Machine Learning Lab: 6 KMeans Clustering

January 2025

Perform Date: January 13-18, 2025

1 Objective

1.1 Implement K Means Clustering Algorithm on the given dataset

2 Description

K-Means Clustering is an unsupervised learning algorithm that is used to solve clustering problems in machine learning or data science. A cluster refers to a collection of data points aggregated together because of certain similarities. We will define a target number k, which refers to the number of centroids you need in the dataset. A centroid is an imaginary or real location representing the center of the cluster. Every data point is allocated to each of the clusters through reducing the in-cluster sum of squares. In other words, the K-means algorithm identifies the k number of centroids, and then allocates every data point to the nearest cluster, while keeping the centroids as small as possible. The 'means' in the K-means refers to averaging of the data; that is, finding the centroid.

3 Implementation Guidelines

3.1 Part A

```
[35]: # Imports
    from sklearn.datasets import make_blobs
    X, _ = make_blobs(n_samples=100, centers=3, n_features=2,
        cluster_std=0.2, random_state=0)

[1]: # Scatter plot of the data points
    import matplotlib.pyplot as plt
    %matplotlib inline

[2]: # Using scikit-learn to perform K-Means clustering
    from sklearn.cluster import KMeans
    # Specify the number of clusters (3) and fit the data X
```

3.2 Part B

Hand Written Digit Recognition

```
[44]: import numpy as np # linear algebra
      import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)
      from sklearn.cluster import KMeans
      from sklearn.datasets import load_digits
      #digits dataset from scikit learn consists of 8x8 pixel images of digits
      #Data plotting and visualization libraries
      import matplotlib.pyplot as plt
      import seaborn as sns
      from scipy.stats import mode
      from sklearn.metrics import accuracy_score, confusion_matrix
[45]: digits = load_digits() #load the dataset in digits
[45]: (1797, 64)
[29]: digits.keys() #Dataset loaded is a dictionary
      # data : flattened arrays/tensors used for clustering
      # target : label associated with flattened array
      #print(digits.target)
[29]: dict_keys(['data', 'target', 'frame', 'feature_names', 'target_names', 'images',
      'DESCR'])
[30]: digits.data[0:3] #flattened data for 3 images of the dataset
```

```
[30]: array([[ 0., 0., 5., 13., 9., 1., 0., 0., 0., 0., 13., 15., 10.,
             15., 5., 0., 0., 3., 15., 2., 0., 11., 8., 0., 0., 4.,
             12., 0., 0., 8., 8., 0., 0., 5., 8., 0., 0., 9., 8.,
             0., 0., 4., 11., 0., 1., 12., 7., 0., 0., 2., 14., 5.,
             10., 12., 0., 0., 0., 6., 13., 10., 0., 0., 0.],
            [ 0., 0., 0., 12., 13., 5., 0., 0., 0., 0., 0., 11., 16.,
              9., 0., 0., 0., 3., 15., 16., 6., 0., 0., 7.,
             15., 16., 16., 2., 0., 0., 0., 1., 16., 16., 3., 0.,
             0., 0., 0., 1., 16., 16., 6., 0., 0., 0., 0., 1., 16.,
             16., 6., 0., 0., 0., 0., 11., 16., 10., 0., 0.],
            [0., 0., 0., 4., 15., 12., 0., 0., 0., 0., 3., 16., 15.,
             14., 0., 0., 0., 0., 8., 13., 8., 16., 0., 0., 0., 0.,
             1., 6., 15., 11., 0., 0., 1., 8., 13., 15., 1., 0.,
             0., 0., 9., 16., 16., 5., 0., 0., 0., 0., 3., 13., 16.,
             16., 11., 5., 0., 0., 0., 3., 11., 16., 9., 0.]])
      #run KMeans clustering on digits.data for 1797 records and 64 features
[31]:
[31]: (10, 64)
[32]: fig, ax = plt.subplots(2, 5, figsize = (8,3)) #Create a figure and a set of \Box
      →subplots( 2 rows and 5 columns)
     centers = k_means.cluster_centers_.reshape(10,8,8)
      #flattened image can't be viewed, re-transform/reshape/inverse transform it to_{f \sqcup}
      →original form to view matrix shaped image
     #reshape 10 rows of clusters (k_means.cluster_centers_ = 10,64) and 64 to 8*8_{\square}
      \rightarrow matrix
     for axi, center in zip(ax.flat, centers): #ax.flat:flattening the image &
      →plotting relevant centers
         axi.set(xticks = [], yticks = [])
         axi.imshow(center, interpolation='nearest',cmap = plt.cm.binary)
      \rightarrow#imshow(matplotlib method) to render the image in notebook
```

```
[33]: labels = np.zeros_like(clusters) # blank labels
     print(f"The labels are : {labels}")
     print(f"\nThe size of labels is : {labels.shape}")
     print("The mask values are : ")
     for i in range(10):
         mask = (clusters == i)
         \textit{\#if a specific digit belongs to/equivalent a specific cluster then its \textit{True}_{\sqcup}
       →else False
         print(mask)
         labels[mask] = mode(digits.target[mask])[0]
     The labels are : [0 0 0 ... 0 0 0]
     The size of labels is: (1797,)
     The mask values are :
     [False False False False False]
     [False False False False False]
     [False False False False False]
     [False False False ... False True True]
     [False False False False False]
     [ True False False ... False False False]
     [False False False False False]
     [False True True ... True False False]
     [False False False False False False]
     [False False False False False]
[10]: #if a specific digit belongs to/equivalent a specific cluster then acuracy is 1
       ⇔else 0
```

4 Exercise

- 1. What is the accuracy, precision, and recall of the model trained on Hand Written Digit Recognition dataset?
- 2. Describe the dataset used in this lab exercise.
- 3. Give insights into the model trained for Hand Written Digit Recognition dataset.
- 4. Use the k-means algorithm in python to cluster the following 8 examples into 3 clusters: A1=(2,10), A2=(2,5), A3=(8,4), A4=(5,8), A5=(7,5), A6=(6,4), A7=(1,2), A8=(4,9).
 - (a) Suppose that the centers of each cluster are A1, A4 and A7. Run the k-means algorithm for 3 epochs only. At the end of this epoch show:
 - i. The new clusters (i.e. the examples belonging to each cluster)(mention the appropriate attribute used to identify the clusters in sklearn)
 - ii. The centers of the new clusters (mention the appropriate attribute used to identify the cluster centers in sklearn)

- 5. Apply Elbow Method on Part A and Part B. How many clusters will you choose according to elbow method?
- 6. Write the silhouette score for model trained in Part A and Part B using sklearn.

5 Reference

- 1. https://scikit-learn.org/stable/modules/generated/sklearn.cluster.KMeans.html
- 2. https://scikit-learn.org/stable/modules/generated/sklearn.cluster.DBSCAN.html
- 3. https://scikit-learn.org/stable/modules/generated/sklearn.cluster. AgglomerativeClustering.html