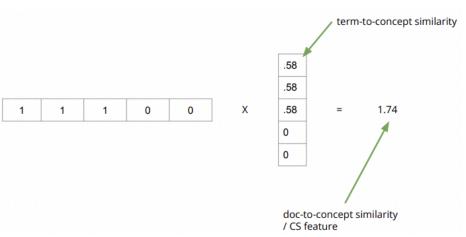
Latent Semantic Analysis

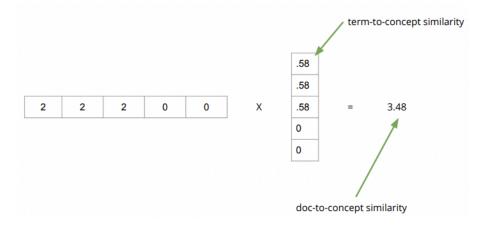
- 1. Latent Semantic Analysis
 - a. Inputs are documents. Each word is a feature. We can represent each document by
 - i. The presence of each word (0/1)

	data	information	retrieval	brain	lung
CS-paper-1	1	1	1	0	0



ii. Count of the word (0, 1, ...)

	data	information	retrieval	brain	lung
CS-paper-1	2	2	2	0	0



2. Latent Semantic Analysis

	data	information	retrieval	brain	lung	
CS-paper-1	1	1	1	0	0	
CS-paper-2	2	2	2	0	0	
CS-paper-3	1	1	1	0	0	
CS-paper-4	5	5	5	0	0	
Med-paper-1	0	0	0	2	2	
Med-paper-2	0	0	0	3	3	
Med-paper-3	0	0	0	1	1	

a.

b.

1	1	1	0	0		0.18	0									
2	2	2	0	0		0.36	0									
1	1	1	0	0		0.18	0		0.64	0		0.50	0.50	0.50		
5	5	5	0	0	=	0.90	0	v	9.64	0	v	0.58	0.58	0.58	0	0
5	3	9	U	U	_	0.90	U	X	0	5.29	X	0	0	0	0.71	0.7
0	0	0	2	2		0	0.53			0.20					0.7 1	0
0	0	0	3	3		0	0.80									
0	0	0	1	1		0	0.27									

doc-to-concep

doc-to-concept similarity matrix

0.18	0										
0.36	0	"	strengtl each c			term-to-concept similarity matrix					
0.18	0				1			IIIatiix			
			9.64	0		0.58	0.58	0.58	0	0	
0.90	0	X		5.29	X	_	_	_	0.74	0.74	
0	0.53		0	5.29		0	0	0	0.71	0.71	
_	0.00										
0	0.80										
0	0.27										

C.

- d. Each document can be better represented by
 - i. Frequency of the word ($n_i / \Sigma n_i$)
 - ii. TfiDf
 - 1. tf: term frequency in the document
 - 2. idf: log(number of documents / number of documents that contain the term)