



ALMA MATER STUDIORUM  
UNIVERSITÀ DI BOLOGNA  
CAMPUS DI FORLÌ

# 91258 / B0385

## Natural Language Processing

### Lesson 3. Vector Space Model

Alberto Barrón-Cedeño  
a.barron@unibo.it

09/10/2024

# Table of Contents

1. Current Status
2. Representations Revisited
3. More Basic Algebra

## Current Status

# Current Status

You know...

- what is natural language processing

*On your own, you have...*

*On your own, you (could) have...*

You can...

# Current Status

You know...

- what is natural language processing
- there are two main paradigms: rule-based and statistical

*On your own, you have...*

*On your own, you (could) have...*

You can...

# Current Status

You know...

- what is natural language processing
- there are two main paradigms: rule-based and statistical

*On your own, you have...*

- setup a Python development environment
  1. command line
  2. PyCharm or any other option (e.g., Eclipse)
  3. Google's Colab
- played with spacy and nltk

*On your own, you (could) have...*

You can...

# Current Status

You know...

- what is natural language processing
- there are two main paradigms: rule-based and statistical

*On your own, you have...*

- setup a Python development environment
  1. command line
  2. PyCharm or any other option (e.g., Eclipse)
  3. Google's Colab
- played with spacy and nltk

*On your own, you (could) have...*

- played with pandas (tutorato)
- found out what is **git** (and perhaps  $\text{\LaTeX}$  as well!)

You can...

# Current Status

You know...

- what is natural language processing
- there are two main paradigms: rule-based and statistical

*On your own, you have...*

- setup a Python development environment
  1. command line
  2. PyCharm or any other option (e.g., Eclipse)
  3. Google's Colab
- played with spacy and nltk

*On your own, you (could) have...*

- played with pandas (tutorato)
- found out what is **git** (and perhaps  $\text{\LaTeX}$  as well!)

You can...

- open a text file (Python intro)



# Current Status

You know...

- what is natural language processing
- there are two main paradigms: rule-based and statistical

*On your own, you have...*

- setup a Python development environment
  1. command line
  2. PyCharm or any other option (e.g., Eclipse)
  3. Google's Colab
- played with spacy and nltk

*On your own, you (could) have...*

- played with pandas (tutorato)
- found out what is **git** (and perhaps  $\text{\LaTeX}$  as well!)

You can...

- open a text file (Python intro)
- tokenise and normalise text

# Current Status

You know...

- what is natural language processing
- there are two main paradigms: rule-based and statistical

*On your own, you have...*

- setup a Python development environment
  1. command line
  2. PyCharm or any other option (e.g., Eclipse)
  3. Google's Colab
- played with spacy and nltk

*On your own, you (could) have...*

- played with pandas (tutorato)
- found out what is **git** (and perhaps  $\text{\LaTeX}$  as well!)

You can...

- open a text file (Python intro)
- tokenise and normalise text
- build some text representations

abhishek  
@abhi1thakur

...



10:23 AM · Sep 29, 2022 · Twitter for iPhone

39 Retweets 3 Quote Tweets 396 Likes



<https://twitter.com/abhi1thakur/status/1575400771541155842>

## Representations Revisited

# Representations Revisited

---

<sup>1</sup><https://www.nltk.org/>

<sup>2</sup><https://spacy.io>

# Representations Revisited

1. Use NLTK<sup>1</sup> or Spacy<sup>2</sup> to tokenise

---

<sup>1</sup><https://www.nltk.org/>

<sup>2</sup><https://spacy.io>

# Representations Revisited

1. Use NLTK<sup>1</sup> or Spacy<sup>2</sup> to tokenise
2. Use `.lower()` to casefold (ignore capitalisation)

---

<sup>1</sup><https://www.nltk.org/>

<sup>2</sup><https://spacy.io>

# Representations Revisited

1. Use NLTK<sup>1</sup> or Spacy<sup>2</sup> to tokenise
2. Use `.lower()` to casefold (ignore capitalisation)
3. Use Porter's stemmer to drop suffixes  
or use a lemmatiser to find the *actual* root of words

---

<sup>1</sup><https://www.nltk.org/>

<sup>2</sup><https://spacy.io>



# Representations Revisited

1. Use NLTK<sup>1</sup> or Spacy<sup>2</sup> to tokenise
2. Use `.lower()` to casefold (ignore capitalisation)
3. Use Porter's stemmer to drop suffixes  
or use a lemmatiser to find the *actual* root of words
4. Discard stopwords from the text\*

---

<sup>1</sup><https://www.nltk.org/>

<sup>2</sup><https://spacy.io>

# Representations Revisited

1. Use NLTK<sup>1</sup> or Spacy<sup>2</sup> to tokenise
2. Use `.lower()` to casefold (ignore capitalisation)
3. Use Porter's stemmer to drop suffixes  
or use a lemmatiser to find the *actual* root of words
4. Discard stopwords from the text\*
5. Build a vectorial representation\*

---

<sup>1</sup><https://www.nltk.org/>

<sup>2</sup><https://spacy.io>

# Stopwords

Common words in a language that occur with a high frequency, but carry much less substantive information about the meaning of a phrase (Lane et al., 2019, p. 51–54)

---

<sup>3</sup>For instance, from NLTK, sklearn, or <https://github.com/stopwords-iso> 

# Stopwords

Common words in a language that occur with a high frequency, but carry much less substantive information about the meaning of a phrase (Lane et al., 2019, p. 51–54)

**Alternative 1** Consider the most frequent tokens in a reference corpus as stopwords (remember Genesis from P4P?)

---

<sup>3</sup>For instance, from NLTK, sklearn, or <https://github.com/stopwords-iso>

# Stopwords

Common words in a language that occur with a high frequency, but carry much less substantive information about the meaning of a phrase (Lane et al., 2019, p. 51–54)

**Alternative 1** Consider the most frequent tokens in a reference corpus as stopwords (remember Genesis from P4P?)

**Alternative 2** Take an existing list of stopwords<sup>3</sup>

en	es	it
i	a	altri
me	ahora	certa
my	alli	della
it	cerca	nessuna
is	el	prima
do	es	quello
the	unas	solito
will	vez	va
other	yo	via

<sup>3</sup>For instance, from NLTK, sklearn, or <https://github.com/stopwords-iso>

# Stopwords

## Discarding stopwords

- They are the most frequent tokens in the documents
- Discarding them reduces the computational effort significantly

# Stopwords

## Discarding stopwords

- They are the most frequent tokens in the documents
- Discarding them reduces the computational effort significantly
- Typical size of a stopwords list: a few hundred words
- For some applications (e.g., **topic clustering**), they can be safely discarded
- For some others (e.g., **dialogue**) they cannot

# Stopwords

## Discarding stopwords

- They are the most frequent tokens in the documents
- Discarding them reduces the computational effort significantly
- Typical size of a stopwords list: a few hundred words
- For some applications (e.g., **topic clustering**), they can be safely discarded
- For some others (e.g., **dialogue**) they cannot

Stopwords have to be considered with a grain of salt  
(as everything in NLP)



# Vector representation

## BoW

- A text is represented as the bag (set) of its words
- It disregards grammar
- It disregards word order
- It (can) consider frequency

From (Lane et al., 2019, p. 41)

## More Basic Algebra

# x and y



<https://twitter.com/miniapeur/status/1710074831079690394>

# Dot product

Algebraically, it is the sum of the products of the corresponding entries of the two sequences of numbers  $a \cdot b$

$$a \cdot b = \sum_{i=1}^n a_i b_i$$

# Dot product

Algebraically, it is the sum of the products of the corresponding entries of the two sequences of numbers  $a \cdot b$

$$\begin{aligned} a \cdot b &= \sum_{i=1}^n a_i b_i \\ &= a_1 b_1 + a_2 b_2 + a_3 b_3 + \cdots + a_n b_n \end{aligned}$$

# Dot product

Algebraically, it is the sum of the products of the corresponding entries of the two sequences of numbers  $a \cdot b$

$$\begin{aligned} a \cdot b &= \sum_{i=1}^n a_i b_i \\ &= a_1 b_1 + a_2 b_2 + a_3 b_3 + \cdots + a_n b_n \end{aligned}$$

```
a = [1,2,3]
b = [3,4,6]
my_sum = 0
for i in range(len(a)):
    my_sum += a[i] * b[i]
```

There are better —more efficient— ways to compute the dot product!

# Dot product

Algebraically, it is the sum of the products of the corresponding entries of the two sequences of numbers  $a \cdot b$

$$\begin{aligned} a \cdot b &= \sum_{i=1}^n a_i b_i \\ &= a_1 b_1 + a_2 b_2 + a_3 b_3 + \cdots + a_n b_n \end{aligned}$$

```
a = [1,2,3]
b = [3,4,6]
my_sum = 0
for i in range(len(a)):
    my_sum += a[i] * b[i]
```

There are better —more efficient— ways to compute the dot product!  
Now, we can use the dot product to compare two documents ( $\sim$  similarity)

# Vector space model

“[...] an **algebraic** model for representing text documents (or more generally, items) as vectors [...]”<sup>4</sup>

---

<sup>4</sup>[https://en.wikipedia.org/wiki/Vector\\_space\\_model](https://en.wikipedia.org/wiki/Vector_space_model) 



# Vector space model

“[...] an **algebraic** model for representing text documents (or more generally, items) as vectors [...]”<sup>4</sup>

## Some applications

- Relevance rankings in keyword-based search
- Document clustering to “discover” structure and relations in a text collection

(not the SOTA for most tasks, but it's a *minimum viable product*)

---

<sup>4</sup>[https://en.wikipedia.org/wiki/Vector\\_space\\_model](https://en.wikipedia.org/wiki/Vector_space_model) 

# Vector space model

“[...] an **algebraic** model for representing text documents (or more generally, items) as vectors [...]”<sup>4</sup>

## Some applications

- Relevance rankings in keyword-based search
- Document clustering to “discover” structure and relations in a text collection

(not the SOTA for most tasks, but it's a *minimum viable product*)

</> Let us see it working

---

<sup>4</sup>[https://en.wikipedia.org/wiki/Vector\\_space\\_model](https://en.wikipedia.org/wiki/Vector_space_model) 

Tomorrow...

VADER

# References

Lane, H., C. Howard, and H. Hapkem  
2019. *Natural Language Processing in Action*. Shelter Island, NY:  
Manning Publication Co.