



# Introduction to NLP

## Chapter 2:

### Text Preprocessing

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## Video course

For those who can not attend the synchronous session or want to relearn this material, you can watch this video for this slide presentation: <https://youtu.be/DSjUpWnkdno>

## Outline

- 2.1 Why we need to preprocess our dataset?
- 2.2 Basic text preprocessing in text dataset
- 2.3 Hands-on: Basic text preprocessing using several NLP libraries (NLTK, Sastrawi, etc.) -> Will be presented using Notebook (Jupyter Notebook or Google Colab)

## 2.1 Why we need to preprocess our dataset?

### Definition and motivation

#### ◎ Definition

Text preprocessing is a process to make our dataset more “clean” so that more "ready" for feature extraction.

#### ◎ Motivation

- Almost all data in a real case is not clean and ready to use
- To reduce memory needs and fasten the computation
- To improve the model performance

## 2.2 Basic text preprocessing in text dataset

### Kind of basic text preprocessing

- ◎ Text Case Uniforming (Uppercasing OR Lowercasing)
- ◎ Emoji and Emoticon Normalization
- ◎ Contraction Normalization
- ◎ Digit Normalization
- ◎ Token Masking
- ◎ Text Cleaning
- ◎ Basic Text Normalization (Using Dictionary)
- ◎ Stop Words Removal
- ◎ Stemming and Lemmatization

## 2.2 Basic text preprocessing in text dataset

### Kind of basic text preprocessing:

#### Text case uniforming (uppercasing or lowercasing)

##### Definition

Text case uniforming is a preprocessing step to convert all characters into the same cases, whether to lowercase or uppercase (almost all NLP research is using lowercase, especially when using a pre-trained word embedding).

##### Example

Lowercasing			Uppercasing	
Before	After		Before	After
Learn NLP	learn nlp		Learn NLP	LEARN NLP
learn NLP			learn NLP	
learn nlp			learn nlp	

## 2.2 Basic text preprocessing in text dataset

### Kind of basic text preprocessing: Emoji and emoticon normalization

#### Definition

Emoji and emoticon normalization is a preprocessing step to convert emoji and emoticon into a “verbal” form.

#### Example

Emoji Normalization		
Before	After (1)	After (2)
♥	heart_suit	heart_suit
💔	heavy_heart_exclamation	

Emoticon Normalization		
Before	After (1)	After (2)
:)	happy_face_or_smiley	happy_face_smiley
:-)	happy_face_smiley	



## 2.2 Basic text preprocessing in text dataset

### Kind of basic text preprocessing:

#### Contraction normalization

##### ◎ Definition

Contraction normalization is a preprocessing step to convert the contraction into the more formal one (just case in several languages).

##### ◎ Example

Contraction Normalization	
Before	After
i've	i have
you're	you are
wanna	want to

## 2.2 Basic text preprocessing in text dataset

### Kind of basic text preprocessing:

#### Digit normalization

#### Definition

Digit normalization is a preprocessing step to convert a digit into a “verbal” form (usually used in text-to-speech task).

#### Example

Digit Normalization	
Before	After
11	eleven
1052	one thousand fifty two
103.5	one hundred and three points five

## 2.2 Basic text preprocessing in text dataset

### Kind of basic text preprocessing:

#### Token masking

##### Definition

Token masking is a preprocessing step to convert a specific “token” (word, phrases, or a special term) into a verbal form.

##### Example

Token Masking		
Before	After (1)	After (2)
<a href="https://langing.ai/">https://langing.ai/</a>	<URL>	url
<a href="http://www.langing.ai">www.langing.ai</a>		
okky@langing.ai	<EMAIL>	email
@okkyibrohim	<USER>	user
+6285123123123	<NO_HANDPHONE>	no_handphone
085123123123		

## 2.2 Basic text preprocessing in text dataset

### Kind of basic text preprocessing:

#### Text cleaning (1)

##### ◎ Definition

Text cleaning is a preprocessing step remove a specific “token” or “character” from our text.

##### ◎ Kind of Text Cleaning

- Extra whitespace (enter, tab, etc.), including trimming
- Punctuation (dot, coma, semicolon, question mark, exclamation mark, etc.)
- Unnecessary token/character (hex, digit, character repetition, etc.)
- Special attribute (HTML tag, RT, username, URL, email, No. Handphone, etc.)

Emoji and emoticon

etc.

## 2.2 Basic text preprocessing in text dataset

### Kind of basic text preprocessing:

#### Text cleaning (2)

#### Example

Text Cleaning		
Type of Cleaning	Before	After
Extra whitespace	i love you. but lies :)	i love you. but lies :)
Punctuation	too, many; 'punctuation' on? "this" text...!!!	to many punctuation on this text
Unnecessary token/character	on 123 you must clean this heeeex \x7f	on you must clean this hex
Special Attribute	<a> hey @okky look at this web <a href="http://www.langing.ai">www.langing.ai</a> or reach me via email good@langing.ai or phone +6285123123123 </a>	hey look at this web or reach me via email or phone
Emoji and Emoticon	i ♥ you 😊 😊	i you

## 2.2 Basic text preprocessing in text dataset

### Kind of basic text preprocessing:

#### Basic text normalization (using dictionary)

#### Definition

Text normalization is a preprocessing step to normalize a text into the “normal” one (usually to get the formal one).

#### Example

Text Normalization		
Language (Example)	Before	After
English	up 2 you	up to you
	i love you 4ever	i love you forever
Indonesian (Bahasa)	gw cnta bgt sama lo	aku cinta banget sama kamu
	gue cinta bgd sama loe	
	aku cinta bgt sm kmu	

## 2.2 Basic text preprocessing in text dataset

### Kind of basic text preprocessing:

#### Stop words removal

#### Definition

Stop words removal is a preprocessing step to stop words i.e. words/phrases that do not have meaning in our research (e.g. conjunction, article, to be, or specific words/phrases that we define based on EDA process) from our text.

#### Example

Stop Words Removal		
Language (Example)	Before	After
English	you <b>and</b> i must live together	you i must live together
	<b>this is a</b> sample <b>of</b> sentence	sample sentence
Indonesian (Bahasa)	kamu suka dia <b>tetapi tidak</b> sebaliknya	kamu suka dia sebaliknya
	kamu <b>dan</b> dia sangat beruntung	kamu dia sangat beruntung

## 2.2 Basic text preprocessing in text dataset

### Kind of basic text preprocessing: Stemming and lemmatization

#### Definition

- Stemming is a preprocessing step to cut affixed words into the base form.
- Lemmatization is a preprocessing step to cut affixed words into the base Verb 1 form.
- In some languages that do not have tenses, stemming is equal to lemmatization.

#### Example

Stemming and Lemmatization		
Type	Before	After
Stemming	i am caring to you	i am car to you
Lemmatization		i be care to you



## 2.2 Basic text preprocessing in text dataset

### Attention in doing text preprocessing (1)

- ◎ Order is matter  
Sequence errors will result in non-optimal preprocessing results. For example, stemming/lemmatization must be in the final step of preprocessing.
- ◎ Follow your feature extraction process  
Not all preprocessing steps must be done before feature extraction. For example, when you want to use an orthography feature (will be explained in the next chapter), the preprocessing step must be done after you extract the orthography feature.

## 2.2 Basic text preprocessing in text dataset

### Attention in doing text preprocessing (2)

#### ◎ Source, domain, and language

Depending on your source (social media, web news, etc.), domain (health, economic, etc.), and languages (English, Indonesian, monolingual, multilingual, etc.) of your dataset, you may need some specific preprocessing steps. For example, preprocessing a news dataset may usually less effort than preprocessing a social media dataset (especially for Twitter). For this reason, a popular text preprocessing library may not be appropriate to your dataset, better you try and look at the output before deciding to use it for your entire dataset.

#### ◎ Follow the pre-trained word embedding you used

If you use a particular pre-trained word embedding (will be explained in the next chapter), you must follow the same pre-processing step used by the authors that build the pre-trained word embedding you used.

## 2.2 Basic text preprocessing in text dataset

### Advanced text preprocessing examples

#### ◎ Text translation

Some dataset may contain multilinguality whether it is switch-coded or mixed-coded. If we do not have multilingual pre-trained word embedding that cover a particular language of our dataset, text translation often used to normalize it.

#### ◎ Context aware text normalization

Actually, a text normalization often can simply done just using dictionary matching. For example, an abbreviation has several interpretations depends on the context. For this case, context aware text normalization e.g. using a seq2seq deep learning model may will be good.

A decorative graphic in the top-left corner featuring a network of interconnected nodes and lines. Some nodes are solid blue circles, while others are white circles with blue outlines. The lines are thin and gray.

**Thank You!**

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A decorative graphic in the bottom-right corner, similar to the one in the top-left, showing a network of nodes and lines with some blue and some white nodes.