#### **Course Overview**

He He



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Logistics

A brief history

Challenges in NLP

Logistics



- Best way to communicate with us: Campuswire (link and code on Brightspace).
- Office hours will be on Zoom (details on website), or in person by appointment.
- Midterm will be online through Gradescope.
- Let us know if you have accessibility needs.

#### What this course is (not) about

- It's not about specific NLP applications (QA, dialogue etc.)
  - Unified approaches to various NLP problems
  - Hands-on experience in building NLP systems through assignments and the course project
- It's not about fundamental machine learning
  - Focus on unique challenges in language data
  - Formalize NLP tasks as statistical learning problems
- Changes from previous offerings
  - Structured prediction (HMM, PCFG, CRF)
  - + Large language models

#### What we expect you to know

- Linear algebra: vector space, vector norm, dot product, gradient etc.
- **Probability and statistics**: conditional probability, expectation, Bayes rule etc.
- **Basic machine learning**: loss function, gradient descent, logistic regression etc.
- **Programming**: read and write Python code, use Numpy, HPC, and deep learning libraries (Pytorch, Huggingface etc.)

#### **Course project**

An important component of the course (more on this later)

- Related to NLP (doesn't have to be in the scope of this course)
- New algorithms or models for existing problems
- Applications of NLP or ML techniques to a problem
- Analysis of well-known approaches that leads to new insight
- ML Reproducibility Challenge 2021 (https://paperswithcode.com/rc2021)

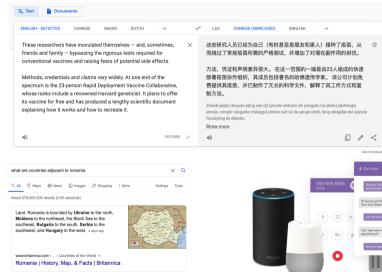
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#### Products powered by NLP technologies





## A single natural language interface for everything

what's the best way to learn NLP?

There is no one-size-fits-all answer to this question as the best way to learn NLP (Natural Language Processing) will depend on your background and goals. However, some ways to get started with learning NLP include:

- Taking online courses: There are many online courses available on NLP, such as those offered by Coursera, edX, and Udemy.
- Reading books: There are many books available on NLP, both for beginners and advanced learners. Some popular books include "Speech and Language Processing" by Daniel Jurafsky and James H. Martin, and "Natural Language Processing with Python" by Steven Bird, Ewan Klein, and Edward Loper.
- Practicing with tutorials and projects: There are many tutorials and open-source projects available online that can help you learn NLP by implementing and experimenting with different algorithms and techniques.
- Joining a community: Joining a community of NLP enthusiasts, such as those on Reddit, Slack, or LinkedIn, can provide you with support, resources, and opportunities to collaborate on projects.
- Participate in NLP Challenges: Participating in NLP challenges such as those hosted by Kaggle or other platforms can be a great way to learn and improve your skills.

# A single natural language interface for everything



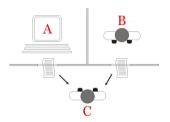
Write an excel formula to compute the average of all scores larger than 80 in column C1



=AVERAGEIF(C1,">80")

6 V

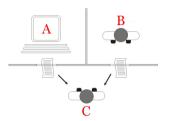
# Language is at the core of AI: the imitation game





*"I believe that in about fifty years' time it will be possible to programme computers, with a storage capacity of about* 10<sup>9</sup>, *to make them play the imitation game so well that an average interrogator will not have more than 70 percent chance of making the right identification after five minutes of questioning."* Turing (1950)

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Is humanlikeness the ultimate goal?

- Built by Joseph Weizenbaum at MIT in 1964 to demonstrate the *superficiality* of human-machine communication.
- Surprisinly, people were convinced that ELIZA had human intelligence.

Human:	Well, my boyfriend made me come here.

- ELIZA: Your boyfriend made you come here?
- Human: He says I'm depressed much of the time.
- ELIZA: I am sorry to hear you are depressed.

Human: It's true. I'm unhappy.

ELIZA: Do you think coming here will help you not to be unhappy?

## Early rule-based systems: the Georgetown-IBM experiment

• The Russian-English machine translation program:

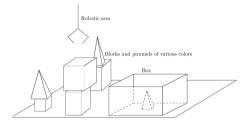


- A vocabulary of 250 words
- Using 6 grammar rules, e.g.,

If first code is 110, is third code associated with preceding complete word equal to 21? If so, reverse order of appearance of words in output (i.e., word carrying 21 should follow that carrying 110)---otherwise, retain order.

# Approaching AI as a whole: SHRDLU

- Built by Terry Winograd at MIT in 1968.
- A person instructs the computer to build structures in a blocks world.
- Require many capabilities: grounding, coreference resolution, question answering, semantic parsing etc.



Person: Pick up a big red block.

Computer: OK.

Person: Grasp the pyramid.

Computer: I don't understand which pyramid you mean.

Person: (changing their mind) Find a block which is taller than the one you are holding and put it into the box.

Computer: By "it", i assume you mean the block which is taller than the one i am hold-ing.

## Limitations of early systems

• Optimism in the 50's and 60's: working on tasks that are too complex at that time

"Within the very near future—much less than twenty-five years—we shall have the technical capability of substituting machines for any and all human functions in organizations."

- Disappointing results due to
  - Limited computation: hardware has limited speed and memory
  - Combinatorial explosion: algorithms are intractable in realistic settings
  - Underestimated complexity: ambiguity, commonsense knowledge etc.

## The rise of statistical learning in the 80's

- Notable progress in MT from IBM (neglected knowlege of linguistics).
- HMMs widely used for speech recognition. *"Every time I fire a linguist, the performance of the speech recognizer goes up."*—Frederick Jelinek.
- The paradigm shift: expert knowledge + rules  $\rightarrow$  data + features
- Statistical learning is the main driving force of NLP today.

## The deep learning tsunami

- Before deep learning (around 2015), NLP is mostly about structured prediction and feature engineering.
- Neural networks can automatically learn good features/representations for a task
- The paradigm shift: features  $\rightarrow$  network architectures + embeddings
- Almost all NLP models are neural networks nowadays.

## Models and data keep getting larger

- Since around 2018, Transformer-based pretrained models have become the standard.
- Pre-training on large data provides useful representations for many downstream tasks.
- The paradigm shift: architecture design  $\rightarrow$  transfer learning (fine-tuning)
- More recently, a single natural language interface for all tasks (e.g., ChatGPT by OpenAl).
- The paradigm shift: transfer learning  $\rightarrow$  instructing / prompting

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#### Discrete

• How to define metrics?

I work at NYU. vs I work for NYU. This is good. vs This is actually good.

• How to define transformations?

The food is okay. They made a brief return to Cambridge to  $\rightarrow$  They returned. drop the book.

- $\rightarrow$  The food is awesome!
- In general, hard to represent text as mathematical objects.

#### • Compositional

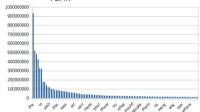
- The whole is built from parts (chars, words, sentences, paragraphs, documents...)
- How to generalize when we don't see all possible combinations?
- An example from [Lake et al., 2018] Vocabulary:

{jump, walk, turn, once, twice, left, right, before, after, and} Sentences:

jump jump left jump left and walk right jump left after walk right once before turn left twice

#### • Sparse

- How to handle the long tail?
- Zipf's law: word frequency  $\propto \frac{1}{rank}$



• Many linguistic phenomena follow Zipf's law BoA's financial assistant Erica:

*The bank "learned [that] there are over 2,000 different ways to ask us to move money."*<sup>1</sup>

<sup>&</sup>lt;sup>1</sup>https://www.aiqudo.com/2019/06/28/voice-success-story-erica-bank-america/

#### • Ambiguous

• How to interpret meaning in context?

Bass: fish? guitar? frequency? (word sense disambiguiation)

I shot an elephant in my pajamas: who is in the pajamas? (PP attachment)

The spirit is willing but the flesh is weak.  $\rightarrow$  The vodka is strong but the meat is rotten.