

Tutorial 1

1. There are a number of ways to compare two objects (vectors) \mathbf{x} and \mathbf{y} that consist of n binary attributes. The comparison of two such vectors leads to the following four quantities:

f_{00} = the number of attributes with a value of 0 in both \mathbf{x} and \mathbf{y} .

f_{01} = the number of attributes with a value of 0 in \mathbf{x} and 1 in \mathbf{y} .

f_{10} = the number of attributes with a value of 1 in \mathbf{x} and 0 in \mathbf{y} .

f_{11} = the number of attributes with a value of 1 in both \mathbf{x} and \mathbf{y} .

Based on these quantities, we can define the following two measures:
Simple Matching Coefficient (SMC):

$$SMC = \frac{f_{11} + f_{00}}{f_{01} + f_{10} + f_{11} + f_{00}}$$

Jaccard coefficient

$$J = \frac{f_{11}}{f_{01} + f_{10} + f_{11}}$$

- a. Calculate the value of the Simple Matching Coefficient and the Jaccard coefficient for the two vectors $\mathbf{x}=(1,0,0,0,0,1,1,0)$ and $\mathbf{y}=(0,0,1,0,1,0,1,0)$.
 - b. What is the main difference between these two measures?
2. The cosine similarity for two vectors \mathbf{x} and \mathbf{y} with continuous attributes is defined as follows:

$$\cos(\mathbf{x}, \mathbf{y}) = \frac{\mathbf{x} \cdot \mathbf{y}}{\|\mathbf{x}\| \|\mathbf{y}\|}$$

where \cdot indicates the dot product between two vectors, $\sum_{k=1}^n x_k y_k$ (x_k and y_k are the k -th attributes of \mathbf{x} and \mathbf{y} respectively), and $\|\mathbf{x}\|$ is the length of vector \mathbf{x} ,

$$\|\mathbf{x}\| = \sqrt{\sum_{k=1}^n x_k^2} = \sqrt{\mathbf{x} \cdot \mathbf{x}}$$

- a. Calculate the value of the cosine similarity for the two vectors $\mathbf{x} = (3,5,0,1,0,1)$ and $\mathbf{y} = (2,6,0,2,3,0)$.

- b. If two vectors have a cosine similarity of 1, are they identical?
 - c. What is the geometric interpretation of the cosine similarity?
3. We consider the problem of document data analysis. Let t_{ij} be the frequency of the i -th word (term) in the j -th document and m be the number of documents. Consider the variable transformation defined by

$$t'_{ij} = t_{ij} \log \frac{m}{n_i}$$

where n_i is the number of documents in which the i -th term appears.

- a. What is the effect of this transformation if a term occurs in one document?
In every document?
- b. What is the purpose of this transformation?