

# **Machine Learning**

Natural Language Processing

Karol Przystalski

April 3, 2019

Department of Information Technologies, Jagiellonian University

# **Agenda**

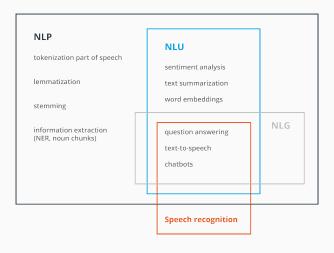
- 1. Introduction
- 2. Use cases
- 3. Text processing
- 4. Text understanding
- 5. Text generation
- 6. Quality metrics

Introduction

#### **Timeline**

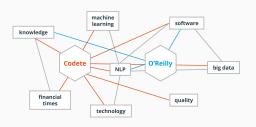


## **Taxonomy**



There are many use cases of NLP methods, just to mention a few:

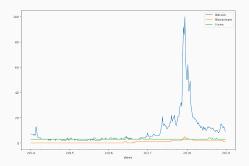
• term relationship,



- term relationship,
- text summarization,



- term relationship,
- text summarization,
- time series analysis,



- term relationship,
- text summarization,
- time series analysis,
- intent recognition,



There are many use cases of NLP methods, just to mention a few:

- term relationship,
- text summarization,
- time series analysis,
- intent recognition,
- sentiment analysis,



Are we facing a confidence gap?
While women outperform men
in school and are entering
the workforce in record numbers,
half of female managers in the UK
reported self doubt about their job
performance and careers



Are we facing a confidence gap?
While women outperform men
in school and are entering
the workforce in record numbers,
half of female managers in the UK
reported self doubt about their job
performance and careers

- term relationship,
- text summarization,
- time series analysis,
- intent recognition,
- sentiment analysis,
- text generation.



# Text processing

# Regular expressions in Python for string comparison

Regular expressions in Python are almost the same as in any other programming languages. We can use regex methods to:

- search finds only the first occurrence of expression in text,
- match finds all occurrences of expression in text,
- fullmatch matches only if the whole string matches the regular expression,
- split splits into a list based on the splitting expression,
- escape replaces all characters in the pattern.

Regular expressions are used within many methods that we go through in the next slides.

# Word and sentence comparison methods

String comparison methods available in Python:

- Levenshtein distance,
- Damerau-Levenshtein distance,
- Jaro distance,
- Jaro-Winkler distance,
- Match rating approach comparison,
- · Hamming distance,
- Gestalt pattern matching.

You can use at least two libraries:

- Difflib –
   https://docs.python.org/3.6/library/difflib.html,
- Jellyfish https://pypi.org/project/jellyfish/.

# String comparison – Levenshtein Distance

The Levenshtein distance is a number of insertion, deletion or replacement changes that needs to be done to get the same strings.

It is a number that is equal or higher than 0. It can be normalized to get a number from 0 to 1.

compared words	word length
training	8
trains	6

The distance for both words is 3. After the normalization the distance is  $\frac{3}{8}=0.375.$ 

# String comparison – Gestalt pattern matching

This solution can be formulated as:

$$G_{PM} = \frac{\# \text{same characters}}{\# \text{total characters}}.$$

For the same example we have 5 same characters in each word and four that are different. This makes the  $G_{PM}$  value:

$$G_{PM} = \frac{10}{14} = 0.7142.$$

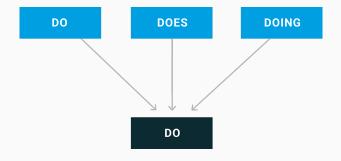
# NLP methods used for sentence comparison

There are three popular methods that are used in NLP for text processing:

- tokenization,
- lemmatization,
- stemming.

Tokenization divides a sentence into separate words.

# **Lemmatization and stemming**



#### More advanced NLP methods

There are three popular NLP methods that make it easier to understand written text:

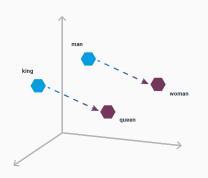
- part of speech,
- noun chunk,
- named entity recognition.

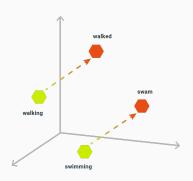
## More advanced NLP methods



**Text understanding** 

# Word vectorization





## Word vectorization – concept



#### Word vectorization - methods

The most popular methods that are used to create a space of vectorized words are:

- bag of words,
- tf-idf.
- transfer learning,
- n-gram model,
- skip-thought vectors.

# Bag of words

#### **BAG OF WORDS**

This is a paper on machine learning. Deep learning is a recent trend.



{ This, is, a, paper, on, machine, learning, deep, recent, trend }

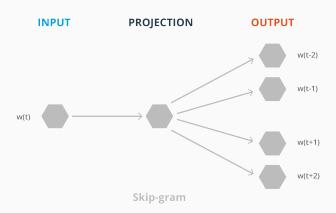
- [11111111000]
- 2 [0110001111]

## **Distance metrics**

# Also known as similarity or dissimilarity measures.

Measure name	equation	
Manhattan distance	$ \rho_{Man}(x_r, x_s) = \sum_{i=1}^{n}  x_{ri} - x_{si}  $	(1)
Chebyshew distance	$\rho_{Ch}(x_r, x_s) = \max_{1 \le i \le n}  x_{ri} - x_{si} $	(2)
Frecht distance	$\rho(x_r, x_s) = \sum_{i=1}^d \frac{ x_{ri} - x_{si} }{1 +  x_{ri} + x_{si} } \frac{1}{2^i}$	(3)
Canberra distance	$\rho(x_r, x_s) = \sum_{i=1}^d \frac{ x_{ri} - x_{si} }{ x_{ri} + x_{si} }$	(4)
Post office distance	$\rho_{\textit{pos}}(x_{\textit{r}},x_{\textit{s}}) = \left\{ \begin{array}{l} \rho_{\textit{Min}}(x_{\textit{r}},0) + \rho_{\textit{Min}}(0,x_{\textit{s}}), \; \text{for} \; x_{\textit{r}} \neq x_{\textit{s}}, \\ \\ 0, \; \text{for} \; x_{\textit{r}} = x_{\textit{s}} \end{array} \right.$	(5)
Bray-Curtis distance	$\rho_{bc}(x_r, x_s) = \frac{\sum_{i=1}^{d}  x_{ri} - x_{si} }{\sum_{i=1}^{d} (x_{ri} - x_{si})}$	(6)

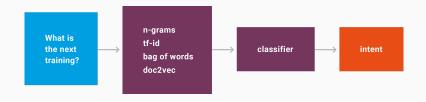
## Word vectorization – concept



## **Entities and intents**



## **Entities and intents**



**Text generation** 

#### **NLG**

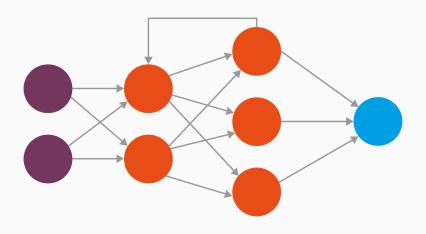
Natural Language Generation is a part of Natural Language Processing. The goal of NLG is to generate a sentence or the whole document that has a logical sense, follows the grammar and answers the question properly if we deal with a bot.

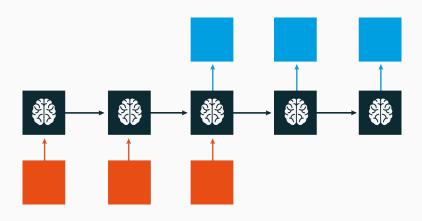
There are plenty of methods that can be used for text generation. The most popular are:

- n-gram model,
- recurrent neural network,
- autoencoders,
- generative adversarial network.

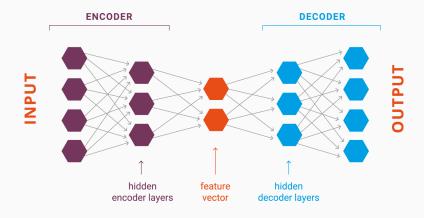
# N-gram model



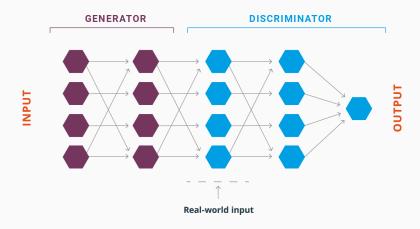




## **Autoencoders**



## **Generative Adversarial Networks**



# Working examples

There are many open source trained models available. Here are a few worth mentioning:

- chatterbot a chatbot implementation http://chatterbot.readthedocs.io/
- DeepQA uses RNN and has a web interface https://github.com/Conchylicultor/DeepQA
- Generative Conversational Agents uses LSTM, RNN and GAN https://github.com/oswaldoludwig/ Adversarial-Learning-for-Generative-Conversational-Agents.

#### Research datasets

A few datasets useful for your research:

- SQuAD reading comprehension dataset, consists of questions posed by crowdworkers on a set of Wikipedia articles, where the answer to each question is a segment of text, https://rajpurkar.github.io/SQuAD-explorer/,
- Cornell Movie Dialogs Corpus movie dialogs, https://www.cs.cornell.edu/~cristian/Cornell\_ Movie-Dialogs\_Corpus.html,
- DeepMind datasets AQua is a dataset of questions and answers, https://github.com/deepmind/AQuA, more datasets from DeepMind: https://deepmind.com/research/open-source/ open-source-datasets/,
- DMQA Daily Mail and CNN articles data sets, https://cs.nyu.edu/~kcho/DMQA/,
- MS MARCO Microsoft MAchine Reading COmprehension Dataset, http://www.msmarco.org/dataset.aspx.

**Quality metrics** 

# **Quality metrics for NLP**

In most cases we use the same quality metrics as in regular machine learning models, if any machine learning model is used for NLP.

Other methods that can be used for NLP methods validation, especially, text generation:

- spell check,
- grammer check,
- typo check.

