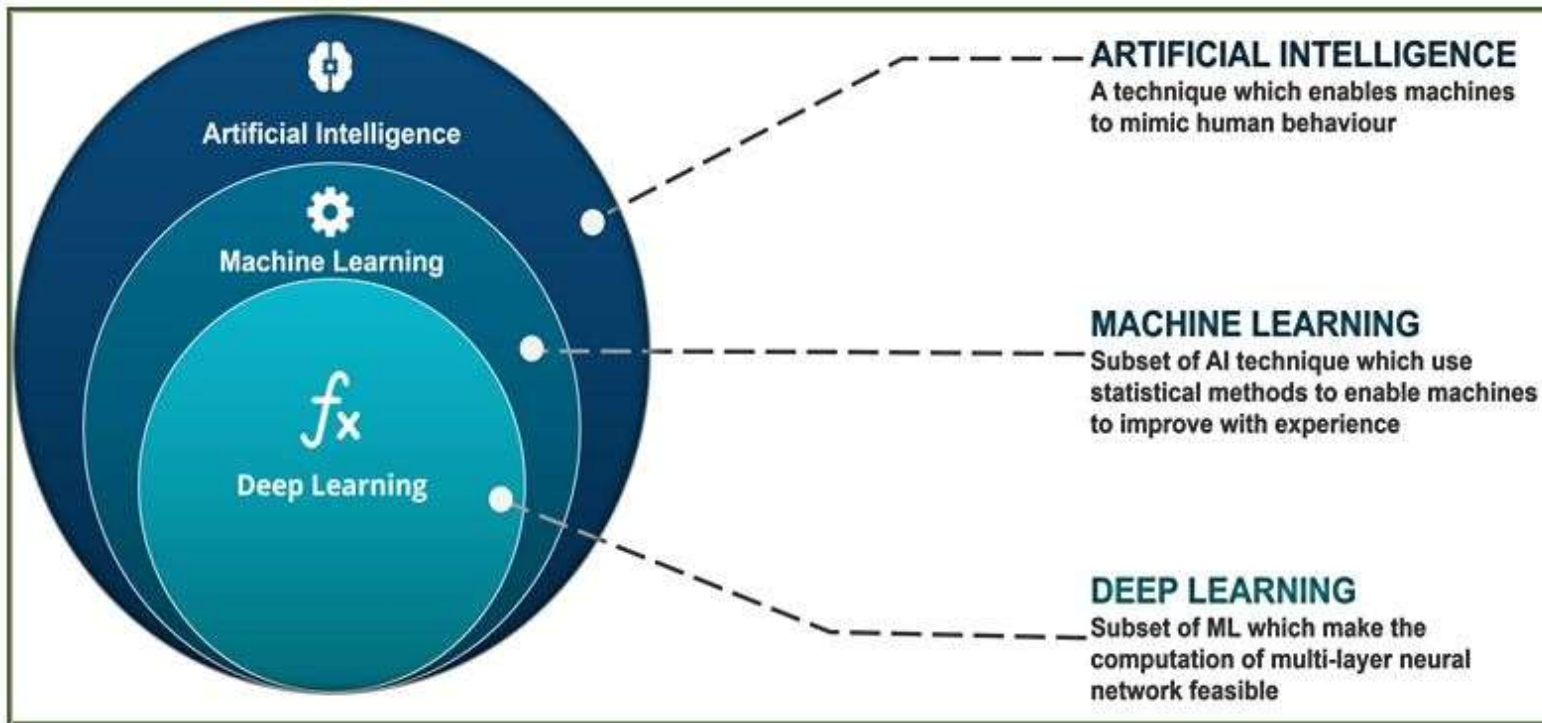


# AI – ML – DL

- What is AI, ML and DL?



Machine Learning

Supervised Learning

Unsupervised Learning

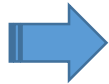
Semi supervised Learning

Reinforcement Learning

# ML

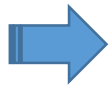
- Supervised Learning
  - Technique to achieve AI through algorithm trained with data
  - Function that maps and input to an output based on a series of examples
  - Examples: input-output pairs

## Regression



- To find/predict an output based on independent predictors
- The output is continuous
- Methods:
  - Linear regression; Decision tree; Random forest; NNETs; ...

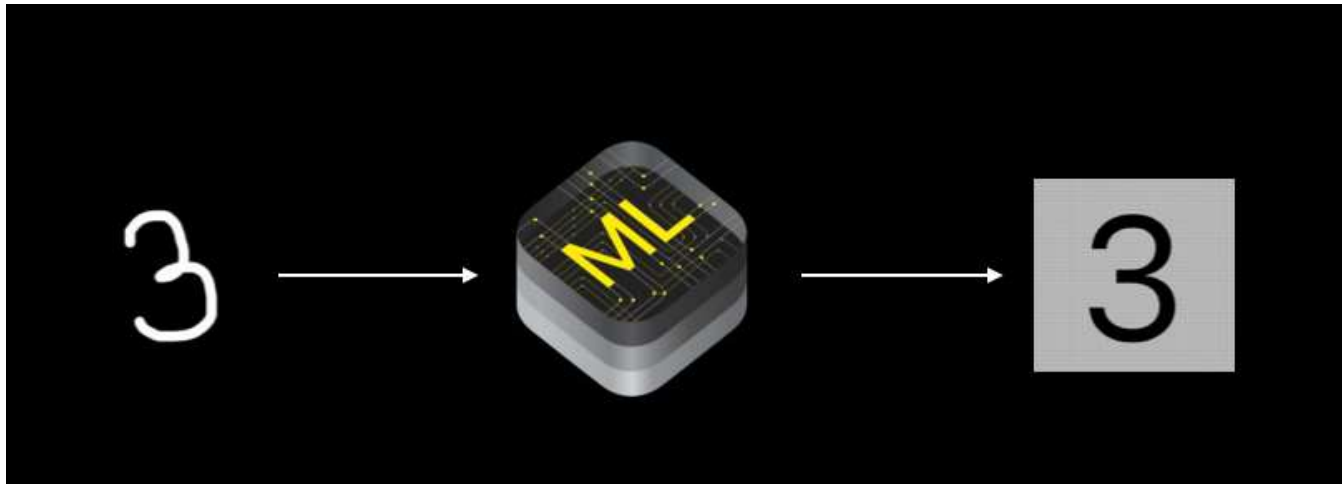
## Classification



- To assign a class to an output (discriminate, separate)
- The output is discrete
- Methods:
  - Logistic regression; Linear discriminant; Support Vector Machines; Naïve Bayes; NNETs; DL based models; ...

# Examples

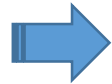
- Recognition of handwriting digits



# ML

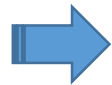
- Unsupervised Learning
  - Find patterns from input data without references to the outcomes

Clustering



- Methods:
  - K-means; Hierarchical; Mean-shift; Density-based ...

Dimensionality  
reduction



- Elimination or extraction of variables
- Methods:
  - PCA; ...

# K-means clustering

Javier M. Antelis

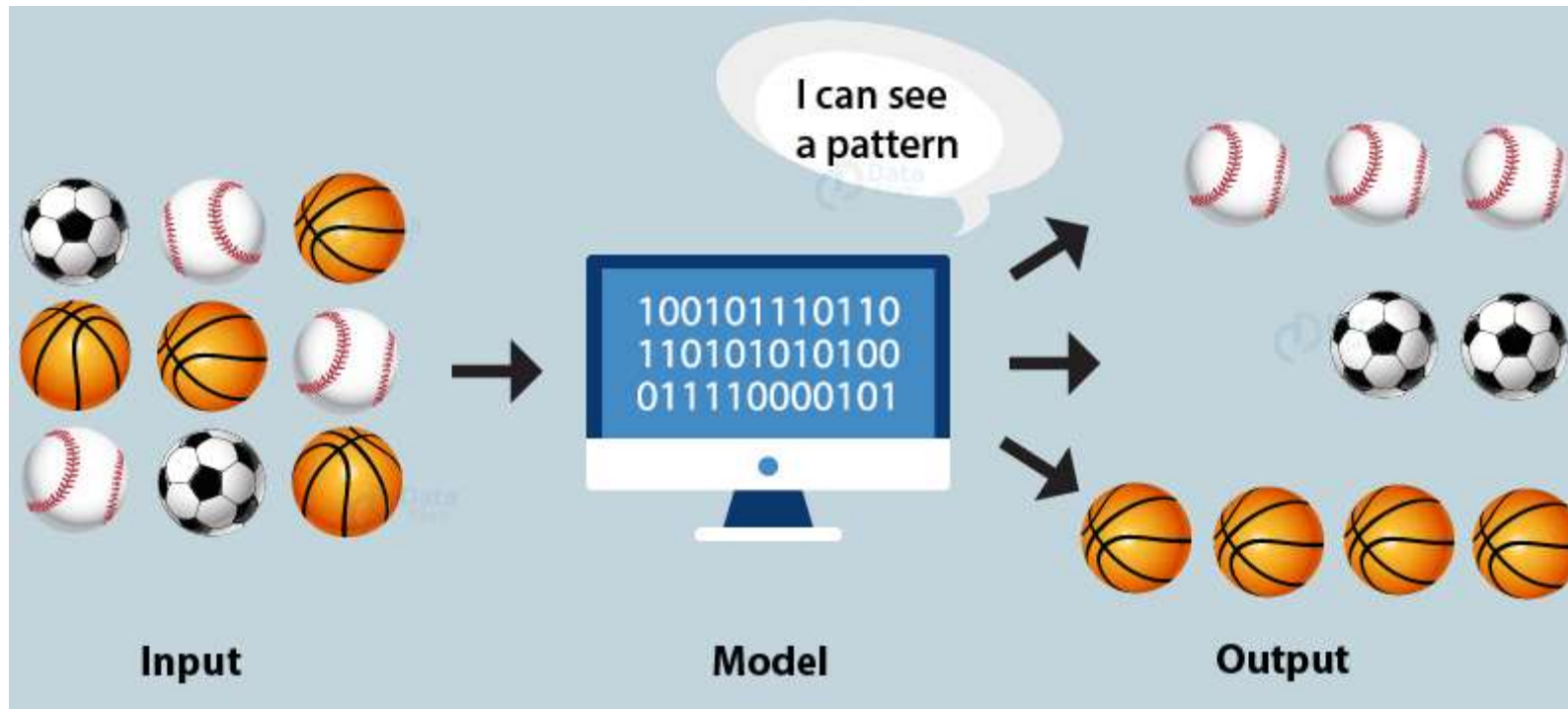
[mauricio.antelis@tec.com](mailto:mauricio.antelis@tec.com)

Escuela de Ingeniería y Ciencias

Tecnológico de Monterrey

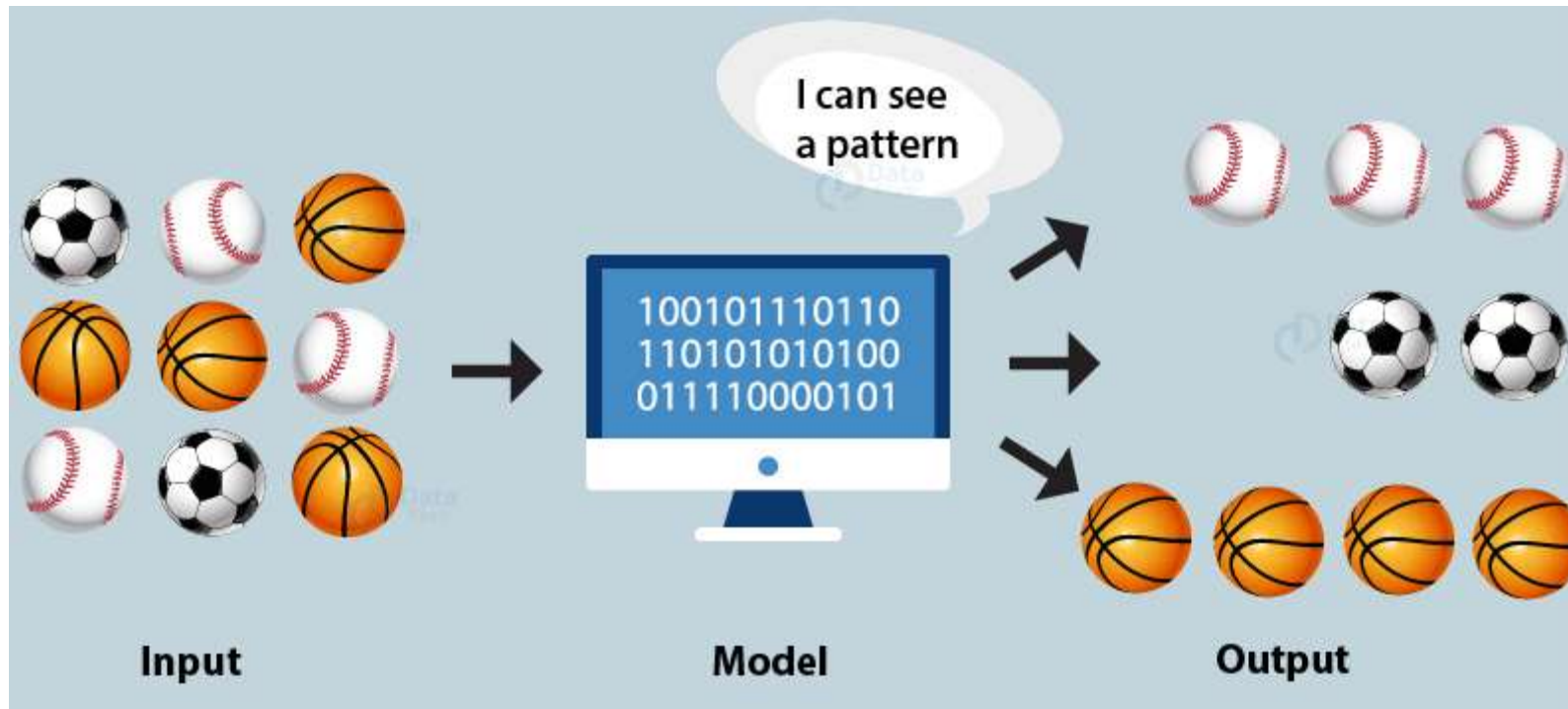
Guadalajara, México

# What is Clustering?



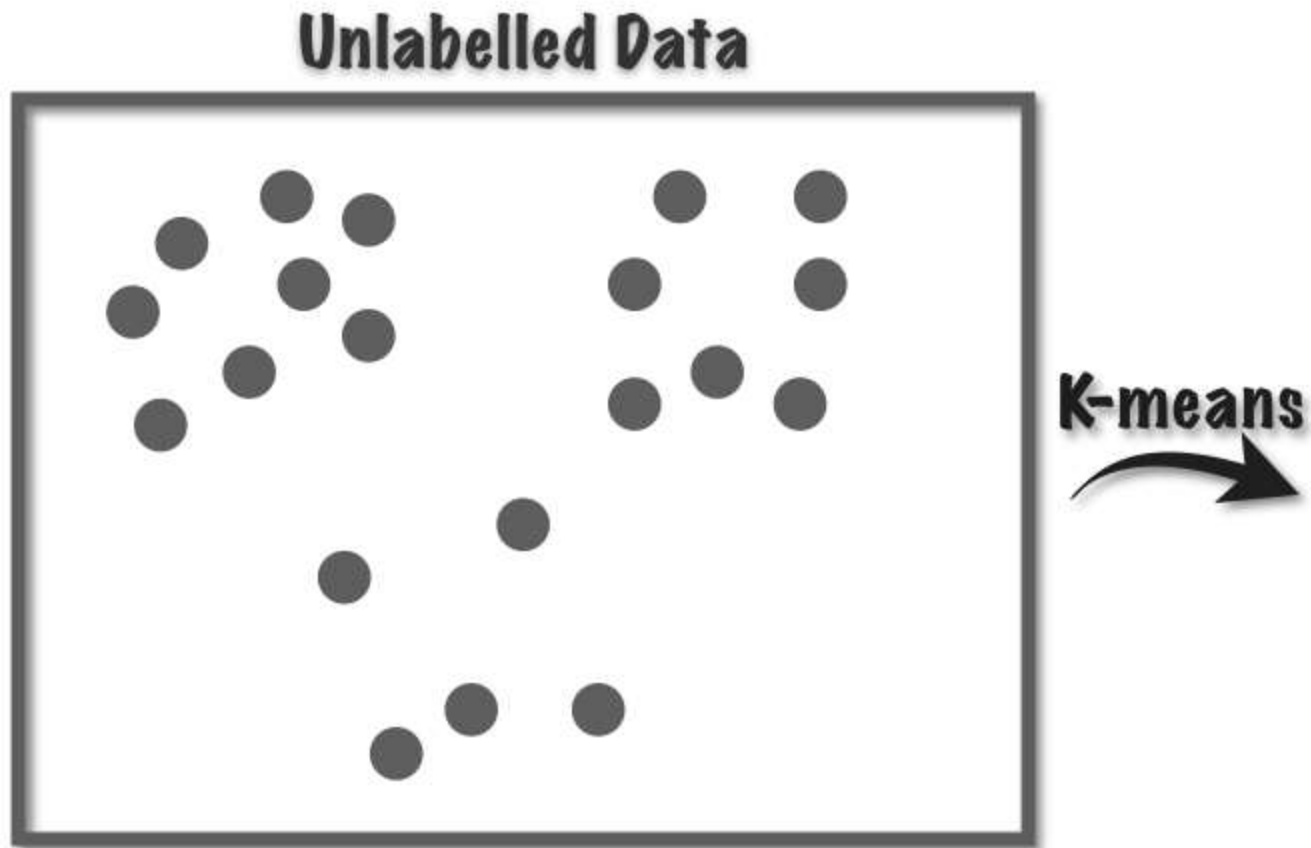
- Clustering is a set of “computational” techniques used to partition data objects into groups (or clusters)

# What is Clustering?



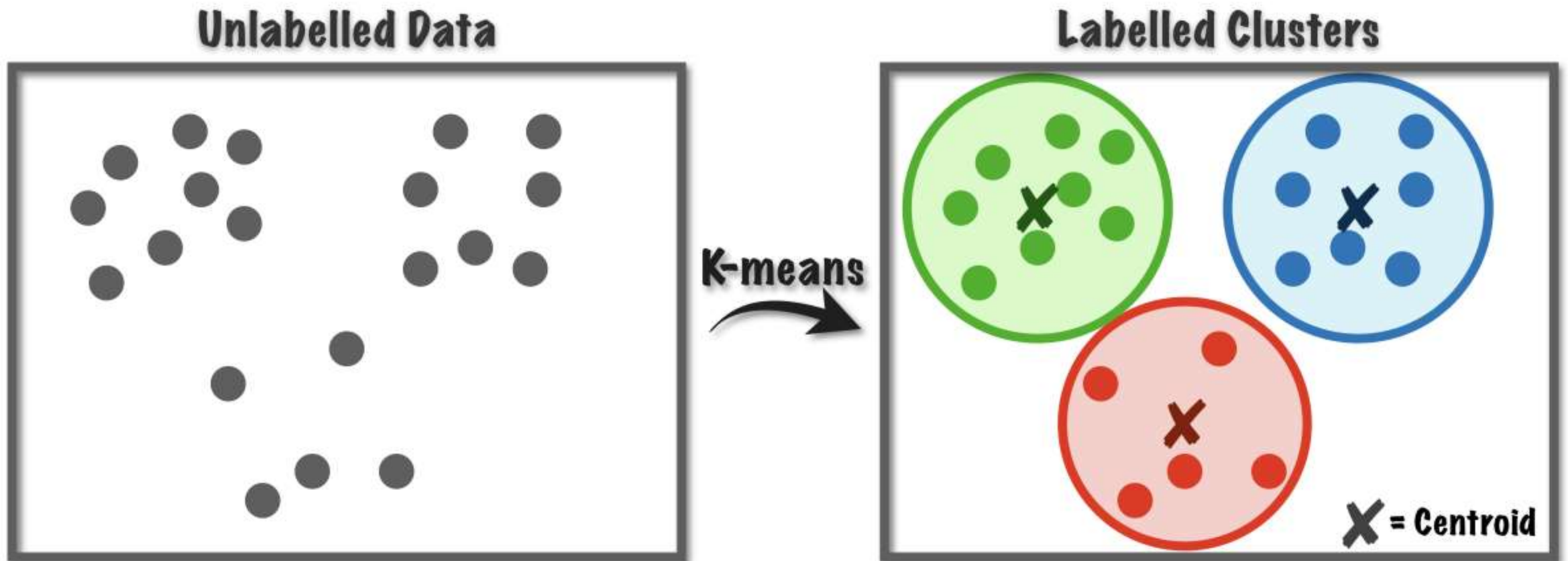
- Clusters are defined as groups of data objects that are more similar to other objects in their cluster than they are to data objects in other clusters

# Clustering algorithms





# Clustering algorithms



# Types of clustering algorithms

- **K-means Clustering:** Using this algorithm, we classify a given data set through a certain number of predetermined clusters or “k” clusters.
- **Hierarchical Clustering:** follows two approaches Divisive and Agglomerative.
- **Fuzzy C means Clustering:** The working of the FCM Algorithm is almost similar to the k-means clustering algorithm, the major difference is that in FCM a data point can be put into more than one cluster.
- **Density-Based Spatial Clustering:** Useful in the application areas where we require non-linear cluster structures, purely based on density

# What is K-means?

- K-means clustering method is an **unsupervised machine learning technique** used **to identify clusters in data**.
- K-means is an **iterative algorithm** that divides a group of data points into **K subgroups/clusters** based on the **similarity of their mean distance** from the **centroid** of that particular subgroup/formed.
- **K is the pre-defined number of clusters** to be formed by the algorithm

# What is K-means?

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**Algorithm 1** *k*-means algorithm

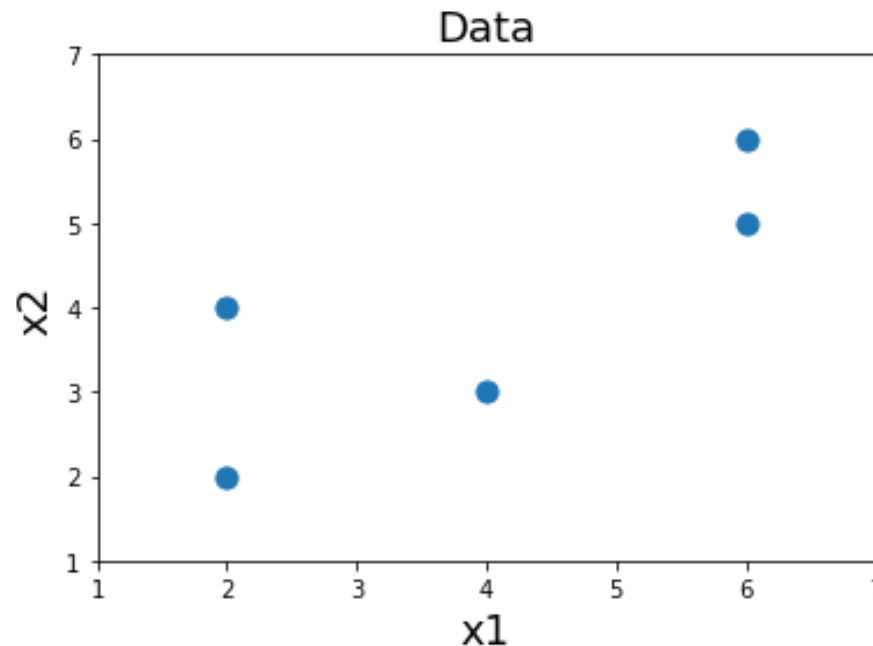
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- 1: Specify the number  $k$  of clusters to assign.
  - 2: Randomly initialize  $k$  centroids.
  - 3: **repeat**
  - 4:     **expectation:** Assign each point to its closest centroid.
  - 5:     **maximization:** Compute the new centroid (mean) of each cluster.
  - 6: **until** The centroid positions do not change.
-

# Intuitive explanation on K-means clustering

- We have a set of data points that we want to cluster
- We do not know the number of clusters in advance, that is, the data is not labelled

	x1	x2
0	2	2
1	4	3
2	2	4
3	6	6
4	6	5



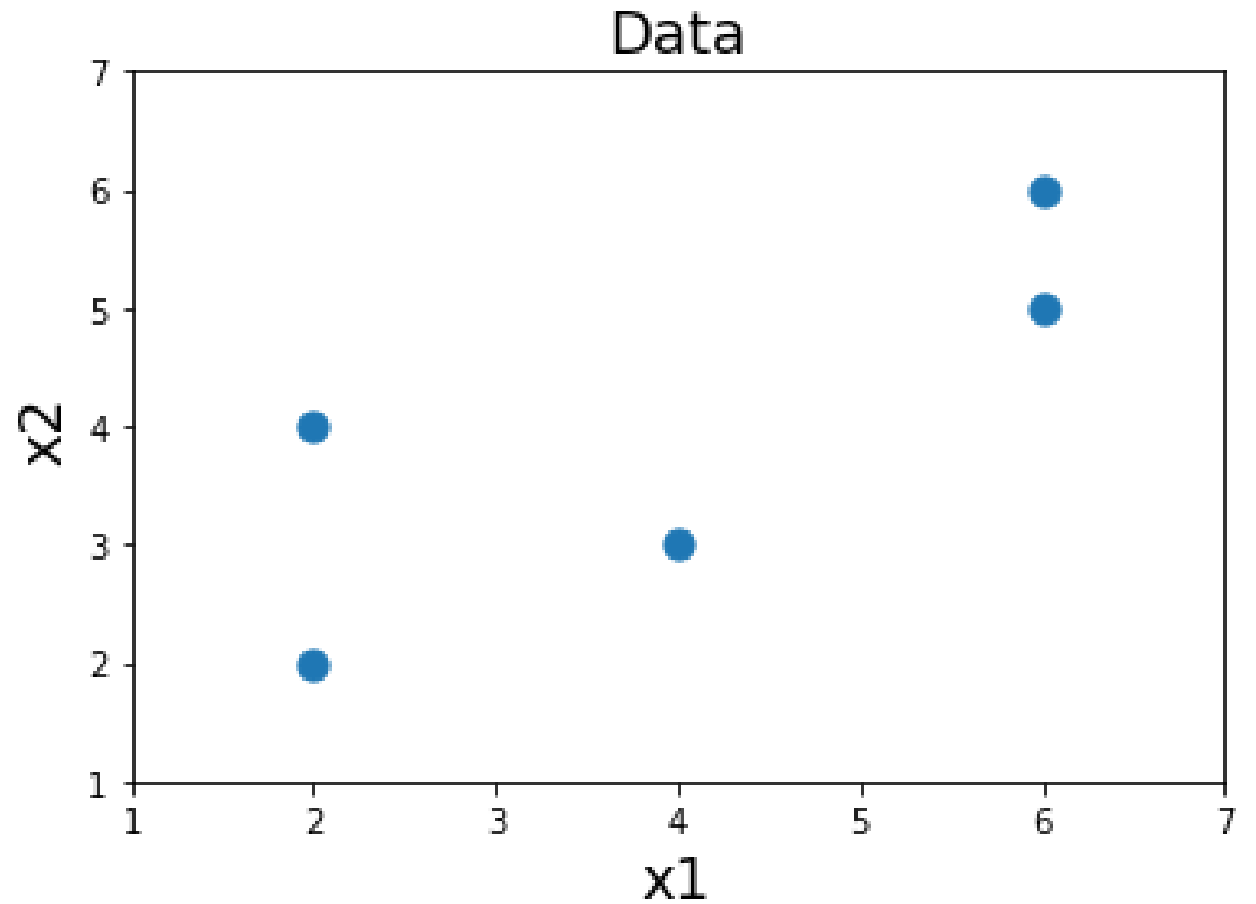
# Graphical illustration

- Data

	<b>x1</b>	<b>x2</b>
<b>0</b>	2	2
<b>1</b>	4	3
<b>2</b>	2	4
<b>3</b>	6	6
<b>4</b>	6	5

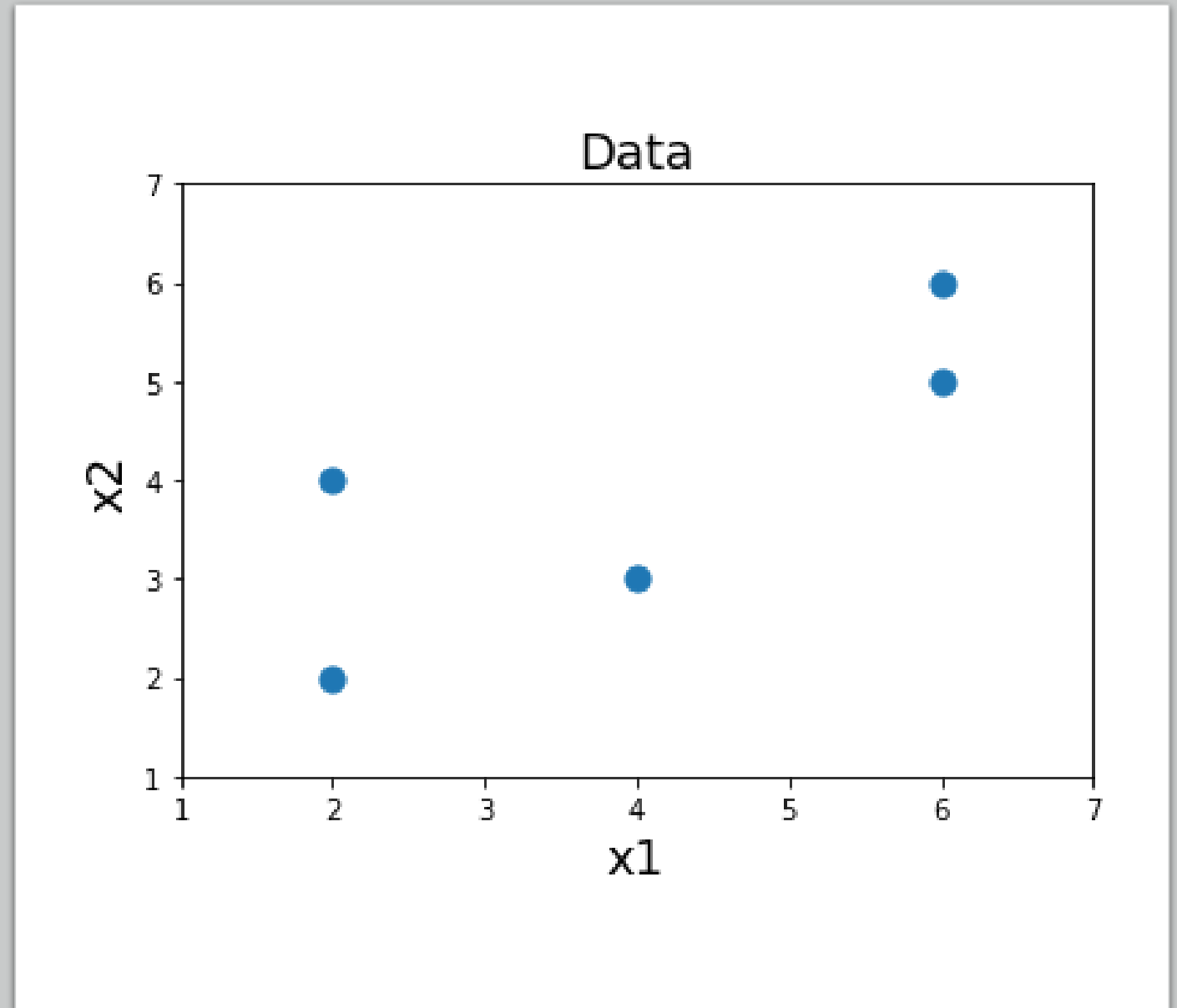
# Graphical illustration

- Data



# Graphical illustration

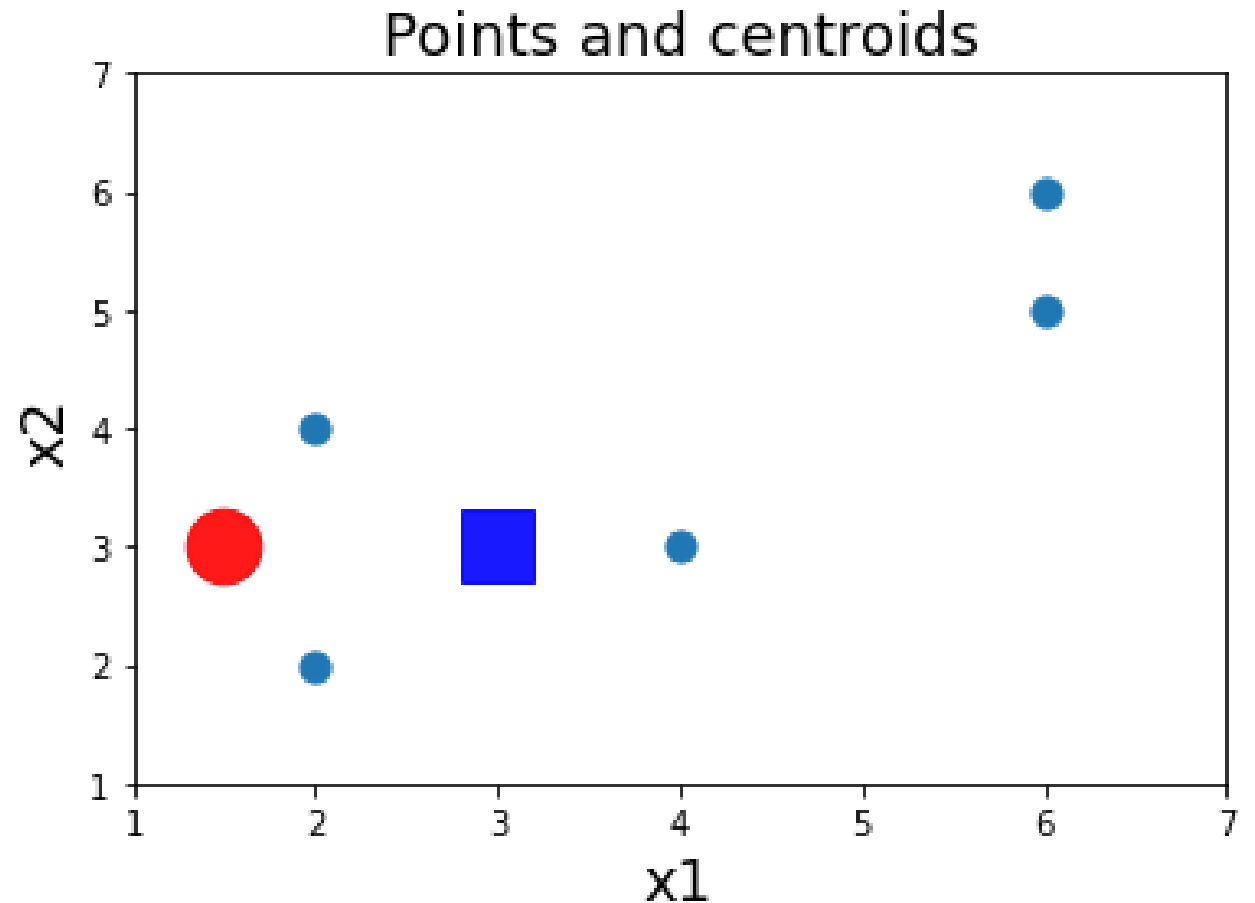
- 1: Specify the number of clusters.  $K=?$





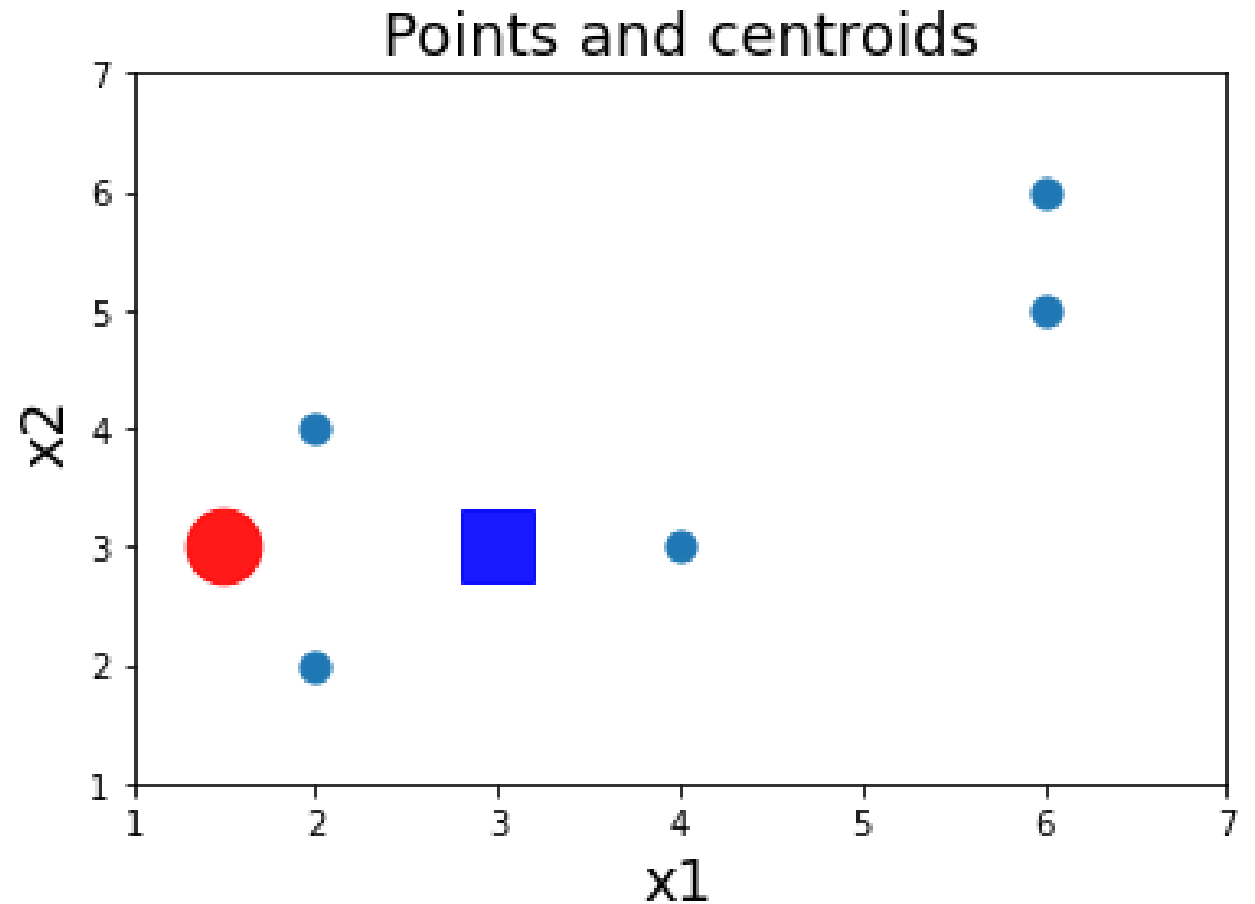
# Graphical illustration

- 2: Initialize the centroids of the clusters (randomly)



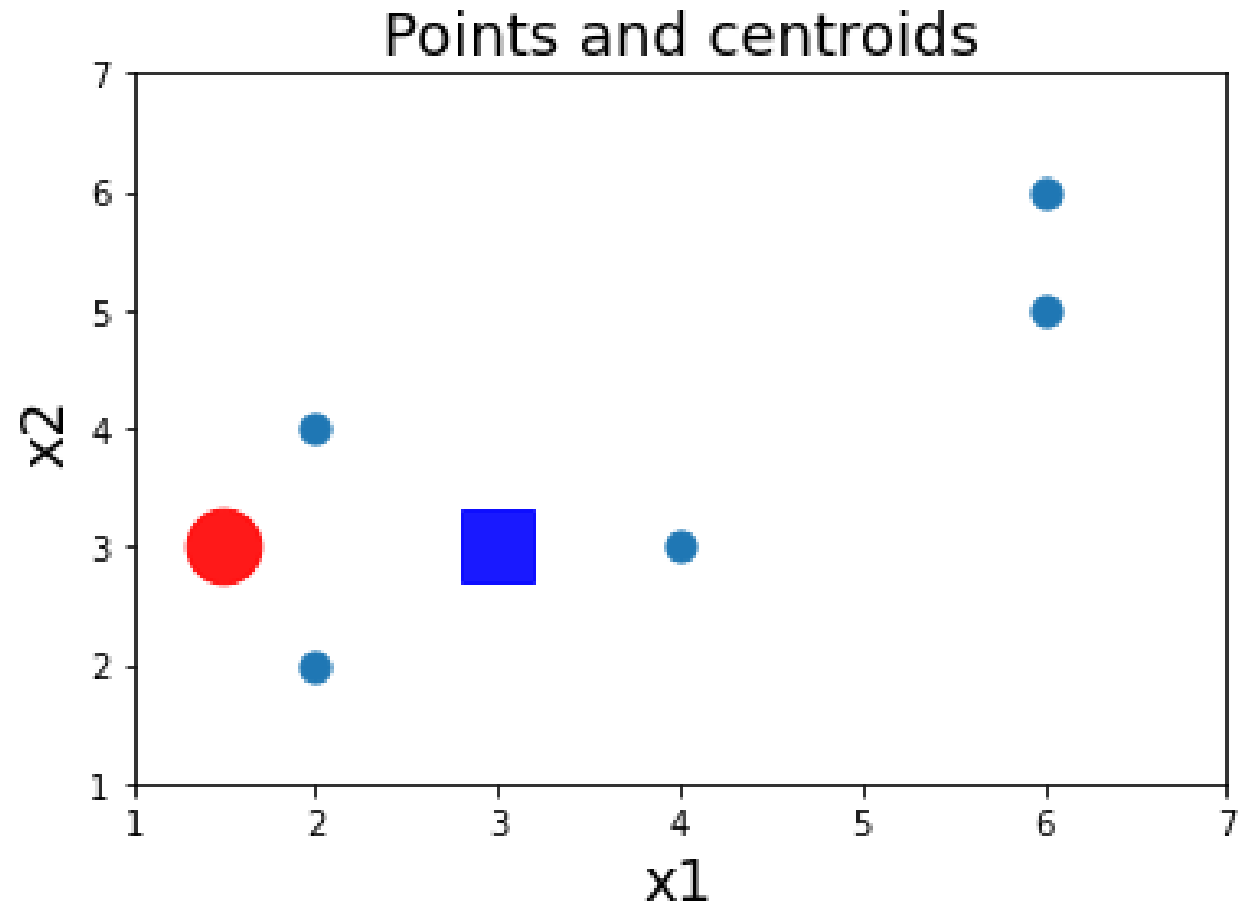
# Graphical illustration

- 3: Repeat the following (with the current centroids)



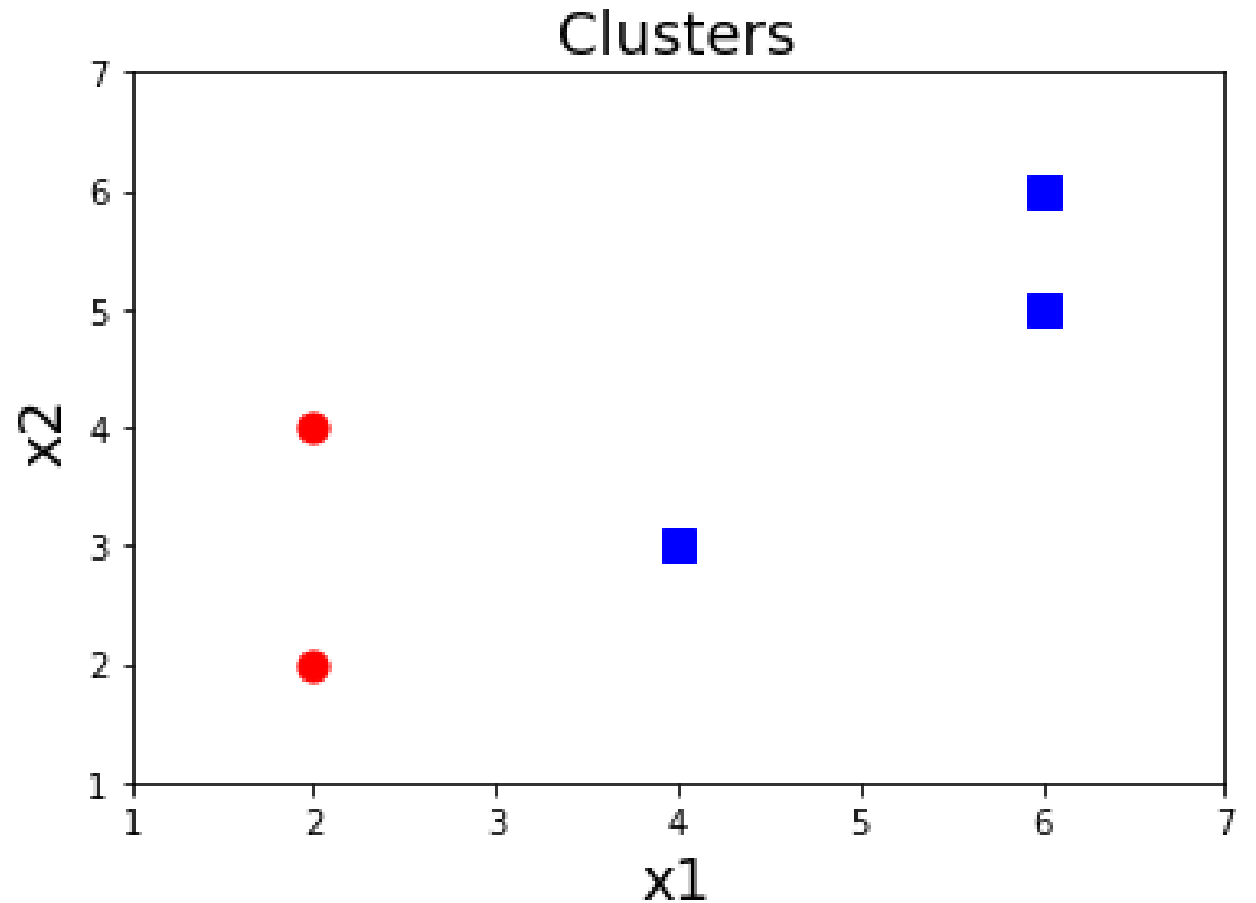
# Graphical illustration

- 4: Assign each point to its closest centroid



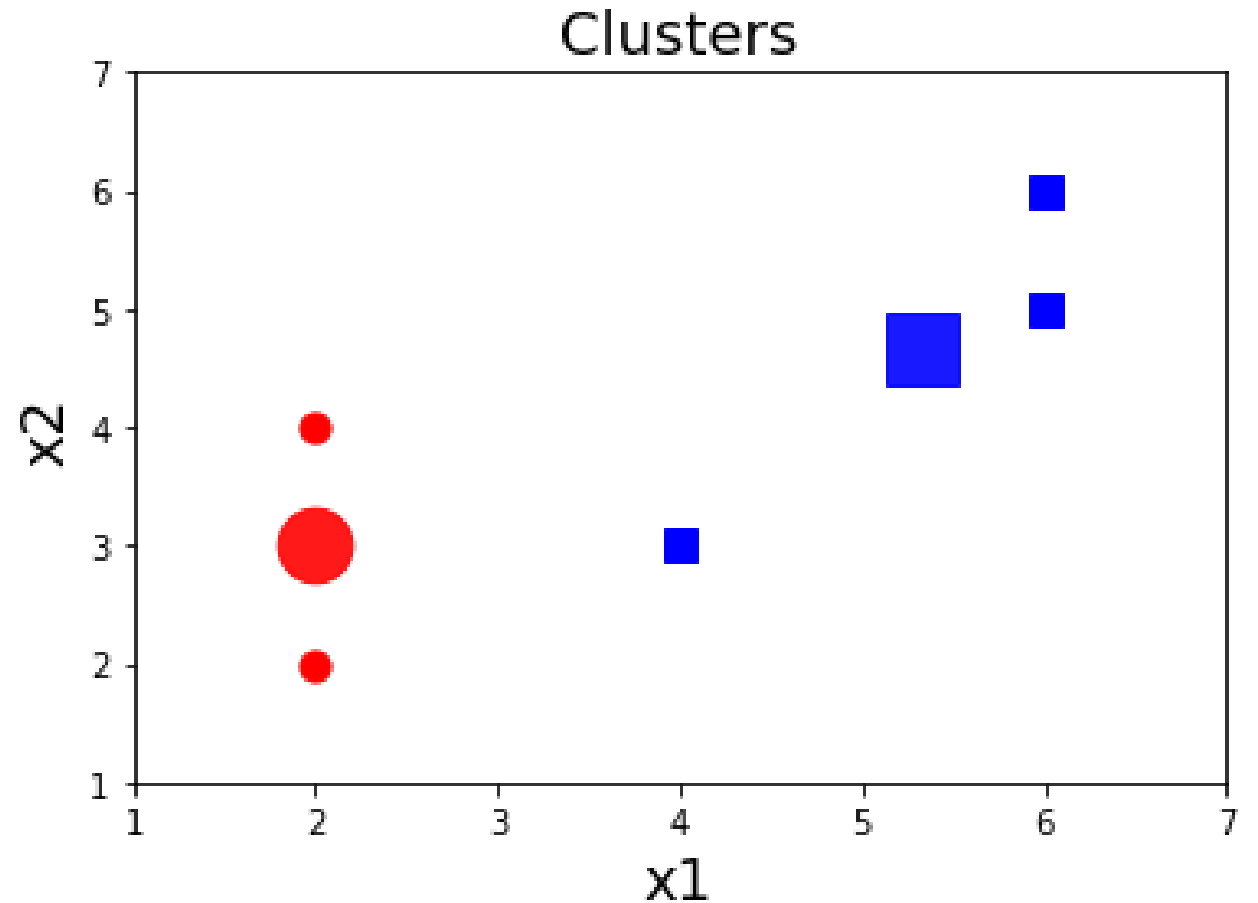
# Graphical illustration

- 4: Assign each point to its closest centroid



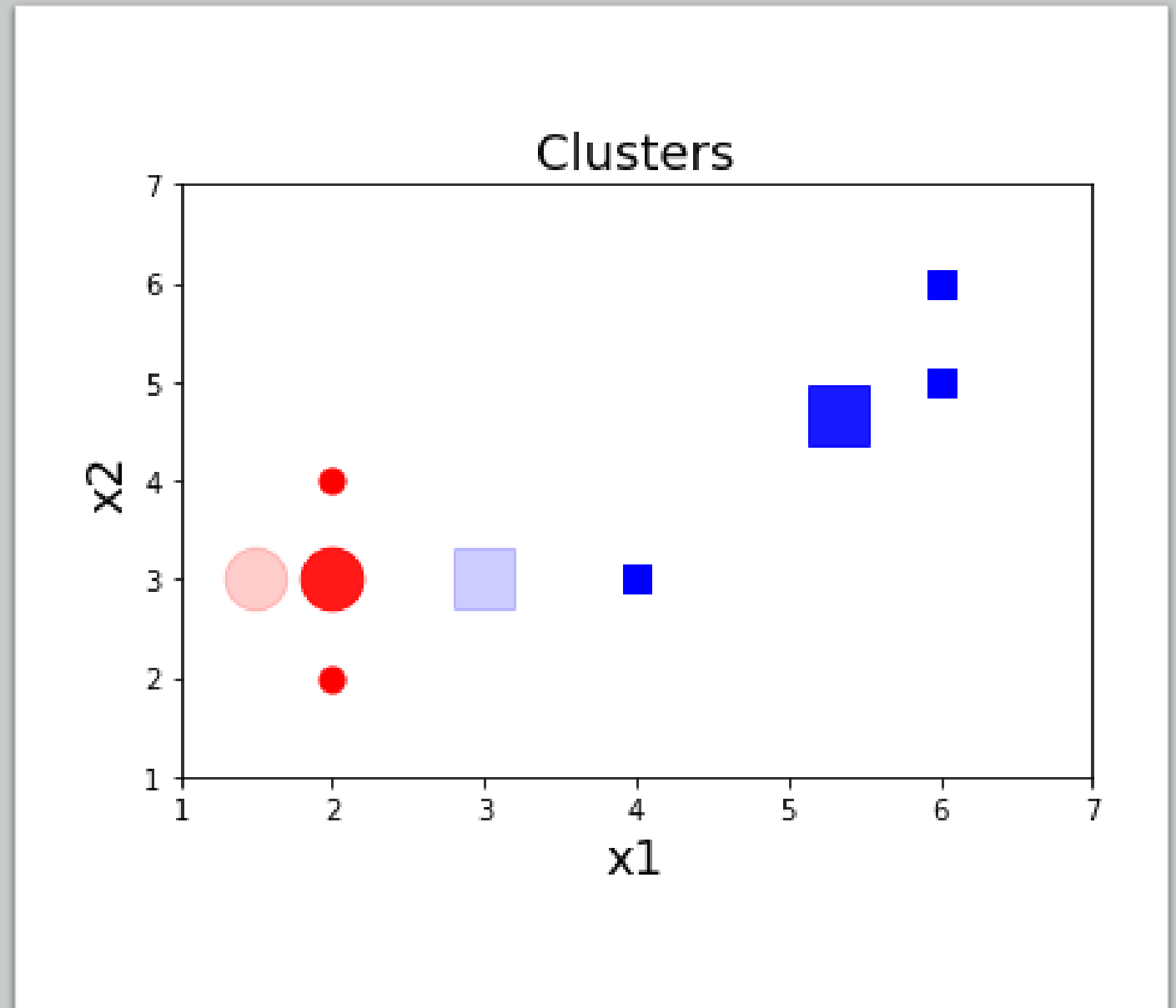
# Graphical illustration

- 5: Compute the new centroid (mean) of each cluster



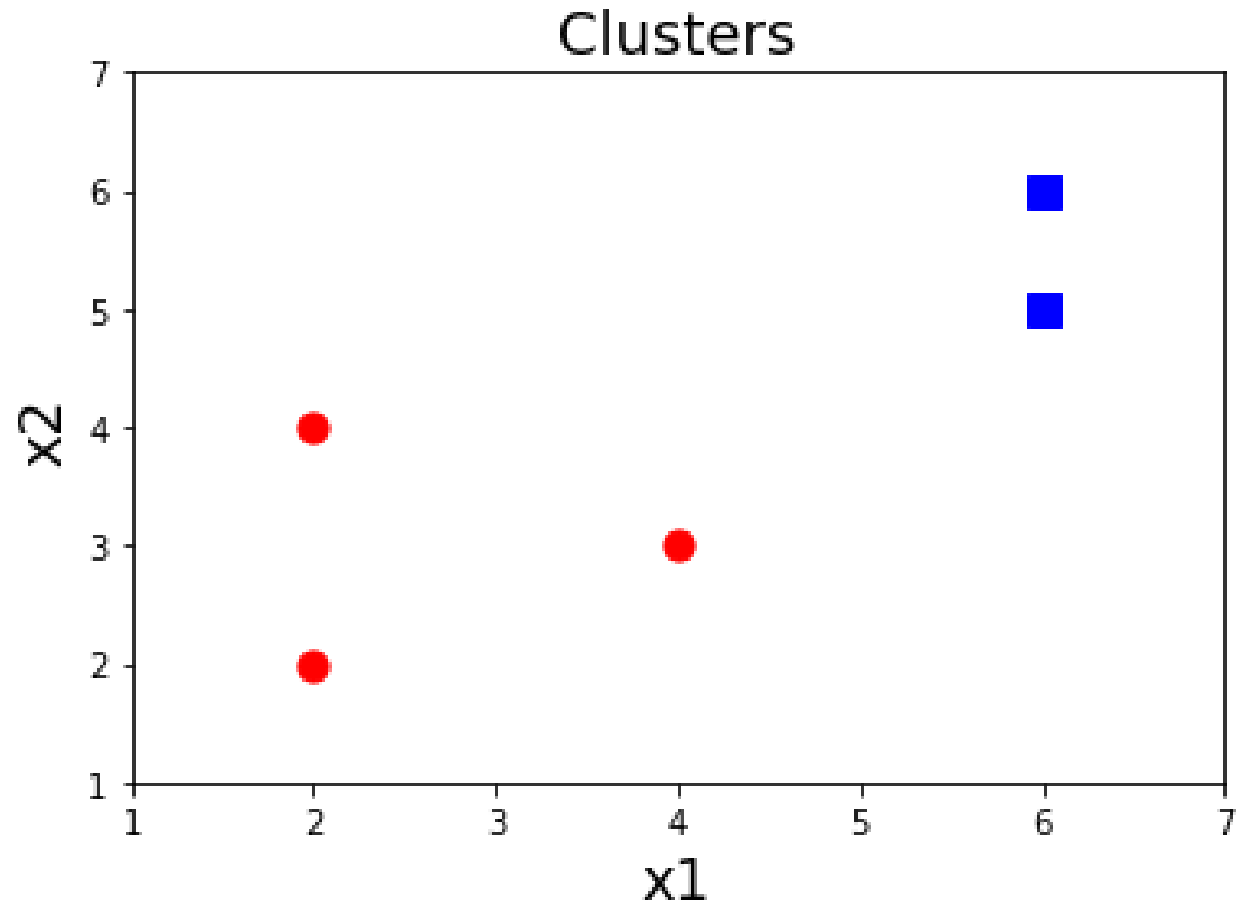
# Graphical illustration

- 6: Until the centroids do not change



# Graphical illustration

- Final solution



# How to choose the number of clusters K

- Elbow method gives us an idea on what a good k number of clusters would be based on the sum of squared distance (SSE) between data points and their assigned clusters' centroids.

