

Computational Psycholinguistics

Lecture 5: Reanalysis

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Pritchett (1992)

■ Theta-Attachment:

- The theta-criterion attempts to be maximally satisfied at every point during processing, given the maximal theta-grid

■ Theta-Criterion:

- Each argument must receive exactly one theta-role, and each theta role must be assigned to exactly one argument

■ Theta Reanalysis Constraint:

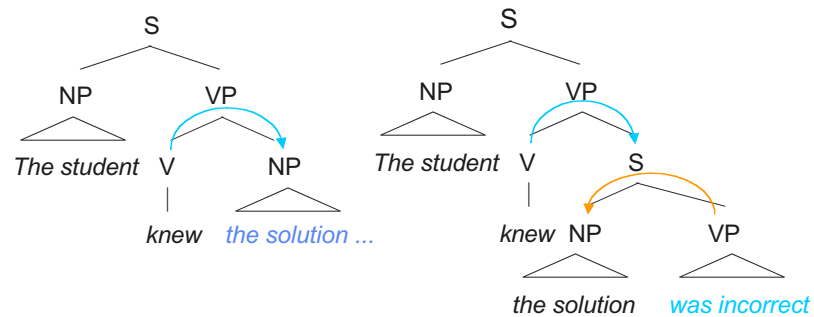
- Reanalysis of a constituent out of its theta-domain results in a garden-path effect

■ Generalised Theta Attachment:

- Every principle of the Syntax attempts to be maximally satisfied at every point during processing.

Theta-Reanalysis: Easy

- Reanalysis to a position within the original theta-domain is easy.



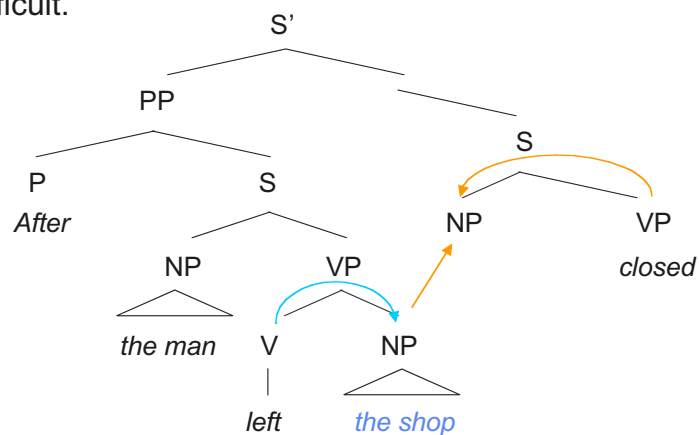
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Theta-Reanalysis: Difficult

- Reanalysis to a position outside the original theta-domain is difficult.



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Ambiguities revisited: [preferred/dis-preferred]

■ NP/VP Attachment Ambiguity:

- ❑ “The cop [saw [the burglar] [with the binoculars]]”
- ❑ “The cop saw [the burglar [with the gun]]”

■ NP/S Complement Attachment Ambiguity:

- ❑ “The athlete [realised [his goal]] last week”
- ❑ “The athlete realised [[his shoes] were across the room]”

■ Clause-boundary Ambiguity:

- ❑ “Since Jay always [jogs [a mile]] the race doesn’t seem very long”
- ❑ “Since Jay always jogs [[a mile] doesn’t seem very long]”

■ Red. Relative-Main Clause Ambiguity:

- ❑ “[The woman [delivered the junkmail on Thursdays]]”
- ❑ “[The woman [delivered the junkmail]] threw it away”

■ Relative/Complement Clause Ambiguity:

- ❑ “The doctor [told [the woman [that he was in love with]] [to leave]]”
- ❑ “The doctor [told [the woman] [that he was in love with her]]”

Grammar-Based

■ Theta-Attachment: relies on theta-grids

- ❑ Head driven
- ❑ Fine for English
- ❑ Not incremental for head-final languages
- ❑ Same problem for head-driven parsers, e.g. Marcus (1981), Abney (1989)

■ Crocker: A-Attachment

- ❑ Prefer attachment into ‘potential’ A-positions, over A'-positions
- ❑ A-position: a position which is visible for theta-role assignment

Experienced-based Models

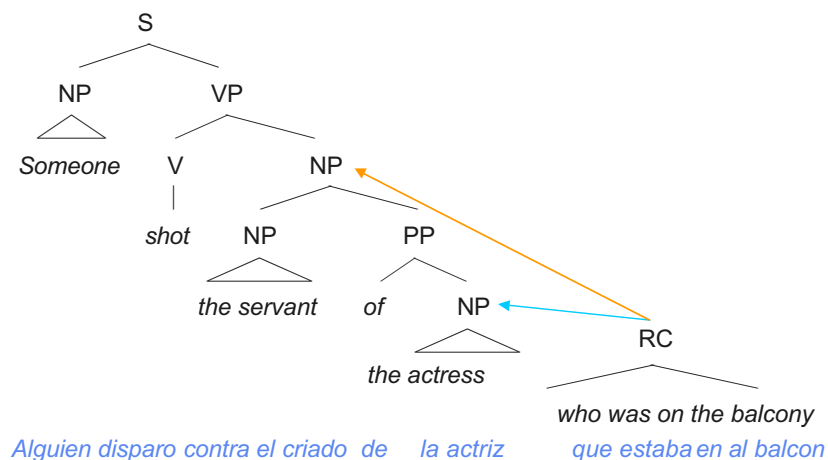
■ Resolve ambiguities according to linguistic experience:

- ❑ Lexical Guidance Hypothesis: (Ford et al)
 - + Resolve subcategorisation ambiguities in favour of the most likely frame for the ambiguous verb
- ❑ Linguistic Tuning Hypothesis: (Mitchell et al)
 - + Resolve structural ambiguities according to the structure which has previously prevailed

■ Relative clause attachment

- ❑ Someone shot the servant of the actress who was on the balcony

Relative Clause Attachment



Cross-linguistic RC Preferences

	Off-Line	On-Line
Spanish	high	low
French	high	low
Italian	high	low
Dutch	high	
German	high	low(early), high(late)
English	low	??, probably low
Arabic	low	
Norwegian	low	
Swedish	low	
Romanian	low	

- Immediate low attachment, possibly revised quickly (even on-line) ... seems the best account

The Grain Problem

- Experience-based models rely on frequency of prior linguistic exposure to determine preferences.
- There are many ways to realise experience-based models

- Possibilities: What kinds of things do we count?
 - + Actual sentence/structure occurrences? Data too sparse?
 - + Head driven: I.e. verb subcategorization frequencies
 - ▲ Do we distinguish tenses? Senses?
 - + Word level, part-of-speech
 - + Tuning is structural: NP P NP RC vs NP P NP RC
 - High Low

- Interesting issues:

- Does all experience have equal weight (old vs. new)?
- Are more frequent “words” or “strings” (idioms) dealt with using finer grain statistics than less frequent?

Summary

■ Shared features:

- ❑ Parsing is a distinct process
- ❑ Decisions are resolved using only syntactic information:
 - + structural principles
 - + grammar constraints
 - + frequency
- ❑ “Dependency formation” seems to be important:
 - + Associate argument to heads ASAP
 - + Associate fillers with gaps/heads ASAP
- ❑ Argument/Adjunct asymmetries

■ Are these models successful?

■ Do they support the notion of a modular architecture?

Reanalysis Revisited

■ Incrementality leads to local ambiguity

■ Local ambiguity leads to reanalysis:

- ❑ how does the HSPM recover?

■ Two possibilities

- ❑ Serial: Deterministic, Backtracking or ‘repair’
- ❑ Parallel: Re-rank or change activation

■ Are there different ‘sorts’ of reanalysis:

- ❑ “John knows the truth **hurts**” easy
- ❑ “While John was walking the dog **barked**” hard
- ❑ “The boat sailed down the river **sank**” v. hard

Deterministic Parsing

- Alternative: make sure there is only one possible parsing action at any point in the derivation.
 - ❑ Avoid/delay rule selection until it is fully determined
 - ❑ How: fully bottom up (e.g. S/R) plus lookahead
 - ❑ Marcus, Berwick and Weinberg, and Abney
- Advantage: very fast, clear predictions
- Disadvantages:
 - ❑ not fully incremental
 - ❑ unsuccessful for head-final languages
 - ❑ wrong predictions!

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Monotonic Parsing

- Inspired by determinism and reanalysis:
 - ❑ many local ambiguities seem to cause little difficulty
 - ❑ contra predictions of naive backtracking
 - ❑ Gorrell, Weinberg, Sturt & Crocker
- Provides a richer set of 'tree-building' operations which means destructive backtracking is not always required
- Predicts 'reanalysis' outwith these operations to be difficult

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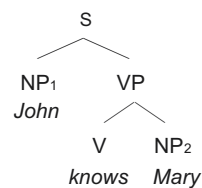
Talking about Talking about Trees

- Traditional theories locate the reanalysis process in the parser: “re-parsing”
- Can local ambiguity be handled using *underspecified* representations?
 - *Representations which allows some ambiguity to remain, and be later removed without reparsing.*
- Description-Theory: (Marcus, Hindle & Fleck, 1983)
 - Uses tree descriptions: dominance, rather than immediate dominance
 - Allows subsequent insertion of nodes & branches
[Gorrell (1995), Weinberg (1993), Sturt & Crocker (1996)]

Monotonic Parsing (Gorrell; Sturt & Crocker)

- Trees are described as a set of nodes, and a set of precedence and dominance relations:

- *John knows Mary*



- {dom(S,NP₁), dom(S,VP), dom(S,V), dom(S,NP₂), prec(NP₁,VP), dom(VP,V), dom(VP,NP₂), prec(V,NP₂) ...}

Properties of Trees

- Single root condition:

$$\exists x \forall y \cdot \text{dom}(x, y)$$

- Exclusivity condition:

$$\forall x, y \cdot \text{prec}(x, y) \vee \text{prec}(y, x) \leftrightarrow \neg \text{dom}(x, y) \wedge \text{dom}(y, x)$$

- Inheritance

$$\forall w, x, y, z \cdot \text{prec}(x, y) \wedge \text{dom}(x, w) \wedge \text{dom}(y, z) \rightarrow \text{prec}(w, z)$$

- *dom* and *prec* are transitive relations
- *dom* is reflexive, *prec* is irreflexive

Constraints on the Model

- Strict incrementality:

- ☐ words are connected to the tree description as they are encountered

- Coherence:

- ☐ tree properties must always be satisfied, the tree must be grammatically licensed

- Full specification of nodes:

- ☐ no features on nodes (e.g. bar-level) can be left unspecified

- Informational monotonicity:

- ☐ the tree description at state_{*n*} is a subset of the description at state_{*n+1*}

- Obligatory assertion of precedence:

- ☐ precedence must be specified for sisters

Monotonic Parsing and Reanalysis

■ Easy: *Monotonic* reanalysis

□ *John knows Mary is smart*

■ Hard: *Non-monotonic* reanalysis

□ *While John walked the dog barked*

■ How does the parser work:

- The monotonic parser is realised via specialised “macro”-like operations, which meet the constraints.
- This simplifies the parsers search process

■ Non-monotonic operations are not permitted

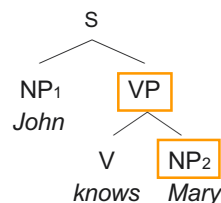
■ General Predictions:

- Easy reanalysis is simply parsing
- Hard reanalysis is not possible

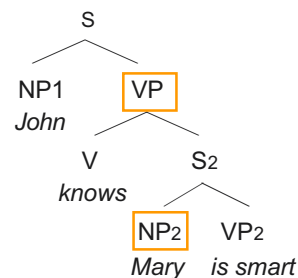
Monotonic Parsing Operation

■ Tree-lowering

□ John knows Mary ...



John knows Mary is smart

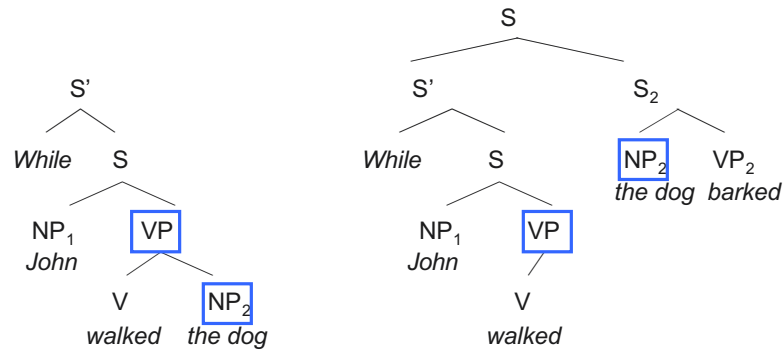


- {dom(S,NP1), dom(S,VP), dom(S,V), dom(S,NP2), prec(NP1,VP), dom(VP,V), dom(VP,NP2), prec(V,NP2), dom(VP,S2), dom(S2,NP2), prec(NP2,VP2) ...}

Non-Monotonic Parsing

■ Predicting difficult reanalysis

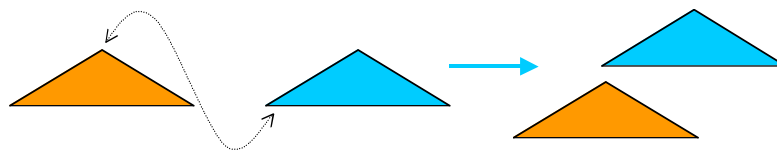
□ While John walked the dog ... barked.



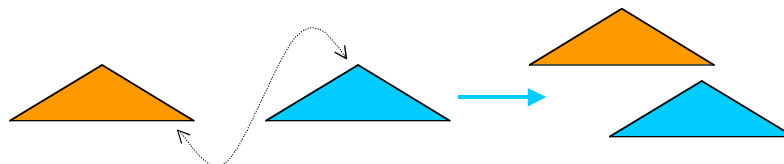
□ ... $dom(VP, NP_2)$... **but** ... $dom(VP, NP_2) \ \& \ prec(VP, NP_2)$...

Parsing Operations: Attachment

■ Left attachment:

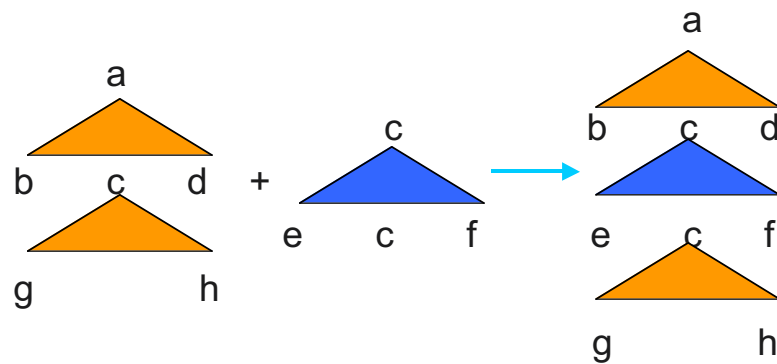


■ Right attachment:



TAG Adjunction

- The operations of the monotonic parser resemble those of Tree Adjoining Grammar.

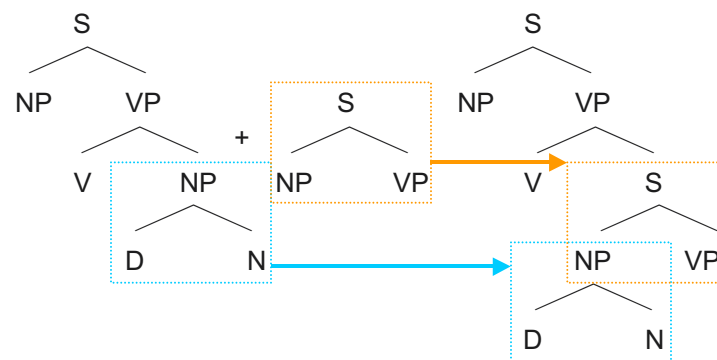


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Tree Lowering



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Psycholinguistic Evidence

■ Are there really two types of reanalysis?

- ❑ NP/S Ambiguous:
The woman saw the famous doctor had been drinking
- ❑ NP/Z Ambiguous:
Before the woman visited the famous doctor had been drinking
- ❑ NP/S Unambiguous:
The woman saw that the famous doctor had been drinking
- ❑ NP/Z Unambiguous:
Before the woman visited, the famous doctor had been drinking

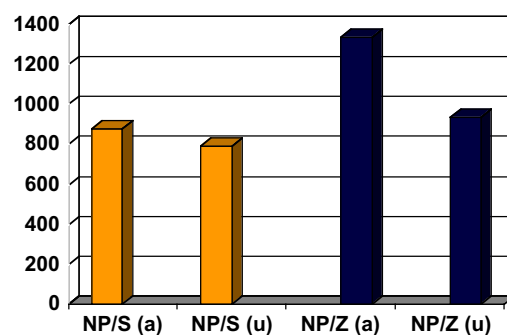
■ All verbs are biased (BNC) towards NP complement

- ❑ To make sure the object attachment is initially adopted, forcing reanalysis

■ Plausibility of the the direct object analysis is similar (pre-test).

Results

■ Reading times:

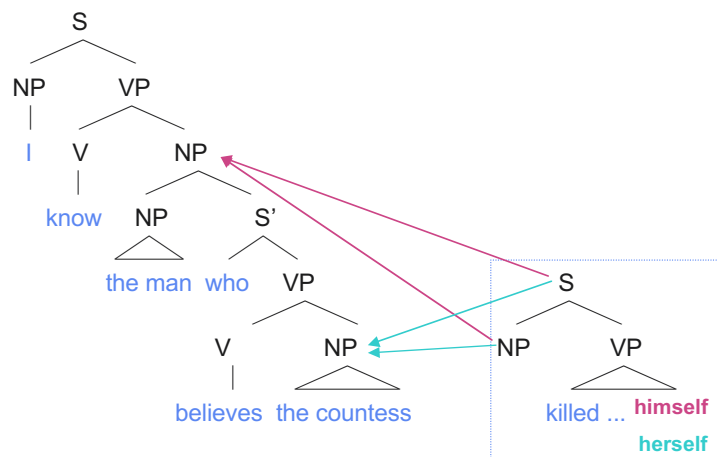


- ❑ Main effects of construction type, ambiguity, and a significant interaction
- ❑ GP effect: NP/Z (400ms) vs. NP/S (87ms)

Search in Parsing/Reanalysis

- (Some) reanalysis is simply monotonic attachment: e.g. tree-lowering.
- What if there are multiple such attachments
- Consider a double NP/S ambiguity:
 - I know the man who believes the countess killed *herself*
 - I know the man who believes the countess killed *himself*
- Which is easier?
- How does the parser search for an attachment?

The Trees



Search

- English appears to use a bottom-up search strategy for attachment

- ☐ late closure, recency

*Someone shot **the servant** of **the actress** who was on the balcony*

- Possibly other influences:

- ☐ verb bias? predicate proximity (Gibson)?

- Japanese seems to be top-down

- ☐ Head final language
- ☐ Sturt & Crocker(1996)