part_1_green

January 23, 2025

1 Imports

```
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
from sklearn.preprocessing import StandardScaler
from scipy.stats import zscore
from sklearn.cluster import KMeans, AgglomerativeClustering, DBSCAN,
SpectralClustering
from sklearn.mixture import GaussianMixture
from sklearn.metrics import silhouette_score, davies_bouldin_score,
calinski_harabasz_score
from sklearn.neighbors import kneighbors_graph
from sklearn.decomposition import PCA
```

2 Load Data

```
[2]: # Load the data
file_path = "fic_epita_kantar_codes.csv"
data_codes = pd.read_csv(file_path, sep=';')
```

```
[4]: # Check for missing values
missing_values = subset_data.isnull().sum()
print("Missing Values:\n", missing_values)
```

```
# Data type analysis
data_types = subset_data.dtypes
print("\nData Types:\n", data_types)

# Summary statistics
summary_stats = subset_data.describe()
print("\nSummary Statistics:\n", summary_stats)

# Check for outliers
outliers = subset_data.apply(zscore).abs() > 3
print("\nOutliers:\n", outliers)

Missing Values:
A11      0
A12      0
```

0	
A11	0
A12	0
A13	0
A14	0
A4	0
A5	1443
A5bis	3557
A8_1_slice	0
A8_2_slice	0
A8_3_slice	0
A8_4_slice	0
B1_1_slice	0
B1_2_slice	0
B2_1_slice	0
B2_2_slice	0
B3	0
B4	0
B6	0
C1_1_slice	0
C1_2_slice	0
C1_3_slice	0
C1_4_slice	0
C1_5_slice	0
C1_6_slice	0
C1_7_slice	0
C1_8_slice	0
C1_9_slice	0
dtype: int64	

Data Types:

A11	int6
A12	int64
A13	int64
A14	int64
A4	int64

A5	float64
A5bis	float64
A8_1_slice	int64
A8_2_slice	int64
A8_3_slice	int64
A8_4_slice	int64
B1_1_slice	int64
B1_2_slice	int64
B2_1_slice	int64
B2_2_slice	int64
B3	int64
B4	int64
B6	int64
C1_1_slice	int64
C1_2_slice	int64
C1_3_slice	int64
C1_4_slice	int64
C1_5_slice	int64
C1_6_slice	int64
C1_7_slice	int64
C1_8_slice	int64
C1_9_slice	int64
dtype: object	

Summary Statistics:

	A11	A12	A13	A14	A4 \		
count	5000.000000	5000.000000	5000.000000	5000.0 500	00.00000		
mean	0.781800	0.225000	0.402200	0.0	1.421600		
std	0.413065	0.417624	0.490391	0.0	0.714112		
min	0.000000	0.000000	0.000000	0.0	1.000000		
25%	1.000000	0.000000	0.000000	0.0	1.000000		
50%	1.000000	0.000000	0.000000	0.0	1.000000		
75%	1.000000	0.000000	1.000000	0.0	2.000000		
max	1.000000	1.000000	1.000000	0.0	3.000000		
	A5	A5bis	A8_1_slice	A8_2_slice	e A8_3_slice	•••	\
count	3557.000000	1443.000000	5000.000000	5000.000000	5000.000000	•••	
mean	2.326399	1.821206	5.122200	5.521400	3.791200		
std	1.243436	0.987393	6.738067	7.00106	5.524971	•••	
min	1.000000	1.000000	0.000000	0.00000	0.000000	•••	
25%	1.000000	1.000000	1.000000	1.000000	1.000000	•••	
50%	2.000000	2.000000	3.000000	3.000000	2.000000	•••	
75%	3.000000	2.000000	6.000000	7.00000	5.000000		
max	5.000000	4.000000	50.000000	50.00000	50.000000	•••	
	В6	C1_1_slice	C1_2_slice	C1_3_slice	e C1_4_slice	\	
count	5000.000000	5000.000000	5000.000000	5000.000000	5000.000000		
mean	3.269000	4.192400	4.131200	4.550600	4.614200		

std min	0.875438 1.000000	0.876091 1.000000	0.897635 1.000000	0.773278 1.000000	0.784781 1.000000
25%	3.000000	4.000000	4.000000	4.000000	4.000000
50%	4.000000	4.000000	4.000000	5.000000	5.000000
75%	4.000000	5.000000	5.000000	5.000000	5.000000
max	4.000000	5.000000	5.000000	5.000000	5.000000
	C1_5_slice	C1_6_slice	C1_7_slice	C1_8_slice	C1_9_slice
count	5000.000000	5000.000000	5000.000000	5000.000000	5000.000000
mean	4.493000	4.682200	4.541000	4.336600	4.352400
std	0.884368	0.726711	0.788951	0.822336	0.849446
min	1.000000	1.000000	1.000000	1.000000	1.000000
25%	4.000000	5.000000	4.000000	4.000000	4.000000
50%	5.000000	5.000000	5.000000	5.000000	5.000000
75%	5.000000	5.000000	5.000000	5.000000	5.000000
max	5.000000	5.000000	5.000000	5.000000	5.000000

[8 rows x 27 columns]

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	A11	A12	A13	A14	A4	. A5	A5bis	A8_1_sl	ice A8_2_	slice
\										
0	False	False	False	False	False	False	False	Fal	se F	alse
1	False	False	False	False	False	False	False	Fal	se F	alse
2	False	False	False	False	False	False	False	Fal	se F	alse
3	False	False	False	False	False	False	False	Fal	se F	alse
4	False	False	False	False	False	False	False	Fal	se F	alse
•••		•••			•••	•••		•••		
4995	False	False	False	False	False	False	False	Fal	se F	alse
4996	False	False	False	False	False	False	False	Fal	se F	alse
4997	False	False	False	False	False	False	False	Fal	se F	alse
4998	False	False	False	False	False	False	False	Fal	se F	alse
4999	False	False	False	False	False	False	False	Fal	se F	alse
									C1_4_slic	e \
0	F	alse …	False	•	False	Fa	lse	True	Tru	е
1		alse …						False		е
2	F	alse …	False	•	False	Fa	lse	False	Fals	е
3		alse …						False	Fals	
4	F	alse …	False		False	Fa	lse	False	Fals	е
•••	•••	•••		•••	•••	•		•••		
4995	F	alse	False		False	Fa	lse	False	Fals	е
4996	F	alse	False		False	Fa	lse	False	Fals	е
4997	F	alse …	False		False	Fa	lse	False	Fals	е
4998	F	alse	False		False	Fa	lse	False	Fals	е
4999	F	alse …	False		False	Fa	lse	False	Fals	е

 ${\tt C1_5_slice} \quad {\tt C1_6_slice} \quad {\tt C1_7_slice} \quad {\tt C1_8_slice} \quad {\tt C1_9_slice}$

0	False	True	True	False	False
1	False	False	False	False	False
2	False	False	False	False	False
3	False	False	False	False	False
4	False	False	False	False	False
•••	•••		•••	•••	
 4995	 False	 False	 False	 False	False
					False False
4995	False	False	False	False	
4995 4996	False False	False False	False False	False False	False

[5000 rows x 27 columns]

```
[5]: # Handle NaN values in columns A5 and A5bis
subset_data['A5'].fillna(subset_data['A5'].mean(), inplace=True)
subset_data['A5bis'].fillna(subset_data['A5bis'].mean(), inplace=True)
```

/tmp/ipykernel_21221/3626455305.py:2: FutureWarning: A value is trying to be set on a copy of a DataFrame or Series through chained assignment using an inplace method.

The behavior will change in pandas 3.0. This inplace method will never work because the intermediate object on which we are setting values always behaves as a copy.

For example, when doing 'df[col].method(value, inplace=True)', try using 'df.method({col: value}, inplace=True)' or df[col] = df[col].method(value) instead, to perform the operation inplace on the original object.

```
subset_data['A5'].fillna(subset_data['A5'].mean(), inplace=True)
/tmp/ipykernel_21221/3626455305.py:2: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame
```

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy subset_data['A5'].fillna(subset_data['A5'].mean(), inplace=True) /tmp/ipykernel_21221/3626455305.py:3: FutureWarning: A value is trying to be set on a copy of a DataFrame or Series through chained assignment using an inplace method.

The behavior will change in pandas 3.0. This inplace method will never work because the intermediate object on which we are setting values always behaves as a copy.

For example, when doing 'df[col].method(value, inplace=True)', try using 'df.method({col: value}, inplace=True)' or df[col] = df[col].method(value) instead, to perform the operation inplace on the original object.

```
subset_data['A5bis'].fillna(subset_data['A5bis'].mean(), inplace=True)
/tmp/ipykernel_21221/3626455305.py:3: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame
```

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy subset_data['A5bis'].fillna(subset_data['A5bis'].mean(), inplace=True)

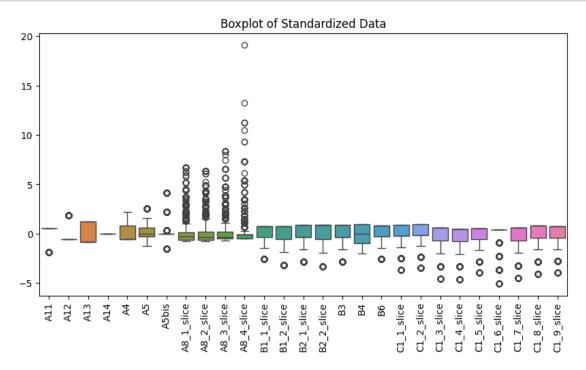
```
[6]: # Correlation heatmap
plt.figure(figsize=(15, 10))
sns.heatmap(subset_data.corr(), annot=True, cmap="coolwarm")
plt.title("Correlation Matrix")
plt.show()
```

Correlation Matrix

```
1.0
                                             -0.68<mark>9e-17<mark>0.22 0.23 0.23 0.21 0.1-</mark>0.0680.17-0.15-0.23-0.16-0.31-0.0320.08<del>2</del>0.0920.0440.020.036.00140.0520.0820.074</mark>
         A11 - 1 -0.6 0.046
         A12 -
                  -0.6 1 -0.18
                                               0.310.011-0.31-0.13-0.13-0.120.04D.0470.110.0950.140.0630.160.03D.0150.0410.01D.0260.0020.0280.0020.018.0088
         A13 -0.046-0.18 1
                                              0.3 0.079 0.3 0.0530.0440.0520.05-0.0340.0590.0640.0840.060.06450.0760.12-0.130.0840.0840.0840.0930.076-0.1 -0.13-0.12
                                                                                                                                                                                                                                      0.8
           A4 -<mark>-0.68</mark> 0.31 0.3
                                             1 .5e-16.21 -0.24-0.22-0.21-0.110.0350.12 0.09 0.14 0.11 0.27.7e-050340.0350.0070.00840.010.0260.0170.0290.015
           A5 -9e-170.0110.079 -1.5e-101 2e-170.16 0.14 0.16 0.150.034 0.086.0040.0430.0630.0950.0580.0330.0640.0130.0130.0130.0130.0130.0130.0350.0630.031
       A5bis -0.22 -0.31 0.3
                                              0.212e-17 1 0.0410.0440.0370.0160.0460.0750.0860.0970.067-0.1-0.0430.080.0860.0650.0550.06-0.050.0780.0860.081
                                                                                                                                                                                                                                      - 0.6
                                             -0.24<mark>0.160.041 1 0.82 0.8 0.59-</mark>0.0150.0240.12-0.13-0.27-0.29-0.18-0.14-0.110.0890.0680.0780.0280.0820.13-0.13
A8_1_slice - 0.23 - 0.130.053
                                             -0.22 <mark>0.140.044 0.82 1 0.74 0.55-</mark>0.0390.0540.13-0.13-0.27 -0.3 -0.17-0.15-0.12-0.11-0.11-0.1-0.0710.12-0.15-0.15
A8_2_slice - 0.23 - 0.13 0.044
                                             -0.21<mark>0.16</mark>0.037 0.8 0.74 1 0.7 -0.01<del>-0</del>.01<del>-0</del>.01<del>2</del>0.0990.11-0.24-0.25-0.16-0.13-0.110.0830.050.0640.01<del>-0</del>.0720.12-0.12
A8_3_slice - 0.21 - 0.120.052
                                                                                                                                                                                                                                      - 0.4
A8 4 slice - 0.1-0.0410.05
                                             -0.11 <mark>0.150.016 0.59 0.55 0.7 1 0.042</mark>0.0370.090.095-0.2 -0.19-0.16-0.13-0.110.0960.0760.0790.0460.0890.13-0.13
B1_1_slice -0.0680.0470.039
                                             0.0350.0340.0460.0150.0390.0120.042 1 0.69 0.56 0.41 0.21 0.15 0.19 0.27 0.24 0.3 0.32 0.27 0.33 0.3 0.3 0.29
B1_2_slice -0.17 0.11-0.059 0.120.0086.075.0240.0540.0120.0370.65 1 0.49 0.54 0.26 0.22 0.19 0.31 0.29 0.36 0.35 0.31 0.36 0.35 0.35 0.35 0.35
                                                                                                                                                                                                                                     - 0.2
B2_1_slice -0.150.0950.069 0.090.0040.0860.12-0.130.0990.09 0.56 0.49 1 0.75 0.36 0.31 0.26 0.35 0.34 0.36 0.36 0.32 0.37 0.36 0.38 0.36
B2_2_slice -0.23 0.140.084 0.140.0430.0970.13-0.13-0.110.0950.41 0.54 0.75 1 0.37 0.34 0.25 0.36 0.35 0.38 0.37 0.34 0.37 0.38 0.4 0.37
          B6 -0.0320.0320.076 -1.7e-950580.0430.18-0.17-0.16-0.16 0.19 0.19 0.26 0.25 0.39 0.29 1 0.39 0.36 0.38 0.33 0.32 0.33 0.38 0.41 0.42
- -0.2
C1_5_slice -0.03\( 6\).002\( 0.03\( 0.020\).003\( 0.03\( 0.03\( 0.03\) 0.02\( 0.03\( 0.03\) 0.02\( 0.03\) 0.003\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 0.03\) 0.03\( 
                                                                                                                                                                                                                                       -0.4
C1_7_slice -0.0570.002-0.1 0.0170.0350.0760.0820.120.0720.089 0.3 0.35 0.36 0.38 0.36 0.26 0.38 0.65 0.61 0.75 0.72 0.63 0.71 1 0.6
AB_1_sitce
AB_2_sitce
AB_3_sitce
AB_1_sitce
B1_1_sitce
B2_1_sitce
B2_1_sitce
B2_1_sitce
B2_1_sitce
B2_1_sitce
B2_2_sitce
B3_1_sitce
B4_1_sitce
B6_1_sitce
C1_2_sitce
C1_2_sitce
C1_2_sitce
C1_2_sitce
C1_2_sitce
C1_2_sitce
```

```
[7]: # Standardize data
scaler = StandardScaler()
scaled_data = scaler.fit_transform(subset_data)
```

```
[8]: # Check if clustering is appropriate (variance and distribution)
plt.figure(figsize=(10, 5))
sns.boxplot(data=pd.DataFrame(scaled_data, columns=columns_of_interest))
plt.xticks(rotation=90)
plt.title("Boxplot of Standardized Data")
plt.show()
```



3 Clustering

```
[9]: # Function to evaluate clustering
def evaluate_clustering(labels, data):
    if len(set(labels)) > 1: # At least 2 clusters
        silhouette = silhouette_score(data, labels)
        davies_bouldin = davies_bouldin_score(data, labels)
        calinski_harabasz = calinski_harabasz_score(data, labels)
    else:
        silhouette, davies_bouldin, calinski_harabasz = None, None
    return silhouette, davies_bouldin, calinski_harabasz
results = {}
```

3.1 Kmeans

```
[10]: from sklearn.model_selection import ParameterGrid
      from sklearn.decomposition import PCA
      def grid_search_kmeans(data, param_grid):
          results = []
          for params in ParameterGrid(param_grid):
              kmeans = KMeans(**params, random_state=42)
              kmeans_labels = kmeans.fit_predict(data)
              score = evaluate_clustering(kmeans_labels, data)
              results.append({'params': params, 'score': score})
          return results
      # Try multiple values of PCA components
      pca results = {}
      pca_components = [2, 3, 4, 5] # List of PCA components to try
      for n_components in pca_components:
          pca = PCA(n_components=n_components)
          reduced_data = pca.fit_transform(scaled_data)
          # Define the parameter grid for KMeans
          param_grid = {
              'n_clusters': [3, 4, 5, 6, 7, 8, 9],
              'init': ['k-means++', 'random'],
              'max_iter': [100, 300],
              'n_init': [10, 20]
          }
          pca_results[n_components] = grid_search_kmeans(reduced_data, param_grid)
      results['K-Means'] = pca_results
```

3.2 Hierarchical Clustering

```
[11]: from sklearn.model_selection import ParameterGrid
from sklearn.decomposition import PCA

def grid_search_agglo(data, param_grid):
    results = []
    for params in ParameterGrid(param_grid):
        agglo = AgglomerativeClustering(**params)
        agglo_labels = agglo.fit_predict(data)
        score = evaluate_clustering(agglo_labels, data)
        results.append({'params': params, 'score': score})
    return results
```

```
# Try multiple values of PCA components
pca_results = {}
pca_components = [2, 3, 4, 5]  # List of PCA components to try

for n_components in pca_components:
    pca = PCA(n_components=n_components)
    reduced_data = pca.fit_transform(scaled_data)

# Define the parameter grid for Agglomerative Clustering
param_grid = {
        'n_clusters': [3, 4, 5, 6, 7, 8, 9],
        'linkage': ['ward', 'complete', 'average', 'single']
    }

    pca_results[n_components] = grid_search_agglo(reduced_data, param_grid)

results['Agglomerative'] = pca_results
```

3.3 DBSCAN

```
[12]: from sklearn.model selection import ParameterGrid
      def grid_search_dbscan(data, param_grid):
          results = []
          for params in ParameterGrid(param_grid):
              dbscan = DBSCAN(**params)
              dbscan_labels = dbscan.fit_predict(data)
              score = evaluate_clustering(dbscan_labels, data)
              results.append({'params': params, 'score': score})
          return results
      # Try multiple values of PCA components
      pca_results = {}
      pca components = [2, 3, 4, 5] # List of PCA components to try
      for n_components in pca_components:
          pca = PCA(n_components=n_components)
          reduced_data = pca.fit_transform(scaled_data)
          # Define the parameter grid for Agglomerative Clustering
          param_grid = {
              'eps': [0.5, 1.0, 1.5, 2.0],
              'min_samples': [3, 5, 10, 15],
          }
          pca_results[n_components] = grid_search_dbscan(reduced_data, param_grid)
```

```
results['DBSCAN'] = pca_results
```

3.4 Gaussian Mixture Model

```
[13]: from sklearn.model_selection import ParameterGrid
      def grid_search_gmm(data, param_grid):
          results = []
          for params in ParameterGrid(param_grid):
              gmm = GaussianMixture(**params)
              gmm_labels = gmm.fit_predict(data)
              score = evaluate clustering(gmm labels, data)
              results.append({'params': params, 'score': score})
          return results
      # Try multiple values of PCA components
      pca results = {}
      pca_components = [2, 3, 4, 5] # List of PCA components to try
      for n_components in pca_components:
          pca = PCA(n_components=n_components)
          reduced_data = pca.fit_transform(scaled_data)
          # Define the parameter grid for Agglomerative Clustering
          param_grid = {
              'n_components': [3, 4, 5, 6, 7, 8, 9],
              'covariance_type': ['full', 'tied', 'diag', 'spherical'],
              'random_state': [42]
          }
          pca_results[n_components] = grid_search_gmm(reduced_data, param_grid)
      results['GMM'] = pca_results
```

3.5 Spectral Clustering

```
[14]: from sklearn.model_selection import ParameterGrid

def grid_search_spectral(data, param_grid):
    results = []
    for params in ParameterGrid(param_grid):
        spectral = SpectralClustering(**params)
        spectral_labels = spectral.fit_predict(data)
        score = evaluate_clustering(spectral_labels, data)
        results.append({'params': params, 'score': score})
    return results
```

```
# Try multiple values of PCA components
pca_results = {}
pca_components = [2, 3, 4, 5] # List of PCA components to try
for n_components in pca_components:
    pca = PCA(n_components=n_components)
    reduced_data = pca.fit_transform(scaled_data)
    # Define the parameter grid for Agglomerative Clustering
    param_grid = {
         'n_clusters': [3, 4, 5, 6, 7, 8, 9],
         'affinity': ['nearest_neighbors'],
         'random_state': [42]
    }
    pca_results[n_components] = grid_search_spectral(reduced_data, param_grid)
results['Spectral'] = pca_results
/home/floflo/Documents/epita/epita-ml-scia/lib/python3.12/site-
packages/sklearn/manifold/_spectral_embedding.py:329: UserWarning: Graph is not
fully connected, spectral embedding may not work as expected.
  warnings.warn(
/home/floflo/Documents/epita/epita-ml-scia/lib/python3.12/site-
packages/sklearn/manifold/_spectral_embedding.py:329: UserWarning: Graph is not
fully connected, spectral embedding may not work as expected.
  warnings.warn(
/home/floflo/Documents/epita/epita-ml-scia/lib/python3.12/site-
packages/sklearn/manifold/_spectral_embedding.py:329: UserWarning: Graph is not
fully connected, spectral embedding may not work as expected.
  warnings.warn(
/home/floflo/Documents/epita/epita-ml-scia/lib/python3.12/site-
packages/sklearn/manifold/_spectral_embedding.py:329: UserWarning: Graph is not
fully connected, spectral embedding may not work as expected.
  warnings.warn(
/home/floflo/Documents/epita/epita-ml-scia/lib/python3.12/site-
packages/sklearn/manifold/_spectral_embedding.py:329: UserWarning: Graph is not
fully connected, spectral embedding may not work as expected.
  warnings.warn(
/home/floflo/Documents/epita/epita-ml-scia/lib/python3.12/site-
packages/sklearn/manifold/_spectral_embedding.py:329: UserWarning: Graph is not
fully connected, spectral embedding may not work as expected.
  warnings.warn(
/home/floflo/Documents/epita/epita-ml-scia/lib/python3.12/site-
packages/sklearn/manifold/_spectral_embedding.py:329: UserWarning: Graph is not
fully connected, spectral embedding may not work as expected.
  warnings.warn(
```

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/home/floflo/Documents/epita/epita-ml-scia/lib/python3.12/site-
packages/sklearn/manifold/_spectral_embedding.py:329: UserWarning: Graph is not
fully connected, spectral embedding may not work as expected.
  warnings.warn(
/home/floflo/Documents/epita/epita-ml-scia/lib/python3.12/site-
packages/sklearn/manifold/_spectral_embedding.py:329: UserWarning: Graph is not
fully connected, spectral embedding may not work as expected.
  warnings.warn(
/home/floflo/Documents/epita/epita-ml-scia/lib/python3.12/site-
packages/sklearn/manifold/_spectral_embedding.py:329: UserWarning: Graph is not
fully connected, spectral embedding may not work as expected.
  warnings.warn(
/home/floflo/Documents/epita/epita-ml-scia/lib/python3.12/site-
packages/sklearn/manifold/_spectral_embedding.py:329: UserWarning: Graph is not
fully connected, spectral embedding may not work as expected.
  warnings.warn(
/home/floflo/Documents/epita/epita-ml-scia/lib/python3.12/site-
packages/sklearn/manifold/_spectral_embedding.py:329: UserWarning: Graph is not
fully connected, spectral embedding may not work as expected.
  warnings.warn(
/home/floflo/Documents/epita/epita-ml-scia/lib/python3.12/site-
packages/sklearn/manifold/_spectral_embedding.py:329: UserWarning: Graph is not
fully connected, spectral embedding may not work as expected.
  warnings.warn(
/home/floflo/Documents/epita/epita-ml-scia/lib/python3.12/site-
packages/sklearn/manifold/_spectral_embedding.py:329: UserWarning: Graph is not
fully connected, spectral embedding may not work as expected.
  warnings.warn(
/home/floflo/Documents/epita/epita-ml-scia/lib/python3.12/site-
packages/sklearn/manifold/_spectral_embedding.py:329: UserWarning: Graph is not
fully connected, spectral embedding may not work as expected.
  warnings.warn(
/home/floflo/Documents/epita/epita-ml-scia/lib/python3.12/site-
packages/sklearn/manifold/_spectral_embedding.py:329: UserWarning: Graph is not
fully connected, spectral embedding may not work as expected.
  warnings.warn(
/home/floflo/Documents/epita/epita-ml-scia/lib/python3.12/site-
packages/sklearn/manifold/_spectral_embedding.py:329: UserWarning: Graph is not
fully connected, spectral embedding may not work as expected.
  warnings.warn(
/home/floflo/Documents/epita/epita-ml-scia/lib/python3.12/site-
packages/sklearn/manifold/_spectral_embedding.py:329: UserWarning: Graph is not
fully connected, spectral embedding may not work as expected.
  warnings.warn(
/home/floflo/Documents/epita/epita-ml-scia/lib/python3.12/site-
packages/sklearn/manifold/_spectral_embedding.py:329: UserWarning: Graph is not
fully connected, spectral embedding may not work as expected.
 warnings.warn(
```

```
/home/floflo/Documents/epita/epita-ml-scia/lib/python3.12/site-
     packages/sklearn/manifold/_spectral_embedding.py:329: UserWarning: Graph is not
     fully connected, spectral embedding may not work as expected.
       warnings.warn(
     /home/floflo/Documents/epita/epita-ml-scia/lib/python3.12/site-
     packages/sklearn/manifold/_spectral_embedding.py:329: UserWarning: Graph is not
     fully connected, spectral embedding may not work as expected.
       warnings.warn(
     /home/floflo/Documents/epita/epita-ml-scia/lib/python3.12/site-
     packages/sklearn/manifold/_spectral_embedding.py:329: UserWarning: Graph is not
     fully connected, spectral embedding may not work as expected.
       warnings.warn(
     /home/floflo/Documents/epita/epita-ml-scia/lib/python3.12/site-
     packages/sklearn/manifold/_spectral_embedding.py:329: UserWarning: Graph is not
     fully connected, spectral embedding may not work as expected.
       warnings.warn(
     /home/floflo/Documents/epita/epita-ml-scia/lib/python3.12/site-
     packages/sklearn/manifold/_spectral_embedding.py:329: UserWarning: Graph is not
     fully connected, spectral embedding may not work as expected.
       warnings.warn(
     /home/floflo/Documents/epita/epita-ml-scia/lib/python3.12/site-
     packages/sklearn/manifold/_spectral_embedding.py:329: UserWarning: Graph is not
     fully connected, spectral embedding may not work as expected.
       warnings.warn(
     /home/floflo/Documents/epita/epita-ml-scia/lib/python3.12/site-
     packages/sklearn/manifold/_spectral_embedding.py:329: UserWarning: Graph is not
     fully connected, spectral embedding may not work as expected.
       warnings.warn(
     /home/floflo/Documents/epita/epita-ml-scia/lib/python3.12/site-
     packages/sklearn/manifold/_spectral_embedding.py:329: UserWarning: Graph is not
     fully connected, spectral embedding may not work as expected.
       warnings.warn(
     /home/floflo/Documents/epita/epita-ml-scia/lib/python3.12/site-
     packages/sklearn/manifold/_spectral_embedding.py:329: UserWarning: Graph is not
     fully connected, spectral embedding may not work as expected.
       warnings.warn(
[15]: print(results)
     {'K-Means': {2: [{'params': {'init': 'k-means++', 'max_iter': 100, 'n_clusters':
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```

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```

```
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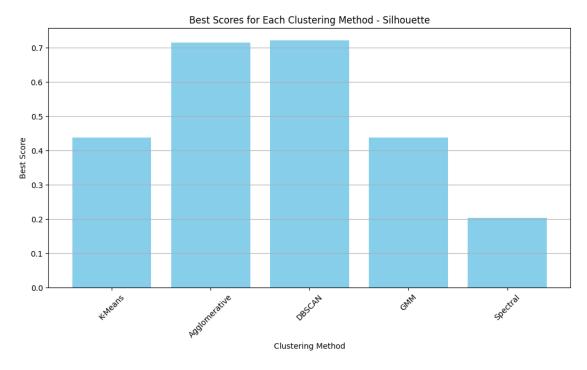
```
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```

3.6 Results

```
[16]: # Display best score and hyperparameters for each method
     best_hyperparameters = {}
     for method, pca_comps in results.items():
         for pca in pca_comps:
             best_score = max(pca_comps[pca], key=lambda x: x['score'][0]) # Get_\( \)
       → the score with the highest first value
              if method not in best_hyperparameters:
                 best_hyperparameters[method] = {'pca': pca, 'params':
       sbest_score['params'], 'score': best_score['score']}
              else:
                  current_best_score = best_hyperparameters[method]['score']
                 if best_score['score'] > current_best_score:
                     best_hyperparameters[method] = {'pca': pca, 'params':__
       ⇔best_score['params'], 'score': best_score['score']}
     print("Best Hyperparameters:")
     for method, params in best_hyperparameters.items():
         print(f"Method: {method}, PCA Components: {params['pca']}, Parameters: ____
       Best Hyperparameters:
     Method: K-Means, PCA Components: 2, Parameters: {'init': 'k-means++',
     'max_iter': 100, 'n_clusters': 4, 'n_init': 10}, Score: (0.43696184323503295,
     0.800807992091704, 4382.46484845105)
     Method: Agglomerative, PCA Components: 2, Parameters: {'linkage': 'single',
     'n clusters': 3}, Score: (0.7144848175386341, 0.16920588504384718,
     24.292275620344395)
     Method: DBSCAN, PCA Components: 2, Parameters: {'eps': 2.0, 'min_samples': 3},
     Score: (0.7214246011996667, 0.6788983701172336, 76.81002251018943)
     Method: GMM, PCA Components: 2, Parameters: {'covariance_type': 'tied',
     'n components': 4, 'random state': 42}, Score: (0.4375197819795308,
     0.7369848464880445, 3188.3251681475836)
     Method: Spectral, PCA Components: 2, Parameters: {'affinity':
     'nearest_neighbors', 'n_clusters': 7, 'random_state': 42}, Score:
     (0.2025093367159689, 0.9011720995719121, 1702.3151644956567)
[17]: import matplotlib.pyplot as plt
     # Data to plot
     methods = list(best_hyperparameters.keys())
     pca components = [best hyperparameters[method]['pca'] for method in methods]
     scores = [best_hyperparameters[method]['score'][0] for method in methods] #__
       ⇒Using the first score for plotting
```

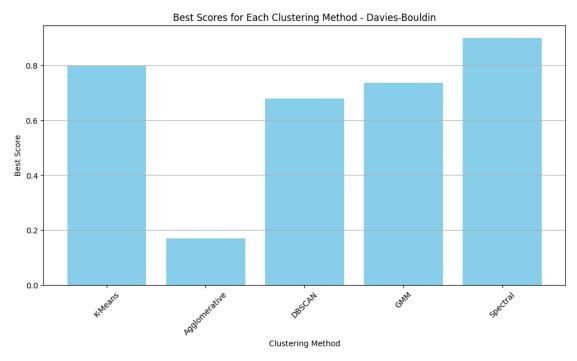
```
# Create a bar plot
plt.figure(figsize=(12, 6))
plt.bar(methods, scores, color='skyblue')
plt.title('Best Scores for Each Clustering Method - Silhouette')
plt.xlabel('Clustering Method')
plt.ylabel('Best Score')
plt.xticks(rotation=45)
plt.grid(axis='y')

# Show the plot
plt.show()
```

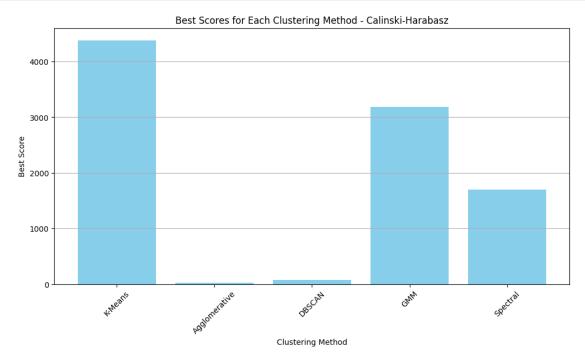


```
plt.xlabel('Clustering Method')
plt.ylabel('Best Score')
plt.xticks(rotation=45)
plt.grid(axis='y')

# Show the plot
plt.show()
```



```
# Show the plot plt.show()
```



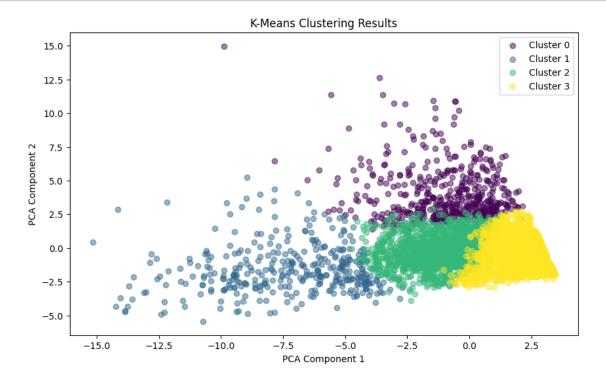
```
[20]: # Best clustering method: K-Means
      best_method = "K-Means"
      best_pca_components = 2
      best_parameters = {'init': 'k-means++', 'max_iter': 100, 'n_clusters': 4, |
      best_score = (0.43696184323503295, 0.800807992091704, 4382.46484845105)
      # Apply K-Means with the best parameters
      kmeans = KMeans(init=best_parameters['init'],__
       →max_iter=best_parameters['max_iter'],
                      n_clusters=best_parameters['n_clusters'],_
      on_init=best_parameters['n_init'], random_state=42)
      labels = kmeans.fit_predict(reduced_data)
      # Calculate intra-cluster variances
      intra_variances = []
      n_clusters = best_parameters['n_clusters'] # Use the number of clusters from_
       \hookrightarrow parameters
      for k in range(n_clusters):
          cluster_data = reduced_data[labels == k]
```

Variance Intra-Groupes: 2.1261409538972154 Variance Inter-Groupes: 3.32245739214457 Ratio: 1.562670332865047

```
[21]: import matplotlib.pyplot as plt
      # Plot the reduced data with cluster labels
      plt.figure(figsize=(10, 6))
      unique_labels = set(labels)
      colors = plt.cm.viridis(np.linspace(0, 1, len(unique_labels))) # Updated\ to_{\square}
       ⇔use viridis directly
      for i, label in enumerate(unique_labels):
          if label == -1: # Noise points
              color = 'k' # Black for noise
          else:
              color = colors[i] # Use the index to get the color
          plt.scatter(reduced_data[labels == label, 0], reduced_data[labels == label,__
       ⊶1],
                      color=color, label=f'Cluster {label}' if label != -1 else_

¬'Noise', alpha=0.5)
      plt.title('K-Means Clustering Results')
      plt.xlabel('PCA Component 1')
      plt.ylabel('PCA Component 2')
      plt.legend()
```

plt.show()



4 Descriptif

```
# Convert the summary to a DataFrame for better visualization
importance_df = pd.DataFrame(importance_summary).T
# Create a summary of the most important features for each cluster
importance_summary_str = {}
for label in unique_labels:
    if label != -1: # Skip noise points
        # Get the importance of features for the current cluster
        importance = importance df.loc[label]
        # Sort features by importance
        sorted_importance = importance.sort_values(ascending=False)
        # Create a summary string for the top features
        summary = ', '.join([f'{feature} : {value:.1f}' for feature, value in_
  ⇔sorted_importance.items()])
        importance_summary_str[label] = summary
# Display the importance summary for each cluster
print("Importance Summary for Each Cluster:")
for label in importance summary str.keys():
    # Get the top 5 features for the current cluster
    top_features = importance_df.loc[label].nlargest(5)
    # Check the most common value for these features in the cluster
    most_common_values = {feature: subset_data[subset_data['Cluster'] ==__
  →label][feature].mode()[0] for feature in top_features.index}
    print(f'Cluster {label}: {top features.index.tolist()} with most common,
  →values: {most common values}')
              A11
                                        A12
                                                                  A13
             mean
                        std count
                                       mean
                                                  std count
                                                                 mean
Cluster
0
        0.976395 0.151979
                              466 0.107296 0.309822
                                                        466 0.450644
1
        0.763587 0.425457
                              368 0.277174 0.448212
                                                        368 0.546196
2
        0.921525 0.268994 1784 0.117713 0.322358 1784 0.446749
        0.641898   0.479543   2382   0.320319   0.466697
                                                       2382 0.337112
                         A14 ... C1_6_slice C1_7_slice
              std count mean ...
                                     count
                                                            std count
                                                 mean
Cluster
                                             4.628755 0.643870
0
        0.498093
                    466
                         0.0
                                       466
                                                                  466
1
                         0.0 ...
                                                                  368
        0.498539
                    368
                                       368
                                             2.644022 0.916644
        0.497296
                         0.0 ...
                                      1784
                                             4.409753 0.636112 1784
                   1784
3
        0.472823
                   2382
                         0.0 ...
                                      2382
                                             4.915197 0.287548 2382
        C1_8_slice
                                   C1_9_slice
                                         mean
              mean
                         std count
                                                    std count
```

```
Cluster
0
          4.283262 0.698092
                               466
                                     4.317597 0.757750
                                                          466
1
          2.532609 0.831283
                               368
                                     2.532609 0.824702
                                                          368
2
          4.122197 0.628970 1784
                                     4.119395 0.691900 1784
3
          4.786314 0.426069
                              2382
                                     4.814861 0.415650 2382
[4 rows x 81 columns]
Importance Summary for Each Cluster:
Cluster 0: ['A8_1_slice', 'A8_2_slice', 'A8_3_slice', 'A8_4_slice',
'C1_6_slice'] with most common values: {'A8_1_slice': 20, 'A8_2_slice': 20,
'A8_3_slice': 10, 'A8_4_slice': 5, 'C1_6_slice': 5}
Cluster 1: ['A8_2 slice', 'A8_1 slice', 'A8_3 slice', 'C1_6 slice',
'C1_3_slice'] with most common values: {'A8_2_slice': 1, 'A8_1_slice': 1,
'A8_3_slice': 1, 'C1_6_slice': 2, 'C1_3_slice': 3}
Cluster 2: ['A8_2_slice', 'C1_6_slice', 'C1_4_slice', 'C1_3_slice',
'C1_7_slice'] with most common values: {'A8_2_slice': 2, 'C1_6_slice': 5,
'C1_4_slice': 5, 'C1_3_slice': 5, 'C1_7_slice': 5}
Cluster 3: ['C1_6_slice', 'C1_4_slice', 'C1_3_slice', 'C1_7_slice',
'C1_5_slice'] with most common values: {'C1_6_slice': 5, 'C1_4_slice': 5,
'C1_3_slice': 5, 'C1_7_slice': 5, 'C1_5_slice': 5}
/tmp/ipykernel_21221/2211454646.py:2: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead
See the caveats in the documentation: https://pandas.pydata.org/pandas-
docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
  subset_data['Cluster'] = labels # Convert labels to a list before adding to
```

Cluster 0:

the DataFrame

Features principales

- A8 1 slice (Temps passé à l'entretien au printemps) : 20 heures
- A8_2_slice (Temps passé à l'entretien en été) : 20 heures
- A8 3 slice (Temps passé à l'entretien en automne) : 10 heures
- A8 4 slice (Temps passé à l'entretien en hiver) : 5 heures
- C1 6 slice (Fréquence consultation : Profils Instagram sur le jardinage) : Jamais (5)

Description Ce cluster regroupe des individus qui :

- Consacrent un temps significatif à l'entretien de leurs espaces extérieurs, surtout au printemps et en été, mais beaucoup moins en automne et hiver.
- Ne s'intéressent pas à des contenus numériques liés au jardinage sur Instagram ou n'en consultent jamais.

Cluster 1:

Features principales

- A8_2_slice (Temps passé à l'entretien en été) : 1 heure
- A8 1 slice (Temps passé à l'entretien au printemps) : 1 heure
- A8_3_slice (Temps passé à l'entretien en automne) : 1 heure
- C1_6_slice (Fréquence consultation : Profils Instagram sur le jardinage) : Au moins une fois par semaine (2)
- C1_3_slice (Fréquence consultation : Blogs d'experts en aménagement) : *Une à trois fois* par mois (3)

Description Ce cluster regroupe des individus qui :

- Passent très peu de temps à l'entretien de leurs espaces extérieurs, quelle que soit la saison.
- Consultent régulièrement des profils Instagram dédiés au jardinage.
- Lisent des blogs d'experts en aménagement quelques fois par mois pour s'inspirer ou s'informer.

Cluster 2:

Features principales

- A8 2 slice (Temps passé à l'entretien en été) : 2 heures
- C1_6_slice (Fréquence consultation : Profils Instagram sur le jardinage) : Jamais (5)
- C1_4_slice (Fréquence consultation : Pages/groupes Facebook sur le jardinage) : Jamais (5)
- C1 3 slice (Fréquence consultation : Blogs d'experts en aménagement) : Jamais (5)
- C1_7_slice (Fréquence consultation : Forums de discussion sur le jardinage) : Jamais (5)

Description Ce cluster regroupe des individus qui :

- Passent très peu de temps à entretenir leurs espaces extérieurs, même en été.
- Ne consultent jamais de contenus numériques liés au jardinage ou à l'aménagement des espaces extérieurs (Instagram, Facebook, blogs, forums).
- Manifestent peu d'intérêt pour les échanges ou discussions autour de l'entretien des espaces extérieurs.

Cluster 3:

Features principales

- C1_6_slice (Fréquence consultation : Profils Instagram sur le jardinage) : Jamais (5)
- C1_4_slice (Fréquence consultation : Pages/groupes Facebook sur le jardinage) : Jamais (5)
- C1_3_slice (Fréquence consultation : Blogs d'experts en aménagement) : Jamais (5)
- C1_7_slice (Fréquence consultation : Forums de discussion sur le jardinage) : Jamais (5)
- C1_5_slice (Fréquence consultation : Pages Pinterest sur le jardinage) : Jamais (5)

Description Ce cluster regroupe des individus qui :

- Ne consultent jamais de contenus numériques en rapport avec le jardinage, que ce soit sur Instagram, Facebook, Pinterest, des blogs ou des forums.
- Peuvent être caractérisés par un désintérêt total pour les ressources numériques ou en ligne dédiées à l'aménagement ou à l'entretien des espaces extérieurs.

5 Conclusion

Dans le cadre de notre analyse de clustering, nous avons réalisé un benchmark en utilisant plusieurs métriques de performance : Silhouette, Davies-Bouldin et Calinski-Harabasz. Nous avons appliqué différentes méthodes de clustering, notamment KMeans, Agglomerative Clustering, DB-SCAN, Gaussian Mixture Model et Spectral Clustering.

Les résultats de cette évaluation ont montré que le Kmeans, encore une fois offrait les meilleures performances globales selon les métriques analysées (surtout pour 4 clusters). En conséquence, nous avons décidé de nous concentrer sur cette méthode.

En termes de variance, nous avons obtenu les résultats suivants : - Variance Intra-Groupes: 2.126 - Variance Inter-Groupes: 3.322 - Ratio: 1.562

Ces résultats soulignent l'efficacité du KMeans dans notre analyse.