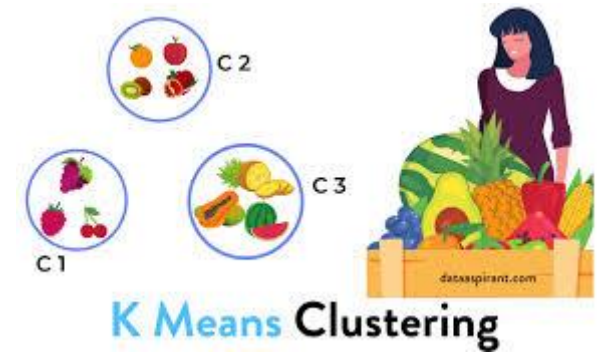


# Data Science




## Lecture 09: *K-means Clustering*

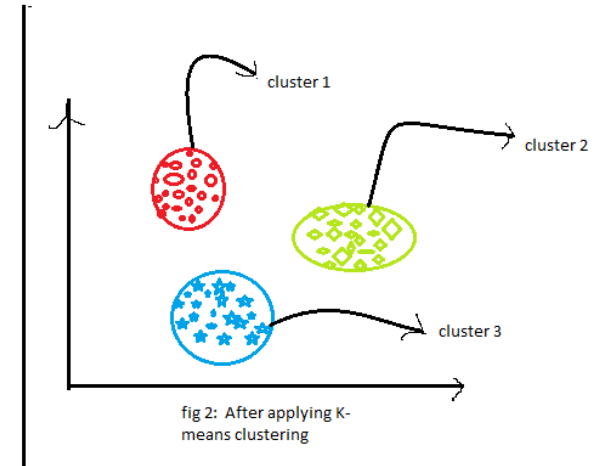
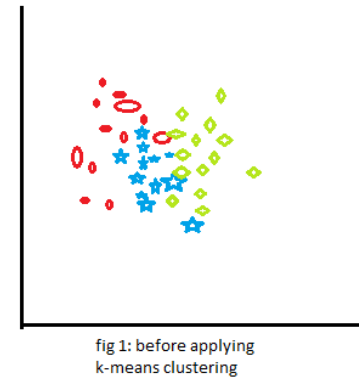
Dr.Fatema Nafa

Fall 2022

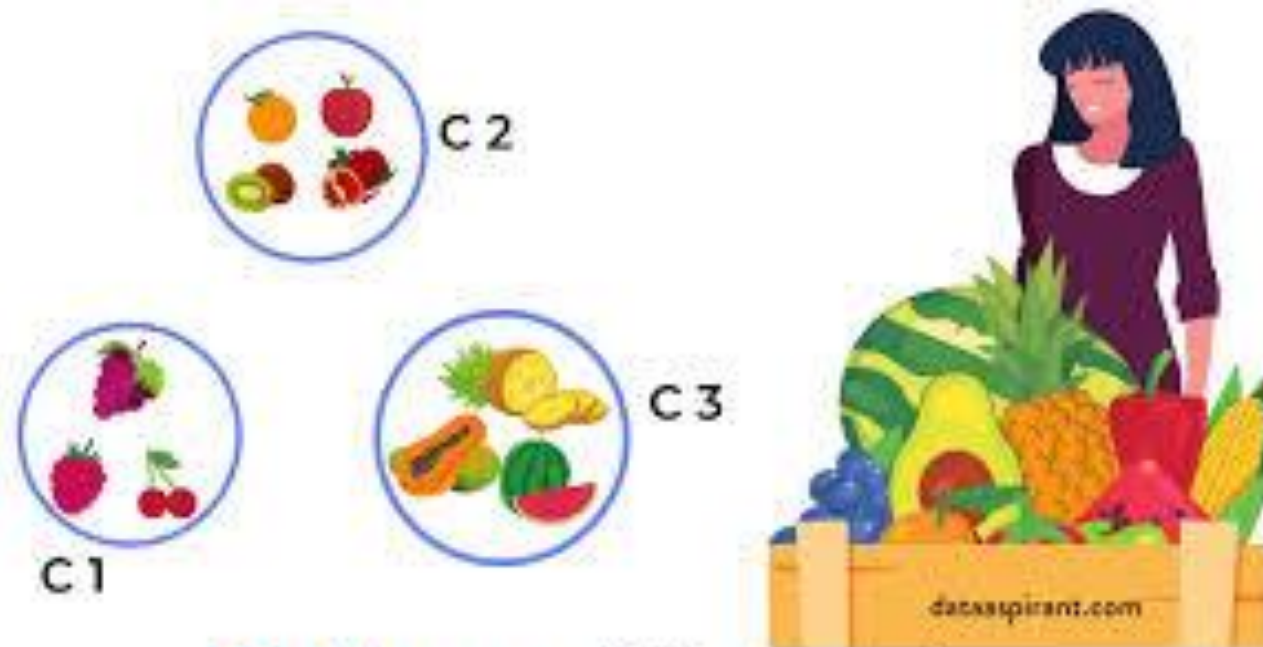


# Learning Objectives

- 
- **Motivating Example**
  - What is clustering?
  - Why would we want to cluster?
  - How would you determine clusters?
  - How can you do this efficiently?







## K Means Clustering

# Machine Learning Problems

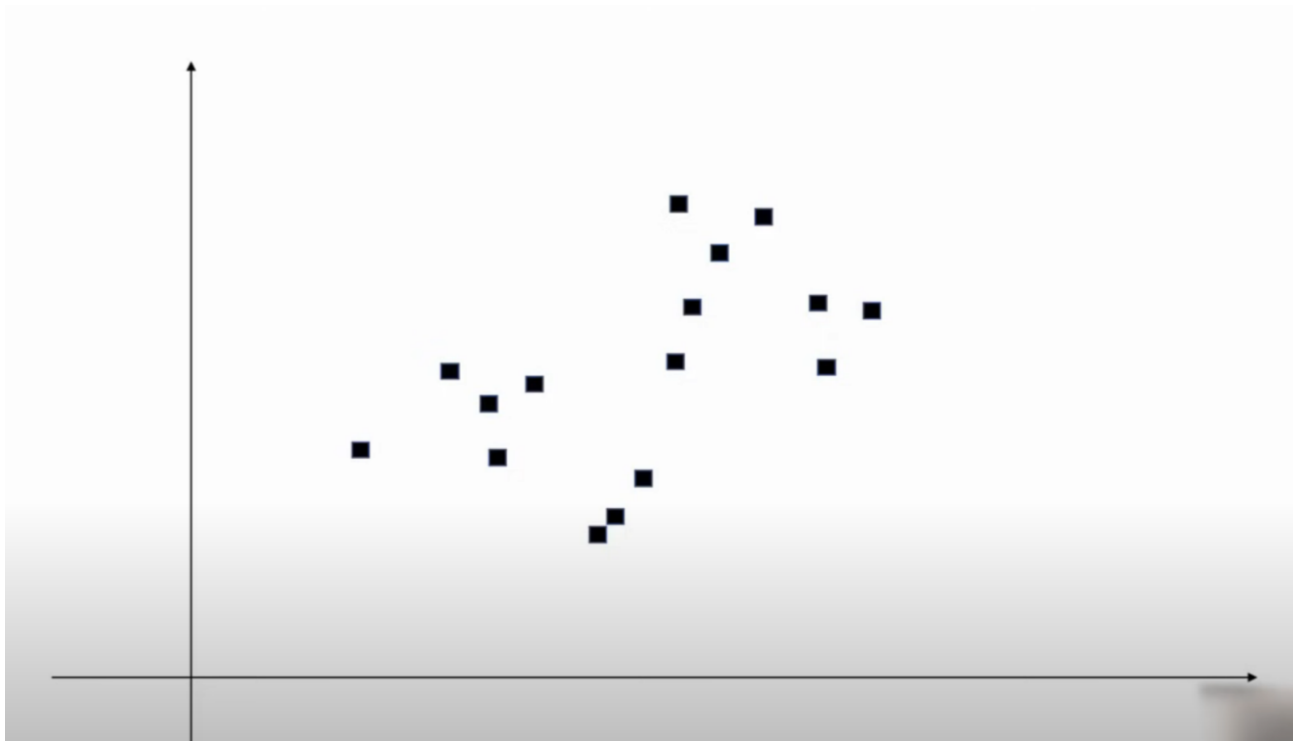
	<i>Supervised Learning</i>	<i>Unsupervised Learning</i>
<i>Discrete</i>	classification or categorization	clustering
<i>Continuous</i>	regression	dimensionality reduction

# Lecture Map

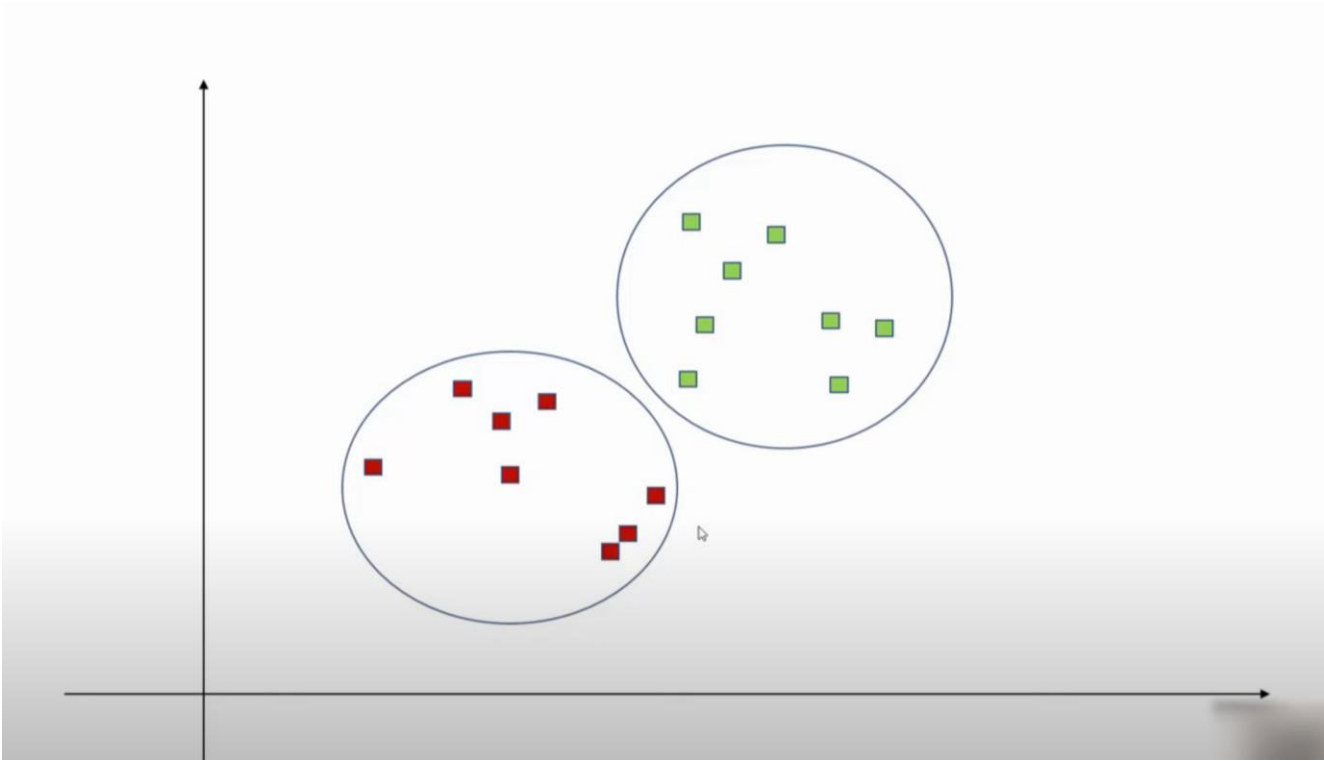
- Theory
- Coding
- Exercise

# K-means Clustering

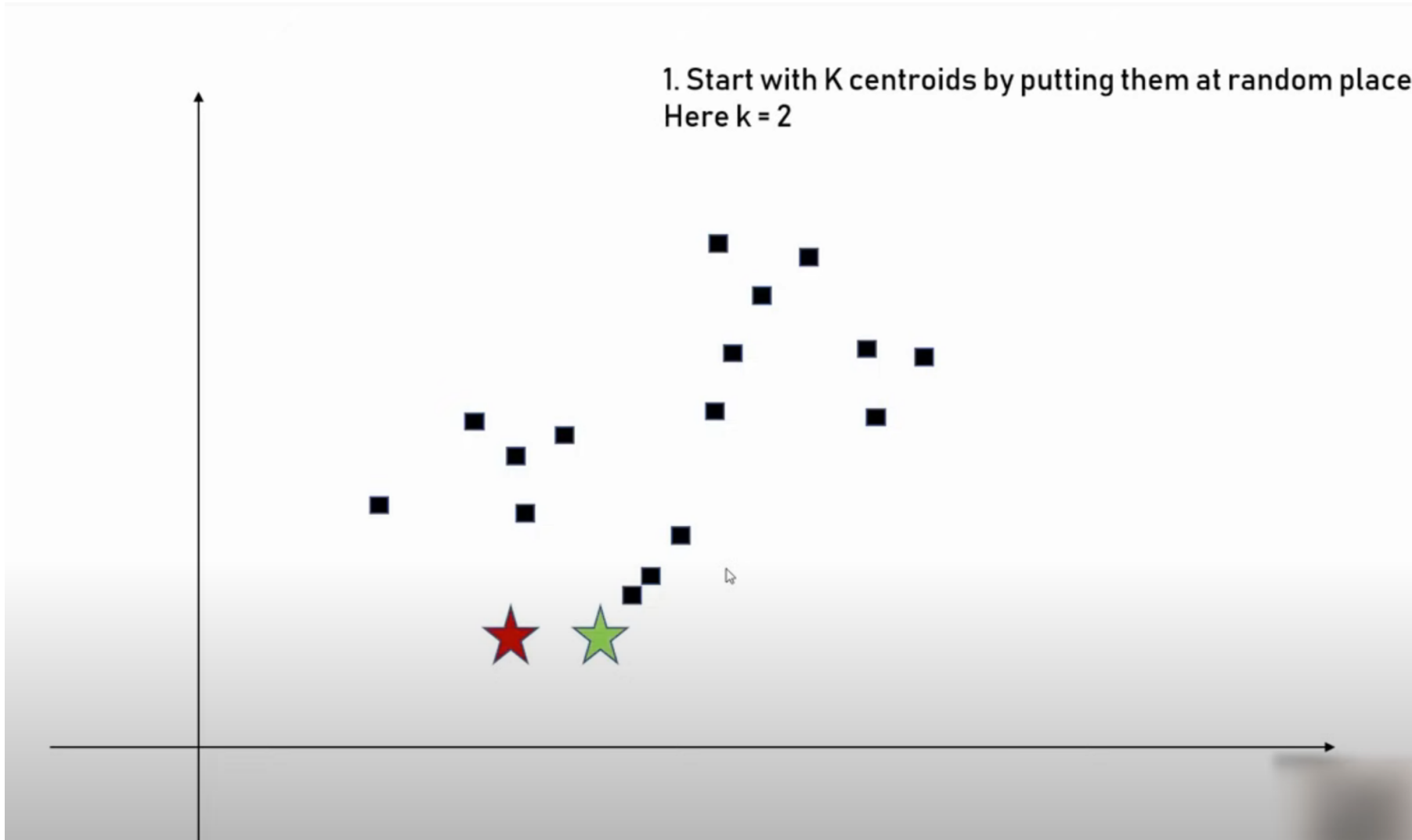
- Unsupervised learning
- Requires data, but no labels
- Detect patterns e.g. in
  - Group emails or search results
  - Customer shopping patterns
  - Regions of images
- Useful when don't know what you're looking for
- But: can get gibberish



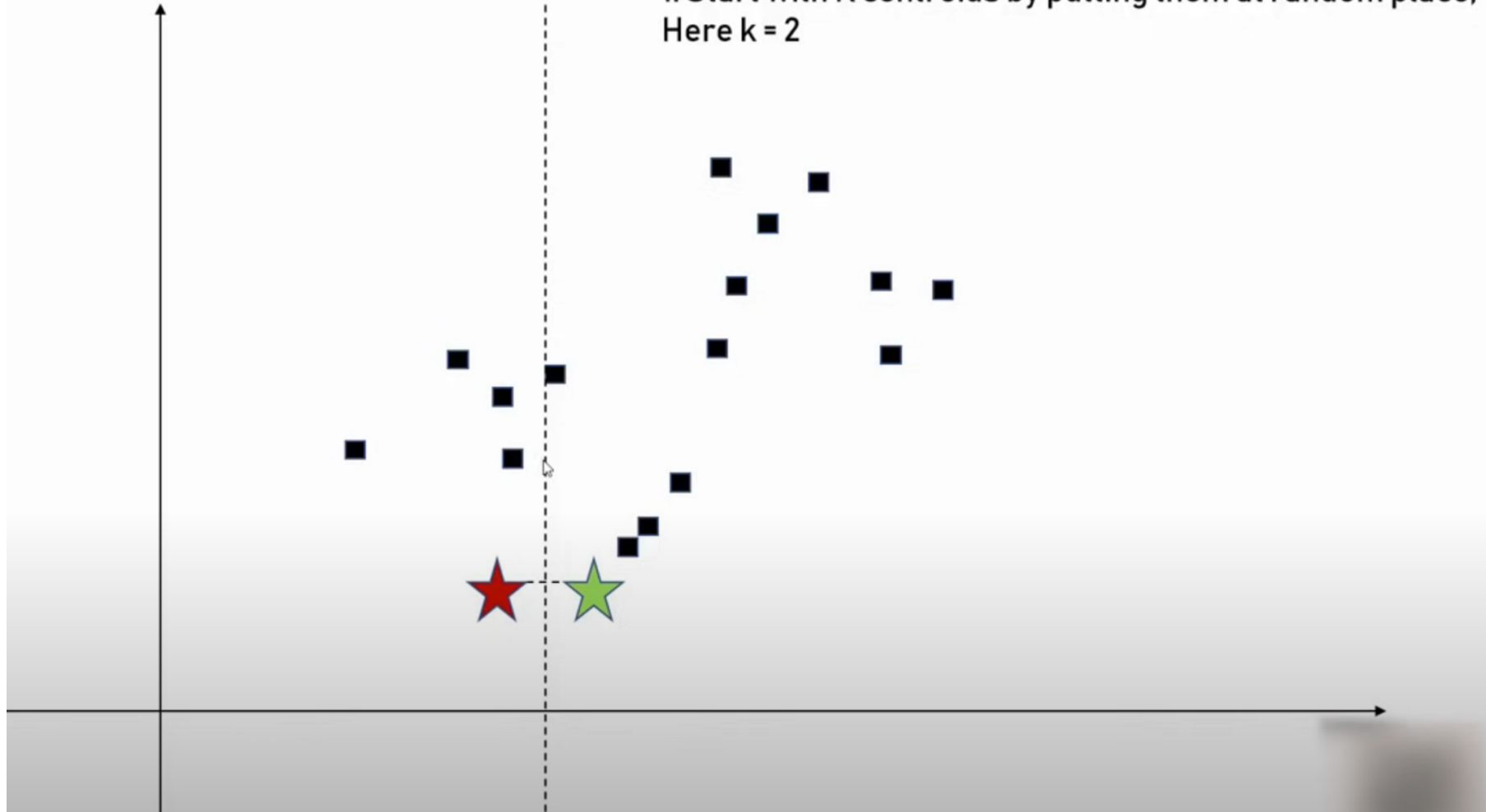




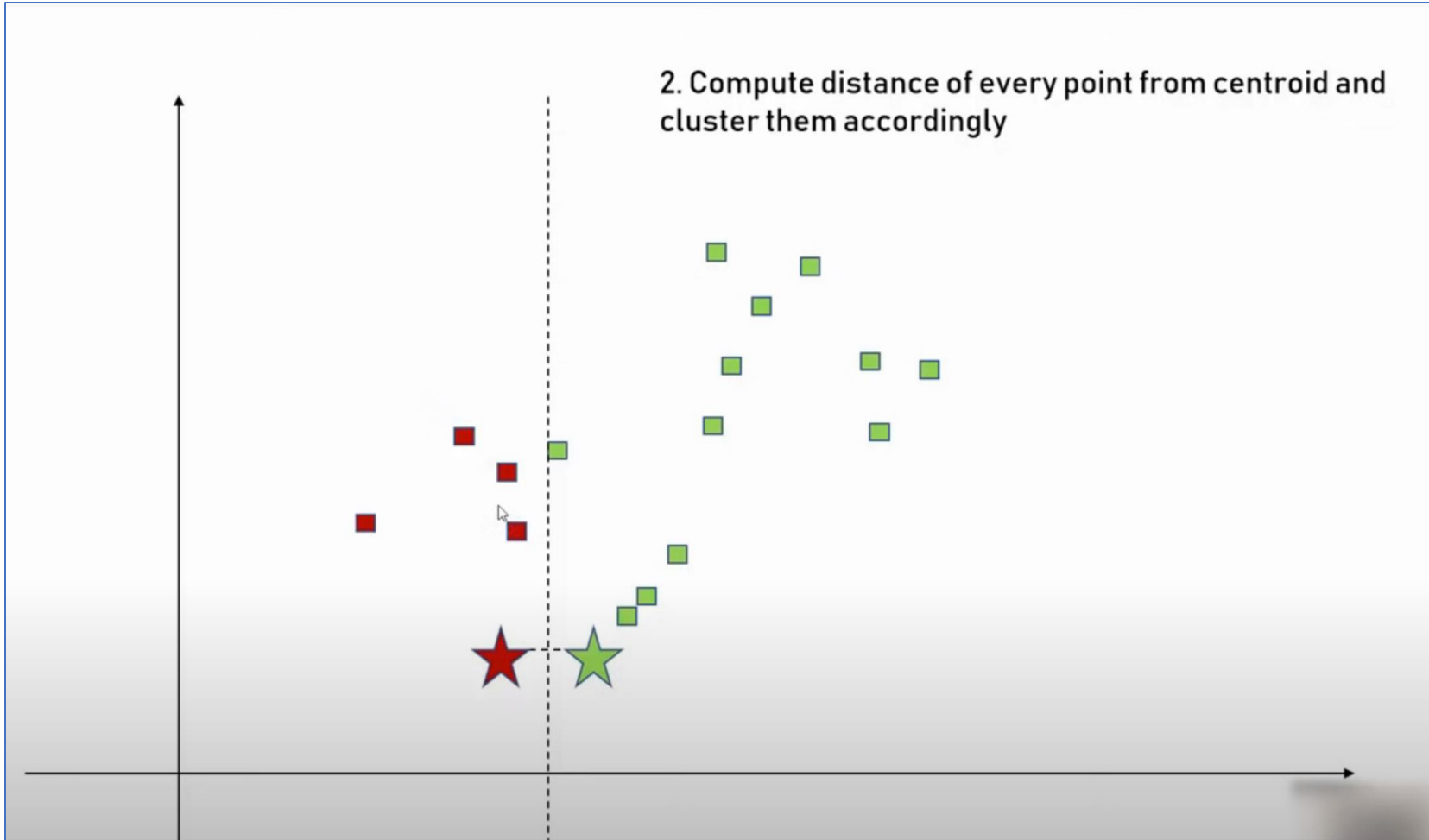
1. Start with K centroids by putting them at random place,  
Here  $k = 2$



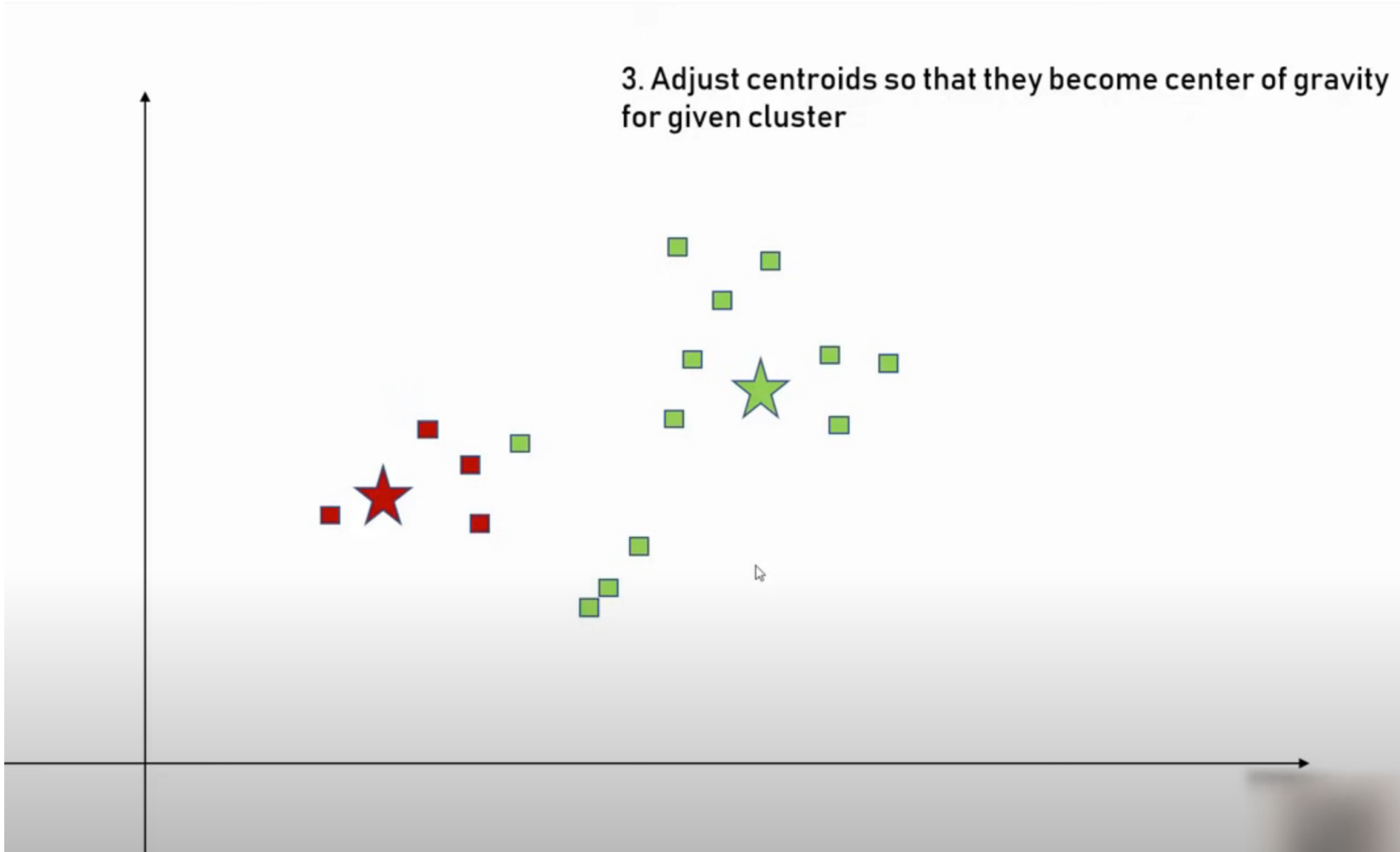
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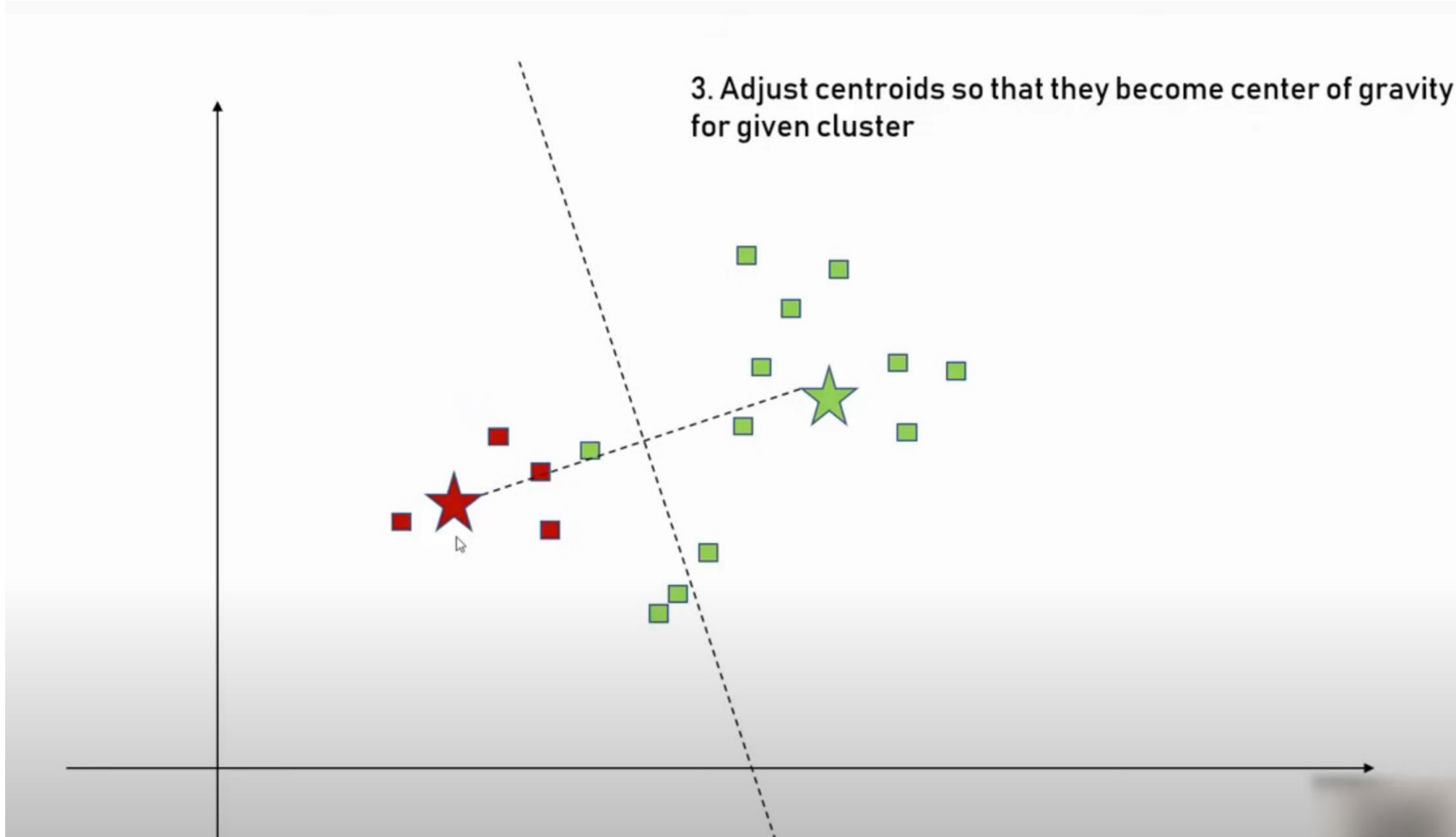
2. Compute distance of every point from centroid and cluster them accordingly



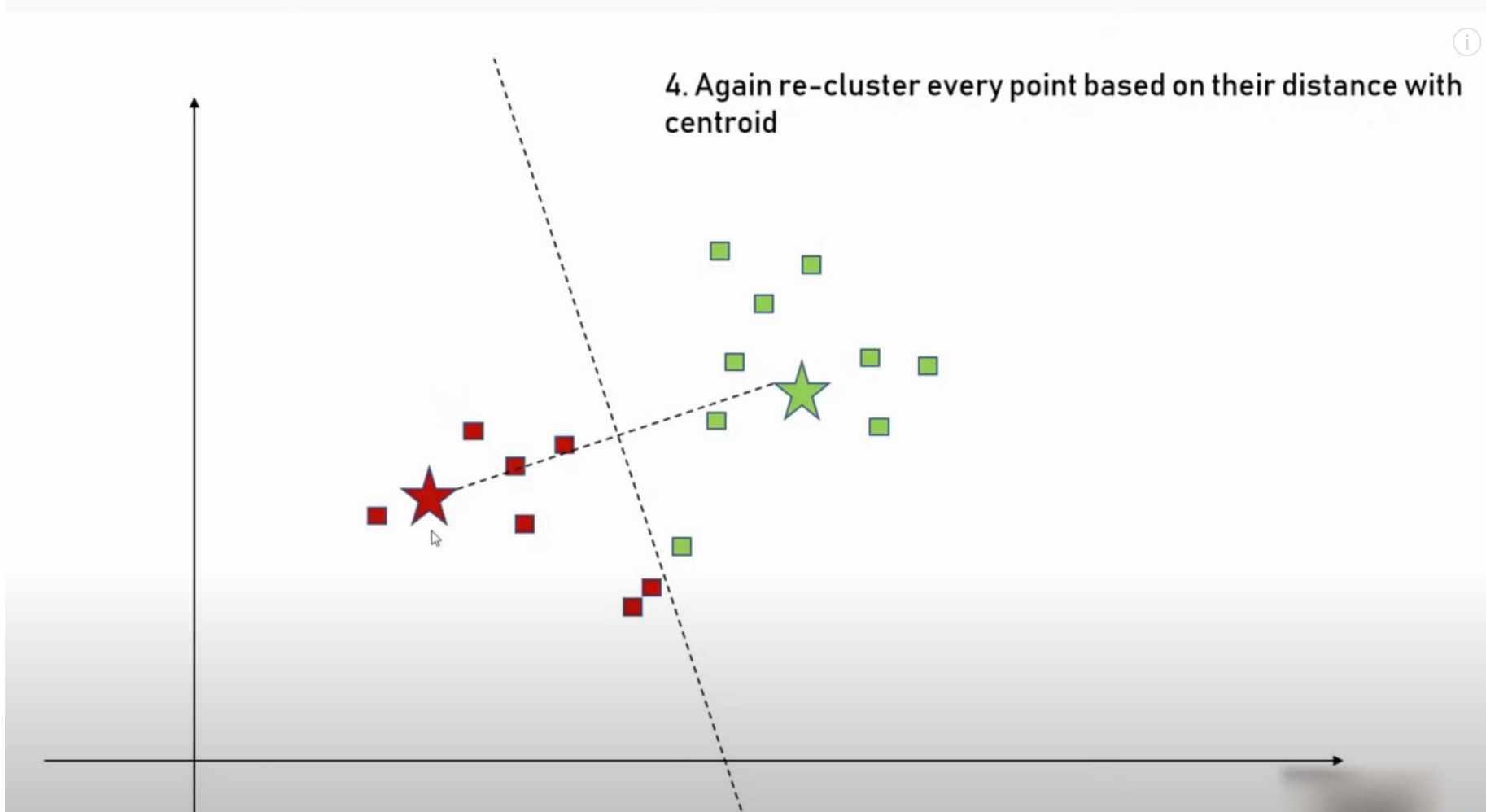
3. Adjust centroids so that they become center of gravity for given cluster



3. Adjust centroids so that they become center of gravity for given cluster



4. Again re-cluster every point based on their distance with centroid



5. Again adjust centroids

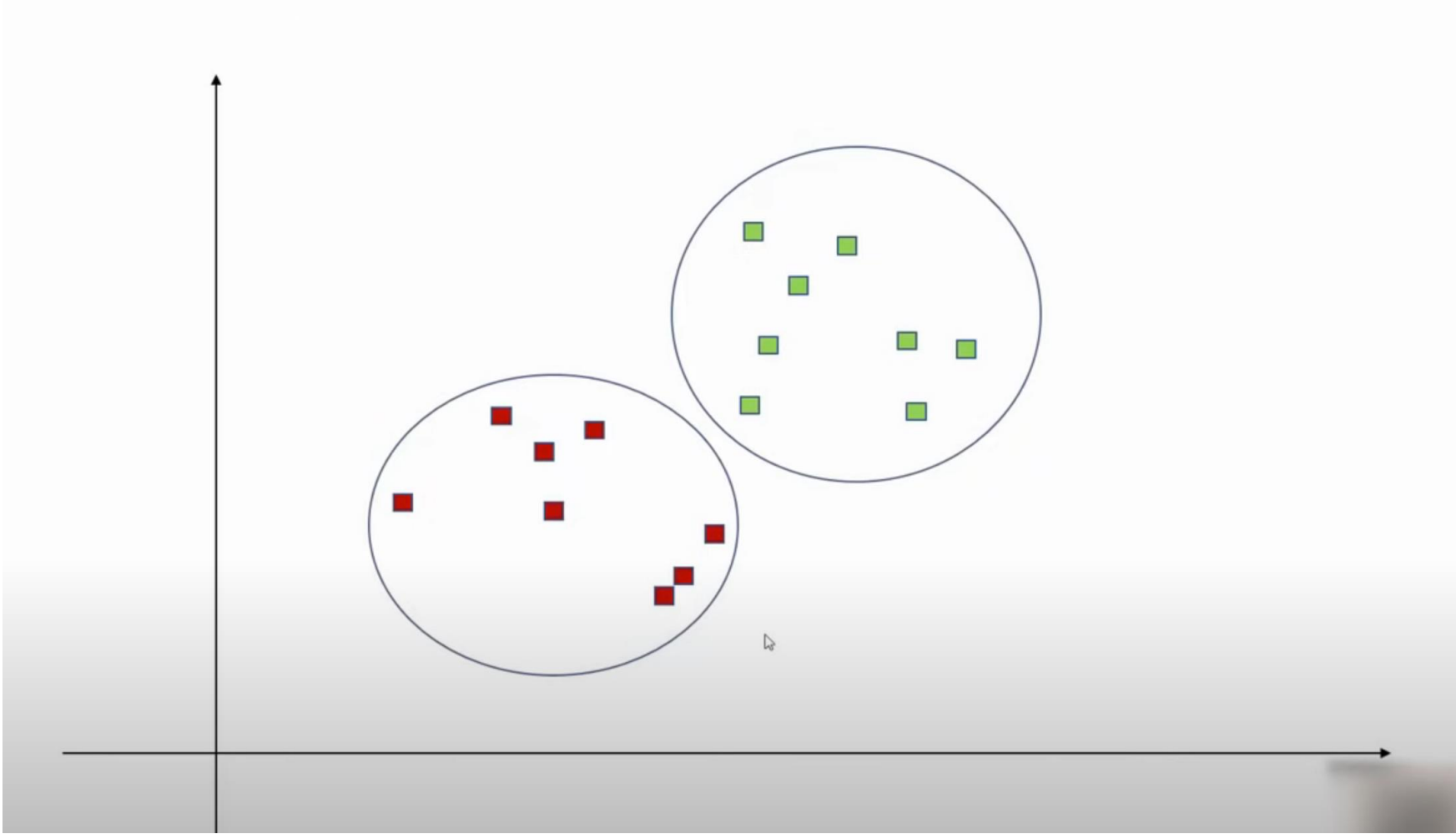
The diagram illustrates the iterative adjustment of centroids in a K-means clustering algorithm. It shows two clusters of data points (red squares and green squares) and their centroids (red and green stars). A dashed line represents the decision boundary. The centroids are being adjusted based on the current cluster membership.





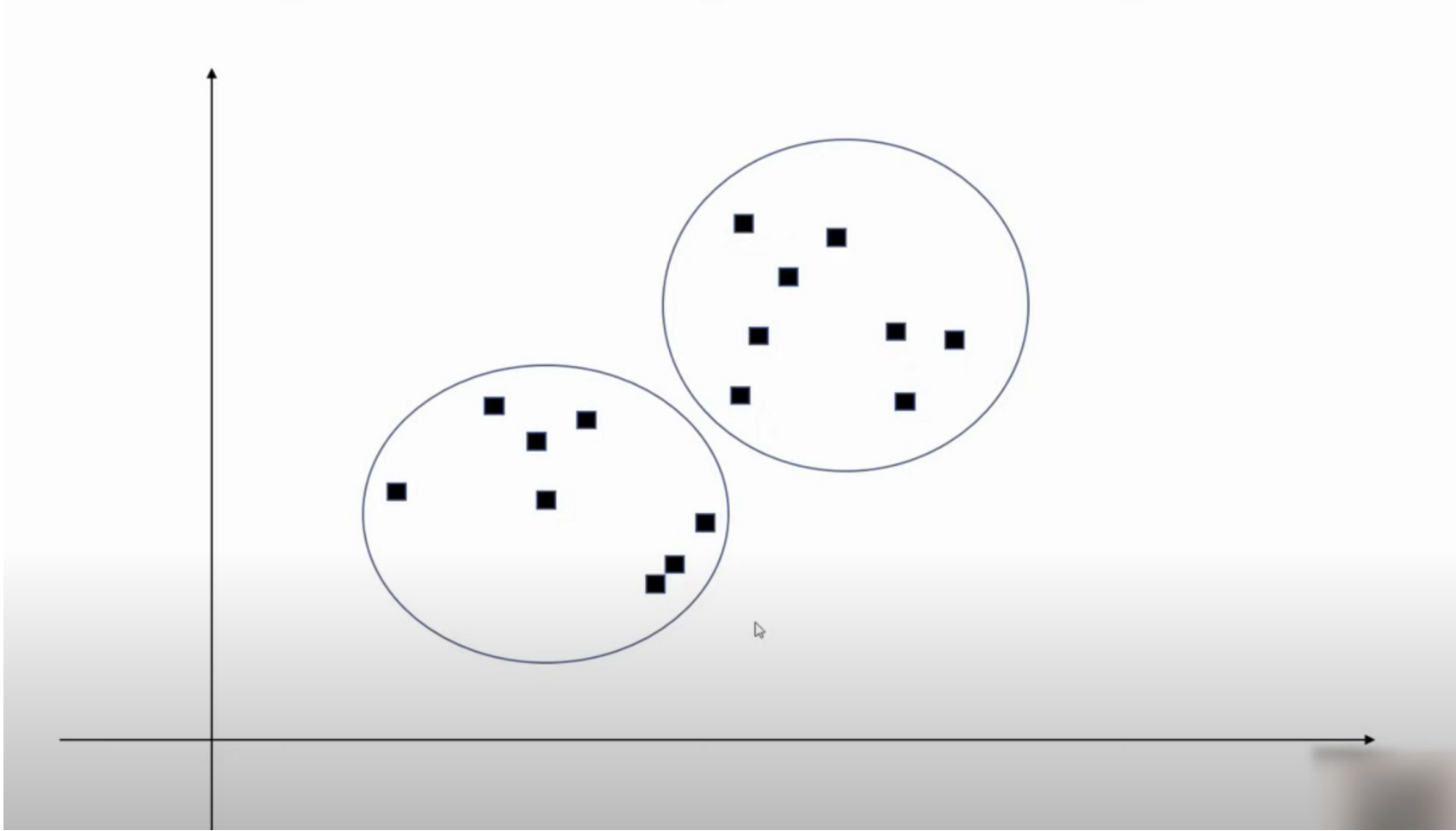
6. Recompute clusters and repeat this till data points stop changing clusters

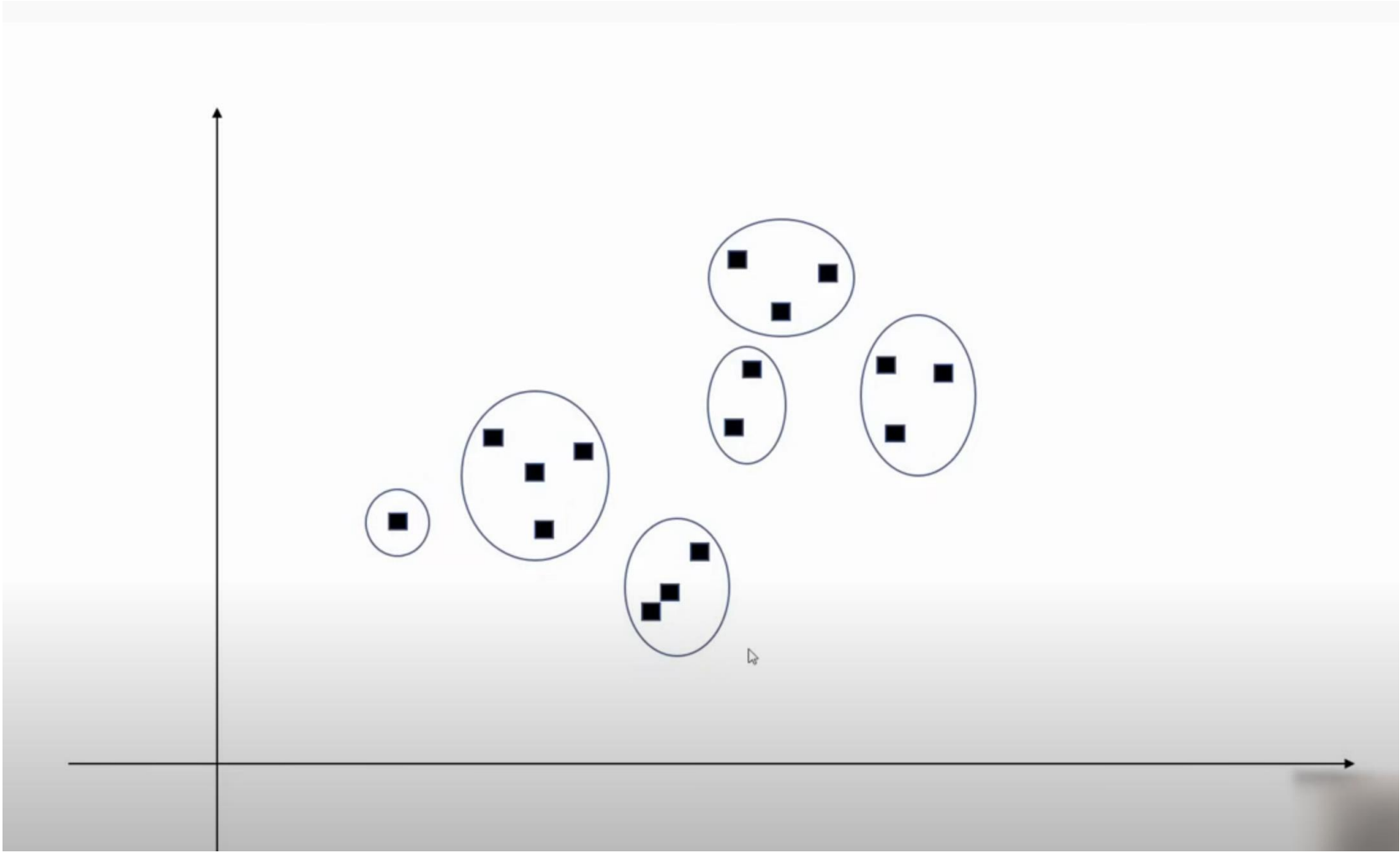




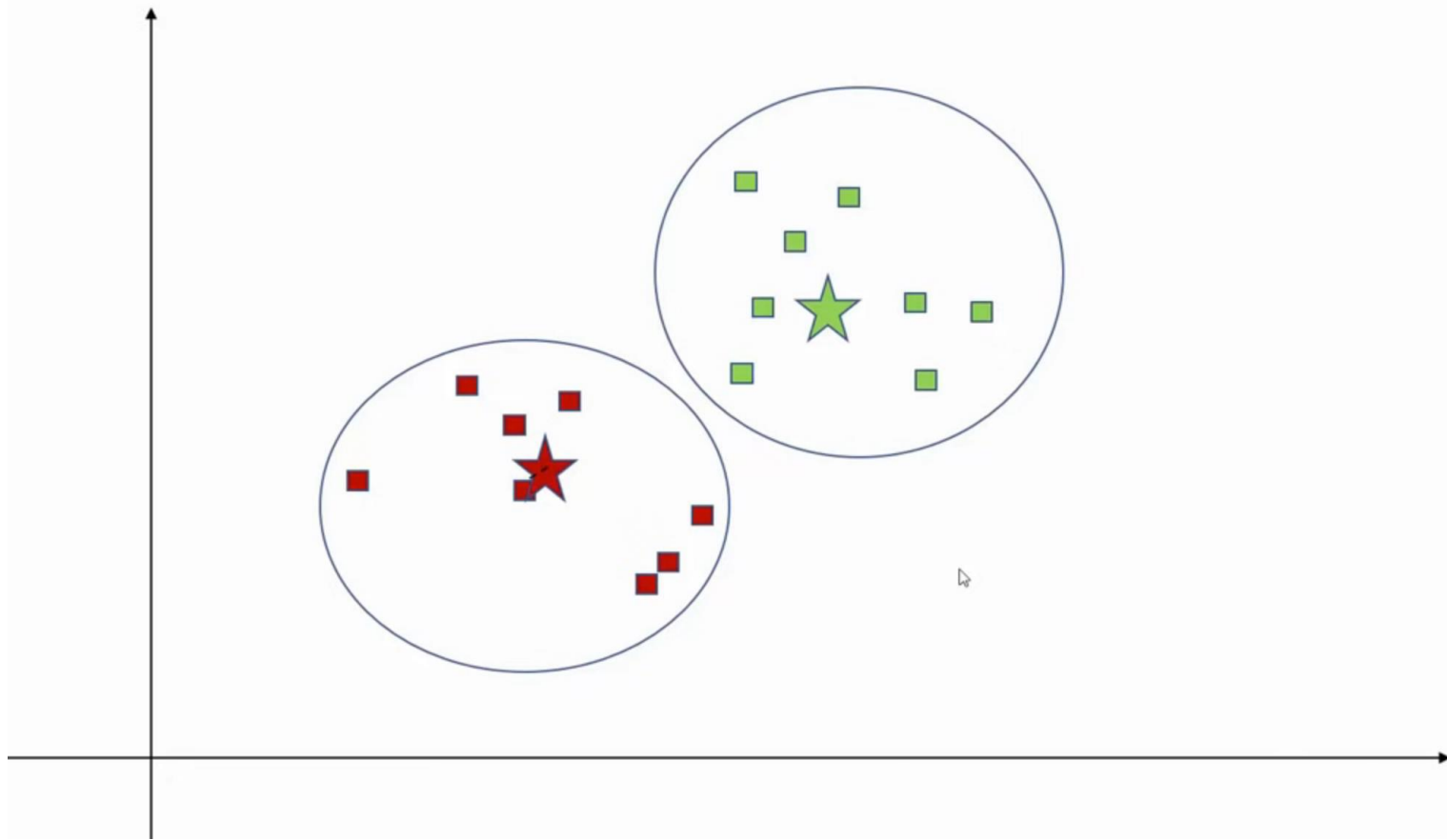


**How to determine correct  
number of clusters ( $k$ )?**

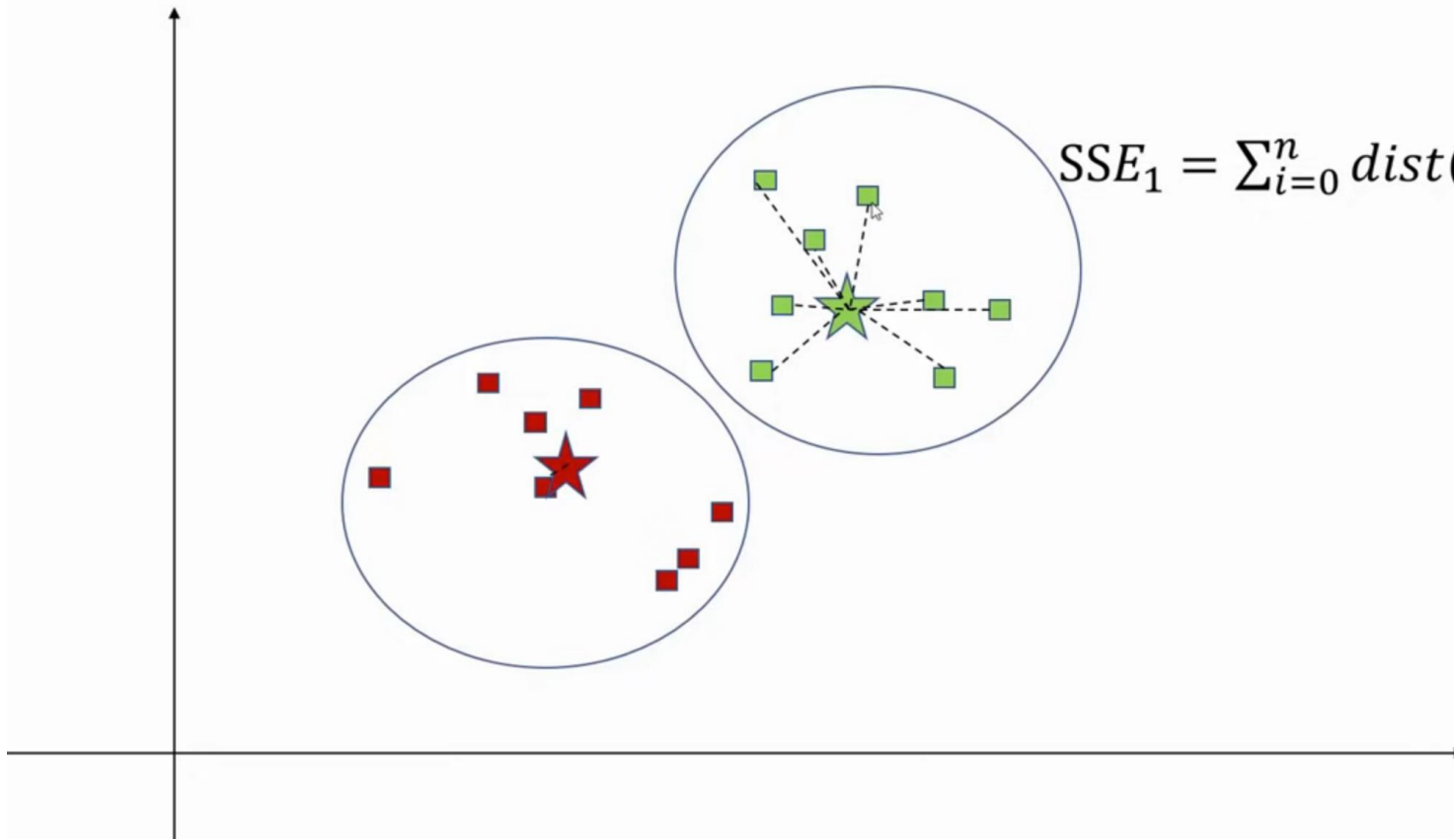




SSE = Sum of Squared Errors

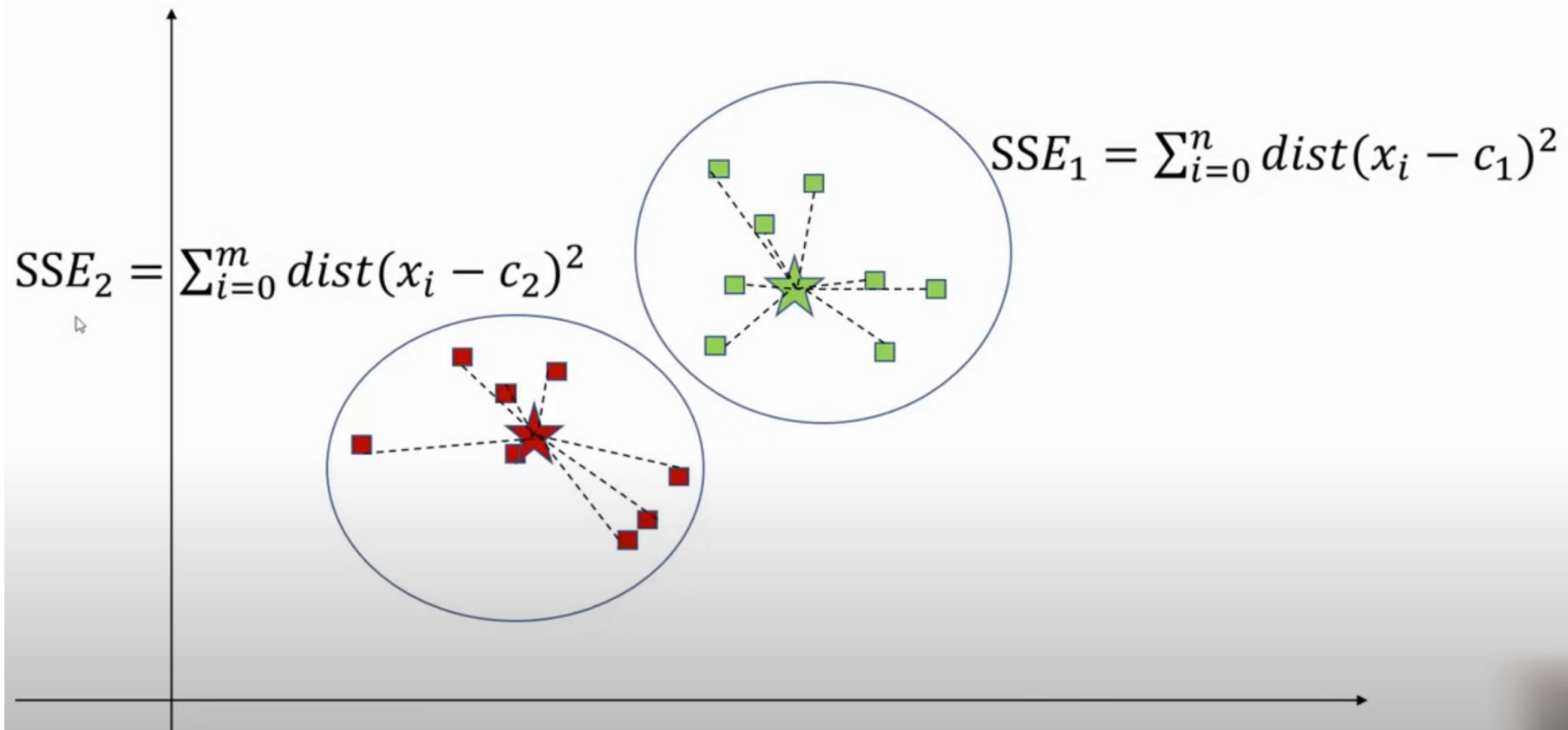


SSE = Sum of Squared Errors



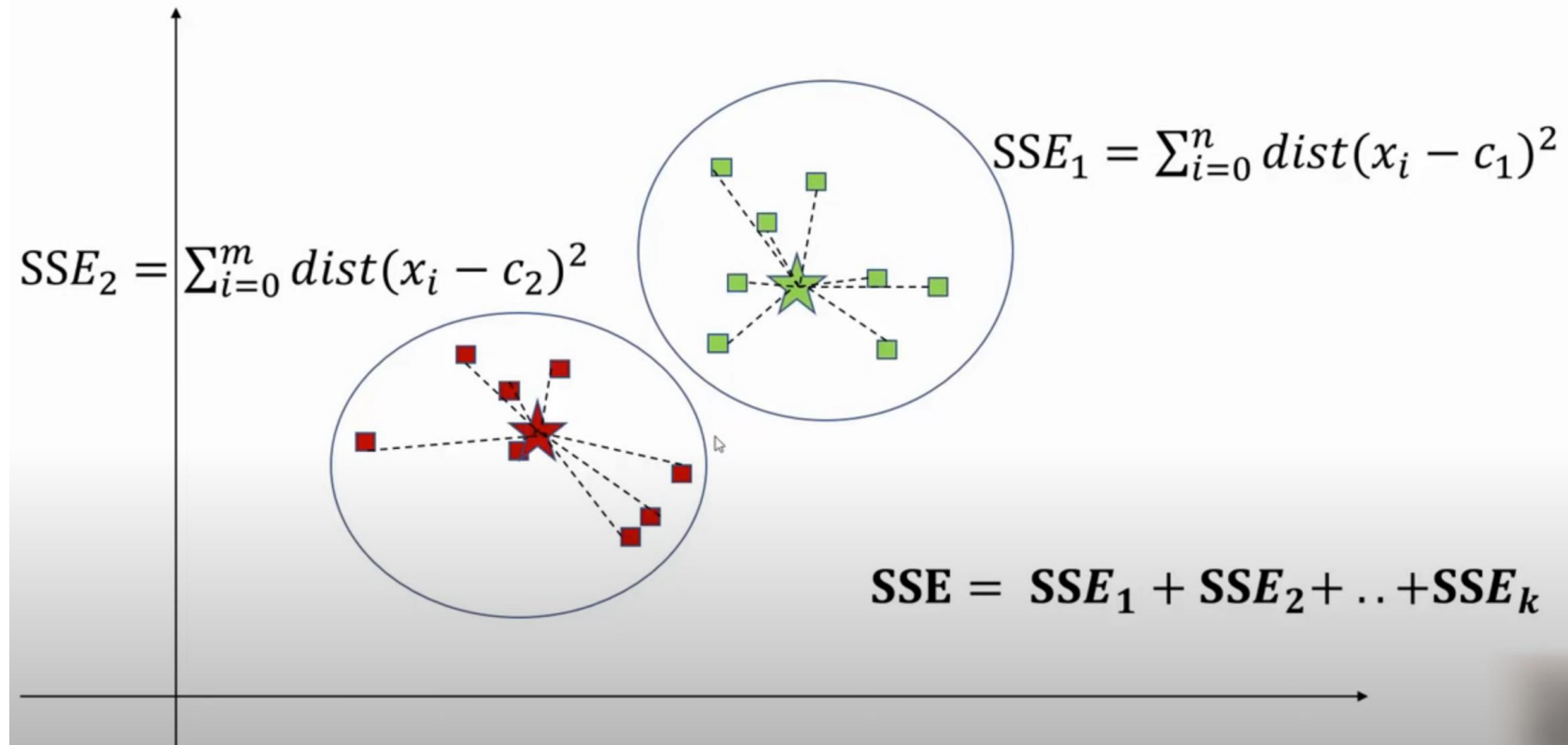
$$SSE_1 = \sum_{i=0}^n dist(x_i - c_1)^2$$

SSE = Sum of Squared Errors

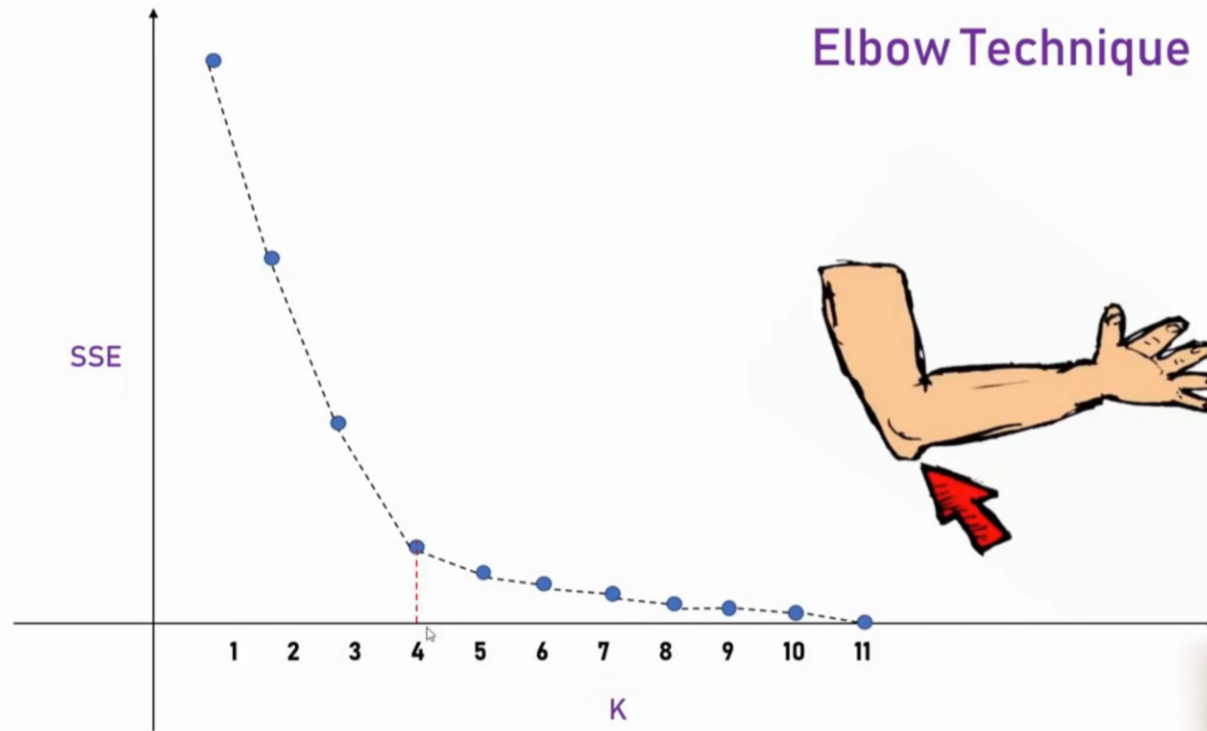




SSE = Sum of Squared Errors



## Elbow Technique



# Lecture Map

- Theory
- Coding
- Exercise

# K-means Clustering

- **Strengths**

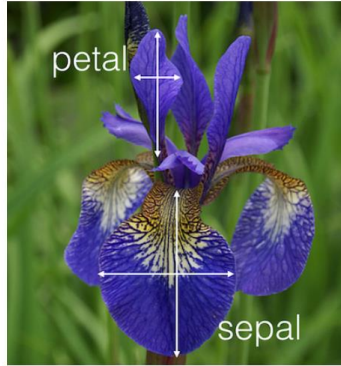
- Simple iterative method
- User provides “K”

- **Weaknesses**

- Often too simple → bad results
- Difficult to guess the correct “K”

# K-means Clustering Exercise

## Exercise



1. Use iris flower dataset from sklearn library and try to form clusters of flowers using petal width and length features. Drop other two features for simplicity.
2. Figure out if any preprocessing such as scaling would help here
3. Draw elbow plot and from that figure out optimal value of k



# References

- <http://brokerstir.com/logistic-regression-model-intuition/>
- <https://www.geeksforgeeks.org/implement-sigmoid-function-using-numpy/>
- <http://ieeexplore.ieee.org/document/6914146/>
- <http://www.svms.org/disadvantages.html>
- <https://www.mit.edu/~9.520/spring09/Classes/multiclass.pdf>
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