

Day 1

Corpus

I am happy because I am learning NLP

I am happy

I am sad, I am not learning NLP

I am sad

Vocabulary

I

am

happy

because

learning

NLP

sad

not

Positive tweets

I am happy because I am learning NLP

I am happy

Negative tweets

I am sad, I am not learning NLP

I am sad

Positive tweets

I am happy because I am learning NLP

I am happy

Vocabulary	PosFreq (1)
I	3
am	3
happy	2
because	1
learning	1
NLP	1
sad	0
not	0

Vocabulary	NegFreq (0)
I	3
am	3
happy	0
because	0
learning	1
NLP	1
sad	2
not	1

Negative tweets

I am sad, I am not learning NLP

I am sad

How does Bag of Words work ?

Suppose we have three sentences :

Sentence 1 : the cat sat

Sentence 2 : the cat sat in the hat

Sentence 3 : the cat with the hat

After removing stop words we are left with :

Sentence 1 : cat sat cat

Sentence 2 : cat sat hat


Sentence 3 : cat hat

When we convert frequency table to vectors it is known as **BOW**

The f1,f2,f3 features will be independent features for the model building.



Words	Frequency
cat	4
sat	2
hat	2



	f1	f2	f3
	cat	sat	hat
Sent 1	2	1	0
Sent 2	1	1	1
Sent 3	1	0	1

Disadvantage of Bag of Words

In this technique, the words are given equal importance and does not have any semantic difference .

Like in our case , in **Sentence 1** cat and sat are represented by 1 .

We can not classify which word is important than the other in this technique so to tackle this problem we have another feature engineering technique which is **TF-IDF**

TF - IDF

TF-IDF stands for
Term Frequency – Inverse Document Frequency .

This is a technique to quantify a word in documents, we generally compute a weight to each word which signifies the importance of the word in the document .

This technique converts text into numerical values , giving more importance to some words by multiplying TF and IDF i.e final result is got by $TF * IDF$

How does TF- IDF work ?

Lets First understand TF i.e. **Term Frequency**

TF = No. of repetition of words in the sentence
/ No. of words in the sentence

So our , TF table will look like this for our example :

Sentence 1 : cat sat

Sentence 2 : cat sat hat

Sentence 3 : cat hat



	f1	f2	f3
	cat	sat	hat
Sent 1	1/2	1/2	0
Sent 2	1/3	1/3	1/3
Sent 3	1/2	0	1/2

How does TF- IDF work ?

Now understand IDF i.e. **Inverse Document Frequency**.

IDF = $\log(\text{No. of reviews} / \text{No. of reviews containing words})$

So our , IDF table will look like this for our example :

Sentence 1 : cat sat

Sentence 2 : cat sat hat

Sentence 3 : cat hat

Here , no. of documents means no.of sentences.



Words	IDF
cat	$\log(3/3) = 0$
sat	$\log(3/2)$
hat	$\log(3/2)$

How does TF- IDF work ?

Finally ,

TF * IDF table in our case is :

Here we can see that the word sat is given the most importance in Sentence 1 out of other words .

Similarly in Sentence 2 and 3 as well .



	f1	f2	f3
	cat	sat	hat
Sent 1	$\frac{1}{2} * 0 = 0$	$\frac{1}{2} * \log(3/2)$	$0 * \log(3/2) = 0$
Sent 2	$\frac{1}{3} * 0 = 0$	$\frac{1}{3} * \log(3/2)$	$\frac{1}{3} * \log(3/2)$
Sent 3	$\frac{1}{2} * 0 = 0$	$0 * \log(3/2) = 0$	$\frac{1}{2} * \log(3/2)$