Chapter 7. Cluster Analysis

1. What is Cluster Analysis?



- 2. Types of Data in Cluster Analysis
- 3. A Categorization of Major Clustering Methods
- 4. Partitioning Methods
- 5. Hierarchical Methods
- 6. Density-Based Methods
- 7. Clustering High-Dimensional Data
- 8. Constraint-Based Clustering
- Outlier Analysis
- 10. Summary



What is Cluster Analysis?

- Cluster: a collection of data objects
 - Similar to one another within the same cluster
 - Dissimilar to the objects in other clusters
- Cluster analysis
 - Finding similarities between data according to the characteristics found in the data
 - Grouping similar data objects into clusters



What is Cluster Analysis?

- Unsupervised learning: no predefined classes
- Typical applications
 - As a stand-alone tool to get insight into data distribution
 - As a preprocessing step for other algorithms



Clustering: Rich Applications and Multidisciplinary Efforts

- Spatial Data Analysis
 - Detect spatial clusters or for other spatial mining tasks
- Economic Science (especially market research)
 - Identify customers whose behaviors are similar
- WWW
 - Cluster documents
 - Cluster Weblog data to discover groups of similar access patterns
- Image Processing & Pattern Recognition



Examples of Clustering Applications

Marketing:

- Help marketers discover distinct groups in their customer bases
- Use this knowledge to develop targeted marketing programs

Land use:

Identification of areas of similar land use in an earth observation database

Examples of Clustering Applications

Insurance:

 Identifying groups of motor insurance policy holders with a high average claim cost

City-planning:

 Identifying groups of houses according to their house type, value, and geographical location



Quality: What Is Good Clustering?

- A good clustering method will produce high quality clusters with
 - high <u>intra-class</u> similarity
 - low <u>inter-class</u> similarity
- The <u>quality</u> of a clustering result depends on both the similarity measure used by the method and its implementation
- The <u>quality</u> of a clustering method is also measured by its ability to discover some or all of the <u>hidden</u> patterns



Measure the Quality of Clustering

- Dissimilarity/Similarity metric: Similarity is expressed in terms of a distance function, typically metric: d(i, j)
- There is a separate "quality" function that measures the "goodness" of a cluster
- The definitions of distance functions
 - Usually very different for interval-scaled, Boolean, categorical, ordinal ratio, and vector variables
 - Weights should be associated with different variables based on applications and data semantics
- Hard to define "similar enough" or "good enough"
 - The answer is typically highly subjective



Requirements of Clustering in Data Mining

- Ability to deal with different types of attributes
- Ability to handle dynamic data
- Discovery of clusters with an arbitrary shape
- Minimal requirements for domain knowledge to determine input parameters
- Able to deal with noises and outliers
- Insensitive to the order of input records
- High dimensionality
- Scalability
- Incorporation of user-specified constraints



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