

importpackages

```
import pandas as pd
import numpy as np
import warnings
from sklearn.cluster import KMeans
from sklearn.decomposition import PCA
import matplotlib
import matplotlib.pyplot as plt
from sklearn.metrics.cluster import silhouette_score
from mpl_toolkits.mplot3d import Axes3D
```

Data pre-processing

전체 데이터셋: train, 서비스업: service, 일반업: normal, 제조업: product

```
train = pd.read_excel('C:/Users/user/Desktop/Statistical_Data_Idea_Contest/DATA/
데이터/기술통계19.xlsx', index_col='global_id')
service =
pd.read_excel('C:/Users/user/Desktop/Statistical_Data_Idea_Contest/DATA/데이터/실
태조사_서비스업19.xlsx', index_col='global_id')
normal =
pd.read_excel('C:/Users/user/Desktop/Statistical_Data_Idea_Contest/DATA/데이터/실
태조사_일반항목19.xlsx', index_col='global_id')
product =
pd.read_excel('C:/Users/user/Desktop/Statistical_Data_Idea_Contest/DATA/데이터/실
태조사_제조업19.xlsx', index_col='global_id')
```

```
print('기술통계자료의 데이터 개수 train: {}'.format(len(train)))
print('실태조사 자료 데이터의 개수 service:{}, normal:{}, product:
{}'.format(len(service), len(normal), len(product)))
```

기술통계자료의 데이터 개수 train: 3994
실태조사 자료 데이터의 개수 service:8500, normal:4000, product:7500

기술통계조사와 실태조사 자료의 중복자료 추출

```
warnings.filterwarnings(action='ignore')
print('기술통계조사와 실태조사 자료의 중복항목 수는 다음과 같다.')
companys = [service, normal, product]
for i in range(len(companys)):
    line = []
    for index in train.index:
        if index in companys[i].index:
            line.append(index)
        else:
            line.append(0)
    if i == 0:
```

```

service_train=train[train.index == line]
service_train['bt'] = 0
print('서비스업종 데이터: {}'.format(len(service_train)))
elif i == 1:
    normal_train=train[train.index == line]
    normal_train['bt'] = 1
    print('일반업종 데이터: {}'.format(len(normal_train)))
else:
    product_train=train[train.index == line]
    product_train['bt'] = 2
    print('제조업종 데이터: {}'.format(len(product_train)))

```

기술통계조사와 실태조사 자료의 중복항목 수는 다음과 같다.

서비스업종 데이터: 195

일반업종 데이터: 183

제조업종 데이터: 834

데이터 병합

```

train = pd.concat([service_train, normal_train, product_train])
index_list = ['A1S1', 'A1N1', 'C2S2', 'H1_1', 'H4_1', 'I4Q1', 'I4Q2', 'I4Q3',
'I4Q4', 'I4Q5', 'I4Q6', 'I4Q7']
my_train = train.loc[:, index_list]

```

결측치 최빈값 처리

```

for index in index_list:
    freq = my_train[index].value_counts(dropna=True).idxmax()
    my_train[index].fillna(freq, inplace = True)

```

응답범주 축소화

```

warnings.filterwarnings(action='ignore')

my_train['A1S1'][(my_train['A1S1']==1) | (my_train['A1S1']==2)] = 1
my_train['A1S1'][my_train['A1S1']==3] = 2
my_train['A1S1'][(my_train['A1S1']==4) | (my_train['A1S1']==5) |
(my_train['A1S1']==6)] = 3
my_train['A1S1'][my_train['A1S1']==7] = 4

my_train['A1N1'] [(my_train['A1N1']==1) | (my_train['A1N1']==2)] = 1
my_train['A1N1'] [(my_train['A1N1']==10) | (my_train['A1N1']==11)] = 2
my_train['A1N1'] [(my_train['A1N1']==5) | (my_train['A1N1']==6) |
(my_train['A1N1']==12) | (my_train['A1N1']==8) | (my_train['A1N1']==9)))] = 3
my_train['A1N1'] [(my_train['A1N1']==3) | (my_train['A1N1']==7)] = 4
my_train['A1N1'] [(my_train['A1N1']==13) | (my_train['A1N1']==14) |
(my_train['A1N1']==15)] = 5

my_train['C2S2'][(my_train['C2S2']==2) | (my_train['C2S2']==4)] = 2

my_train['H1_1'][(my_train['H1_1']==2) | (my_train['H1_1']==3)] = 1
my_train['H1_1'][(my_train['H1_1']==4) | (my_train['H1_1']==5) |
(my_train['H1_1']==6) | (my_train['H1_1']==7)] = 2
my_train['H1_1'][(my_train['H1_1']==1)] = 3

```

```

my_train['H1_1'][(my_train['H1_1']==9) | (my_train['H1_1']==8)] = 4

my_train['H4_1'][(my_train['H4_1']==1) | (my_train['H4_1']==3)] = 1
my_train['H4_1'][(my_train['H4_1']==4)] = 2
my_train['H4_1'][(my_train['H4_1']==5) | (my_train['H4_1']==6)] = 3
my_train['H4_1'][(my_train['H4_1']==2)] = 4
my_train['H4_1'][(my_train['H4_1']==7) | (my_train['H4_1']==9)] = 5
my_train['H4_1'][(my_train['H4_1']==8) | (my_train['H4_1']==10)] = 6

my_train['I4Q1'][(my_train['I4Q1']==3) | (my_train['I4Q1']==4)] = 1
my_train['I4Q1'][(my_train['I4Q1']==2)] = 2
my_train['I4Q1'][(my_train['I4Q1']==1)] = 3

my_train['I4Q2'][(my_train['I4Q2']==3) | (my_train['I4Q2']==4)] = 1
my_train['I4Q2'][(my_train['I4Q2']==2)] = 2
my_train['I4Q2'][(my_train['I4Q2']==1)] = 3

my_train['I4Q3'][(my_train['I4Q3']==3) | (my_train['I4Q3']==4)] = 1
my_train['I4Q3'][(my_train['I4Q3']==2)] = 2
my_train['I4Q3'][(my_train['I4Q3']==1)] = 3

my_train['I4Q4'][(my_train['I4Q4']==3) | (my_train['I4Q4']==4)] = 1
my_train['I4Q4'][(my_train['I4Q4']==2)] = 2
my_train['I4Q4'][(my_train['I4Q4']==1)] = 3

my_train['I4Q5'][(my_train['I4Q5']==3) | (my_train['I4Q5']==4)] = 1
my_train['I4Q5'][(my_train['I4Q5']==2)] = 2
my_train['I4Q5'][(my_train['I4Q5']==1)] = 3

my_train['I4Q6'][(my_train['I4Q6']==3) | (my_train['I4Q6']==4)] = 1
my_train['I4Q6'][(my_train['I4Q6']==2)] = 2
my_train['I4Q6'][(my_train['I4Q6']==1)] = 3

my_train['I4Q7'][(my_train['I4Q7']==3) | (my_train['I4Q7']==4)] = 1
my_train['I4Q7'][(my_train['I4Q7']==2)] = 2
my_train['I4Q7'][(my_train['I4Q7']==1)] = 3

print(my_train.head(3))

```

	A1S1	A1N1	C2S2	H1_1	H4_1	I4Q1	I4Q2	I4Q3	I4Q4	I4Q5	I4Q6	\
global_id												
111016.0	1.0	4.0	2.0	3.0	1.0	3.0	3.0	3.0	3.0	3.0	3.0	
111037.0	1.0	4.0	2.0	3.0	1.0	3.0	3.0	3.0	3.0	3.0	3.0	
111039.0	3.0	4.0	3.0	3.0	5.0	3.0	3.0	3.0	3.0	3.0	3.0	

	I4Q7
global_id	
111016.0	3.0
111037.0	3.0
111039.0	3.0

완성된 데이터원본 저장해놓기

```
my_train_original = my_train[:]
```

K-MEANS 클러스터

```
model = KMeans(n_clusters=3)
model.fit(my_train)
kmeans_predict = model.fit_predict(my_train)
my_train['k_means'] = kmeans_predict
print(my_train.groupby('k_means').size())
```

```
k_means
0      849
1      343
2        20
dtype: int64
```

K-means + dummy

```
my_train_dummy = my_train_original[:]
for index in index_list:
    my_train_dummy[index] = my_train_dummy[index].astype(str)
my_train_dummy = pd.get_dummies(my_train_dummy, columns=[index])
```

```
model = KMeans(n_clusters=3)
model.fit(my_train_dummy)
kmeans_dummy_predict = model.fit_predict(my_train_dummy)
my_train['k_means_dummy'] = kmeans_dummy_predict
print(my_train.groupby('k_means_dummy').size())
```

```
k_means_dummy
0      253
1      598
2      361
dtype: int64
```

K-Means + PCA

```
my_train_pca = my_train_original[:]
pca = PCA(n_components = 10)
pca.fit(my_train_pca)
# print('singular value :', pca.singular_values_)
# print('singular vector :\n', pca.components_.T)
print('explained variance ratio : \n', pca.explained_variance_ratio_)
percentile = []
for i in pca.explained_variance_ratio_:
    if len(percentile) == 0:
        percentile.append(i)
    else:
        percentile.append(percentile[-1]+i)
print(' ')
print('PCA의 데이터반영 누적비율')
print(percentile)
```

```
explained variance ratio :  
[0.318961  0.23951282 0.11738039 0.10428578 0.05561842 0.04536555  
0.03137973 0.02508638 0.02143859 0.01627272]
```

PCA의 데이터반영 누적비율

```
[0.31896099752677104, 0.5584738156795982, 0.6758542050584552, 0.78013998445393,  
0.8357584027054615, 0.8811239543351679, 0.912503682206646, 0.937590057564122,  
0.9590286515428968, 0.975301374379612]
```

따라서 5개부터 시도해보겠음(80%반영! -지극히 주관적)

```
my_train_pca = my_train_original[:]  
pca = PCA(n_components=5).fit(my_train_pca)  
my_train_pca_5 = pca.transform(my_train_pca)
```

```
model = KMeans(n_clusters=3)  
model.fit(my_train_pca_5)  
kmeans_pca_predict = model.fit_predict(my_train_pca_5)  
my_train['k_means_pca'] = kmeans_pca_predict  
print(my_train.groupby('k_means_pca').size())
```

```
k_means_pca  
0      849  
1       20  
2      343  
dtype: int64
```

그때그때 너무 다른결과가 나오는데 어떻게하지 (코드 돌릴때마다 다름)

근데 전체적인 군집에 나눔정도는 비슷(군집번호만바뀜)

K-means + dummy + PCA

```
my_train_dummy_pca = my_train_dummy[:]  
  
pca = PCA(n_components = 10)  
pca.fit(my_train_dummy_pca)  
  
# print('singular value :', pca.singular_values_)  
# print('singular vector :\n', pca.components_.T)  
print('explained variance ratio : \n', pca.explained_variance_ratio_)  
percentile = []  
for i in pca.explained_variance_ratio_:  
    if len(percentile) == 0:  
        percentile.append(i)  
    else:  
        percentile.append(percentile[-1]+i)  
print('PCA의 데이터반영 비율의 누적은 아래와같다!')  
print(percentile)
```

explained variance ratio :

```
[0.23509272 0.12013462 0.09408357 0.08814047 0.06633756 0.06310844
 0.05589778 0.05230321 0.04766146 0.0389537 ]
```

PCA의 데이터반영 비율의 누적은 아래와 같다!

```
[0.2350927195249571, 0.35522734324228655, 0.4493109112940427,
0.5374513815624317, 0.6037889368639298, 0.6668973725927927, 0.7227951562310242,
0.7750983633367577, 0.8227598269764591, 0.8617135297943245]
```

```
my_train_dummy_pca = my_train_dummy[:]
pca = PCA(n_components=9).fit(my_train_dummy_pca)
my_train_dummy_pca_9 = pca.transform(my_train_dummy_pca)
model = KMeans(n_clusters=3)
model.fit(my_train_dummy_pca_9)
kmeans_dummy_pca_predict = model.fit_predict(my_train_dummy_pca_9)
my_train['k_means_dummy_pca'] = kmeans_dummy_pca_predict
print(my_train.groupby('k_means_dummy_pca').size())
```

```
k_means_dummy_pca
0    363
1    572
2    277
dtype: int64
```

스펙트럴 군집

```
from sklearn.cluster import SpectralClustering
```

```
spectral = SpectralClustering(n_clusters=3, n_init=10)
spectral_predict = spectral.fit_predict(my_train)
my_train['Spectral'] = spectral_predict

spectral_dummy_predict = spectral.fit_predict(my_train_dummy)
my_train['Spectral_dummy'] = spectral_dummy_predict

spectral_pca_predict = spectral.fit_predict(my_train_pca)
my_train['Spectral_pca'] = spectral_pca_predict

spectral_dummy_pca_predict = spectral.fit_predict(my_train_dummy_pca)
my_train['Spectral_dummy_pca'] = spectral_dummy_pca_predict

print(my_train.head(3))
print('-----')
print(my_train.groupby('Spectral').size())
# print('-----')
print(my_train.groupby('Spectral_dummy').size())
print('-----')
print(my_train.groupby('Spectral_pca').size())
print('-----')
print(my_train.groupby('Spectral_dummy_pca').size())
```

```
#
my_train.to_csv('C:/Users/user/Desktop/Statistical_Data_Idea_Contest/0423/kyumin
/cluster_result.csv', index=True)
```

	A1S1	A1N1	C2S2	H1_1	H4_1	I4Q1	I4Q2	I4Q3	I4Q4	I4Q5	I4Q6	\
global_id												
111016.0	1.0	4.0	2.0	3.0	1.0	3.0	3.0	3.0	3.0	3.0	3.0	
111037.0	1.0	4.0	2.0	3.0	1.0	3.0	3.0	3.0	3.0	3.0	3.0	
111039.0	3.0	4.0	3.0	3.0	5.0	3.0	3.0	3.0	3.0	3.0	3.0	

	I4Q7	k_means	k_means_dummy	k_means_pca	k_means_dummy_pca	\
global_id						
111016.0	3.0	0		1	0	1
111037.0	3.0	0		1	0	1
111039.0	3.0	1		1	2	1

	Spectral	Spectral_dummy	Spectral_pca	Spectral_dummy_pca
global_id				
111016.0	0		0	2
111037.0	0		0	2
111039.0	0		0	2

```
-----
----
Spectral
0    1192
1      4
2     16
dtype: int64
Spectral_dummy
0     881
1      2
2     329
dtype: int64
-----
Spectral_pca
0    1192
1      4
2     16
dtype: int64
-----
Spectral_dummy_pca
0     329
1      2
2     881
dtype: int64
```

혹시몰라서 int로도 string으로도 해봤으나 결과 값은 똑같다!

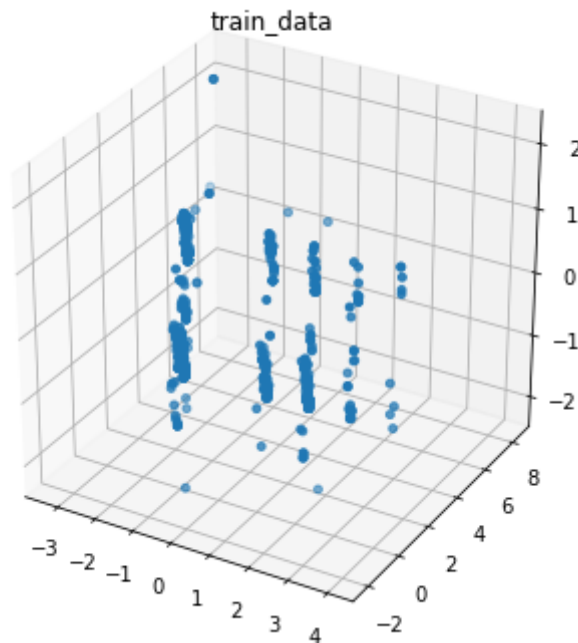
시각화 (결과 판단)

k-means

```

my_train_show = my_train_original[:]
pca = PCA(n_components=3).fit(my_train_show) #2개로 진행해봄
pca_train = pca.transform(my_train_show)
plt.rcParams["figure.figsize"] = (6,6)
fig = plt.figure()
ax = fig.add_subplot(111, projection='3d')
ax.scatter(pca_train[:, 0], pca_train[:, 1], pca_train[:,2], marker='o', s=15)
ax.set_title('train_data')
plt.show()

```



```
<class 'matplotlib.axes._subplots.Axes3DSubplot'>
```

K-means

```

my_train_show = my_train_original[:]
pca = PCA(n_components=3).fit(my_train_show)
pca_train = pca.transform(my_train_show)
plt.rcParams["figure.figsize"] = (6,6)
fig = plt.figure()
ax = fig.add_subplot(111, projection='3d')
ax.scatter(pca_train[:, 0], pca_train[:, 1], pca_train[:,2], c=kmeans_predict,
marker='o', s=15)

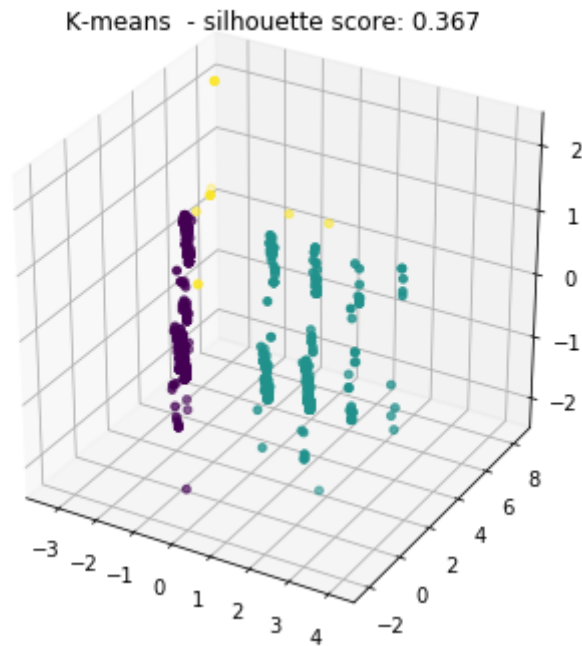
ax.set_title('k-means - silhouette score:
{: .3f}'.format(silhouette_score(my_train_show, kmeans_predict)))
print(my_train.groupby('k_means').size())
plt.show()

```

```

k_means
0      849
1      343
2       20
dtype: int64

```

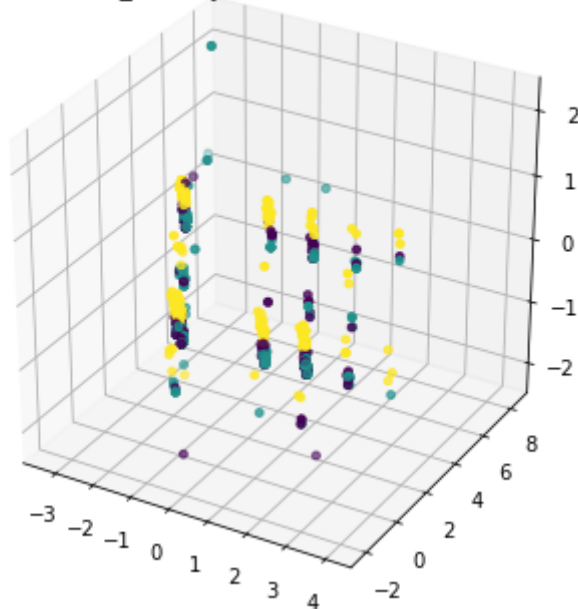
k+dummy

```
my_train_show = my_train_original[:]
pca = PCA(n_components=3).fit(my_train_show)
pca_train = pca.transform(my_train_show)
plt.rcParams["figure.figsize"] = (6,6)
fig = plt.figure()
ax = fig.add_subplot(111, projection='3d')
ax.scatter(pca_train[:, 0], pca_train[:, 1], pca_train[:,2],
          c=kmeans_dummy_predict, marker='o', s=15)

ax.set_title('K-means_dummy - silhouette score:
{: .3f}'.format(silhouette_score(my_train_show, kmeans_dummy_predict)))
print(my_train.groupby('k_means_dummy').size())
plt.show()
```

```
k_means_dummy
0    253
1    598
2    361
dtype: int64
```

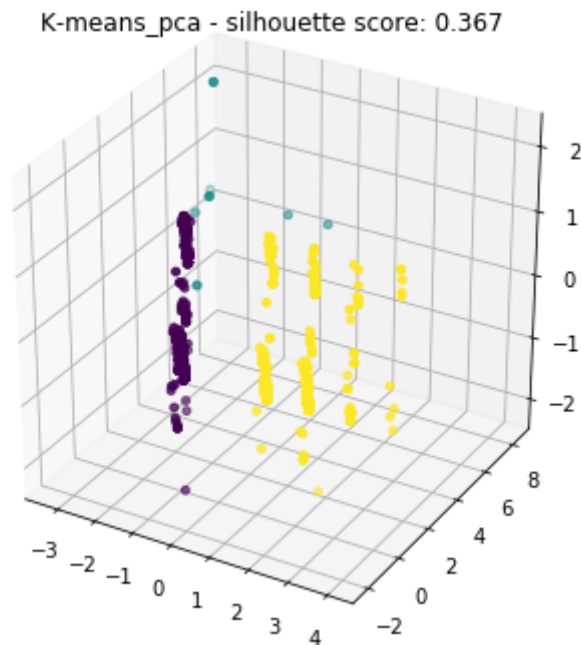
K-means_dummy - silhouette score: 0.110



k+PCA

```
my_train_show = my_train_original[:]  
pca = PCA(n_components=3).fit(my_train_show)  
pca_train = pca.transform(my_train_show)  
plt.rcParams["figure.figsize"] = (6,6)  
fig = plt.figure()  
ax = fig.add_subplot(111, projection='3d')  
ax.scatter(pca_train[:, 0], pca_train[:, 1], pca_train[:,2],  
           c=kmeans_pca_predict, marker='o', s=15)  
ax.set_title('K-means_pca - silhouette score:  
{:.3f}'.format(silhouette_score(my_train_show, kmeans_pca_predict)))  
print(my_train.groupby('k_means_pca').size())  
plt.show()
```

```
k_means_pca  
0    849  
1     20  
2    343  
dtype: int64
```

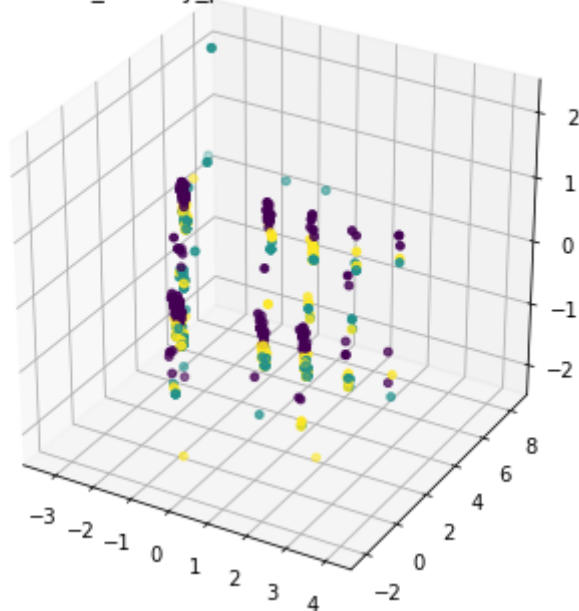


K+PCA+DUMMY

```
my_train_show = my_train_original[:]
pca = PCA(n_components=3).fit(my_train_show)
pca_train = pca.transform(my_train_show)
plt.rcParams["figure.figsize"] = (6,6)
fig = plt.figure()
ax = fig.add_subplot(111, projection='3d')
ax.scatter(pca_train[:, 0], pca_train[:, 1], pca_train[:,2],
           c=kmeans_dummy_pca_predict, marker='o', s=15)
ax.set_title('K-means_dummy_pca - silhouette score:
{: .3f}'.format(silhouette_score(my_train_show, kmeans_dummy_pca_predict)))
print(my_train.groupby('k_means_dummy_pca').size())
plt.show()
```

```
k_means_dummy_pca
0    363
1    572
2    277
dtype: int64
```

K-means_dummy_pca - silhouette score: 0.107

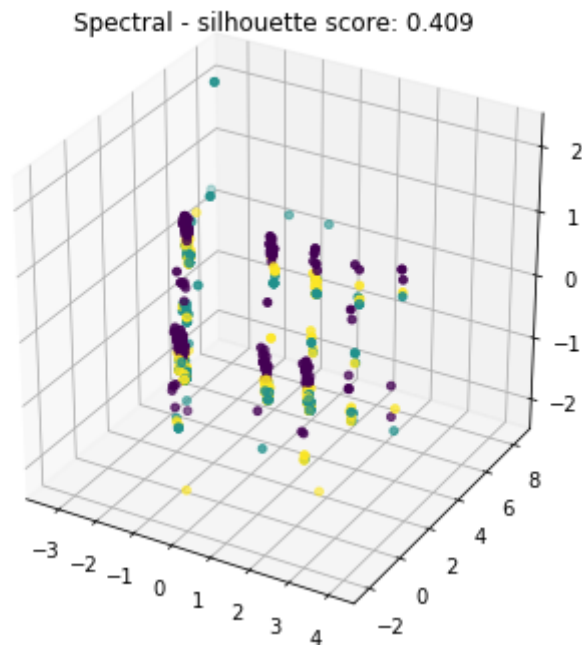


Spectral

sepctral

```
my_train_show = my_train_original[:]  
pca = PCA(n_components=3).fit(my_train_show)  
pca_train = pca.transform(my_train_show)  
plt.rcParams["figure.figsize"] = (6,6)  
fig = plt.figure()  
ax = fig.add_subplot(111, projection='3d')  
ax.scatter(pca_train[:, 0], pca_train[:, 1], pca_train[:,2],  
           c=kmeans_dummy_pca_predict, marker='o', s=15)  
ax.set_title('Spectral - silhouette score:  
{:.3f}'.format(silhouette_score(my_train_show, spectral_predict)))  
print(my_train.groupby('Spectral').size())  
plt.show()
```

```
Spectral  
0    1192  
1         4  
2        16  
dtype: int64
```



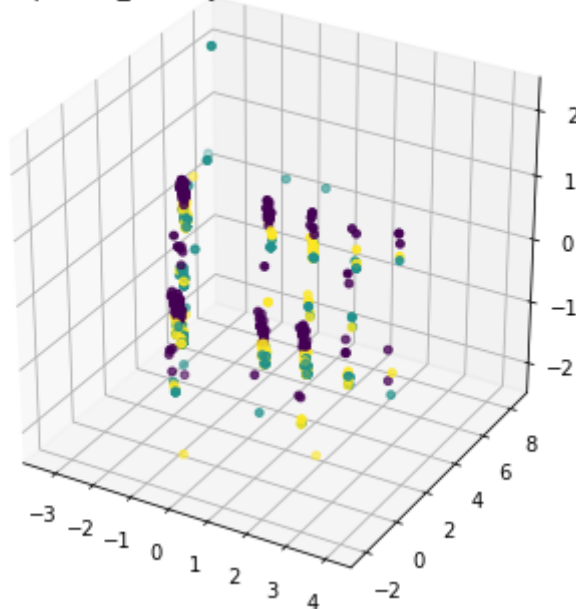
spectral+dummy

```
my_train_show = my_train_original[:]
pca = PCA(n_components=3).fit(my_train_show)
pca_train = pca.transform(my_train_show)
plt.rcParams["figure.figsize"] = (6,6)
fig = plt.figure()
ax = fig.add_subplot(111, projection='3d')
ax.scatter(pca_train[:, 0], pca_train[:, 1], pca_train[:, 2],
           c=kmeans_dummy_pca_predict, marker='o', s=15)
ax.set_title('Spectral_dummy - silhouette score:
{:.3f}'.format(silhouette_score(my_train_show, spectral_dummy_predict)))
print(my_train.groupby('Spectral_dummy').size())

plt.show()
```

```
Spectral_dummy
0      881
1         2
2      329
dtype: int64
```

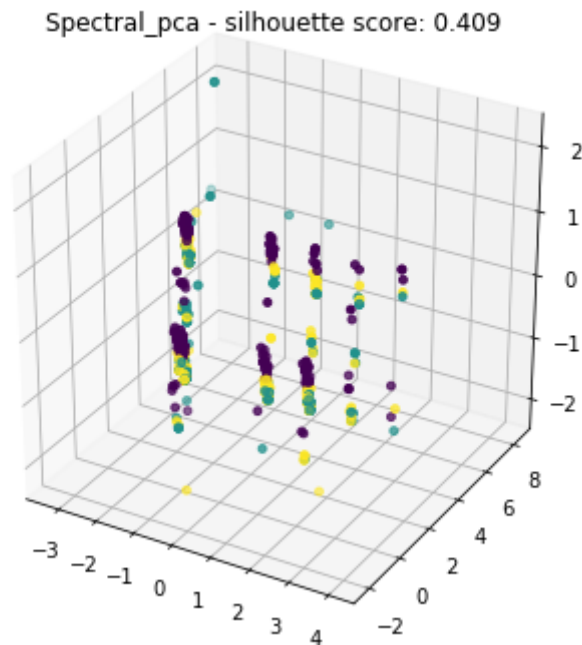
Spectral_dummy - silhouette score: 0.065



spectral+PCA

```
my_train_show = my_train_original[:]  
pca = PCA(n_components=3).fit(my_train_show)  
pca_train = pca.transform(my_train_show)  
plt.rcParams["figure.figsize"] = (6,6)  
fig = plt.figure()  
ax = fig.add_subplot(111, projection='3d')  
ax.scatter(pca_train[:, 0], pca_train[:, 1], pca_train[:,2],  
           c=kmeans_dummy_pca_predict, marker='o', s=15)  
ax.set_title('Spectral_pca - silhouette score:  
{:.3f}'.format(silhouette_score(my_train_show, spectral_pca_predict)))  
print(my_train.groupby('Spectral_pca').size())  
plt.show()
```

```
Spectral_pca  
0    1192  
1         4  
2        16  
dtype: int64
```

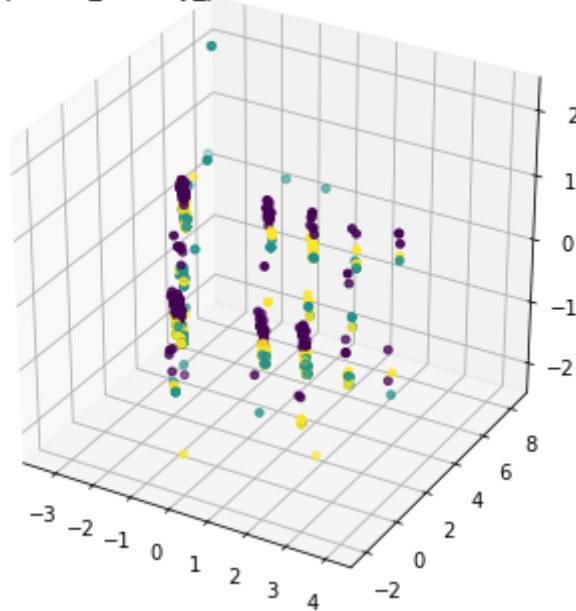


spectral+dummy+PCA

```
my_train_show = my_train_original[:]
pca = PCA(n_components=3).fit(my_train_show)
pca_train = pca.transform(my_train_show)
plt.rcParams["figure.figsize"] = (6,6)
fig = plt.figure()
ax = fig.add_subplot(111, projection='3d')
ax.scatter(pca_train[:, 0], pca_train[:, 1], pca_train[:,2],
           c=kmeans_dummy_pca_predict, marker='o', s=15)
ax.set_title('spectral_dummy_pca - silhouette score:
{:.3f}'.format(silhouette_score(my_train_show, spectral_dummy_pca_predict)))
print(my_train.groupby('spectral_dummy_pca').size())
plt.show()
```

```
Spectral_dummy_pca
0    329
1      2
2    881
dtype: int64
```

spectral_dummy_pca - silhouette score: 0.065



전체결과 출력 및 저장

```
print(my_train.head(3))
my_train.to_csv('C:/Users/user/Desktop/Statistical_Data_Idea_Contest/0423/kyumin/cluster_result.csv', index=True)
```

	A1S1	A1N1	C2S2	H1_1	H4_1	I4Q1	I4Q2	I4Q3	I4Q4	I4Q5	I4Q6	\
global_id												
111016.0	1.0	4.0	2.0	3.0	1.0	3.0	3.0	3.0	3.0	3.0	3.0	
111037.0	1.0	4.0	2.0	3.0	1.0	3.0	3.0	3.0	3.0	3.0	3.0	
111039.0	3.0	4.0	3.0	3.0	5.0	3.0	3.0	3.0	3.0	3.0	3.0	

	I4Q7	k_means	k_means_dummy	k_means_pca	k_means_dummy_pca	\
global_id						
111016.0	3.0	0	1	0	1	
111037.0	3.0	0	1	0	1	
111039.0	3.0	1	1	2	1	

	Spectral	Spectral_dummy	Spectral_pca	Spectral_dummy_pca
global_id				
111016.0	0	0	0	2
111037.0	0	0	0	2
111039.0	0	0	0	2

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