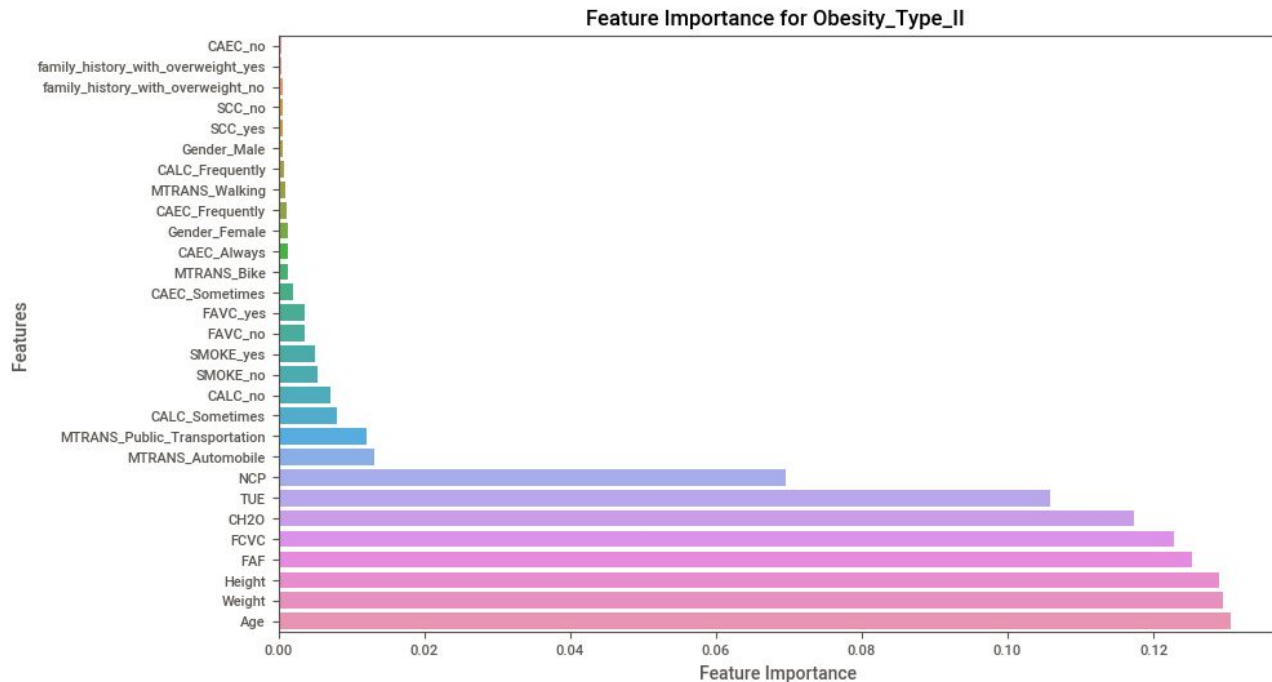
- FAF: 운동량
- CH2O: 물 섭취량
- TUE: 전자기기 사용
- FCVC: 채소섭취량
- NCP: 하루에 먹는 끼니 수

비만 그룹-Obesity_Type_I



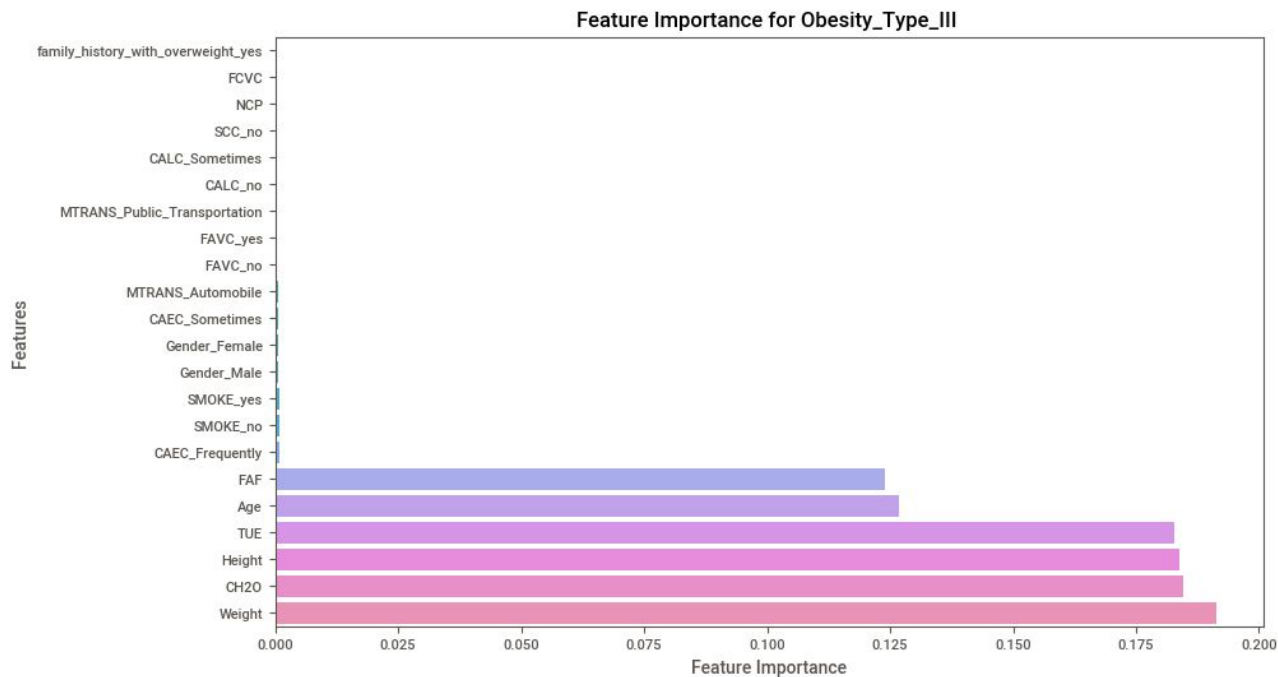
1. CH2O
2. FAF
3. TUE
4. FCVC
5. NCP

비만 그룹-Obesity_Type_II



1. FAF
2. FCVC
3. CH2O
4. TUE
5. NCP

비만 그룹-Obesity_Type_III



1. CH2O
2. TUE
3. FAF