

## Distance measures for clustering

- What is the distance between two objects (rows in the data matrix)?
- (Earlier we talked about distances between sets of objects, given the distance between objects; now we look at the distance between objects.)
- Examples: objects are documents; what is the distance between two documents
- Objects are gene sequences (strings in ACGT); what is the distance?
- Euclidean distances vs. non-Euclidean distances (points in space vs. some other properties)

## Desired properties of a distance measure

- Objects  $X$ , distance measure  $d : X \times X \rightarrow R$
- Should be a metric
- $d(x, y) \geq 0$  for all  $x$  and  $y$
- $d(x, y) = 0$  if and only if  $x = y$
- $d(x, y) = d(y, x)$
- $d(x, z) \leq d(x, y) + d(y, z)$  for all  $x, y, z$  (triangle inequality)

## Examples of metrics

- Assume points are in  $d$ -dimensional space
- $x = (x_1, x_2, \dots, x_d)$  and  $y = (y_1, y_2, \dots, y_d)$
- Normal Euclidean distance ( $L_2$  distance)

$$d_2(x, y) = \sqrt{\sum_{i=1}^d (x_i - y_i)^2}.$$

- This is the  $L_2$  norm of  $x - y$ , also written as  $\|x - y\|_2$
- $L_p$  distance  $p > 0$

$$d_p(x, y) = \left( \sum_{i=1}^d (x_i - y_i)^p \right)^{1/p}$$

- What is  $L_1$  distance? What is  $L_\infty$  distance?

## Distance and similarity

- Similarity  $s(x, y) = M - d(x, y)$ , where  $M$  is the maximal value the distance can obtain
- About equivalent concepts
- Not always

### Other examples for vectors of 0s and 1s

- Jaccard distance: view  $x = (x_1, x_2, \dots, x_d) \in \{0, 1\}^d$  and  $y = (y_1, y_2, \dots, y_d) \in \{0, 1\}^d$  as sets

- 

$$J(x, y) = 1 - \frac{|x \cap y|}{|x \cup y|}.$$

- Cosine distance: view  $x$  and  $y$  as vectors, similarity between the vectors is measured as the cosine of the angle between the vectors

$$s(x, y) = \frac{x \cdot y}{\|x\|_2 \|y\|_2}$$

How does this behave?

### Distance notions for strings

- How close are two strings to each other?
- Edit distance
- Select a set of operations, say insert a character, delete a character, substitute one character for another
- $d(s, t)$ : the smallest number of operations that is needed to transform string  $s$  into string  $t$
- This is a metric (why?)
- Can be computed efficiently

### Similarity for columns

- Similarity for columns of a 0-1 matrix  $M$

- Jaccard coefficient for columns  $A$  and  $B$ :

$$J(A, B) = \frac{|\{t \in M | t(A) = 1 \wedge t(B) = 1\}|}{|\{t \in M | t(A) = 1 \vee t(B) = 1\}|}$$

- How many rows  $t$  of  $M$  have both  $A$  and  $B$ , divided by how many rows have at least one of  $A$  and  $B$
- $|\{t \in M | t(A) = 1 \vee t(B) = 1\}| = f(A) + f(B) - f(AB)$
- Thus  $J(A, B) = f(AB) / (f(A) + f(B) - f(AB))$
- High if  $A$  and  $B$  are strongly associated
- Alternatives: correlation between variables  $A$  and  $B$

### Other measures of similarity

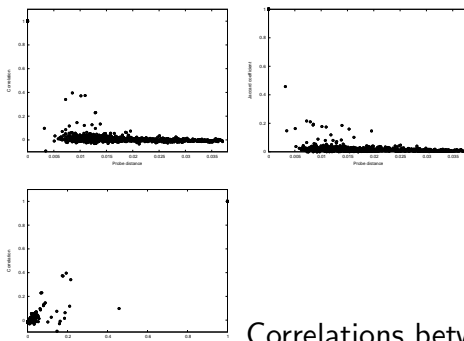
- Two columns  $A$  and  $B$  might actually be similar even if there are only a few rows with  $A = B = 1$
- Example: retail data, two soft drinks  $A$  and  $B$
- $A$  and  $B$  are similar in the sense that the behavior of the customers who buy  $A$  is about the same as the behavior of the customers who buy  $B$
- Can be very few customers who buy both  $A$  and  $B$
- How to formalize this?

## Probe similarity

- Define the distance of  $A$  and  $B$  by the difference in the probabilities of the other variables
- $Pr(C|A = 1)$ : what fraction of the customers who buy  $A$  also buy  $C$  (conditional probability)
- $Pr(C|A = 1) = f(AC)/f(A)$
- Let  $U$  be the set of attributes, and  $U_{AB} = U \setminus \{A, B\}$
- $d(A, B) = \sum_{C \in U_{AB}} |Pr(C|A = 1) - Pr(C|B = 1)|$

G. Das, H. Mannila, P. Ronkainen, KDD 1998

## Example: retail.over1000cols.01



Probe	Corr	Jaccard
1.00000	-0.36364	-0.34927
-0.36364	1.00000	0.98819
-0.34927	0.98819	1.00000

## Contextual similarity

- “Two words are similar, if they occur in similar sentences.”
- “Two sentences are similar, if they contain similar words.”
- How to implement this?