

K MEANS Clustering

- Unsupervised Learning Algorithm
- Attempts to group similar clusters in your data
- A typical Clustering Problem will
- Cluster similar documents
- Cluster customers based on features/behaviour
- Market segmentation
- Product segmentation
- Identify similar physical groups

When to Go for K-Means Clustering

- When your data is not labelled (no headers)
- When you know what your target classes are going to be. For example if have some data, without labels but you know you want only 2 or specific classes as outcome

The overall goal is to divide data into distinct groups such that observations within each group are similar

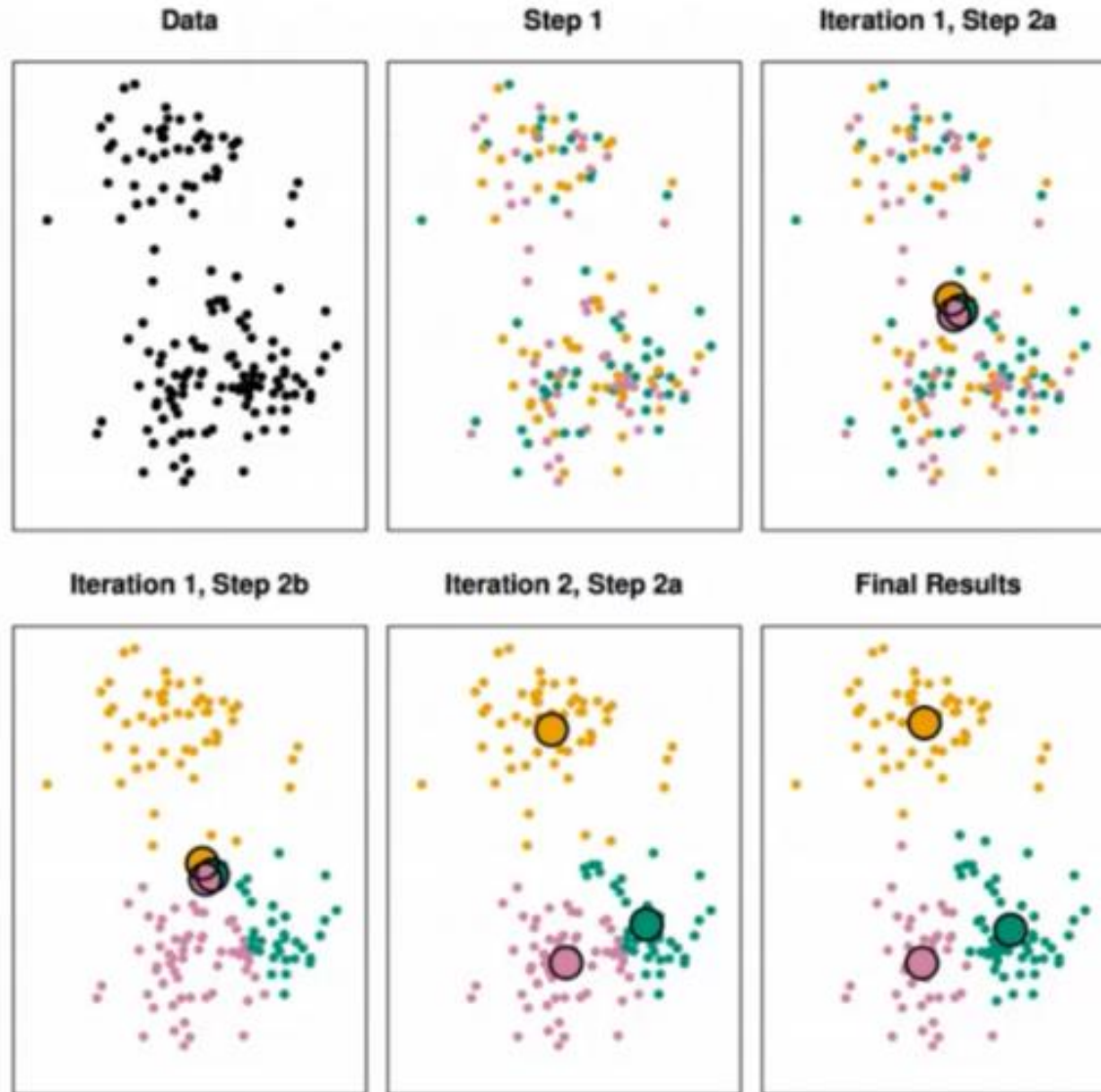
Unlabelled training data
K-means clusters the data into 5 different cluster



K-Means – How Does It Work

- Choose a number of Clusters “K”
- Randomly assign each data point to a Cluster
- Until the clusters stop changing, do the below
 - For each cluster, compute the cluster centroid (Center of cluster) by taking the mean vector of points in the cluster
 - Assign each data point to the cluster for which the centroid is the closest

K-Means Process



☐ Data

All observations are plotted. No groups yet

☐ Step1

Each observation is randomly assigned to a cluster. Observations are shown by different Colors

☐ Iteration1-Step2a

Cluster centroids are shown (large discs orange). Initially these centroids overlap as the initial cluster assignments are chosen at random

☐ Iteration1-Step2b

Each data point is assigned to nearest centroid.

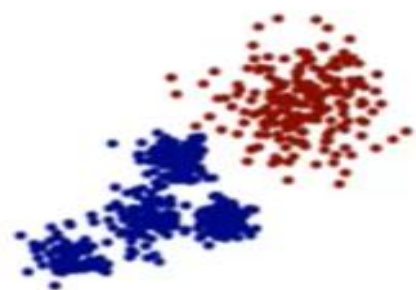
☐ Iteration 2-Step2a

Iterate until no new clusters are assigned. See clusters are already forming

☐ Final results

after 10 or more iterations

Choosing K Value



❑ **Choosing the best K Value** - Elbow Method can be used to choose the value of K (clusters)

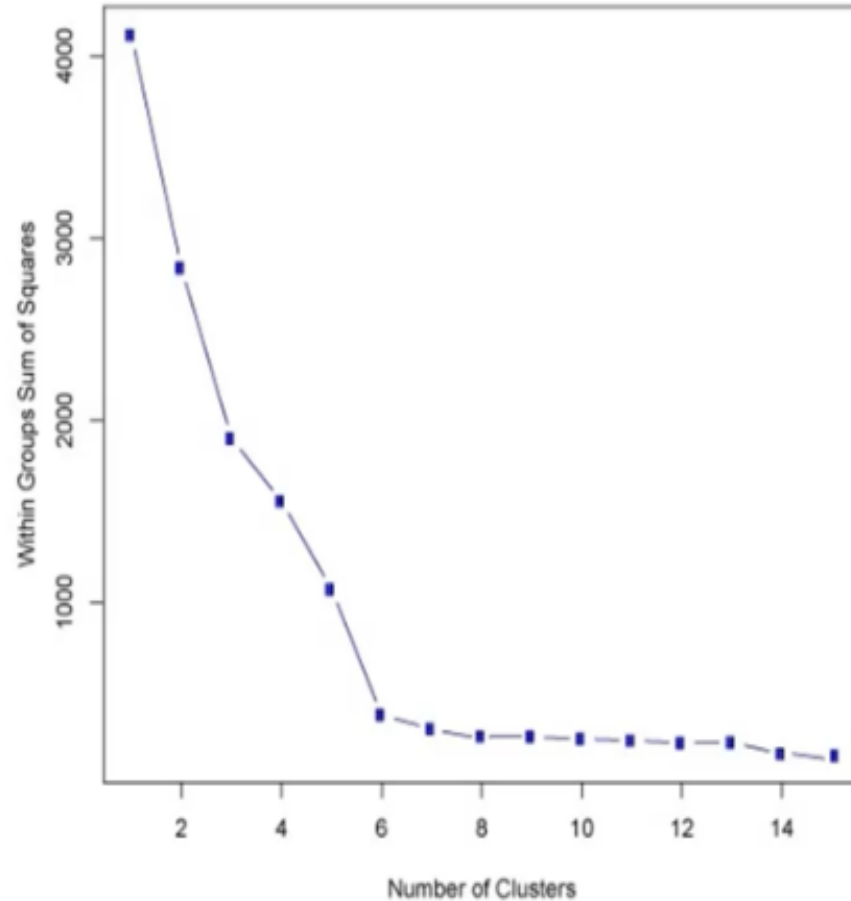
- Compute the Sum of Squared Errors(SSE) for some values of K
- $SSE = \text{Sum of [Square of (Distance between each data point in a cluster and its centroid)]}$

❑ **Plot K against SSE** - Note as error decreases K gets larger.

- When Number of clusters increases the size gets smaller, so the error (distortion) will also be smaller

❑ **Elbow Method**

- ❑ Helps in choosing a value of K where the error decreases suddenly
- ❑ A “elbow effect” (as shown in graph) will be produced



- X Axis indicates the K value or number of clusters
- Y Axis – Within group Sum of Squares (SSE)
- We need a K where the SSE does not decrease any more as K increases
- That point is at a K value of 6 Where we are able to see the error does not significantly decrease anymore.
- That would be the ideal K value for our clustering process
- Always remember that we will not be able to find out the perfect k value through this process.
- But this along with your domain knowledge can prove vital in the clustering