

Example (TD matrix, VSM model)

Let

$D = \{D_1, \dots, D_j, \dots, D_m\}$ denote the documents,

$T = \{t_1, \dots, t_i, \dots, t_n\}$ denote the terms constructed.

matrix $TD = (w_{ij})_{n \times m}$, ($i = 1, \dots, n, j = 1, \dots, m$)

where

w_{ij} the *weight* of term t_i in the document D_j

f_{ij} : the number of times term t_i occurs in document D_j ,

m is the number of documents

F_i is the number of documents where t_i occurs

$$TD := \begin{pmatrix} w_{1,1} & \blacksquare & \blacksquare & w_{1,j} & \blacksquare & \blacksquare & w_{1,m} \\ \blacksquare & \blacksquare & \blacksquare & \blacksquare & \blacksquare & \blacksquare & \blacksquare \\ \blacksquare & \blacksquare & \blacksquare & \blacksquare & \blacksquare & \blacksquare & \blacksquare \\ \blacksquare & \blacksquare & \blacksquare & \blacksquare & \blacksquare & \blacksquare & \blacksquare \\ w_{i,1} & \blacksquare & \blacksquare & w_{i,j} & \blacksquare & \blacksquare & w_{i,m} \\ \blacksquare & \blacksquare & \blacksquare & \blacksquare & \blacksquare & \blacksquare & \blacksquare \\ \blacksquare & \blacksquare & \blacksquare & \blacksquare & \blacksquare & \blacksquare & \blacksquare \\ \blacksquare & \blacksquare & \blacksquare & \blacksquare & \blacksquare & \blacksquare & \blacksquare \\ w_{n,1} & \blacksquare & \blacksquare & w_{n,j} & \blacksquare & \blacksquare & w_{n,m} \end{pmatrix}$$

Collection of book titles with the index terms

($n = 9$ terms, $m = 7$ documents)

Documents:

$D_1 = "$ Infant and Toddler First Aid $"$

$D_2 = "$ Babies and Children's Room (For your Home) $"$

$D_3 = "$ Child Safety at Home $"$

$D_4 = "$ Your Baby's Health and Safety: From Infant to Toddler $"$

$D_5 = "$ Baby Proofing Basics $"$

$D_6 = "$ Your Guide to Easy Rust Proofing $"$

$D_7 = "$ Beanie Babies Collectors Guide $"$

Terms:

$t_1 = "$ Baby $"$

$t_2 = "$ Child $"$

$t_3 = "$ Guide $"$

$t_4 = "$ Health $"$

$t_5 = "$ Home $"$

$t_6 = "$ Infant $"$

$t_7 = "$ Proofing $"$

$t_8 = "$ Safety $"$

$t_9 = "$ Toddler $"$

The *term-by-document* matrix
and the *term-by-query* are the following

using **binary weighting method**:

$$w_{ij} = \begin{cases} 1 & \text{if } t_i \text{ occurs in } D_j \\ 0 & \text{otherwise} \end{cases},$$

or **frequency weighting method**:

$$w_{ij} = f_{ij}.$$

e.g., f_{17} : the number of times term t_1 (*Baby*) occurs in
document D_7 (*Beanie Babies Collectors Guide*),

$$D := \begin{pmatrix} 0 & 1 & 0 & 1 & 1 & 0 & 1 \\ 0 & 1 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 1 & 1 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 1 & 0 \\ 0 & 0 & 1 & 1 & 0 & 0 & 0 \\ 1 & 0 & 0 & 1 & 0 & 0 & 0 \end{pmatrix} \quad Q := \begin{pmatrix} 0 \\ 1 \\ 0 \\ 0 \\ 0 \\ 1 \\ 1 \\ 1 \\ 1 \\ 0 \end{pmatrix}$$

**The *term-by-document* matrix
and the *term-by-query* are the following:**

using *norm-tf*, length-normalized method:

$$w_{ij} = \frac{f_{ij}}{\sqrt{\sum_{k=1}^n f_{kj}^2}} .$$

D = **■**

$$q := \begin{pmatrix} 0 \\ 0.4472 \\ 0 \\ 0 \\ 0.4472 \\ 0.4472 \\ 0.4472 \\ 0.4472 \\ 0 \end{pmatrix}$$

Similarity measures

Cosine measure:

$$\sigma(\mathbf{w}_j, \mathbf{q}) = \frac{\sum_{i=1}^n w_{ij} q_i}{\sqrt{\sum_{i=1}^n w_{ij}^2 \cdot \sum_{i=1}^n q_i^2}}$$

Dice's coefficient:

$$\sigma(\mathbf{w}_j, \mathbf{q}) = 2 \cdot \frac{\sum_{i=1}^n w_{ij} q_i}{\sum_{i=1}^n (w_{ij} + q_i)}$$

Jaccard's coefficient:

$$\sigma(\mathbf{w}_j, \mathbf{q}) = \frac{\sum_{i=1}^n w_{ij} q_i}{\sum_{i=1}^n \left(\frac{w_{ij} + q_i}{2^{w_{ij} q_i}} \right)}$$

Doc.	Similarity values (Rank)		
	Cosine	Jaccard	Dice
D1	0.316 (4.)	0.092 (4.)	0.174 (4.)
D2	0.516 (2.)	0.142 (2.)	0.26 (2.)
D3	0.775 (1.)	0.224 (1.)	0.39 (1.)
D4	0.4 (3.)	0.094 (3.)	0.178 (3.)
D5	0.316 (4.)	0.092 (4.)	0.174 (4.)
D6	0.316 (4.)	0.092 (4.)	0.168 (4.)
D7	0	0	0

EXAMPLE (TD Matrix, VSM model)

Let the set T of index terms be

$$T = \{t_1, t_2, t_3\} = \{ \\ t_1 = \text{Bayes}, \\ t_2 = \text{probability}, \\ t_3 = \text{epistemology} \\ \}.$$

Conceive the documents as sets of terms
(together with their frequencies):

$$D = \{D_1, D_2, D_3\}, \text{ where}$$

$$D_1 = \{ (\text{Bayes}, 1); (\text{probability}, 1); (\text{epistemology}, 0) \}$$

$$D_2 = \{ (\text{Bayes}, 2); (\text{probability}, 1); (\text{epistemology}, 0) \}$$

$$D_3 = \{ (\text{Bayes}, 3); (\text{probability}, 3); (\text{epistemology}, 3) \}$$

Let the query Q be:

$$Q = \{ \text{What is Bayesian epistemology?} \}$$

Let the query Q be (as a set of terms):

$$Q = \{ (t_1 = \text{Bayes}); (t_3 = \text{epistemology}) \}$$

**The *Term-by-Document matrix TD*,
and *term-by-query Q* :**

$TD_{3 \times 3} = (w_{ij})$, where

- Using *Frequency Weighting Method*:

$$TD = \begin{pmatrix} 1 & 2 & 3 \\ 1 & 1 & 3 \\ 0 & 0 & 3 \end{pmatrix} \quad Q = \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}$$

- Using *Binary Weigthing Method*

$$TD = \begin{pmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 0 & 0 & 1 \end{pmatrix} \quad Q = \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}$$

- Using *Binary Length Normalized Weigthing Method*:

$$TD = \begin{pmatrix} \frac{\sqrt{2}}{2} & \frac{\sqrt{2}}{2} & \frac{\sqrt{3}}{3} \\ \frac{2}{\sqrt{2}} & \frac{2}{\sqrt{2}} & \frac{3}{\sqrt{3}} \\ 2 & 2 & \frac{3}{\sqrt{3}} \\ 0 & 0 & \frac{\sqrt{3}}{3} \end{pmatrix} \quad Q = \begin{pmatrix} \frac{\sqrt{2}}{2} \\ 2 \\ 0 \\ \frac{\sqrt{2}}{2} \\ 2 \end{pmatrix}$$

Similarity values:

- *Dot Product:*

$$Dot_1 = 0.5; Dot_2 = 0.5; Dot_3 = 0.816$$

- *Cosine Measure:*

$$Cosine_1 = 0.5; Cosine_2 = 0.5; Cosine_3 = 0.816$$

- *Dice Measure:*

$$Dice_1 = 0.354; Dice_2 = 0.354; Dice_3 = 0.519$$

- *Jaccard Measure:*

$$Jaccard_1 = 0.21; Jaccard_2 = 0.21; Jaccard_3 = 0.32$$

EXAMPLE (different ranking order)

Let

$$D = \{D_1, D_2, D_3, D_4, D_5, D_6\},$$

$$T = \{t_1, t_2, t_3, t_4, t_5, t_6, t_7, t_8, t_9\}$$

Doc.	Similarity values (Rank)		
	Cosine	Jaccard	Dice
D1	0.676 (3.)	0.149 (4.)	0.276 (4.)
D2	0.845 (1.)	0.191 (3.)	0.347 (2-3.)
D3	0.632 (4.)	0.198 (2.)	0.347 (2-3.)
D4	0.775 (2.)	0.224 (1.)	0.39 (1.)
D5	0.258 (6.)	0.068 (6.)	0.13 (6.)
D6	0.316 (5.)	0.092 (5.)	0.173 (5.)