# Sec 1 Homework 9

March 7, 2024

## 1 0.) Import and Clean data

```
[]: import pandas as pd
     import matplotlib.pyplot as plt
     import numpy as np
     from sklearn.preprocessing import StandardScaler
     from sklearn.cluster import KMeans
[]: #drive.mount('/content/qdrive/', force_remount = True)
     df = pd.read_csv("Country-data.csv", sep = ",")
[]:
                       country
                                 child_mort
                                              exports
                                                       health
                                                                imports
                                                                          income
                                                          7.58
                                                                    44.9
     0
                   Afghanistan
                                        90.2
                                                 10.0
                                                                            1610
     1
                       Albania
                                        16.6
                                                 28.0
                                                          6.55
                                                                    48.6
                                                                            9930
     2
                       Algeria
                                        27.3
                                                 38.4
                                                          4.17
                                                                    31.4
                                                                           12900
     3
                                      119.0
                                                 62.3
                                                          2.85
                                                                    42.9
                                                                            5900
                        Angola
     4
                                                          6.03
          Antigua and Barbuda
                                        10.3
                                                 45.5
                                                                   58.9
                                                                           19100
     162
                                        29.2
                                                 46.6
                                                          5.25
                                                                    52.7
                       Vanuatu
                                                                            2950
     163
                     Venezuela
                                        17.1
                                                 28.5
                                                          4.91
                                                                    17.6
                                                                           16500
     164
                       Vietnam
                                        23.3
                                                 72.0
                                                          6.84
                                                                    80.2
                                                                            4490
     165
                         Yemen
                                        56.3
                                                 30.0
                                                          5.18
                                                                    34.4
                                                                            4480
     166
                        Zambia
                                        83.1
                                                 37.0
                                                          5.89
                                                                    30.9
                                                                            3280
          inflation
                     life_expec
                                   total_fer
                                                gdpp
                9.44
     0
                             56.2
                                        5.82
                                                 553
     1
                4.49
                             76.3
                                         1.65
                                                4090
     2
               16.10
                             76.5
                                         2.89
                                                4460
     3
               22.40
                             60.1
                                                3530
                                         6.16
     4
                1.44
                             76.8
                                         2.13
                                               12200
     162
                2.62
                             63.0
                                        3.50
                                                2970
     163
               45.90
                             75.4
                                        2.47
                                               13500
     164
               12.10
                             73.1
                                         1.95
                                                1310
     165
               23.60
                             67.5
                                        4.67
                                                1310
     166
               14.00
                                        5.40
                                                1460
                             52.0
```

## 2 1.) Fit a kmeans Model with any Number of Clusters

```
[]: names = df[['country']].copy()
   X = df.drop('country', axis = 1)

[]: scaler = StandardScaler().fit(X)
   X_scaled = scaler.transform(X)

[]: kmeans = KMeans(n_clusters=5).fit(X_scaled)

   c:\Users\nikpa\anacondafinal\Lib\site-packages\sklearn\cluster\_kmeans.py:1412:
   FutureWarning: The default value of `n_init` will change from 10 to 'auto' in
   1.4. Set the value of `n_init` explicitly to suppress the warning
        super()._check_params_vs_input(X, default_n_init=10)
   c:\Users\nikpa\anacondafinal\Lib\site-packages\sklearn\cluster\_kmeans.py:1436:
   UserWarning: KMeans is known to have a memory leak on Windows with MKL, when
   there are less chunks than available threads. You can avoid it by setting the
   environment variable OMP_NUM_THREADS=1.
        warnings.warn(
```

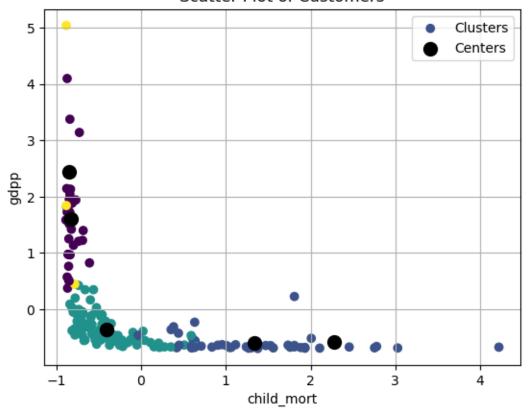
## 3 2.) Pick two features to visualize across

```
plt.ylabel(X.columns[x2_index])
plt.title('Scatter Plot of Customers')

# Generate legend
plt.legend()

plt.grid()
plt.show()
```

#### Scatter Plot of Customers



4 3.) Check a range of k-clusters and visualize to find the elbow. Test 30 different random starting places for the centroid means

```
[]: WCSSs = []
Ks = range(1,15)
for k in Ks:
    kmeans = KMeans(n_clusters=k, n_init=30).fit(X_scaled)
    WCSSs.append(kmeans.inertia_)
```

c:\Users\nikpa\anacondafinal\Lib\site-packages\sklearn\cluster\\_kmeans.py:1436:

UserWarning: KMeans is known to have a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP\_NUM\_THREADS=1.

warnings.warn(

c:\Users\nikpa\anacondafinal\Lib\site-packages\sklearn\cluster\\_kmeans.py:1436: UserWarning: KMeans is known to have a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP\_NUM\_THREADS=1.

warnings.warn(

c:\Users\nikpa\anacondafinal\Lib\site-packages\sklearn\cluster\\_kmeans.py:1436: UserWarning: KMeans is known to have a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP\_NUM\_THREADS=1.

warnings.warn(

c:\Users\nikpa\anacondafinal\Lib\site-packages\sklearn\cluster\\_kmeans.py:1436: UserWarning: KMeans is known to have a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP\_NUM\_THREADS=1.

warnings.warn(

c:\Users\nikpa\anacondafinal\Lib\site-packages\sklearn\cluster\\_kmeans.py:1436: UserWarning: KMeans is known to have a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP\_NUM\_THREADS=1.

warnings.warn(

c:\Users\nikpa\anacondafinal\Lib\site-packages\sklearn\cluster\\_kmeans.py:1436: UserWarning: KMeans is known to have a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP\_NUM\_THREADS=1.

warnings.warn(

c:\Users\nikpa\anacondafinal\Lib\site-packages\sklearn\cluster\\_kmeans.py:1436: UserWarning: KMeans is known to have a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP\_NUM\_THREADS=1.

warnings.warn(

c:\Users\nikpa\anacondafinal\Lib\site-packages\sklearn\cluster\\_kmeans.py:1436: UserWarning: KMeans is known to have a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP\_NUM\_THREADS=1.

warnings.warn(

c:\Users\nikpa\anacondafinal\Lib\site-packages\sklearn\cluster\\_kmeans.py:1436: UserWarning: KMeans is known to have a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP\_NUM\_THREADS=1.

warnings.warn(

c:\Users\nikpa\anacondafinal\Lib\site-packages\sklearn\cluster\\_kmeans.py:1436: UserWarning: KMeans is known to have a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP\_NUM\_THREADS=1.

```
warnings.warn(
```

c:\Users\nikpa\anacondafinal\Lib\site-packages\sklearn\cluster\\_kmeans.py:1436: UserWarning: KMeans is known to have a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP\_NUM\_THREADS=1.

warnings.warn(

c:\Users\nikpa\anacondafinal\Lib\site-packages\sklearn\cluster\\_kmeans.py:1436: UserWarning: KMeans is known to have a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP\_NUM\_THREADS=1.

warnings.warn(

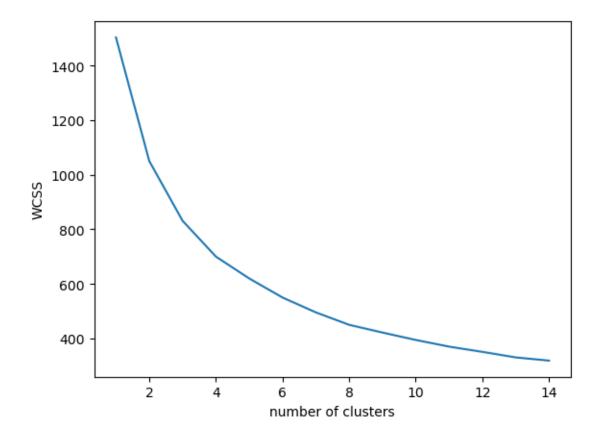
c:\Users\nikpa\anacondafinal\Lib\site-packages\sklearn\cluster\\_kmeans.py:1436: UserWarning: KMeans is known to have a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP\_NUM\_THREADS=1.

warnings.warn(

c:\Users\nikpa\anacondafinal\Lib\site-packages\sklearn\cluster\\_kmeans.py:1436: UserWarning: KMeans is known to have a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP\_NUM\_THREADS=1.

warnings.warn(

```
[]: plt.plot(Ks, WCSSs)
  plt.xlabel('number of clusters')
  plt.ylabel('WCSS')
  plt.show()
```



For interpretability we would use 2 clusters, developed and underdeveloped economies

# 5 4.) Use the above work and economic critical thinking to choose a number of clusters. Explain why you chose the number of clusters and fit a model accordingly.

```
[]: kmeams = KMeans(n_clusters=2, n_init=30).fit(X_scaled)

    c:\Users\nikpa\anacondafinal\Lib\site-packages\sklearn\cluster\_kmeans.py:1436:
    UserWarning: KMeans is known to have a memory leak on Windows with MKL, when
    there are less chunks than available threads. You can avoid it by setting the
    environment variable OMP_NUM_THREADS=1.
        warnings.warn(
[]: preds = pd.DataFrame(kmeams.labels_)

[]: output = pd.concat([preds, df], axis = 1)
```

# 6 6.) Do the same for a silhoutte plot

```
[]: from sklearn.metrics import silhouette_score
```

```
[]: SSs = []
   Ks = range(2,15)
   for k in Ks:
        kmeans = KMeans(n_clusters=k, n_init=30).fit(X_scaled)
        sil = silhouette_score(X_scaled, kmeans.labels_)
        SSs.append(sil)
```

c:\Users\nikpa\anacondafinal\Lib\site-packages\sklearn\cluster\\_kmeans.py:1436: UserWarning: KMeans is known to have a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP\_NUM\_THREADS=1.

warnings.warn(

c:\Users\nikpa\anacondafinal\Lib\site-packages\sklearn\cluster\\_kmeans.py:1436: UserWarning: KMeans is known to have a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP\_NUM\_THREADS=1.

warnings.warn(

c:\Users\nikpa\anacondafinal\Lib\site-packages\sklearn\cluster\\_kmeans.py:1436: UserWarning: KMeans is known to have a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP\_NUM\_THREADS=1.

warnings.warn(

c:\Users\nikpa\anacondafinal\Lib\site-packages\sklearn\cluster\\_kmeans.py:1436: UserWarning: KMeans is known to have a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP\_NUM\_THREADS=1.

warnings.warn(

c:\Users\nikpa\anacondafinal\Lib\site-packages\sklearn\cluster\\_kmeans.py:1436: UserWarning: KMeans is known to have a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP\_NUM\_THREADS=1.

warnings.warn(

c:\Users\nikpa\anacondafinal\Lib\site-packages\sklearn\cluster\\_kmeans.py:1436: UserWarning: KMeans is known to have a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP\_NUM\_THREADS=1.

warnings.warn(

c:\Users\nikpa\anacondafinal\Lib\site-packages\sklearn\cluster\\_kmeans.py:1436: UserWarning: KMeans is known to have a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP\_NUM\_THREADS=1.

warnings.warn(

c:\Users\nikpa\anacondafinal\Lib\site-packages\sklearn\cluster\\_kmeans.py:1436: UserWarning: KMeans is known to have a memory leak on Windows with MKL, when

there are less chunks than available threads. You can avoid it by setting the environment variable OMP\_NUM\_THREADS=1.

warnings.warn(

c:\Users\nikpa\anacondafinal\Lib\site-packages\sklearn\cluster\\_kmeans.py:1436: UserWarning: KMeans is known to have a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP NUM THREADS=1.

warnings.warn(

c:\Users\nikpa\anacondafinal\Lib\site-packages\sklearn\cluster\\_kmeans.py:1436: UserWarning: KMeans is known to have a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP\_NUM\_THREADS=1.

warnings.warn(

c:\Users\nikpa\anacondafinal\Lib\site-packages\sklearn\cluster\\_kmeans.py:1436: UserWarning: KMeans is known to have a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP\_NUM\_THREADS=1.

warnings.warn(

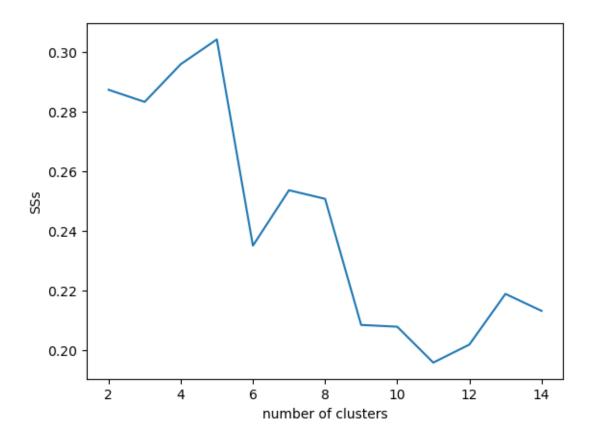
c:\Users\nikpa\anacondafinal\Lib\site-packages\sklearn\cluster\\_kmeans.py:1436: UserWarning: KMeans is known to have a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP\_NUM\_THREADS=1.

warnings.warn(

c:\Users\nikpa\anacondafinal\Lib\site-packages\sklearn\cluster\\_kmeans.py:1436: UserWarning: KMeans is known to have a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid it by setting the environment variable OMP\_NUM\_THREADS=1.

warnings.warn(

```
[]: plt.plot(Ks, SSs)
  plt.xlabel('number of clusters')
  plt.ylabel('SSs')
  plt.show()
```



7 7.) Create a list of the countries that are in each cluster. Write interesting things you notice.

```
[]: print("Cluster 1: ")
    list(output.loc[output[0] == 0, "country"])

Cluster 1:

[]: ['Albania',
    'Algeria',
    'Antigua and Barbuda',
    'Argentina',
    'Armenia',
    'Australia',
    'Austria',
    'Azerbaijan',
    'Bahamas',
    'Bahrain',
    'Barbados',
    'Belarus',
```

```
'Belgium',
'Belize',
'Bhutan',
'Bosnia and Herzegovina',
'Brazil',
'Brunei',
'Bulgaria',
'Canada',
'Cape Verde',
'Chile',
'China',
'Colombia',
'Costa Rica',
'Croatia',
'Cyprus',
'Czech Republic',
'Denmark',
'Dominican Republic',
'Ecuador',
'El Salvador',
'Estonia',
'Fiji',
'Finland',
'France',
'Georgia',
'Germany',
'Greece',
'Grenada',
'Hungary',
'Iceland',
'Iran',
'Ireland',
'Israel',
'Italy',
'Jamaica',
'Japan',
'Jordan',
'Kazakhstan',
'Kuwait',
'Latvia',
'Lebanon',
'Libya',
'Lithuania',
'Luxembourg',
'Macedonia, FYR',
'Malaysia',
'Maldives',
```

```
'Malta',
      'Mauritius',
      'Moldova',
      'Montenegro',
      'Morocco',
      'Netherlands',
      'New Zealand',
      'Norway',
      'Oman',
      'Panama',
      'Paraguay',
      'Peru',
      'Poland',
      'Portugal',
      'Qatar',
      'Romania',
      'Russia',
      'Saudi Arabia',
      'Serbia',
      'Seychelles',
      'Singapore',
      'Slovak Republic',
      'Slovenia',
      'South Korea',
      'Spain',
      'Sri Lanka',
      'St. Vincent and the Grenadines',
      'Suriname',
      'Sweden',
      'Switzerland',
      'Thailand',
      'Tunisia',
      'Turkey',
      'Ukraine',
      'United Arab Emirates',
      'United Kingdom',
      'United States',
      'Uruguay',
      'Venezuela',
      'Vietnam']
[]: print("Cluster 2: ")
     list(output.loc[output[0] == 1, "country"])
    Cluster 2:
```

```
[]: ['Afghanistan',
      'Angola',
      'Bangladesh',
      'Benin',
      'Bolivia',
      'Botswana',
      'Burkina Faso',
      'Burundi',
      'Cambodia',
      'Cameroon',
      'Central African Republic',
      'Chad',
      'Comoros',
      'Congo, Dem. Rep.',
      'Congo, Rep.',
      "Cote d'Ivoire",
      'Egypt',
      'Equatorial Guinea',
      'Eritrea',
      'Gabon',
      'Gambia',
      'Ghana',
      'Guatemala',
      'Guinea',
      'Guinea-Bissau',
      'Guyana',
      'Haiti',
      'India',
      'Indonesia',
      'Iraq',
      'Kenya',
      'Kiribati',
      'Kyrgyz Republic',
      'Lao',
      'Lesotho',
      'Liberia',
      'Madagascar',
      'Malawi',
      'Mali',
      'Mauritania',
      'Micronesia, Fed. Sts.',
      'Mongolia',
      'Mozambique',
      'Myanmar',
      'Namibia',
      'Nepal',
      'Niger',
```

```
'Nigeria',
'Pakistan',
'Philippines',
'Rwanda',
'Samoa',
'Senegal',
'Sierra Leone',
'Solomon Islands',
'South Africa',
'Sudan',
'Tajikistan',
'Tanzania',
'Timor-Leste',
'Togo',
'Tonga',
'Turkmenistan',
'Uganda',
'Uzbekistan',
'Vanuatu',
'Yemen',
'Zambia']
```

#8.) Create a table of Descriptive Statistics. Rows being the Cluster number and columns being all the features. Values being the mean of the centroid. Use the nonscaled X values for interprotation

```
[]: new_df = output.drop('country', axis = 1)
[]: new_df.groupby(0).mean()
[]:
        child mort
                       exports
                                  health
                                             imports
                                                             income
                                                                     inflation \
     0
     0
         12.161616
                    48.603030
                                7.314040
                                          49.121212
                                                      26017.171717
                                                                      5.503545
     1
         76.280882
                    30.198515
                                6.090147
                                          43.642146
                                                       4227.397059
                                                                     11.098750
        life_expec
                    total_fer
                                        gdpp
     0
     0
         76.493939
                                20507.979798
                     1.941111
         61.910294
                                 1981.235294
     1
                     4.413824
```

# 8 9.) Write an observation about the descriptive statistics.

On of the most apparent observation we can make is about the child mortality rRate. The child mortality rate appears to vary significantly between the two groups. The developed countries group indicates a significantly lower child mortality rate compared to the least developed countries, suggesting huge differences in healthcare and socio-economic conditions between the countries represented. Moreover, least developed countries second appear to have lower exports in relation to imports, which suggests trade deficits or a less export-oriented economy.