

Text Mining

2004-2005

Master TKI

Antal van den Bosch en Walter Daelemans
<http://ilk.uvt.nl/~antalb/textmining/>

Dinsdag, 10.45 - 12.30, SZ33

Timeline

- [22 februari 2005]
 - Concept chunking (Sander Canisius)
- [1 maart 2005]
 - Syntactic pipeline 2: chunking, relation finding (WD)
- [8 maart 2005]
 - Named-entity recognition (Toine Borgers)

Outline

- Shallow Parsing
 - (Tokenization)
 - (POS Tagging)
 - Chunking
 - Relation-finding
- Applications
 - Information Extraction [15/3]
 - Ontology Extraction [26/4]
 - Question Answering
 - Factoid Extraction [3/5]

Shallow Parsing

- Steve Abney 1991 (FST)
 - <http://www.vinartus.net/spa/>
- Ramshaw & Marcus 1995 (TBL)
- CoNLL Shared tasks 1999, 2000, 2001
 - <http://cnts.uia.ac.be/signll/shared.html>
- JMLR special issue 2002
 - http://jmlr.csail.mit.edu/papers/special/shallow_parsing02.html

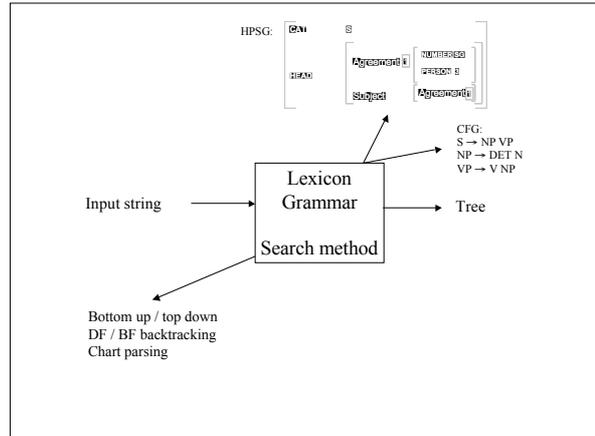
Formalisms for Computational Linguistics

Orthography	finite-state	spelling rules
Phonology	finite-state	text to speech
Morphology	finite-state	synthesis / analysis
	context-free	compounds
Syntax	context-free	parsing
	+ extensions	
Semantics	FOPC / CD	interpretation
Pragmatics		

- Classes of grammars are differentiated by means of a number of restrictions on the type of production rule
 - **Type-0-grammar** (unrestricted rewrite system). Rules have the form $\alpha \rightarrow \beta$
 - **Type-1-grammar** (context-sensitive). Rules are of the type $\alpha \rightarrow \beta$, where $|\alpha| \leq |\beta|$
 - **Type-2-grammar** (context-free). Rules are of the form $A \rightarrow \beta$, where $\beta \neq \epsilon$
 - **Type-3-grammar** (regular, finite). Rules are of the form $A \rightarrow a$ or $A \rightarrow aB$
- A grammar *generates* strings of $L(G)$, an automaton *accepts* strings of $L(M)$. Structure may be assigned as a side-effect.

The problem with full parsing

- Vicious trade-off coverage - ambiguity
 - The larger the grammar (more coverage), the more spurious ambiguity
- Why parsing ?
 - Structure of sentence determines its meaning



Shallow parsing

- Approximate expressive power of CFG and feature-extended CFG by means of a *cascade* of simple transformations
- Advantages
 - deterministic (no recursion)
 - efficient (1600 words per second vs. 1 word per second for a typical comparison)
 - accurate
 - robust (unrestricted text, partial solutions)
 - can be learned

Cascade

- POS tagging
- NP chunking
- XP chunking
- Grammatical relation assignment
- Function assignment
- Parsing

Chunk Parsing

Pierre Vinken, 61 years old, will join the board of directors as a non-executive director November 29.

Pierre/NNP Vinken/NNP ./, 61/CD years/NNS old/JJ ./, will/MD join/VB the/DT board/NN of/IN directors/NNS as/IN a/DT non-executive/JJ director/NN November/NNP 29/CD ./.

[_{NP} Pierre Vinken _{NP}] , [_{NP} 61 years _{NP}] old , [_{VP} will join _{VP}] [_{NP} the board _{NP}] of [_{NP} directors _{NP}] as [_{NP} a non-executive director _{NP}] [_{NP} Nov 29 _{NP}]

Approaches

Deductive

CASS-parser (Abney, 1991)
Finite-State

Fidditch (Hindle, 1994)
Rule-based

Inductive

Ramshaw & Marcus, 1995

Transformation Rules

Daelemans/Buchholz/Veenstra, 1999;
Tjong Kim Sang, 2000

Memory-based

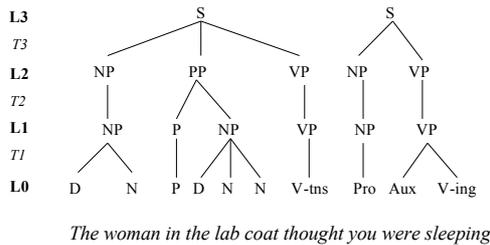
Abney (1991): CASS-parser

- Chunk = maximal, continuous, non-recursive syntactic segment around a head
- Comparable to morphologically complex word in synthetic languages
- Motivation
 - Linguistic (incorporate syntactic restrictions)
 - Psycholinguistic
 - Prosodic (phonological phrases)

Levels and Transformations

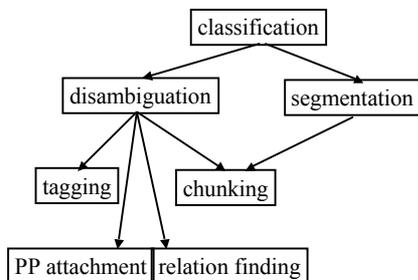
Levels

- words and their part of speech tags
- chunks (kernel NP, VP, AP, AdvP)
 - NP → D? N* N
 - VP → V-tns | Aux V-ing
- simple phrases (transforming embedding to iteration)
 - PP → P NP
- complex phrases
 - S → PP* NP PP* VP PP*



- Pattern = category + regular expression
- Regular expression is translated into FSA
- For each T_i we take the union of the FSAs to construct a recognizer for level L_i
- In case of more than one end state for the same input, choose the longest
- In case of blocking, advance one word
- “Easy-first parsing” (islands of certainty)
- Extensions: add features by incorporating actions into FSAs

MBLP Cascade: shallow parsing



NP Chunking as tagging

[_{NP} Pierre Vinken _{NP}] , [_{NP} 61 years _{NP}] old , [_{VP} will join _{VP}] [_{NP} the board _{NP}] of [_{NP} directors _{NP}] as [_{NP} a non-executive director _{NP}] [_{NP} Nov 29 _{NP}]

Pierre/I Vinken/I /O 61/I years/I old/O /O will/O join/O the/I board/I of/O directors/I as/O a/I non-executive/I director/I Nov/B 29/I /O

I	Inside chunk
O	Outside chunk
B	Between chunks

Memory-Based XP Chunker

Assigning non-recursive phrase brackets (Base XPs) to phrases in context:



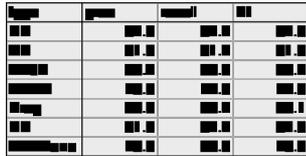
Convert NP, VP, ADJP, ADVP, PrepP, and PP brackets to classification decisions (I/O/B tags) (Ramshaw & Marcus, 1995).

Features:

POS₋₂, IOBtag₋₂, word₋₂,
 POS₋₁, IOBtag₋₁, word₋₁,
 POS_{focus}, word_{focus},
 POS₊₁,
 word₊₁, POS₊₂, word₊₂ → IOB tag

Memory-Based XP Chunker

- Results (WSJ corpus)



- One-pass segmentation and chunking for all XP
- Useful for: Information Retrieval, Information Extraction, Terminology Discovery, etc.

Finding subjects and objects

- Problems

- One sentence can have more than one subject/object in case of more than one VP
- One VP can have more than one subject/object in case of conjunctions
- One NP can be linked to more than one VP
- subject/verb or verb/object can be discontinuous

Task Representation

- From tagged and chunked sentences, extract
 - Distance from verb to head in chunks
 - Number of VPs between verb and head
 - Number of commas between verb and head
 - Verb and its POS
 - Two words/chunks context to left, word + POS
 - One word/chunk context to right
 - Head itself

Memory-Based GR labeling

Assigning labeled Grammatical Relation links between words in a sentence:



GR's of Focus with relation to Verbs (subject, object, location, ..., none)

Features:

Focus: prep, adv-func, word₊₁, word₀, word₋₁, word₋₂, POS₊₁, POS₀, POS₋₁,
 POS₋₂, Chunk₊₁, Chunk₀, Chunk₋₁, Chunk₋₂,
 Verb: POS, word,
 Distance: words, VPs, comma's
 → GRtype

Memory-Based GR labeling

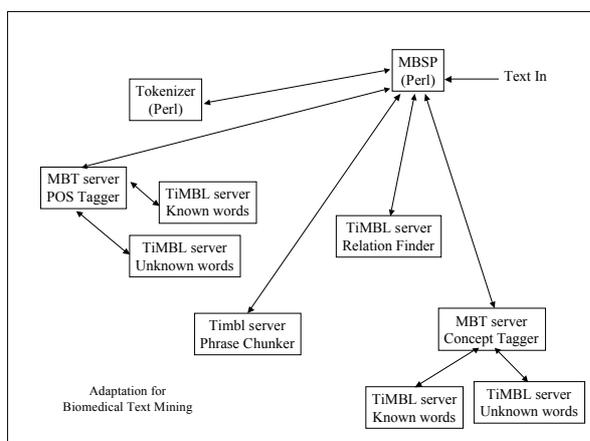
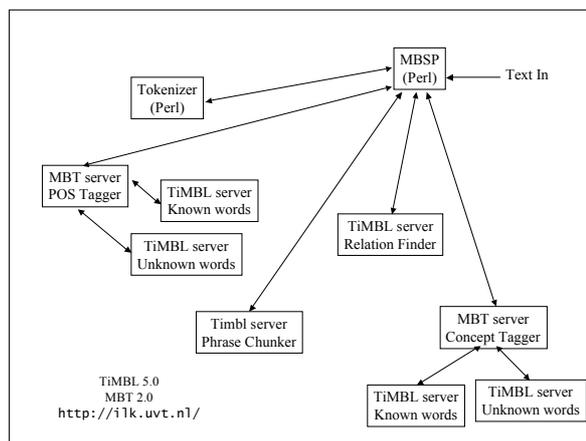
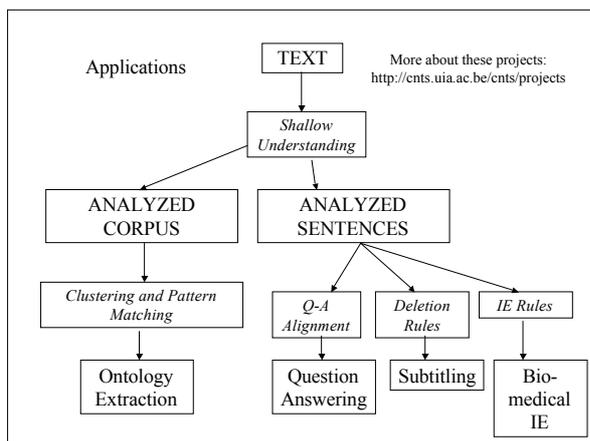
- Results (WSJ corpus)

Subject	Object	Location	Time
83%	87%	47%	63%

- Subjects: 83%, Objects: 87%, Locations:47%, Time:63%
- Completes shallow parser. Useful for e.g. Question Answering, IE etc.

From POS tagging to IE Classification-Based Approach

- POS tagging
The/Det woman/NN will/MD give/VB Mary/NNP a/Det book/NN
- NP chunking
The/I-NP woman/I-NP will/I-VP give/I-VP Mary/I-NP a/B-NP book/I-NP
- Relation Finding
[NP-SUBJ-1 the woman] [VP-1 will give] [NP-I-OBJ-1 Mary] [NP-OBJ-1 a book]
- Semantic Tagging = Information Extraction
[Giver the woman][will give][Givee Mary][Given a book]
- Semantic Tagging = Question Answering
Who will give Mary a book?
[Giver ?][will give][Givee Mary][Given a book]



What should be in?

- Shallow parsing (tagging, chunking, grammatical relations)
- Semantic roles
- Domain semantics (NER / concept tagging)
- Negation, modality, quantification can be solved as classification tasks?

Conclusions

- Text Mining tasks benefit from linguistic analysis (shallow understanding).
- Understanding can be formulated as a flexible hierarchy of classifiers.
- These classifiers can be trained on annotated corpora.

Assignment 1

- <http://ilk.uvt.nl/~antalb/textmining/assignment1.html>