## **Statistical Natural Language Processing**

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November, 2003

# **Applications of NLP**

(NLP = Natural Language Processing)

Natural language and NLP play a central role in systems that

- augment textual or spoken data with information (e.g. automatic transcription of speech signals, part-of-speech tagging, named-entity recognition, parsing/chunking, word-sense disambiguation)
- transform textual or spoken data (e.g. text-to-speech, speech-to-text, spelling correction, text summarization, machine translation)
- extract information from textual or spoken data (e.g. information retrieval, question answering, information extraction, data mining)
- communicate with people (dialog systems)

## The aim of NLP

Scientific: Build models reflecting the human use of language and speech.

**Technological**: Build models that serve in technological applications.

The main NLP questions are:

- 1. What are the kind of things that people say and write?
- 2. What do these things mean?
- 3. How to incorporate the knowledge about these things into algorithms?

## The elements of NLP

**Phonetics/Phonology**: map acoustic signals to phoneme sequences and vice versa (speech recognition/synthesis)

**Morphology**: analyze the structure of words (morphological analysis)

**Syntax**: identify the category of words (POS tagging), analyze the structure of sentences (parsing/generation)

**Semantics**: calculate the meaning of words/sentences (lexical/compositional semantics)

Discourse: analyze the structure of dialog or text (discourse representation)

**Pragmatics**: incorporate world knowledge, cultural conventions, a specific use of language.

### How to build models of NLP?

Nowadays, there are two views on modeling what people write and say:

Competence (Chomsky,  $\sim 1960$ ): The traditional linguistic view on NLP

- A language is a set of word sequences,
- A sentence is a sequence of words in the language,
- A grammar is a device defining the language,
- So grammaticality of sentences is defined via a set membership test.

**Performance** ( $\sim 1990$ ): Given a specific NLP task and a specific domain of language use, the human language-behavior is nowadays modeled by a

#### (black-box) function: input $\longrightarrow$ output

The output that humans perceive as the most plausible for a given input.

## From competence to performance — What changed NLP?

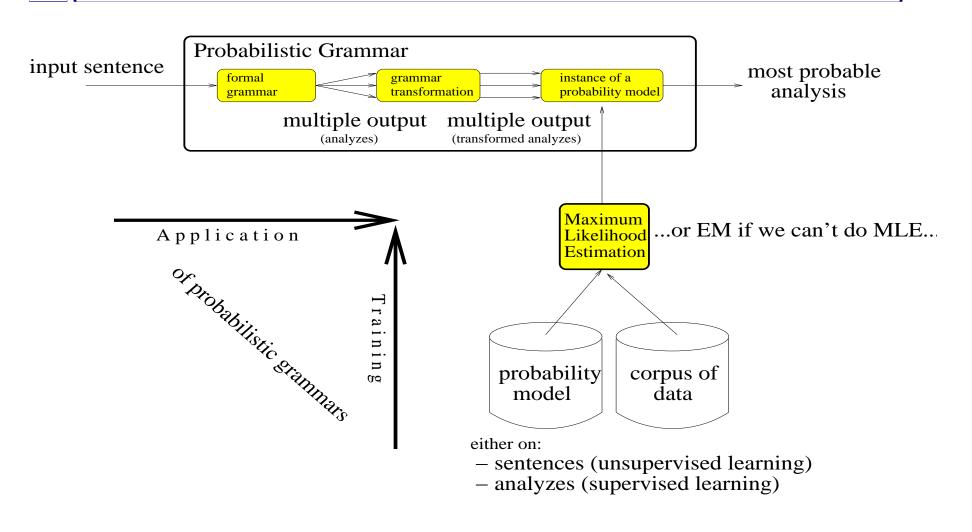
**Competence models:** In contrast to people, the linguistic view of language as a set does not care about problems caused by ambiguity/uncertainty. It

- cannot resolve multiple output
- cannot handle multiple input (resulting e.g. from noisy utterances)
- cannot express multiple levels of grammaticality

**Performance models** mimic people's language behavior and are specifically designed to resolve ambiguity. The majority of these models do the job by

- + generalizing competence models. Competence models serve simply as components of performance models.
- + dealing naturally with uncertainty concerning input/output and grammaticality, exploiting properly **Probability Theory and Statistics**.
- + incorporating the potential to model even extra-linguistic factors.

#### **Statistical NLP**



#### **Past LIT-Research on NLP**

- Computational Linguistics, Probability Theory, Statistics, and Machine Learning is our daily bread.
- **Probabilistic models for NLP**: We developed the world-best or at least state-of-the-art models for
  - Phonology (English, German) [CELEX-based probabilistic grammars]
  - Morphology (Hebrew)
  - Syntax (English, Dutch, German) [phrase-structure grammars]
  - Lexical Semantics (English, German) [semantically annotated lexica]

Beside this, we have a bit of experience in <u>question-answering</u> and <u>machine-translation</u> systems...

### **Future LIT-Research on NLP**

• **Deep Parsing**: Development of probabilistic grammars/parsers for several European languages (Dutch, English, German). Our main interest is to

integrate in our current probabilistic parsers: <u>morphology</u>, predicate-argument structure, and semantically annotated lexica.

- Learning Stochastic Tree Grammars from Treebanks: STGs work on the basis of tree-fragment probabilities. Compared to other grammars (PCFGs)
  - + STGs are more expressive and achieve better empirical results
  - Current STG estimation has several flaws: bias and inconsistency
- Creating NLP Tools for Question Processing
  - + QA systems: Desired information is completely specified by the question
  - Unfortunately, current NLP tools are of limited use for question processing

### **Download: Introduction to Statistical NLP**

- The course material of two courses given at the 15th European Summer School in Logic, Language and Information, Vienna, August 2003:
  - Probabilistic Models of NLP (sponsored by EACL)
  - Probabilistic Parsing

http://staff.science.uva.nl/~simaan/ESSLLI03.html

• A Tutorial on the Expectation-Maximization Algorithm Including Maximum-Likelihood Estimation and EM Training of Probabilistic Context-Free Grammars:

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http://staff.science.uva.nl/~prescher/papers/em/
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