# Statistical Natural Language Processing

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> University of Tübingen Seminar für Sprachwissenschaft

> Summer Semester 2018

# Why study (statistical) NLP

- (Most of) you are studying in a 'computational linguistics' program
- Many practical applications
- Investigating basic questions in linguistics and cognitive science (and more)

# Application examples

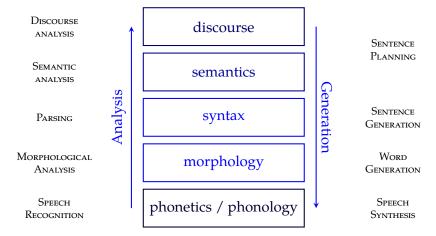
### For profit (engineering):

- Machine translation
- Question answering
- Information retrieval
- Dialog systems
- Summarization
- Text classification
- Text mining/analytics
- Sentiment analysis
- Speech recognition/synthesis
- Automatic grading
- Forensic linguistics

#### For fun (research):

- Modeling cognitive/social behavior
- Authorship attribution
- Investigating language change through time and space
- (Automatic) corpus annotation for linguistic research

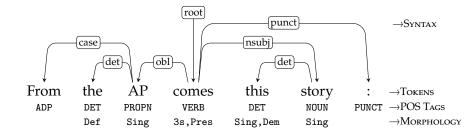
# Layers of linguistic analysis



From the AP comes this story :  $\rightarrow Tokens$ 

From the AP comes this story :  $\rightarrow$ Tokens add details approximately define the AP comes this story :  $\rightarrow$ Tokens had defined a property defined as  $\rightarrow$ Morphology

From	the	AP	comes	this	story	: $\rightarrow$ Tokens
ADP	DET	PROPN	VERB	DET	NOUN	$\mathtt{PUNCT} \ \to \! POS \ Tags$
	Def	Sing	3s.Pres	Sing.Dem	Sing	$\rightarrow$ Morphology



# Typical NLP pipeline

- Text processing / normalization
- Word/sentence tokenization
- POS tagging
- Morphological analysis
- Syntactic parsing
- Semantic parsing
- Named entity recognition
- Coreference resolution

# Do we need a pipeline?

- Most "traditional" NLP architectures are based on a pipeline approach:
  - tasks are done individually, results are passed to upper level
- Joint learning (e.g., POS tagging and syntax) often improves the results
- End-to-end learning (without intermediate layers) is another (recent/trending) approach

### On the word 'statistical'

But it must be recognized that the notion 'probability of a sentence' is an entirely useless one, under any known interpretation of this term. — Chomsky (1968)

- Some linguistic traditions emphasize(d) use of 'symbolic', rule-based methods
- Some NLP systems are based on rule-based systems (esp. from 80's 90's)
- Virtually, all modern NLP systems include some sort of statistical component

### What is difficult with NLP?

- Combinatorial problems computational complexity
- Ambiguity
- Data sparseness

# NLP and computational complexity

- How many possible parses a sentence may have?
- How many ways can you align two (parallel) sentences?
- How to calculate probability of sentence based on the probabilities of words in it?

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- How many possible parses a sentence may have?
- How many ways can you align two (parallel) sentences?
- How to calculate probability of sentence based on the probabilities of words in it?
- Many similar questions we deal with have an exponential search space
- Naive approaches often are computationally intractable

fun with newspaper headlines

• FARMER BILL DIES IN HOUSE

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- TEACHER STRIKES IDLE KIDS

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- DRUNK GETS NINE MONTHS IN VIOLIN CASE
- MINERS REFUSE TO WORK AFTER DEATH

we do not recognize many of them at first read

- Time flies like an arrow
- Outside of a dog, a book is a man's best friend
- One morning I shot an elephant in my pajamas

 Don't eat the pizza with knife and fork

we do not recognize many of them at first read

- Time flies like an arrow; fruit flies like a banana.
- Outside of a dog, a book is a man's best friend
- One morning I shot an elephant in my pajamas

 Don't eat the pizza with knife and fork

we do not recognize many of them at first read

- Time flies like an arrow; fruit flies like a banana.
- Outside of a dog, a book is a man's best friend; inside it's too hard to read.
- One morning I shot an elephant in my pajamas

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- Time flies like an arrow; fruit flies like a banana.
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- One morning I shot an elephant in my pajamas.
   How he got in my pajamas, I don't know.
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- Hearing voices? Then you're not alone!
- No parking on both sides.
- They are canning peas.
- My job was keeping him alive.
- We watched another fly.
- Double job pay.
- He fed her cat food.

## Even more ambiguities

#### with pretty pictures



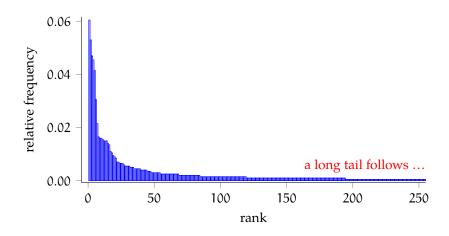
Cartoon Theories of Linguistics, SpecGram Vol CLIII, No 4, 2008. http://specgram.com/CLIII.4/school.gif

# Statistical methods and data sparsity

- Statistical methods (machine learning) are the best way we know to deal with ambiguities
- Even for rule-based approaches, a statistical disambiguation component is necessary
- Machine learning methods require (annotated) data
- But ...

## Languages are full of rare events

word frequencies in a small corpus



- Quick introduction / refreshers on important prerequisites
- The computational linguist's toolbox: basic methods and tools in NLP
- Some applications of NLP

#### Preliminaries

- Linear algebra, some concepts from calculus
- Probability theory
- Information theory
- Statistical inference
- Some topics from machine learning
  - Regression & classification
  - Sequence learning (HMMs)
  - Neural networks and deep learning
  - Unsupervised learning

#### NLP Tools and techniques

- Tokenization, normalization, segmentation
- N-gram language models
- Part of speech tagging
- Statistical parsing
- Distributed representations (of words, and other linguistic objects)

#### Applications

- Text classification
  - sentiment analysis
  - language detection
  - authorship attribution
  - ..

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#### If time allows

- Statistical machine translation
- Named entitiy recognition
- Text summarization
- Dialog systems
- ...

- Cutting edge, latest methods & applications
- In-depth treatment of particular topics
- Introduction to terms / concepts from linguistics

# Logistics

- Lectures: Mon/Fri 12:15 at Hörsaal 0.02
- Practical sessions: Wed 10:15 at Hörsaal 0.02
- Office hours: Wed 12:00-14:00 (room 1.09), or by appointment (email ccoltekin@sfs.uni-tuebingen.de)
- Course web page: http://sfs.uni-tuebingen.de/~ccoltekin/courses/snlp
- We will use GitHub classroom in this class (more on this soon)

# Reading material

- Daniel Jurafsky and James H. Martin (2009). Speech and Language Processing: An Introduction to Natural Language Processing, Computational Linguistics, and Speech Recognition. second. Pearson Prentice Hall, ISBN: 978-0-13-504196-3
  - Draft chapters of the third edition is available at http://web.stanford.edu/~jurafsky/slp3/
- Trevor Hastie, Robert Tibshirani, and Jerome Friedman (2009). The Elements of Statistical Learning: Data Mining, Inference, and Prediction. Second. Springer series in statistics. Springer-Verlag New York, ISBN: 9780387848587, URL:

http://web.stanford.edu/~hastie/ElemStatLearn/

# Grading / evaluation

- Seven graded homework assignments (5 % each)
- Final exam (70 %)
- Attendance
  - -5% (bonus) if you miss only one or two classes
  - you lose one bonus point for each additional class you miss
- Up to 5 % additional bonus points for Easter eggs:
  - first person finding intentional trivial mistakes in the course material gets 1 %

# Assignments

- For distribution and submission of assignments, we will use GitHub Classroom
- The amount of git usage required is low, but learning/using git well is strongly recommended
- You are encouraged to pair up for the assignments, but you cannot pair with the same person twice
- Late assignments up to one week, will be graded up to half points indicated
- The solutions will be discussed in the tutorial session after one week from deadline

# Assignment 0

- Your first assignment is already posted on the web page
- You need to follow the URL on the print version of the syllabus
- By completing assignment 0, you will
  - register for the course
  - have access to the non-public course material
  - exercise with how later assignments will work
  - provide some data for future exercises
- The repository created for assignment 0 is private, and can only be accessed you and the instructors

### Practical sessions

- We will start with two sessions on Python tutorial/refresher
- You need to bring your own computer, make sure you have a working Python interpreter
- You are encouraged to ask questions about the exercises during practical sessions
- You are encouraged to ask questions about the assignments
- The solutions will be discussed during tutorial sessions

# Further git/GitHub usage

- Once you complete Assignment 0, you will be a member of the 'organization' snlp2018
- You will get access to
  - private course material
  - assignment links
  - news and announcements

through the repository at https://github.com/snlp2018/snlp2018

- Make sure to watch this repository
- You are also encouraged to use 'issues' in this repository as a place to discuss course topics, ask questions about the material and assignments

### Next

Fri (this week) a hands-on introduction to Python

Mon Mathematical preliminaries (some linear algebra and bits
from calculus)

Wed Python tutorial (continued)

# References / additional reading material

URL: http://web.stanford.edu/~hastie/ElemStatLearn/.



Bishop, Christopher M. (2006). Pattern Recognition and Machine Learning. Springer. ISBN: 978-0387-31073-2.



Chomsky, Noam (1968). "Quine's empirical assumptions". In: Synthese 19.1, pp. 53-68. DOI: 10.1007/BF00568049.



Hastie, Trevor, Robert Tibshirani, and Jerome Friedman (2009). The Elements of Statistical Learning: Data Mining, Inference, and Prediction. Second. Springer series in statistics. Springer-Verlag New York. isbs: 9780387848587.



Jurafsky, Daniel and James H. Martin (2009). Speech and Language Processing: An Introduction to Natural Language Processing, Computational Linguistics, and Speech Recognition. second. Pearson Prentice Hall. ISBN: 978-0-13-504196-3.



Manning, Christopher D. and Hinrich Schütze (1999). Foundations of Statistical Natural Language Processing. MIT Press. ISBN: 9780262133609.