

### C\$ 412 Intro. to Data Mining

Chapter 10. Cluster Analysis: Basic Concepts and Methods



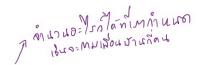
### **What Is Cluster Analysis?**

- What is a cluster?
  - □ A cluster is a collection of data objects which are
    - ☐ Similar (or related) to one another within the same group (i.e., cluster)
    - □ Dissimilar (or unrelated) to the objects in other groups (i.e., clusters)
- □ Cluster analysis (or clustering, data segmentation, ...)
  - ☐ Given a set of data points, partition them into a set of groups (i.e., clusters) which are as similar as possible
- ☐ Cluster analysis is unsupervised learning (i.e., no predefined classes)
- ☐ This contrasts with classification (i.e., supervised learning)
- ☐ Typical ways to use/apply cluster analysis
  - ☐ As a stand-alone tool to get insight into data distribution, or
  - ☐ As a preprocessing (or intermediate) step for other algorithms

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#### What Is Good Clustering?

- ☐ A good clustering method will produce high quality clusters which should have
  - ☐ **High intra-class similarity: Cohesive** within clusters
- □ Low inter-class similarity: Distinctive between clusters
- Quality function
  - There is usually a separate "quality" function that measures the "goodness" of a cluster
  - It is hard to define "similar enough" or "good enough"
  - ☐ The answer is typically highly subjective
- ☐ There exist many similarity measures and/or functions for different applications
- ☐ Similarity measure is critical for cluster analysis

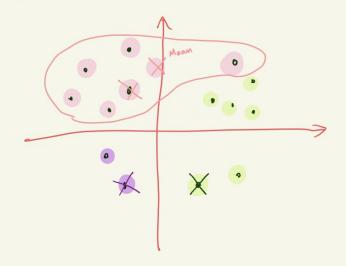


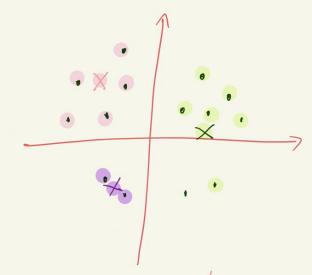
## The K-Means Clustering Method

- □ K-Means (MacQueen'67, Lloyd'57/'82)
  - Each cluster is represented by the center of the cluster
- ☐ Given K, the number of clusters, the K-Means clustering algorithm is outlined as กำแนก หล่อน ว่าเพจะเอาก็กล่ม follows mancorrin in the
  - □ Select K points as initial centroids
  - □ Repeat ทำหัวไปเรื่อย าลแกว่าล = ครบที่กำแส
    - ☐ Form K clusters by assigning each point to its closest centroid
    - Re-compute the centroids (i.e., mean point) of each cluster
  - □ Until convergence criterion is satisfied
- Different kinds of measures can be used

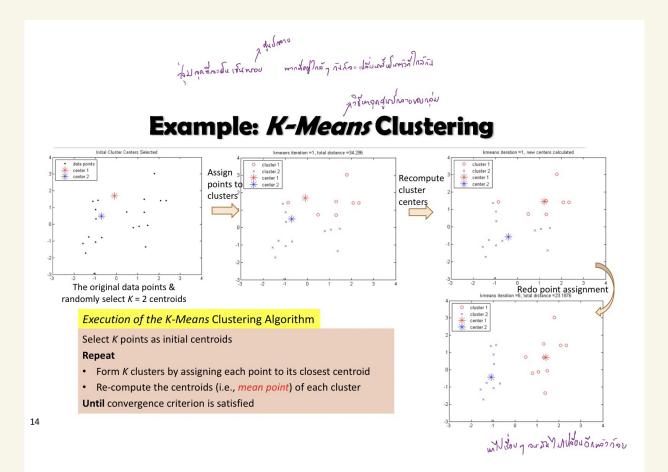
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Manhattan distance (L<sub>1</sub> norm), Euclidean distance (L<sub>2</sub> norm), Cosine similarity





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### Discussion on the K-Means Method

- **Efficiency**: O(tKn) where n: # of objects, K: # of clusters, and t: # of iterations
- □ Normally, *K*, *t* << *n*; thus, an efficient method
- K-means clustering often terminates at a local optimal
- ☐ Initialization can be important to find high-quality clusters
- **Need to specify** *K*, the *number* of clusters, in advance
- ☐ There are ways to automatically determine the "best" K
- ☐ In practice, one often runs a range of values and selected the "best" K value
- Sensitive to noisy data and outliers
- □ Variations: Using K-medians, K-medoids, etc.
- ☐ K-means is applicable only to objects in a continuous n-dimensional space
- ☐ Using the K-modes for *categorical data*
- Not suitable to discover clusters with *non-convex shapes* 
  - □ Using density-based clustering, kernel *K*-means, etc.

# Variations of *K-Means*

☐ There are many variants of the <i>K-Means</i> method, varying in different aspects
Choosing better initial centroid estimates ผ่าลักษณร์เคลาจ์เกาอบ
☐ K-means++, Intelligent K-Means, Genetic K-Means  To be discussed in this lecture
Choosing different representative prototypes for the clusters
□ K-Medoids, K-Medians, K-Modes To be discussed in this lecture
Applying feature transformation techniques
☐ Weighted K-Means, Kernel K-Means  To be discussed in this lecture