LSA 125 - Psycholinguistics and Syntactic Corpora

Today: Extracting and importing

data from syntactic

corpora into a database

LSA Summer Institute 2009, UC Berkeley

Florian Jaeger

TAs: Judith Degen, Alex Fine, and Peter Graff



A forwarded plea for elevatorian consideration

Andrew Garrett:

"[please] avoid the north elevator, the one that requires [you] to walk through the Rhetoric or Classics Department corridor[...]; instead, [...] take the elevator that ends up between 370 and 7205."



Website, Passwords, Confusion

- Our website:
 - Info: http://wiki.bcs.rochester.edu:2525/HlpLab/Lsa09
 - Readings linked to website:
 http://www.hlp.rochester.edu/internal/classes/lsa09/readings/
 - Login: psycho
 - Password: *linguistics*
 - nothing on bSpace
- Our corpus server: (see website for most up-to-date info)
 - 174.129.5.193 (faster CPU)
 - 174.129.205.212 (maybe less traffic)
 - Login: *IsaXX*, e.g. *Isa18*
 - Password: only you know



The C-Team (again)

• Judith Degen

Alex Fine

• Peter Graff





Q&A

- Do you have questions about the processing accounts described in the previous lecture?
 - Accessibility
 - Availability
 - Alignment
 - Audience Design
 - Ambiguity Avoidance
 - Dependency Minimization (MiD, MaOP)
- We will get back to them next Tuesday when I will also discuss
 - Alternative audience design accounts
 - Uniform Information Density a computational account of efficient language production
 - Alternatives to processing accounts



Today

- Get your feet wet:
 - What is a syntactically-annotated corpus?
 - TGrep2 :: a tool to search syntactically-annotated corpora
 - TDTlite: a set of scripts we wrote to combine TGrep2 output into a database that can be handed to Excel or a stats program of your choice (e.g. R).
- Start thinking about your project have you found a work group?



Timeline for Corpus-based Project

- What is the structure of interest?
- What at the mark-up conventions of the corpus?
- Define & refine patterns () (TGrep2; TigerSearch; Tregexp):
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 - avoid over-exclusive (hard)
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that-omission

 Non-subject-extracted relative clauses in English allow optional *that*-omission:



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- Search tools for syntactic corpora developed by Doug Rohde (2005)
 - Downloadable for free: http://tedlab.mit.edu/~dr/Tgrep2/
 - Online tutorial:
 http://www.bcs.rochester.edu/people/fjaeger/teaching/tut
 orials/TGrep2/LabSyntax-Tutorial.html
- Parsed Switchboard in Penn Treebank format
 - 800,000 word syntactically annotated telephone conversation corpus (Switchboard, Treebank III)



A common syntactic annotation standard

- Syntactic structure annotation
 - Hierarchical dependencies
 - Linear order
 - Traces
 - Syntactic categories
- Predicate argument structure annotation
 - Grammatical functions (e.g. SUBJ, TOP, ADV, ...)
 - Modification types (e.g. NP-TEMP, ADV-LOC, ...)
 - Case marking preposition (e.g. PP-DTV)
- Part-of-speech (POS) annotation
- In Switchboard: disfluency (reparandum, repair)
- Genre, speaker, etc. information



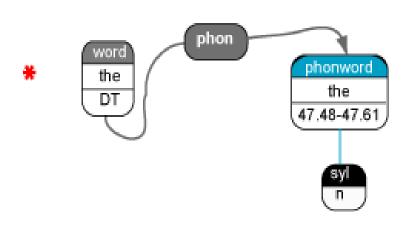
```
(TOP (S (NP-SBJ (NP (NNP Pierre)
                                                                WSJ
                     (NNP Vinken))
                 (,,)
                 (ADJP (NP (CD 61)
                           (NNS years))
                       (JJ old))
                 (\ ,\ ,\ )
        (VP (MD will)
             (VP (VB join)
                 (NP (DT the)
                     (NN board))
                 (PP-CLR (IN as)
                          (NP (DT a)
                              (JJ nonexecutive)
                              (NN director)))
                 (NP-TMP (NNP Nov.)
                          (CD 29))))
        (...)))
(TOP (S (NP-SBJ (NNP Mr.)
                 (NNP Vinken))
        (VP (VBZ is)
             (NP-PRD (NP (NN chairman))
                     (PP (IN of)
                          (NP (NP
                                 (NNP Elsevier)
                                  (NNP N.V.)) ...
```



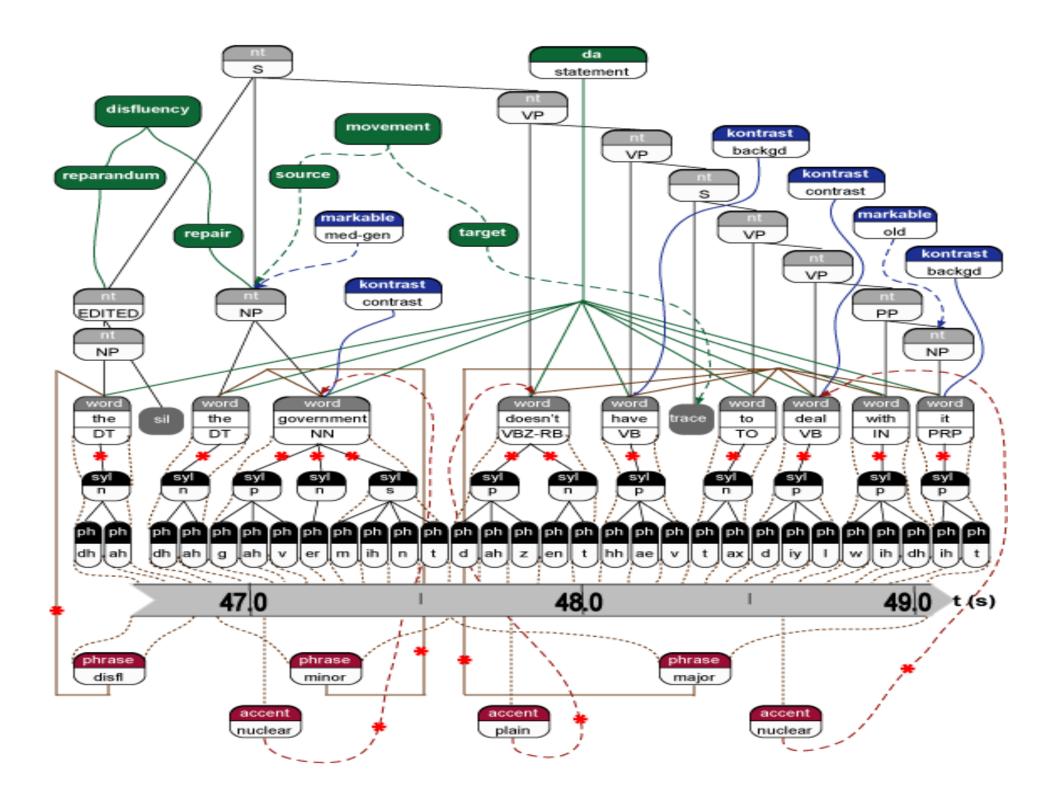
```
(TOP (CODE (SYM SpeakerA1)
                                                        SWBD
           (...))
(TOP (INTJ (UH Okay)
           (...)
           (-DFL- E_S)))
(TOP (S (INTJ (UH Uh))
        (,,)
        (ADVP-TMP (RB first))
        (,,)
        (INTJ (UH um))
        (,,)
        (NP-SBJ-1 (PRP I))
        (VP (VBP need)
            (S (NP-SBJ (-NONE- *-1))
                (VP (TO to)
                    (VP (VB know)
                        (,,)
                        (INTJ (UH uh))
                        (,,)
                        (SBARO (WHADVP-2 (WRB how))
                                (SQ (VBP do)
                                    (NP-SBJ (PRP you))
                                    (VP (VB feel)
                                         (ADVP (-NONE- *T*-2))
                                         (EDITED (RM (-DFL- \[))
LSA 125 - Psycholinguistics and Syntactic Corpora
                                                  (PP-UNF (IN about))
```

[13]

Annotations in SWBD: NITE XML



- Combination of annotations from different projects in one big data structure
- Nodes can
 - have children (hierarchical relationship)
 - point at other nodes (arbitrary relationship)
- Some nodes have timing information from original sound files



```
(WHADVP (N 400B34)
                    (WDT that))
13 3
          (S (NP-SBJ MARKABLE human (N 400B21)
                                        (PRP we))
              (VP (VBD had)
                  (S (NP-SBJ MARKABLE (-NONE- (N 400B21)))
                      (VP (TO to)
                          (VP (VB do)
                               (NP MARKABLE nonconc (PRP it))
                               (ADVP-TMP (-NONE- (N 400B34)))))))))
     (S (NP-SBJ_MARKABLE_human (PRP they))
                                 TGrep2 search pattern for RC*s
                                  ^SBAR
                                         > /^NP/
    < (/^WH/ != /PP/)
                                         < (/^S/ < (/-SBJ/ !< ``-NONE-''))
                                         ! < IN | WDT | DT
--Hore--
                                         !< ``-NONE-''
```

Data

 Over 3,700 RC*s (RCs with obligatory that were excluded) from approximately 350 different speakers



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Extracting all RC*s with a pronoun subject:

5:73

21:68

31:28

41:25

236:62

331:168

589:30

651:9

•••



Variables in the model

 Use a set of scripts (TGrep2 Database Tools; cf. class webpage as of today) to combine the output of many TGrep2 searches into a database of cases.

Probabilities:

- RC Predictability; Predictability of RC onset
- Frequency of words immediately preceding and following RC onset



Variables in the model

- Continuous syntactic variables, e.g.
 - Lengths of each of 3 regions (pre-NP, between head noun and RC, & RC)
- Categorical structural variables, e.g.
 - Embedding within the RC
 - Properties of RC subject (NP type, animacy)
 - Properties of matrix clause (negation, verb)
- Structural priming, e.g.
 - Within speakers
 - Across speakers
 - Distance-based; Lemma-based; etc ...



Variables in the RC* model

- Phonological variables, e.g.
 - segmental properties of preceding segment
 - stress structure of preceding segment
- Speech variables, e.g.
 - Speech rate, Pauses
 - Rate of disfluency in different regions
 - (Prosodic phrases & accents)
- Social variables, e.g.
 - Age
 - Speaker gender
 - Education



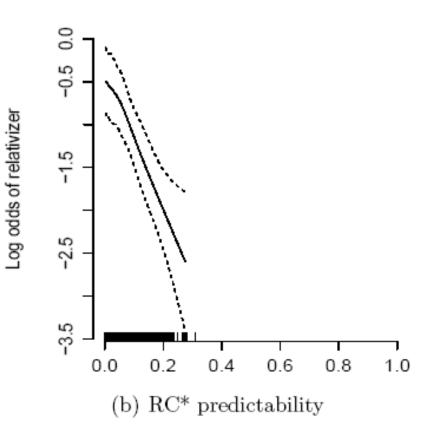
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Results of model

- Predictability one of the most influential factors
 - Both RC* predictability and the predictability of the RC* onset affect thatrates even when many other factors are considered



 As predicted by Uniform Information Density



Contemporary American English with Penn Treebank III annotation – Text

- Parts of ATIS-3
- Parsed Brown corpus, release 3
 - approx. 24,000 sentences & 396,000 words
 - 15 different written text categories of (good standard reference; like BNC).
- Parts of Wall Street Journal corpus (WSJ), release 3
 - approx. 24k sentences & 505,000 words [1 million out of 30 million]
 - Newspaper articles
 - Also available:
 - RST discourse annotation (for parts)
 - Propositional/event structure annotation (113,000 verb tokens; 3,200 verb types)
 - Automatically annotated extension to 30 million words



Contemporary American English with Penn Treebank III annotation – Speech

- International Corpus of English (ICE-GB)
 - approx. 84,000 sentences & 1 million words
 - Speech and written language
 - Not quite Treebank III annotation style
- Parts of Switchboard corpus (Swbd), release 3
 - approx. 100k sentences & 800,000 words [1 million out of 2 million]
 - Spontaneous speech
 - Also available:
 - Disfluency annotation (all)
 - Sound files (all)
 - Phonetic & phonological annotation (~38,000 words)
 - Animacy annotation (~140,000 NPs)
 - Information Structure annotation (~60,000 NPs)



Diachronic American English with Penn Treebank III annotation

- The York-Toronto-Helsinki Parsed Corpus of Old English Prose (YCOE)
 - approx. 110,000 sentences & 1.5 million words
 - Also available:
 - Text source, genre, dialect, and publication date information
- Helsinki Parsed Corpus of Middle English, second edition (PPCME2)
 - Over 100,000 sentences & 1.3 million words
 - Prose text samples of Middle English
 - Also available:
 - Text source, genre, dialect, and publication date information



POS & Syntactically annotated corpora of other languages - (1)

- Parsed NEGRA corpus, version 2
 - German
 - approx. 200,000 sentences
 - Newspaper articles (Frankfurter Rundschau)
 - Also available:
 - Morphological analysis (first 60,000 words)
- Parsed TIGER corpus
 - German
 - approx. 40,000 sentences & 700,000 words
 - same source as NEGRA
- Prague Dependency Treebank, version 1.0
 - Czech
 - approx. 1.8 million words



POS & Syntactically annotated corpora of other languages - (2)

- Penn Chinese Treebank, version 6
 - approx. 600,000 words
 - Newswire text
- Penn Arabic Treebank, Part 3, version 1.0
 - approx. 340,000 words
 - Newswire text
 - Also available:
 - Vocalization and Lemmatization information
 - Aligned translations into English (for parts)
- Penn Korean Treebank,
 - approx. 5,000 sentence & 55,000 words
 - 33 constructed texts in Korean (translated into English) for purposes of language training in a military setting.



Let's do some practice

Login to/login into/log into the corpus server

• If your username is **Isa1** to **Isa30** log onto the corpus server:

ssh <username>@174.129.5.193

• If your username is **Isa31** to **Isa99** log onto the corpus server:

ssh <username>@174.129.205.212



Sanity check

- Type env (and press enter):
 - TGREP2_CORPUS=/corpora/TGrep2able//swbd.t2c.gz
 - TGREP2ABLE=/corpora/TGrep2able/
 - TDTlite=/corpora/TDTlite/
 - TDT_DATABASES=/corpora/TDTlite/databases/
 - PATH=...:/corpora/TDTlite



- Type *tgrep2*
- tgrep2 -c <corpus> -af <output-options | outputformating> <macro-file> <pattern | pattern-file>
 - -c <corpus> defaults to TGREP2_CORPUS
 - -af gives all matches exactly once
 - -i makes TGrep2 case-insensitive (default is case-sensitive)

<output-options> and <macro-file> are optional



• ... a very simple call: let's find sentences in the default corpus (Switchboard)

tgrep2 "TOP" | more

[more gives output page-by-page – press ENTER or SPACE]



let's find NPstgrep2 "NP" | more

• Now let's count:

tgrep2 "NP" | wc -l

[wc -l counts lines of the output; TGrep2 defaults to one match per line]



TGrep2 – Different outputs

We can format the output:

```
tgrep2 -l "NP" | more
tgrep2 -t "NP" | more
tgrep2 -u "NP" | more
```

[be cautious with the tgrep2 -I | wc -I]

There are more options for later ...

TGrep2 – Regular Expressions

• Let's count *all* instances of *any type of* NP in the corpus:

```
tgrep2 -af "NP" | wc -l 
tgrep2 -af "/^NP/" | wc -l
```

- Investigate why there is a difference:
 tgrep2 -af "/^NP/" | more
- →Each node can be described as a regular expression /.../

[/^ .../ means that the node has to start with whatever follows the ^]



Across Corpora

Count all instances of any type of NP in the Wall Street
 Journal, Brown, and Switchboard corpus

Is \$TGREP2ABLE

brown.t2c.gz wsj_mrg.t2c.gz swbd.t2c.gz

tgrep2 -c \$TGREP2ABLE/<corpus-file> -af "/^NP/" | wc -l

 What's the ration of NPs (/^NP) to VPs (/^VP/) in the three corpora?



How many of these NPs have lexical content (as opposed to traces)?

$$tgrep2$$
 -af "/^NP/ << (/^'{0,1}[a-zA-Z].*/!<*)" | wc -I

- NB:
 - Left-headedness



Time to get real: PP-ordering in English

(Hawkins, 1999; taken from Hawkins, 2007:97)

(19) a. The man vp[waited pp1[for his son] pp2[in the cold but not unpleasant wind]]

1 2 3 4 5

b. The man vp[waited pp2[in the cold but not unpleasant wind] pp1[for his son]]

1 2 3 4 5 6 7 8 9

Structures like (19) were selected from a corpus on the basis of a permutation test (Hawkins, 2000, 2001): the two PPs had to be permutable with truth-conditional equivalence (i.e. the speaker had a choice). Only 15% (58/394) of these English sequences had long before short. Among those with at least a one-word weight difference (excluding 71 with equal weight), 82% had short before long, and there was a gradual reduction in the long before short orders, the bigger the weight difference (PPS = shorter PP, PPL = longer PP):

(22) PPL>PPS by 1 word by 2 4 by 5 6 by 7 + [V PPS PPL] 60% (58) 86% (108) 94% (31) 99% (68) [V PPL PPS] 40% (38) 14% (17) 6% (2) 1% (1)

Time to get real ...

 What should be the cases we extract to get all and only the relevant cases? (avoid inclusion and exclusion errors)

- VPs
- VPs with PPs
- VPs with PPs that are sisters to each other
- VPs with adjacent PPs that are sisters to each other
- VPs with exactly two adjacent PPs that are sisters to each other



Cheat sheet

- TGrep2 is left-headed!
- Syntactic relations: < > << >> \$ ~ =
- Linear relations: , .
- Labeling of nodes: =xx
- Disjunction | []
- Negation: !

```
/^VP/=VP1 < (/^PP/=PP1
$.. (/^PP/=PP2 !$ (/^PP/!= =PP1)
!,, (* !< * ,, =PP1
!>> (EDITED|UH|PRN|/-UNF/
>> =VP1))))
```



Macros

 Macros keep those precious fingers soft and smooth by avoiding to much typing

```
/^NP/;
@ NP
               /^VP/;
@ VP
               /^PP/;
@ PP
               /^(ADJ|ADV)P/;
@ AP
               /^WH/;
 WH
@ SBJ_ZERO
          (@SBJ) < (@ZERO);
          (@SBJ) !< (@ZERO);
@ SBJ_NERO
 SSBJ ZERO S < (@SBJ ZERO);
 SSBJ NERO
               S < (@SBJ_NERO);
```

