Term-frequency * Inverse-document-frequency

Tuesday, August 8, 2023 7:53 PM

TF-IDF → TF × IDF

↓ ↓

F() × F() → Cantidad → Matriz (documento-término) (document-term) → $\begin{bmatrix} \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot \end{bmatrix}$ → SVM, NB, MLP

Term frequency : Cantidad de ocurrencias de una palabra w_i en el documento d_i

document frequency : Cantidad de documentos que contienen a la palabra w_i

CountVectorizer() → tfidf()

	and	document	first	is	one	second	the	third	this
D0 This is the first document.	0	$1 = \log(\frac{4}{3})$	1	1	0	0	1	0	1
D1 This document is the second document.	0	1	0	1	0	1	1	0	1
D2 And this is the third one.	1	0	0	1	1	0	1	1	1
D3 Is this the first document?	0	1	1	1	0	0	1	0	1
df_i	1	3	2	4	1	1	4	1	4
idf_i	$\log(\frac{4}{1})$	$\log(\frac{4}{3})$	$\log(\frac{4}{2})$	$\log(\frac{4}{4})$	. . . .				
$tf \times idf$									

(bag-of-words)

w_i = and

$tf = count(t_i, d_i)$

$idf = \log(\frac{N}{df_i})$ N: Cantidad de documentos en el corpus N=4

para "and" → $tf("and", D1) = 0$

$idf(\frac{4}{df_{and}}) = \log(\frac{4}{1})$

$0 \times \log(\frac{4}{1}) = 0$

Matrices con muchos ceros → sparse matrix

sin → dense matrix

TFIDF vs TF (Bag-of-words)

ponderaciones → enfoque por contador

Palabras:

menos repetidas: ↑

mas repetidas ↓

$$\text{idf}(t) = \log \frac{1+n}{1+\text{df}(t)} + 1, \quad \text{(sklearn)}$$

$$\log\left(\frac{1+4}{1+3}\right)+1 = \log\left(\frac{5}{4}\right)+1$$