# Alchemist 3기 아이디어톤

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

<class 'pandas.core.frame.DataFrame'> RangeIndex: 2111 entries, 0 to 2110 Data columns (total 17 columns): Non-Null Count Dtype Column Gender object 2111 non-null Age 2111 non-null float64 Height 2111 non-null float64 Weight 2111 non-null float64 family history with overweight 2111 non-null object FAVC object 2111 non-null FCVC float64 2111 non-null NCP 2111 non-null float64 CAEC 2111 non-null object 2111 non-null object SMOKE 2111 non-null float64 CH20 object 11 SCC 2111 non-null 12 FAF 2111 non-null float64 13 TUE 2111 non-null float64 14 CALC 2111 non-null object 15 MTRANS 2111 non-null object 2111 non-null object 16 NObeyesdad dtypes: float64(8), object(9)

----- 300 Ft MR

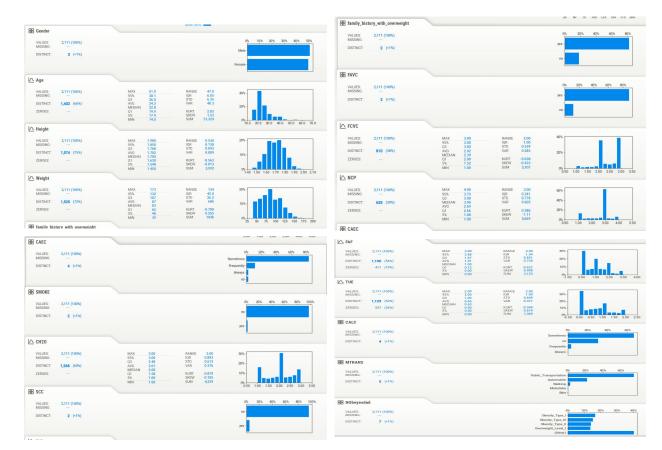
2111 rows × 17 columns

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()
          FCVC
                         TUE
Age
                NCP
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
21.274628 3.0
               3.489918 0.128394
21.243142 3.0
               1.726260 0.000000
21.238416
         3.0
               3.000000 0.890527
                                      1
61.000000 3.0 3.000000 1.000000
Name: count, Length: 1784, dtype: int64
```

#### 2. 데이터 전처리

1. 실수형 데이터 스케일링 (StandardScaler)

```
#실수형 데이터인 피쳐 세계 정규화
from sklearn.preprocessing import StandardScaler
scaler = StandardScaler()
float_list=['Height','Weight','CH2O']
#타겠라 인코딩
from sklearn.preprocessing import LabelEncoder
label_encoder = LabelEncoder()
pig['NObeyesdad'] = label_encoder.fit_transform(pig['NObeyesdad'])
#실수형 데이터만 선택한 데이터프레일
pig_float=pig[float_list]
```

#### 데이터 전처리

2. smote로 인해 변환된 함수 round() 적용

```
#smote로 인해 변환된 함수 반을림 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

```
#폐처와 타켓 분리
X_features = pig.drop(['NObeyesdad'], axis=1, inplace=False)
y_target = pig['NObeyesdad']
```

4. 비정형 데이터에 원핫인코딩 적용

X\_features\_ohe = pd.get\_dummies(X\_features, columns=['Gender','family\_history\_with\_overweight','FAVC','CAEC','SMOKE','SCC','CALC','MTRANS'], dtype=int)
X features ohe

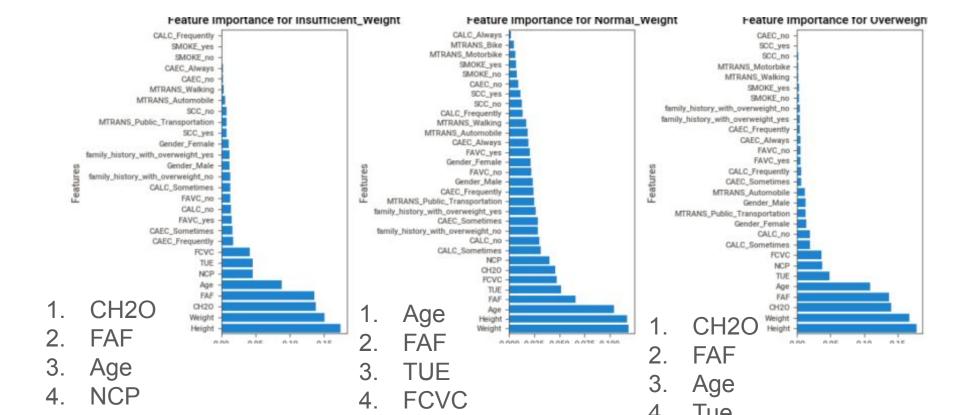
## 데이터 전처리

	Age	FCVC	NCP	FAF	TUE	Height	Weight	CH20	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

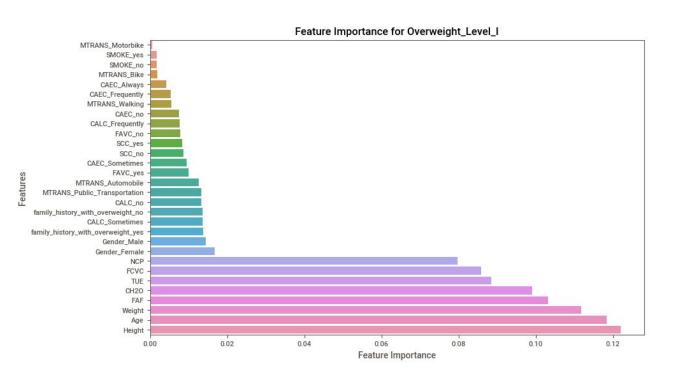
target	1.000000	Age - 100594316 200 258.04649-12121059639112-74652909114-0201093-0100050006-10.0634
Weight	0.387643	FCVCI.00 1 0 THE GUT DEADED 4728 2502902 BB THE TO B SHE BY SHE
CAEC Sometimes	0.316962	NOR-O, COMB 1 11). 133 1324 0.06 137 Q2 235 0330 05 D032 40 130 47 04 21 46 5 179 05 05 05 06 05 05 05
family history with overweight yes	0.313667	FAF -0. 15502.1 1 1 0.06 290 511-7 19 1999 73 57-9 . 0 10 2 30 300 90 4 10 00 1074 7400 265 1040 10 0020 6 740 6 49+0 . 13
Age	0.235660	TUE 40.290070 TEOR 1 088008000000020288250698 80 00805001 0080500 301 94 863080402 46 4020 620 008050
CALC Sometimes	0.114104	Height-0, 02.0 9498, 2005 10 45 2 1 02.6 2, 2628, 1821 904612-10409-658-551-60100009-4614, 02007-0219-03.008-7839
CH20	0.108868	Weight -0.20 130-0 95 0 296 1 0.20 151 1 0 20 272 0 0 358 4 0 202 26 20 2 20 20 20 26 26 26 272 0 29 0 20 26 26 272 0 20 20 20 20 20 20 20 20 20 20 20 20
CAEC no	0.066715	Gender_Female-0.0493607821000730 16.1 1 3 0.100.06965090207356943/45 10-0.023056307440657945602925 -0.50
SCC no	0.050679	Gender_Male 9.049267721900(36) 16.11-1 1 0.10-9.0668819920728378646.100.028850287297005.845.908925
The state of the s		family_history_with_overweight_no-0-2/19236687/2525-0-18.10.11-11.20.207/0328-34/1909/2-07/1919-0.002/29/30/2909.88/0596/731
CALC_Frequently	0.047318	family_history_with_overweight_yes
FAVC_yes	0.044582	FAVC_no-0.0060000640.04634062.70906506520.2 1 -1.03516.1940005050.19409.0006071041205682950290045
MTRANS_Automobile	0.041170	FAVC_yes 0.062030640 063162 0630066505212 1 1 0631819 0336109119 0 0439674 0206982286230 0345
Height	0.038986	CAEC . Always-0, 03 CD 82 02 00 00 00 00 00 00 00 00 00 00 00 00
Gender_Male	0.024908	CAEC_Sometimes 9.0.005994440-04716-006307072-343-9-1010-3-1011-3-3-0-5-0-5-0-0-0-0-7-0-0-0-0-0-0-0-0-0-0-0
SMOKE_no	0.023256	CAEC_no-0.062082022H7100631115GB643319_1908_43325673_1_001DD3454569084536305A1649X110560267
FCVC	0.012068	SMOKE_no-0.4998 287/0104885 52:63:2652650 20 7550 57-040446520 11 -1 0430450420 01:9630 1382 63289 2010/23
MTRANS Public Transportation	-0.003748	SMOKE_yes 0.0000700700100098559260280695000070B93051040838500 -1 1 0040480820.0030560.0030560.00808502907023
MTRANS Bike	-0.017351	SDC_no
CALC Always	-0.022484	SOC_YES - <b>0.02050.00-0151.00008.10.0.10.10.10.0000010.00000844-11.00-006</b> 10 <b>4200402.00.00051</b> CALC_AAMAYS-0.00.01-0-0500.000000000000000000000000
SMOKE yes	-0.023256	CALC_Frequently 0.0830.20.0655.200004090383.2202950508.99.0000050050.00.0608.0010.20.20.20.20.20.20.20.20.20.20.20.20.20
Gender Female	-0.024908	CALC_Sometimes 0.0000.1-0.002.0.20089803960.02514.400610.0030.00.002248.02.12.10.0.070.00.99.0611.1
MTRANS Motorbike	-0.034293	CALC., no. 4. 429569/12069 to 7559379941893314, 2013 to 47059630307292041, 931, 08 20 6941563213
	-0.044582	MTRANS_Automobile -0.30.07/EB093059/80.446917140999058EBE8999-0-0600.902903830.089079-410.4300.92
FAVC_no	-0.050679	MTRANS_Bix40-04284284284702293292845558050580530280339080688282940160.00.0150 100489909517
SCC_yes		MTRANS_Motorbike 0.02004880.642040.8980.9844.981050830.9950.00888.3940404080.80.80.9940.380 10.0088.3040404080.80.80.9040.380 10.0088.3040404080.80.8040.80
TUE	-0.059050	MTRANS_Public_Transportation - 0.07/060494.09591816.00695929294963662622281001038889309.0491 1048037
MTRANS_Walking	-0.073823	MTRANS, Walking-0, USBID 2054/1058/3/07 DEBES 03 08 76 78 808 76 707 00 709 45 000 900 000 000 000 000 000 000 000 00
NCP	-0.085367	-1.0
CAEC_Always	-0.099028	Age RAPE FARE THE THE THE THE THE THE THE THE THE TH
FAF	-0.129564	y y y y y y y y y y y y y y y y y y y
CALC_no	-0.134716	Gender, Gender
family_history_with_overweight_no	-0.313667	with with A G G G G G G G G G G G G G G G G G G
CAEC Frequently	-0.351827	Port (Port)
		Part Print

Name: target, dtype: float64

#### 저체중 그룹, 정상체중 그룹, 과체중 1 그룹



## 과체중 그룹-Overweight\_Level\_I

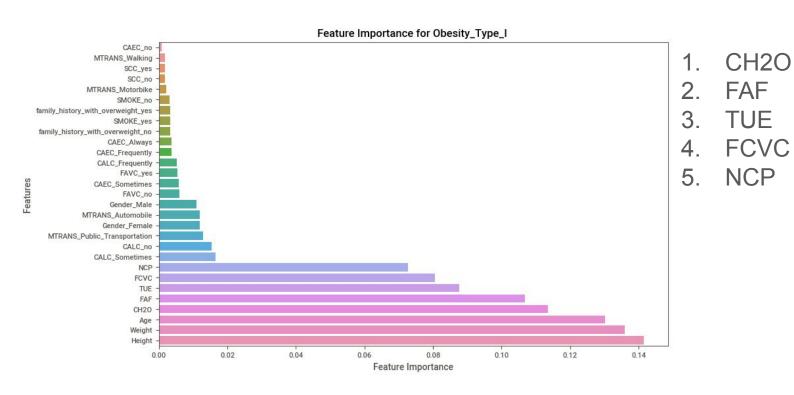


- 1. FAF
- 2. CH2O
- 3. TUE
- 4. FCVC
- 5. NCP

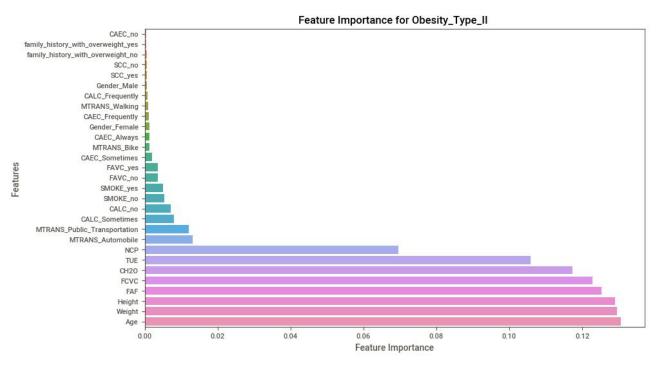
중요 인자

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

# 비만 그룹-Obesity\_Type\_I



## 비만 그룹-Obesity\_Type\_II



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

## 비만 그룹-Obesity\_Type\_III

