

91258 / B0385 Natural Language Processing

Lesson 1. Introduction

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30/09/2024

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Materials

Core Bibliography

- 1. Lane et al. (2019)'s Natural Language Processing in Action¹
- 2. Numerous Wikipedia articles on relevant topics
- 3. Multiple online forums



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 $^{^{1}} https://www.manning.com/books/natural-language-processing-in-action \, \circ \, \circ \, (-1) \, decomposition \, \circ \, (-1) \, decomposit$

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Complementary Bibliography

- 1. Intro to computing for text
 - K.W. Church's Unix for poets²
- 2. For social media analysis
 - Hovy (2021)'s Text Analysis in Python for Social Scientists*3
- 3. A basic intro in Italian
 - Nissim and Pannitto (2022)'s Che cos'è la linguistica computazionale
- 4. From linguistics
 - Bender (2013)'s Linguistic fundamentals for natural language processing: 100 essentials from morphology and syntax⁴
- 5. Advanced
 - Koenigstein (2024)'s Transformers in Action*5

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²https://web.stanford.edu/class/cs124/kwc-unix-for-poets.pdf

³https://doi.org/10.1017/9781108873352

⁴https://doi.org/10.2200/S00493ED1V01Y201303HLT020

⁵https://www.manning.com/books/transformers-in-action

The NLP environment this year

1. Tutorato

K. Korre (3rd year PhD student) will offer 10 lessons of tutorato.

https://moodle.dipintra.it

2. Selected Topics in NLP (3 cfu)

Optional lesson covering complementary (non mandatory) topics

https://www.unibo.it/it/studiare/

dottorati-master-specializzazioni-e-altra-formazione/insegnamenti/insegnamento/2024/508811

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Lesson coordinates

Slides, code, calendar⁶ and more are all available at:

albarron.github.io/teaching/natural-language-processing

⁶For all three initiatives.

Tools

Essential

Python 3 development framework on any modern OS

- 1. Command line or
- 2. Integrated development Environment; e.g., Pycharm⁷, Eclipse⁸ or
- 3. Jupyter notebook; e.g., Google's colab⁹, local Jupyter¹⁰

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⁷https://www.jetbrains.com/pycharm/

⁸https://www.eclipse.org

⁹https://colab.research.google.com

¹⁰https://jupyter.org

¹¹Could be part of Selected topics/tutorato

¹²https://gitlab.com

¹³https://github.com

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Desirable¹¹

- 2. LATEX system for document preparation

```
<sup>7</sup>https://www.jetbrains.com/pycharm/
```

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⁸https://www.eclipse.org

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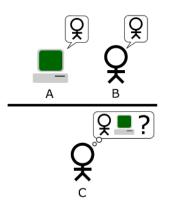
¹⁰https://jupyter.org

¹¹Could be part of Selected topics/tutorato

¹²https://gitlab.com

¹³https://github.com

Natural language as a measure of intelligence



Turing (1950). "Computing machinery and intelligence". Mind. 59(236)

Picture from

upload.wikimedia.org/wikipedia/commons/e/e4/Turing_Test_version_3.png

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CL vs NLP

Computational linguistics¹⁴

• Interdisciplinary field concerned with the computational (it used to say "statistical or rule-based"!) modeling of natural language

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¹⁴https://en.wikipedia.org/wiki/Computational_linguistics

¹⁵ https://en.wikipedia.org/wiki/Natural_language_processing

CL vs NLP

Computational linguistics¹⁴

- Interdisciplinary field concerned with the computational (it used to say "statistical or rule-based"!) modeling of natural language
- Study of appropriate computational approaches to linguistic questions

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• Interdisciplinary subfield of computer science and artificial intelligence (it used to say "linguistics"!)[...] concerned with providing computers the ability to process data encoded in natural language

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- Study of appropriate computational approaches to linguistic questions

Natural Language Processing¹⁵

- Interdisciplinary subfield of computer science and artificial intelligence (it used to say "linguistics"!)[...] concerned with providing computers the ability to process data encoded in natural language
- Data is collected in text corpora, using either rule-based, statistical or neural-based approaches in machine learning and deep learning

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¹⁵ https://en.wikipedia.org/wiki/Natural_language_processing

CL vs NLP

Natural Language Processing (Lane et al., 2019, p. 4)

- Area of research in computer science and artificial intelligence concerned with processing natural languages
- This processing generally involves translating natural language into data (numbers) that a computer can use to learn about the world

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CL vs NLP

Natural Language Processing (Lane et al., 2019, p. 4)

- Area of research in computer science and artificial intelligence concerned with processing natural languages
- This processing generally involves translating natural language into data (numbers) that a computer can use to learn about the world

The term natural language processing is nowadays considered to be a near-synonym of computational linguistics and (human) language technology. ¹⁶

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Rule-based vs Statistical NLP

Rule-based NLP

Models are based on a number of hand-crafted rules or grammars

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Rule-based NLP

Models are based on a number of hand-crafted rules or grammars



Diagram borrowed from L. Moroney's Introduction to TensorFlow for Artificial Intelligence, Machine Learning, and Deep Learning

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Models are based on a number of hand-crafted rules or grammars

Derived from

https://stackabuse.com/python-for-nlp-creating-a-rule-based-chatbot/

Statistical NLP

Models are tuned on annotated data

Statistical NLP

Models are tuned on annotated data

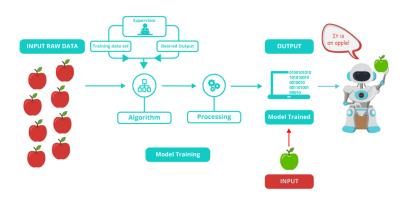


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Statistical NLP

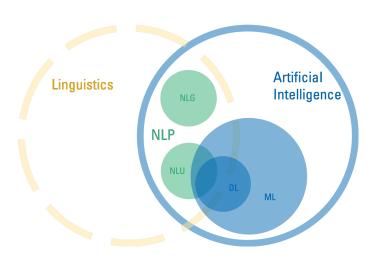
Models are tuned on annotated data



Borrowed from https://www.edureka.co/blog/machine-learning-tutorial

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The NLP neighborhood



Borrowed from

https://www.retresco.de/en/how-to-ai-natural-language-processing/

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Non-exhaustive list of NLP applications with examples

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Q Search

web search engines \cdot text autocompletion

Non-exhaustive list of NLP applications with examples

Q Search Editing

web search engines · text autocompletion grammar issues identification

Non-exhaustive list of NLP applications with examples

Q Search	web search engines · text autocompletion
☑ Editing	grammar issues identification

○ Dialogue chatbot creation

Non-exhaustive list of NLP applications with examples

Q Search	web search engines · text autocompletion
Editing	grammar issues identification
□ Dialogue	chatbot creation
☑ Email	spam filtering \cdot message classification

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Non-exhaustive list of NLP applications with examples

Q	Search	web search engines · text autocompletion
	Editing	grammar issues identification
\wp	Dialogue	chatbot creation
\square	Email	spam filtering · message classification
	Text mining	(multi-)document summarisation

Non-exhaustive list of NLP applications with examples

Q Search web search engines · text autocompletion

 Editing grammar issues identification

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| ■ Text mining (multi-)document summarisation News analysis

event identification · fact checking

Non-exhaustive list of NLP applications with examples

Q Search web search engines · text autocompletion

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 $lue{}$ **News analysis** event identification \cdot fact checking

♣ Forensics plagiarism detection · authorship attribution

Introduction

Non-exhaustive list of NLP applications with examples

Q Search web search engines · text autocompletion Editing grammar issues identification

Dialogue chatbot creation

Email spam filtering · message classification **|** ■ Text mining (multi-)document summarisation

News analysis event identification · fact checking

Forensics

plagiarism detection · authorship attribution

Sentiment analysis product review ranking · opinion mining

Introduction

Non-exhaustive list of NLP applications with examples

- Q Search
- Editing
- Dialogue
- **Email**
- **|** Text mining
- News analysis
- Forensics
- **Sentiment** analysis
- Creative writing

web search engines · text autocompletion

grammar issues identification

chatbot creation

spam filtering · message classification

(multi-)document summarisation

event identification · fact checking

plagiarism detection · authorship attribution

product review ranking · opinion mining text generation with a narrative and style

Introduction

Non-exhaustive list of NLP applications with examples

Q	Search	web search engines · text autocompletion
	Editing	grammar issues identification
\wp	Dialogue	chatbot creation
\square	Email	spam filtering · message classification
	Text mining	(multi-)document summarisation
	News analysis	event identification · fact checking
2	Forensics	plagiarism detection · authorship attribution
Ô	Sentiment analysis	product review ranking · opinion mining
	Creative writing	text generation with a narrative and style
ZA.	Translation	translation · quality estimation

The philosophy of this lesson

- 1. Concept understanding
- 2. Know what you are doing and why
- 3. In position to go forward

Requirements & Evaluation

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Requirements

Necessary

- Linguistics
- Algebra
- Programming in Python

Requirements

Necessary

- Linguistics
- Algebra
- Programming in Python

Desirable

- Intermediate programming (e.g., object-oriented, testing)
- High-performance computing (e.g., slurm)¹⁷

Final project: 80%

You will address a relevant problem...

- within the range of your own (research) interests or
- participating (formally) in a shared task or
- proposed by me, if you prefer

Homework: 20%

mostly programs addressing relatively small problems

Typical final project pipeline

1. You propose a topic/problem. We assess if it is reasonable, doable. . .

¹⁸Talk to me well in advance; it would require my heavy involvement

¹⁹I do not like (student) novels

Typical final project pipeline

- 1. You propose a topic/problem. We assess if it is reasonable, doable. . .
- 2. You compile data, study the problem, design experiments, code. . .

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Typical final project pipeline

- 1. You propose a topic/problem. We assess if it is reasonable, doable. . .
- 2. You compile data, study the problem, design experiments, code... IF you plan for a publication 18
 - We meet regularly to see the advances and shape the experiments, submissions, and/or paper towards the submission deadline

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ELSE

• We could meet sporadically, if you need it

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- 3. You submit a written report (\sim 7 pages), ¹⁹ your implementation, results 1 week before the *appello*

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- 4. We meet on the date of the appello to discuss about your project, in the context of the lecture

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Final mark

(Beside the 20% for homework) a combination of the quality of the experiments, report, code, and oral discussion

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Targeting 30L?

If I let you submit a paper, it is very likely. But it is not the only way...

Final mark

(Beside the 20% for homework) a combination of the quality of the experiments, report, code, and oral discussion

Targeting 30L?

If I let you submit a paper, it is very likely. But it is not the only way. . .

$$p(30L \mid \text{paper submitted} == True) \approx 0.85$$
 (1)

$$p(30L \mid paper submitted == False) \approx 0.15$$
 (2)

Previous final projects

2023-2024

- Multilingual Persuasion Technique Detection in News
- W Tracing source languages in Wikipedia articles

2022-2023

- Sentiment analysis of video game reviews
- Authorship attribution: machine vs human

2021-2022

- ★ Hate Speech Detection in Incel Online Spaces
- \vec{Q} Fishing for catfishes: predicting the author gender in Reddit

^{*} student with previous programming skills

[•] turned into (part of a) thesis • turned into a publication

Previous final projects

2020-2021

- ★ Semantic similarity between originals and machine translations
- Definition extraction on food-related Wikipedia articles
- Identifying Characters' Lines in Original and Translated Plays
- Classifying an Imbalanced Dataset with CNN, RNN, and LSTM

2019-2020

- AriEmozione: Identifying Emotions in Opera Verses
- UniBO@AMI: A Multi-Class Approach to Misogyny and Aggressiveness Identification on Twitter Posts Using AIBERTo
- * students with previous programming skills
- turned into (part of a) thesis turned into a publication

Visit the projects section of the class website for details, reports and papers

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References

Bender, E. M.

2013. Linguistic Fundamentals for Natural Language Processing: 100 Essentials from Morphology and Syntax. Morgan & Claypool Publishers.

Hovy, D.

2021. Text Analysis in Python for Social Scientists: Discovery and Exploration, Elements in Quantitative and Computational Methods for the Social Sciences. Cambridge University Press.

Koenigstein, N.

2024. *Transformers in Action* . Shelter Island, NY: Manning Publication Co.

Lane, H., C. Howard, and H. Hapkem

2019. *Natural Language Processing in Action*. Shelter Island, NY: Manning Publication Co.

Nissim, M. and L. Pannitto

2022. Che cos'è la linguistica computazionale. Carocci editore.

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