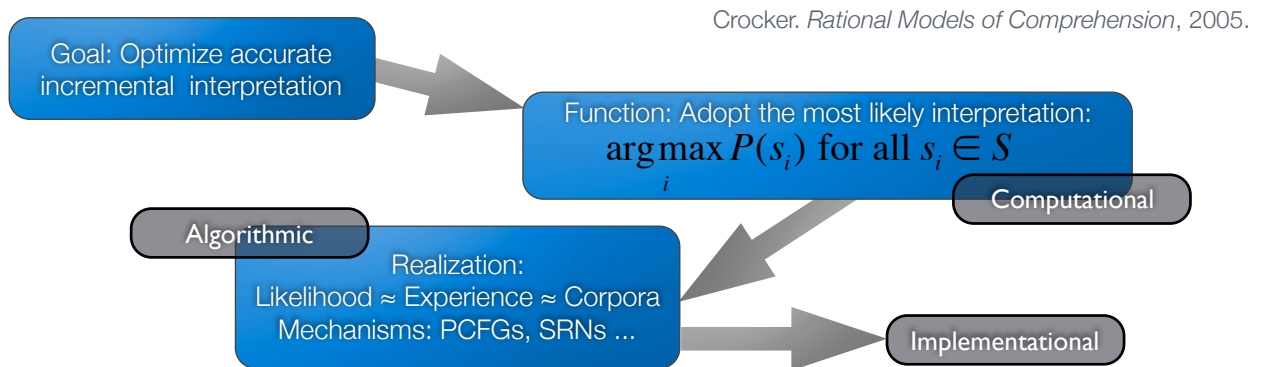


Computational Psycholinguistics

Lecture 6: **Category Disambiguation**

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$$\arg \max_i P(s_i) \text{ for all } s_i \in S$$

- Empirical: lexical access, word category/sense, subcategorization
- Rational: accurate, robust, broad coverage
- Rational Models:
 - explain accurate performance in general: i.e. rational behaviour
 - explain specific observed human behavior: e.g. for specific phenomena

Lexical Category Disambiguation

- Sentence processing involves the resolution of lexical, syntactic, and semantic ambiguity.

- Solution 1: These are not distinct problems

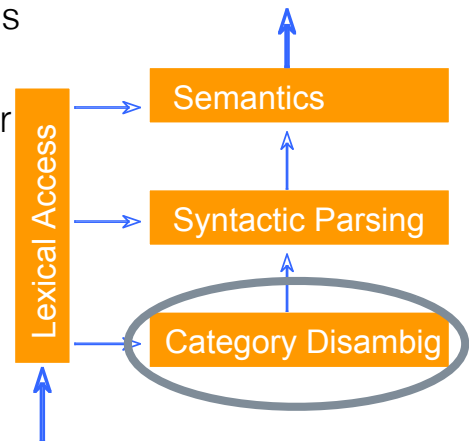
- Solution 2: Modularity, divide and conquer

- Category ambiguity:

- *Time flies like an arrow.*

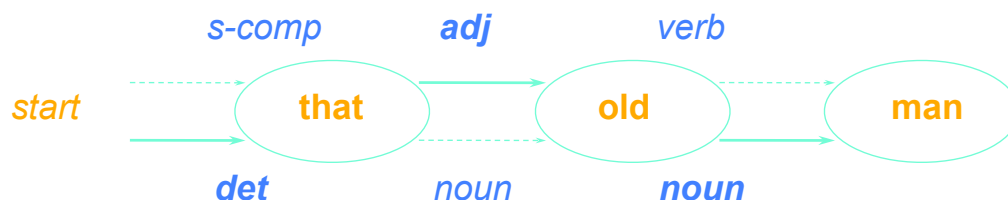
- Extent of ambiguity:

- **10.9%** (types) **65.8%** (tokens) (Brown Corpus)



The Model: A Simple POS Tagger

- Find the best category path ($t_1 \dots t_n$) for an input sequence of words ($w_1 \dots w_n$):
$$P(t_0, \dots, t_n, w_0, \dots, w_n)$$
- Initially preferred category depends on two parameters:
 - Lexical bias: $P(w_i | t_i)$
 - Category context: $P(t_i | t_{i-1})$
- Categories are assigned incrementally: Best path may require revision



Viterbi explained!

2 Predictions

- The Statistical Hypothesis:
 - Lexical word-category frequencies, $P(w_i|t_i)$, are used for initial category resolution
- The Modularity Hypothesis:
 - Initial category disambiguation is modular, and not determined by (e.g. syntactic) context beyond $P(t_i|t_{i-1})$.
- Two experiments investigate
 - The use word-category statistics
 - Autonomy from syntactic context

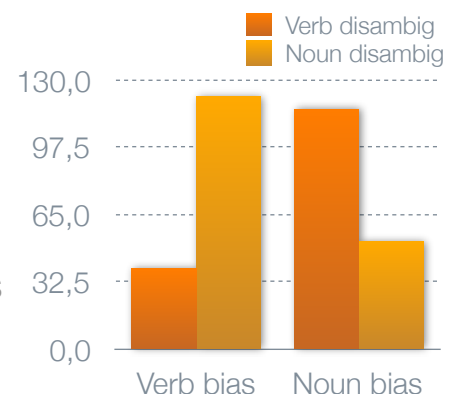
Statistical Lexical Category Disambiguation

- Initially preferred category depends on: $P(t_0, \dots, t_n, w_0, \dots, w_n) \approx \prod_{i=1}^n P(w_i | t_i) P(t_i | t_{i-1})$

- Categories are assigned incrementally
 - the warehouse prices the beer very modestly
 - DET N N / V V!
 - the warehouse prices are cheaper than the rest
 - DET N N / V N ...
 - the warehouse makes the beer very carefully
 - DET N N / V V
 - the warehouse makes are cheaper than the rest
 - DET N N / V N! ...

- Lexical bias: $P(w_i|t_i)$
- Category context: $P(t_i|t_{i-1})$ – constant!
- Trained on the Susanne corpus

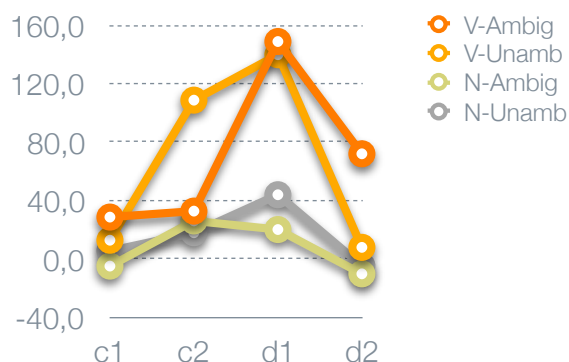
- Interaction between bias and disambiguation
- Category frequency determines initial decisions



Modular Disambiguation?

- Do initial decisions reflect integrated use of both lexical and syntactic constraints/biases or just (modular) lexical category biases?
 - N/V bias with immediate/late syntactic disambiguation as noun

- Main effect of bias at disambiguation:
 - Initial decisions ignore syntactic context.
 - Problematic for lexicalist syntactic theories
 - At c2, VA/VU difference is significant
 - Implies lexical category doesn't include number (!)



- [V-bias, N-disamb]** The warehouse **makes are** cheaper than the rest.
- [V-bias, N-unamb]** The warehouse **make is** cheaper than the rest.
- [N-bias, N-disamb]** The warehouse **prices are** cheaper than the rest.
- [N-bias, N-unamb]** The warehouse **price is** cheaper than the rest.

‘That’ Ambiguity (Juliano & Tanenhaus)

A. *That* experienced diplomat(s) would be very helpful ... [DET]

B. The lawyer insisted *that* experienced diplomat(s) would be very helpful [Comp]

- Corpus based estimates:
Initially: **det=.35 comp=.11** Post-verbally: **comp=.93 det=.06**
- Found increased RT when dispreferred POS for “that” (according to context) is forced in the disambiguation region “diplomat(s)”
- Advocates bigram over unigram: $P(t_i|t_{i-1})$

$$P(\text{that}|\text{comp}) = 1, P(\text{that}|\text{det}) = .171$$

$$P(\text{comp}|\text{verb}) = .0234, P(\text{det}|\text{verb}) = .0296$$

$$P(\text{comp}|\text{start}) = .0003, P(\text{det}|\text{start}) = .0652$$

t_i	Comp	Det
$t_{i-1} = \text{verb}$	.0234	.0051
$t_{i-1} = \text{start}$	.0003	.0111

Internal Reanalysis

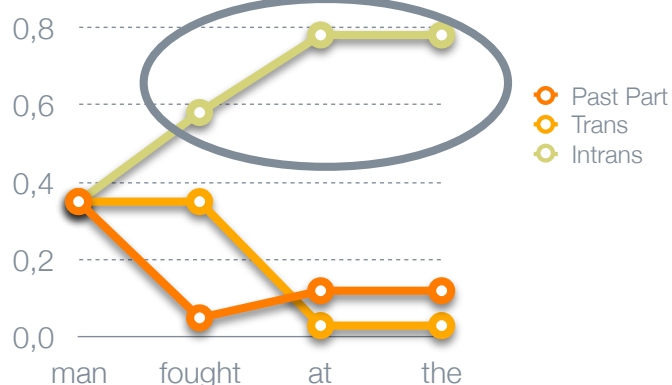
- The tagger model predicts internal reanalysis for some sequences.
- Viterbi: revise most likely category sequence based on next transition
- Right context in RR/MV ambiguities: [MacDonald 1994]
 - The sleek greyhound raced *at the track* won the event
 - The sleek greyhound admired *at the track* won the event
- **raced** = intrans bias, **admired** = trans bias
- Increased RT (blue) indicate transitivity bias is used

An SLCM Account

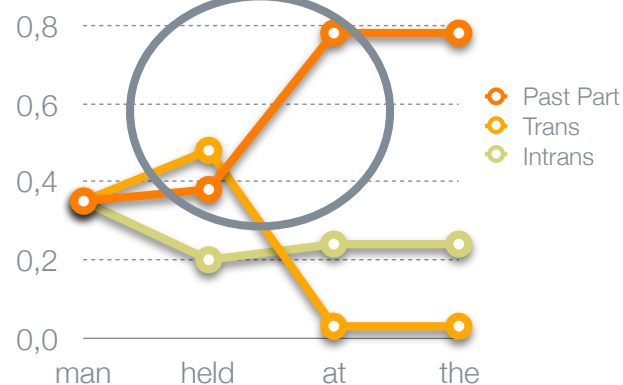
- Assume transitive/intransitive POS categories, extract frequencies from the Susanne corpus:

The man fought at the police station fainted [intransitive]
The man held at the police station fainted [transitive]

Predicts garden path for intransitives

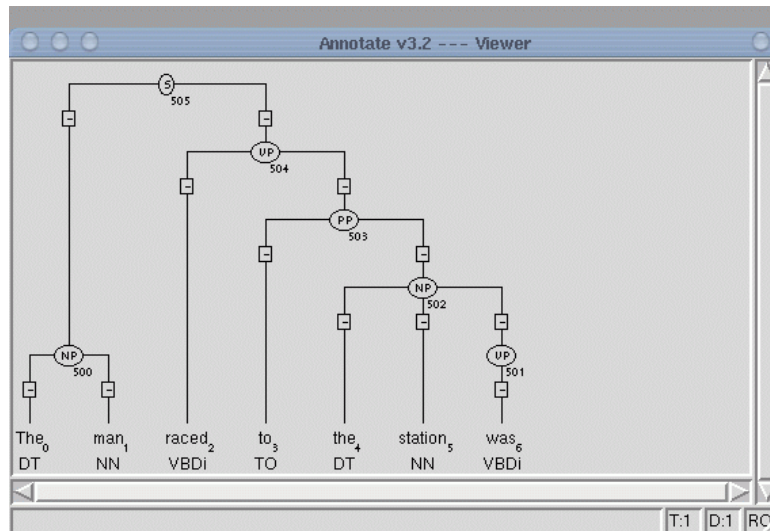


Predicts rapid reanalysis for transitives



Reduced Relative Clause

- Parsers can make wrong decisions that lead them up the garden path

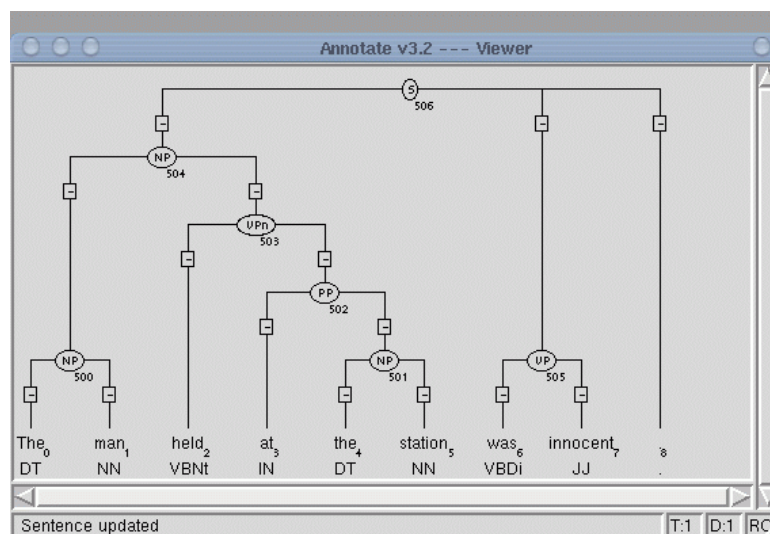


“The man raced to the station was innocent”

Crocker & Brants, *Journal of Psycholinguistic Research*, 2000.

The Problem

- In some cases it may be possible to recover from the error earlier



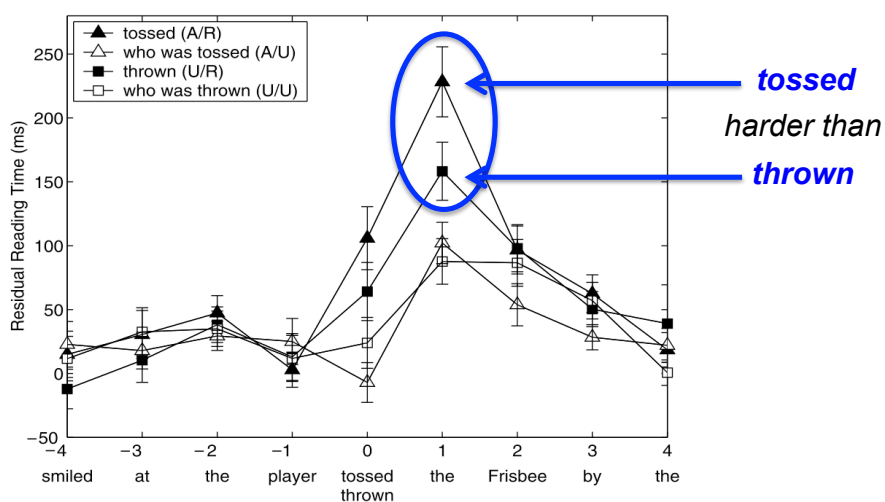
“The man held at the station was innocent”

Crocker & Brants, *Journal of Psycholinguistic Research*, 2000.

A Puzzle

- Sometimes local dependencies appears to violate the global parse:
 - **[A/R]** The coach smiled at the player tossed a frisbee by the ...
 - **[U/R]** The coach smiled at the player thrown a frisbee by the ...
 - **[A/U]** The coach smiled at the player who was tossed a frisbee by the ...
 - **[U/U]** The coach smiled at the player who was thrown a frisbee by the ...
- Syntactically, “tossed” is must a past-participle in this context, but what do people do?
- We might expect to see:
 - *Main effect of verb ambiguity:* if ambiguous verbs are difficult
 - *Main effect of structure ambiguity:* if ambiguous RRCs are difficult

Results:



- These results are problematic for theories requiring global syntactic wellformedness (e.g. Frazier, 1987; Gibson, 1991, 1998)

An SLCM Account

Note: explains the difficulty, but doesn't construct the local active clause interpretation! Is this right?

- Initially preferred category depends on two parameters:
 - Lexical bias: $P(w_i|t_i)$ Category context: $P(t_i|t_{i-1})$
- [AS-AV]** The coach smiled at the player **tossed** a frisbee ... [slowest]
 - $P(\text{tossed}|V_{\text{past}}) * P(V_{\text{past}}|\text{noun}) > P(\text{tossed}|V_{\text{part}}) * P(V_{\text{part}}|\text{noun})$
 - So: assign $\text{tossed}=V_{\text{past}}$, but can't integrate into parse, so reanalyse
- [US-AV]** The coach smiled at the player who was **tossed** a frisbee ... [fast]
 - $P(\text{tossed}|V_{\text{past}}) * P(V_{\text{past}}|\text{Aux}) < P(\text{tossed}|V_{\text{part}}) * P(V_{\text{part}}|\text{Aux})$
 - So: assign $\text{tossed}=V_{\text{part}}$, integrate into parse, no difficulty

SLCM Summary

- Psychologically plausible: lower statistical complexity than other models
- High accuracy in general: explains why people perform well overall
- Explains where people have difficulty
 - Statistical: category frequency **drives** initial category decisions
 - Modular: syntax structure **doesn't determine** initial category decisions
 - Bigram evidence: "that" ambiguity [Juliano and Tanenhaus]
 - Reanalysis of verb transitivity for 'reduced relatives' [MacDonald]

Comments on the SLCM

- Evidence category preference appears truly frequency-based
- Indication of which features are exploited [e.g. transitivity, not number]
 - But this is subject to further empirical investigation & verification
- Combines optimality of probabilities with advantages of modularity
 - psychological plausibility due to tractable parameter space
- Implications for the **Grain Problem**?
 - Bigrams used, but not tri-grams, or syntactic structure ?
 - Transitivity but not number ? More/less syntactically-rich POS tags ?

Probabilistic Syntax

- The SLCM is only a model of lexical category assignment
 - But note: these category decisions underlie many “syntactic” ambiguities
- Some ambiguities are purely syntactic, however:
 - Relative clause attachment, or other modifier attachment
 - NP/S complement ambiguity (unless subcat is encoded in the POS tags)
- Also evidence that compositional interpretation influences parsing
 - Can’t be modeled in the SLCM alone
- Apply probabilistic approaches to modeling human syntactic parsing