# Clustering problem



### Clustering problem

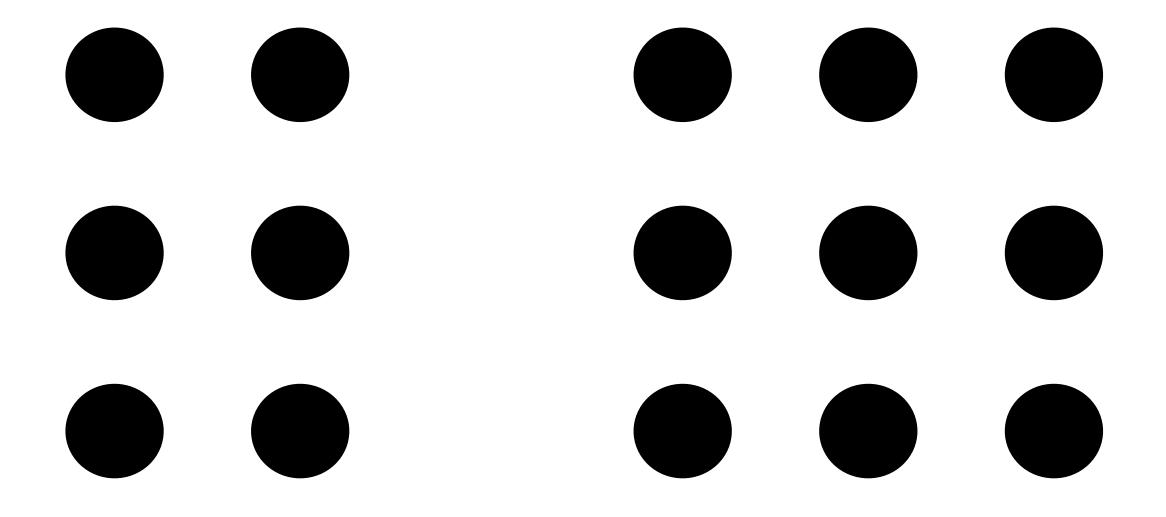
- Given a dataset, partition its objects into sets (clusters)  $C_1, C_2, \ldots, C_k$  such that the objects in each cluster are "similar" to one another.
- Specific definitions depend on how the notion of similarity is defined.

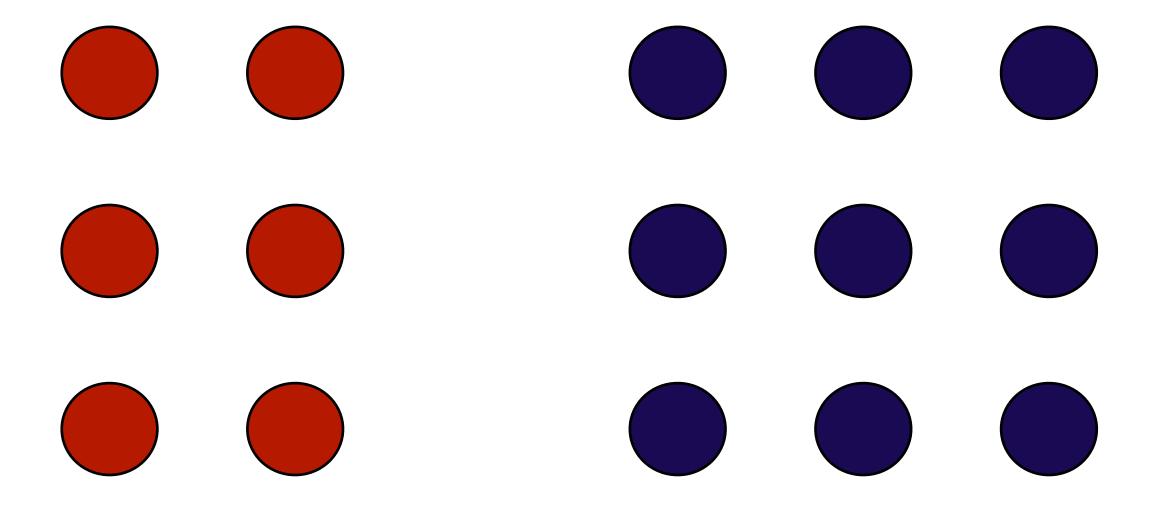
## Why cluster data?

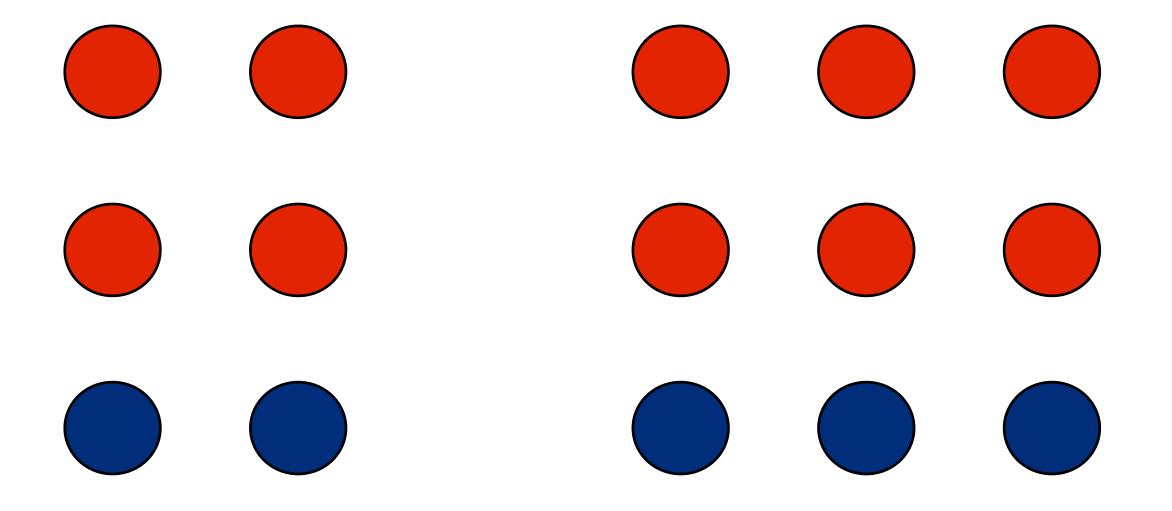
- Data summarization
- Topic detection
- Visualisation
- Outlier detection
- Community detection

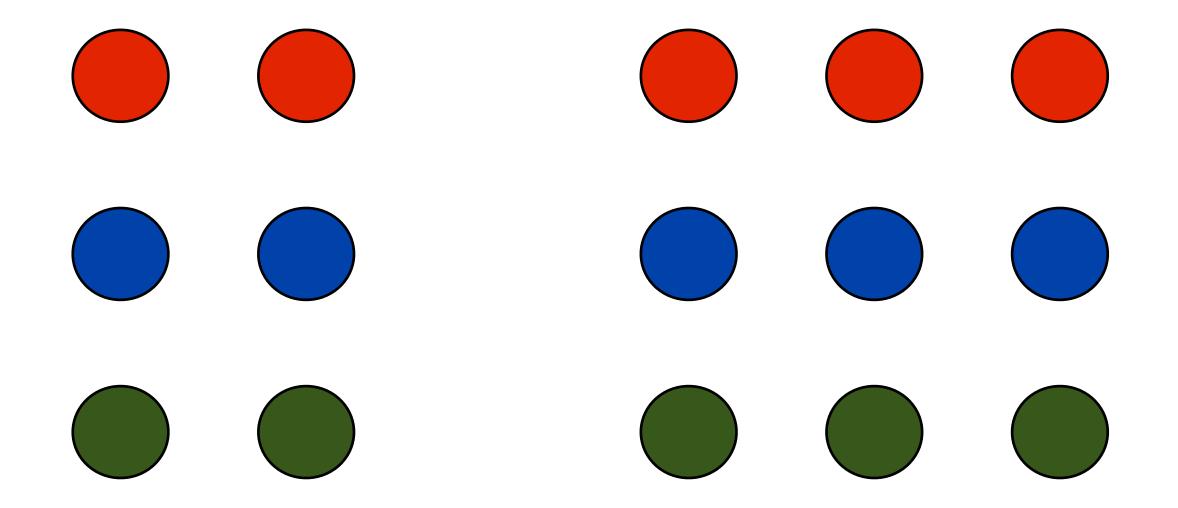
### Clustering is unsupervised learning

- Supervised learning
  - Labels for training instances are provided
- Unsupervised learning
  - No labels for training instances are provide
- Semi-supervised learning
  - Both labeled and unlabeled training instances are provided
- What can we learn about training data if we do not have any labels?
  - The similarity and distribution of the features can still be learned and this can be used to create rich feature spaces for supervised learning (if required)









How many clusters?

#### General Remarks

- A single dataset can be clustered in several ways
- There is no single right or wrong clustering
  - Simply different views on the same data
- Then how can we measure the quality of a clustering algorithm?
  - Extrinsic methods: Compare the clusters produced by a clustering algorithm against some reference (gold standard or ground truth) set of clusters
  - Intrinsic methods: only the partition of objects into clusters is used

### Clustering Algorithms

#### Representative-based

- ullet Choose k representatives, assign each element in the dataset to a representative, and iteratively update the partition
  - k-Means, k-Medoids

#### Hierarchical

- Create a hierarchy of clusters (dendrogram)
  - Agglomerative clustering (bottom-up)
  - Conglomerative clustering (top-down)

#### Graph-based clustering

- Community detection (Modularity optimisation)
- Graph-cut algorithms (Spectral Clustering)

#### Many other types