What is Clustering?



Organizing data into clusters such that there is:

- √ High intra-cluster similarity
- ✓ Low inter-cluster similarity
- ✓ Informally, finding natural groupings among objects.



Why do we want to do it??

Why Clustering?

- Organizing data into clusters shows internal structure of the data
 Ex. Clusty and clustering genes
- ✓ Sometimes the partitioning is the goal Ex. Market segmentation
- ✓ Prepare for other AI techniques
 - Ex. Summarize news (cluster and then find centroid)
- √ Techniques for clustering is useful in knowledge
- √ Discovery in data

Ex. Underlying rules, reoccurring patterns, topics, etc.

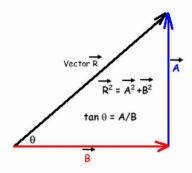
Clustering - Example

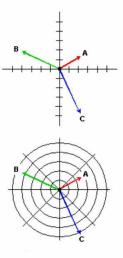
A sample news grouping from Google News:



A **vector** is a quantity or phenomenon that has two independent properties: magnitude and direction.

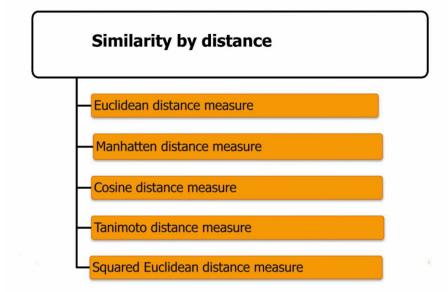
The term also denotes the mathematical or geometrical representation of such a quantity.





Similarity measurement definition —Similarity by Correlation —Similarity by Distance

Distance measures



Euclidean distance measure

Mathematically, Euclidean distance between two n-dimensional vectors

$$d = \sqrt{(a_1 - b_1)^2 + (a_2 - b_2)^2 + ... + (a_n - b_n)^2}$$

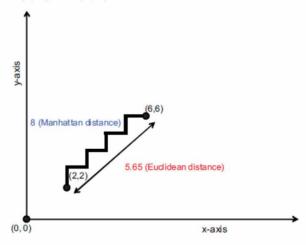
Manhatten distance measure

Mathematically, the Manhattan distance between two n-dimensional vectors

$$d = |a1 - b1| + |a2 - b2| + ... + |an - bn|$$

Difference between Euclidean and Manhattan

From this image we can say that, The Euclidean distance measure gives 5.65 as the distance between (2, 2) and (6, 6) whereas the Manhattan distance is 8.0



Cosine distance measure

The formula for the cosine distance between *n*-dimensional vectors (a1, a2, ..., an) and (b1, b2, ...,bn) is

$$d = 1 - \frac{(a_1b_1 + a_2b_2 + \dots + a_nb_n)}{(\sqrt{(a_1^2 + a_2^2 + \dots + a_n^2)}\sqrt{(b_1^2 + b_2^2 + \dots + b_n^2)})}$$

Tanimoto distance measure

The formula for the Tanimoto distance between two n-dimensional vectors (a1, a2, ..., an) and (b1, b2, ..., bn) is

$$d = 1 - \frac{(a_1|b_1 + a_2b_2 + ... + a_nb_n)}{\sqrt{(a_1^2 + a_2^2 + ... + a_n^2)} + \sqrt{(b_1^2 + b_2^2 + ... + b_n^2)} - (a_1b_1 + a_2b_2 + ... + a_nb_n)}$$