What is (E)NLP?

Nathan Schneider ~ 11 January 2024

Some slides adapted from Sharon Goldwater, Philipp Koehn, Alex Lascarides

What do YOU think?

- Team up with a partner you don't already know.
- Take 5 min. to discuss:
 - What have you heard lately about NLP & AI?
 - What do you expect to learn in this course?

Introductions

- · Say your name, program, year, natural languages spoken
- and what you discussed with your partner



Original text:

Istotą instytucji wyłączenia organu podatkowego od załatwienia sprawy dotyczącej zobowiązania podatkowego lub innej sprawy normowanej przepisami prawa podatkowego jest utrata właściwości danego organu do załatwienia danei snrawy

Detect language ▼ >> Finnish ▼

Translate

Translation: Polish (automatically detected) » Finnish

Pelkät vapautusta veron käsittelevälle viranomaiselle tapauksissa, joissa verovelan tai muita aineita, normowanej vero-oikeuden menetys kiinteistöä kyseisen viranomaisen ratkaista asian erityinen veronmaksajille.

Suggest a better translation









Applications & Core Tasks

- Machine Translation
- Information Retrieval
- Question Answering
- Dialogue Systems
- Information Extraction
- Summarization
- Sentiment Analysis
- •

- Language modeling/text generation
- Part-of-speech tagging
- Syntactic parsing
- Named entity recognition
- Coreference resolution
- Word sense disambiguation
- Semantic role labeling
- •

NLP as a Field

- NLP lies at the intersection of computational linguistics and artificial intelligence.
- NLP is (to various degrees) informed by linguistics, but with practical/engineering rather than purely scientific aims.
- Processing speech (i.e., the acoustic signal) is separate.

This Course

- NLP is a big field! This course focuses mainly on foundational ideas and methods to answer the question: "How can we formulate computation for natural language?"
 - Linguistic facts and issues
 - Computational models and algorithms, especially using data ("empirical")
 - More emphasis on representations and tasks than applications

What are your goals?

Why are you here? Perhaps you want to:

- work at a company that uses NLP (perhaps as the sole language expert among engineers)
- use NLP tools for research in linguistics (or other domains where text data is important: social sciences, humanities, medicine, ...)
- conduct research in NLP (or IR, MT, etc.)

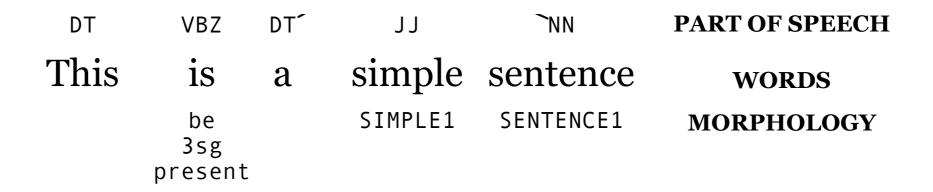
- An important insight of linguistics is that language consists of many levels of structure
- Humans fluently integrate all of these in producing/ understanding language
- Ideally, so would a computer!

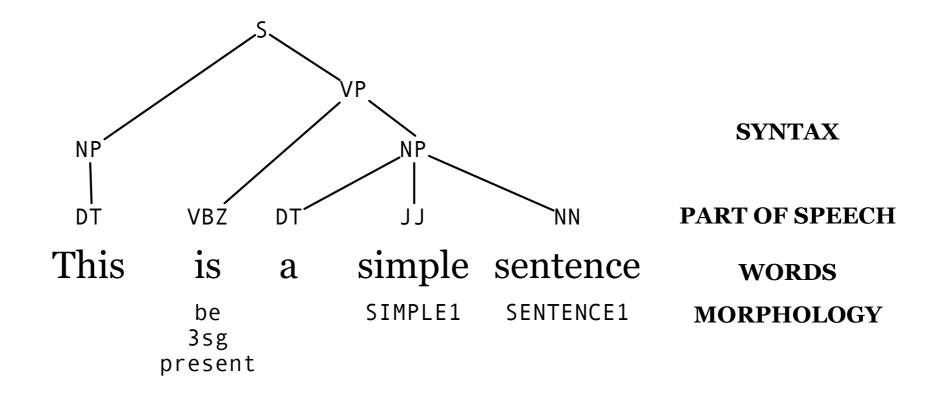
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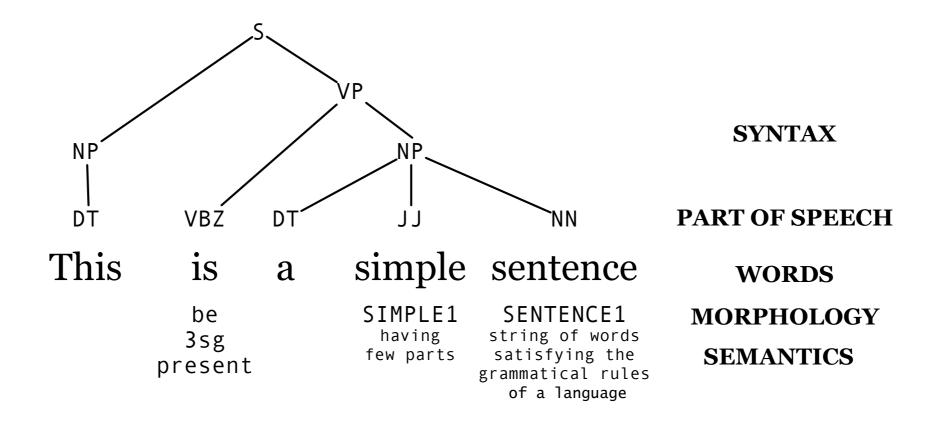
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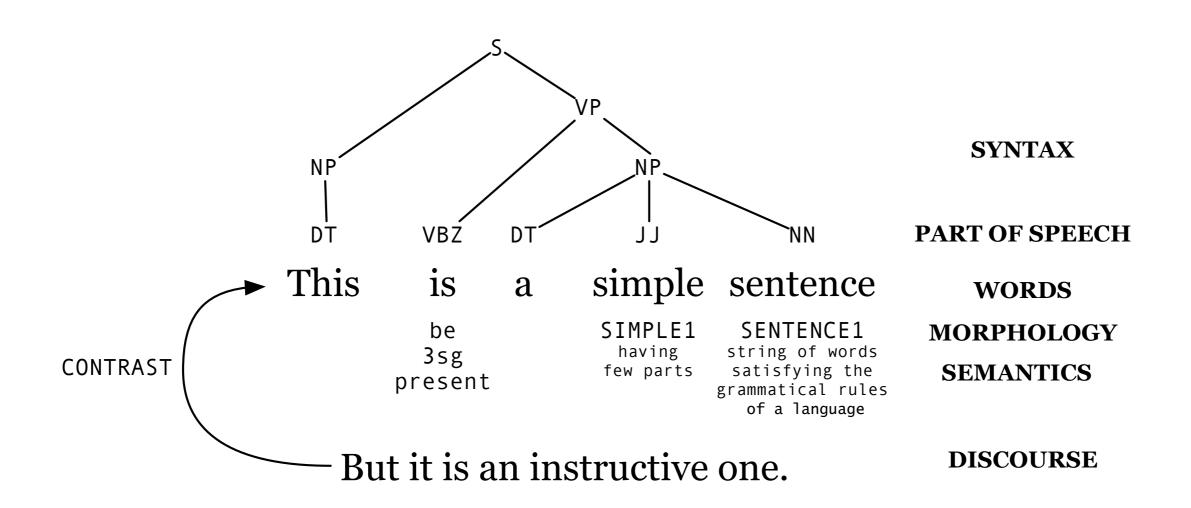
be SIMPLE1 SENTENCE1 MORPHOLOGY

3sg present









- 1. Ambiguity at many levels:
- Word senses: bank (finance or river?)
- Part of speech: chair (noun or verb?)
- Syntactic structure: I saw a man with a telescope
- Quantifier scope: Every child loves some movie
- Multiple: I saw her duck

How can we model ambiguity, and choose the correct analysis in context?

Ambiguity

What can we do about ambiguity?

- non-probabilistic methods (FSMs for morphology, CKY parsers for syntax) return all possible analyses.
- probabilistic models (HMMs for POS tagging, PCFGs for syntax) and algorithms (Viterbi, probabilistic CKY) return the *best possible analysis*.

But the "best" analysis is only good if our probabilities are accurate. Where do they come from?

Statistical NLP

Like most other parts of AI, NLP is dominated by statistical methods.

- Typically more robust than earlier rule-based methods.
- Relevant statistics/probabilities are *learned from data*.
- Normally requires lots of data about any particular phenomenon.

- 2. Sparse data due to Zipf's Law.
- To illustrate, let's look at the frequencies of different words in a large text corpus.
- Assume "word" is a string of letters separated by spaces (a great oversimplification, we'll return to this issue...)

Word Counts

Most frequent words in the English Europarl corpus (out of 24m word tokens)

any word			nouns	
Frequency	Token	F	requency	Token
1,698,599	the		124,598	European
849,256	of		104,325	Mr
793,731	to		92,195	Commission
640,257	and		66,781	President
508,560	in		62,867	Parliament
407,638	that		57,804	Union
400,467	is		53,683	report
394,778	\mathbf{a}		53,547	Council
263,040	I		45,842	States

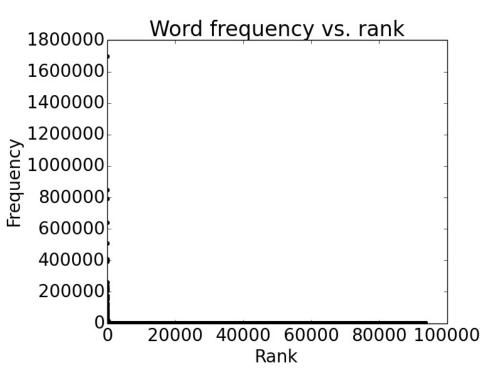
Word Counts

But also, out of 93,638 distinct words (**word types**), 36,231 occur only once. Examples:

- cornflakes, mathematicians, fuzziness, jumbling
- pseudo-rapporteur, lobby-ridden, perfunctorily,
- Lycketoft, UNCITRAL, H-0695
- policyfor, Commissioneris, 145.95, 27a

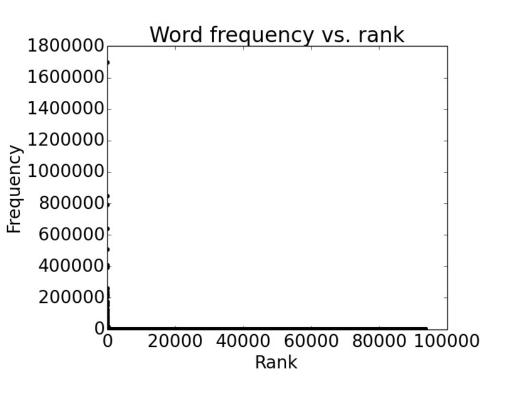
Plotting word frequencies

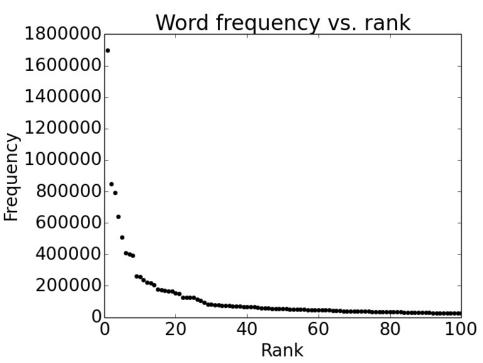
Order words by frequency. What is the frequency of nth ranked word?



Plotting word frequencies

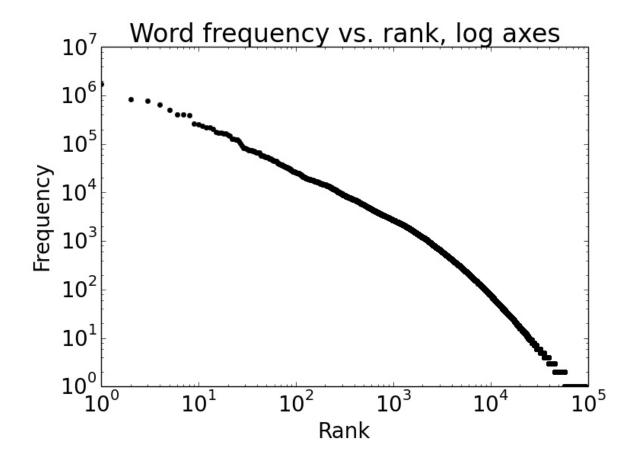
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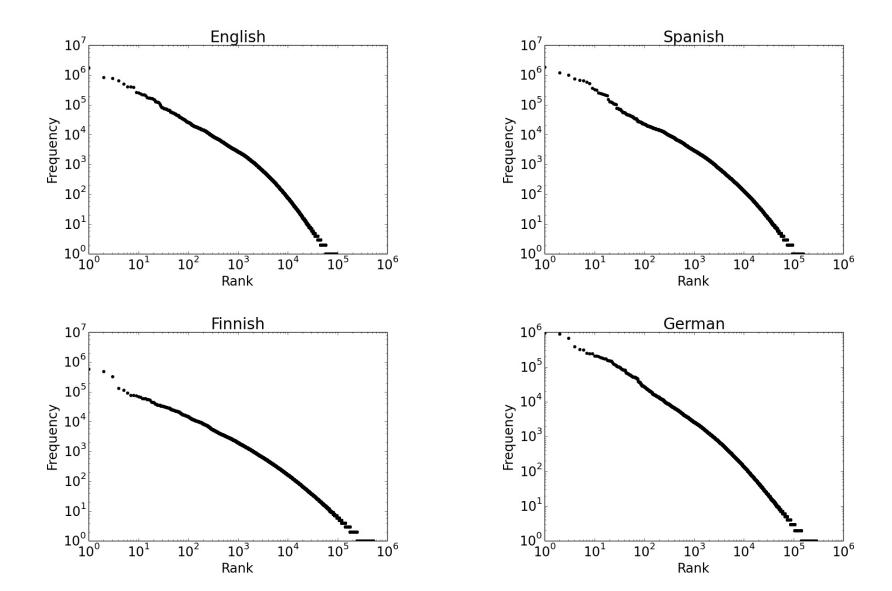




Rescaling the axes

To really see what's going on, use logarithmic axes:





Zipf's law

Summarizes the behaviour we just saw:

$$f \times r \approx k$$

- ullet f = frequency of a word
- r = rank of a word (if sorted by frequency)
- k = a constant

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Why a line in log-scales? $fr = k \implies f = \frac{k}{r} \implies \log f = \log k - \log r$

Implications of Zipf's Law

- Regardless of how large our corpus is, there will be a lot of infrequent (and zero-frequency!) words.
- In fact, the same holds for many other levels of linguistic structure (e.g., syntactic rules in a CFG).
- This means we need to find clever ways to estimate probabilities for things we have rarely or never seen.

3. Variation

• Suppose we train a part of speech tagger on the Wall Street Journal:

Mr./NNP Vinken/NNP is/VBZ chairman/NN of/IN Elsevier/NNP N.V./NNP ,/, the/DT Dutch/NNP publishing/VBG group/NN ./.

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• What will happen if we try to use this tagger for social media??

ikr smh he asked fir yo last name

Twitter example due to Noah Smith

4. Expressivity

 Not only can one form have different meanings (ambiguity) but the same meaning can be expressed with different forms:

> She gave the book to Tom vs. She gave Tom the book Some kids popped by vs. A few children visited Is that window still open? vs Please close the window

5 and 6. Context dependence and Unknown representation

- Last example also shows that correct interpretation is context-dependent and often requires world knowledge.
- Very difficult to capture, since we don't even know how to represent the knowledge a human has/needs: What is the "meaning" of a word or sentence? How to model context? Other general knowledge?

Organization of Topics

- Introduction, N-grams: Some basics of text processing, linguistics, and probabilistic models of word sequences.
- Classification, Lexical Semantics with Classical Approaches
- Sequential Prediction, Part-of-Speech Tagging with Classical Approaches, Annotation
- Word Embeddings and Neural Networks
- Hierarchical Sentence Structure: Syntax, Grammars, and Parsing
- Neural Text Generation and Large Language Models

Backgrounds

- This course has enrollment from multiple programs!:
 - Linguistics
 - Computer Science
 - possibly: Data Science; ...
- This means that there will be a diversity of backgrounds and skills, which is a fantastic opportunity for you to learn from fellow students.
- It also requires a bit of care to make sure the course is valuable for everyone.

What's NOT in this course

- Formal language theory
- Computational morphology
- Compositional semantics
- Speech/signal processing, phonetics, phonology

Related Courses

https://gucl.georgetown.edu/courses.php

Course Organization

- Instructor: Nathan Schneider
- TA: Tatsuya Aoyama (+ a 2nd TA to be announced)
- Lectures: TuTh 11:00–12:15 ET, White-Gravenor 211
- Office hours: stay tuned for times.
- Website: for syllabus, schedule (lecture slides/readings/assignments): https://people.cs.georgetown.edu/cosc5402/
 - Make sure to read the syllabus!
 - No hard-copy textbook; readings will be posted online.
- We will also use Canvas for communication, submitting assignments.

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- Registration:
 - Several of you are on the waitlist.
 - If you are not yet enrolled, bring a paper add form to the next class.
 - There will be room for some of the people on the waitlist, but I cannot guarantee a seat for everyone who wants one.