Introduction to Discourse

LING 571 — Deep Processing Methods in NLP November 21st, 2018 Ryan Georgi







Clarification

- In pseudocode from Monday:
 - incrementing support is done *after* determination of MI-LCS
 - That is, each probe word only increments support for one target sense.







-

Alternative Resnik WSD Pseudocode

for input word w_0 and probe words $\{p_1, \dots, p_n\}$ for $sense_w$ in NUMSENSES(W_0): most informative lcs = null most information = 0.0 for $sense_p$ in NUMSENSES(p_n): $lcs_{synset} = LowestCommonSubsumer(sense_w, sense_p)$ $lcs_{info} = INFORMATIONCONTENT(lcs_{synset})$ if $lcs_{info} > most$ information: most informative lcs = lcs_{synset} most information = lcs_{info} **increment** support[sense_w] by most information







Alternative Resnik WSD Pseudocode

for input word w_0 and probe words $\{p_1, \dots, p_n\}$ for $sense_w$ in NUMSENSES(W_0): most informative lcs = null most information = 0.0 for $sense_p$ in NUMSENSES(p_n): $lcs_{synset} = LowestCommonSubsumer(sense_w, sense_p)$ $lcs_{info} = INFORMATIONCONTENT(lcs_{synset})$ if $lcs_{info} > most$ information: most informative lcs = lcs_{synset} most information = lcs_{info} endfor

increment support[*sense*_w] **by** *most information*



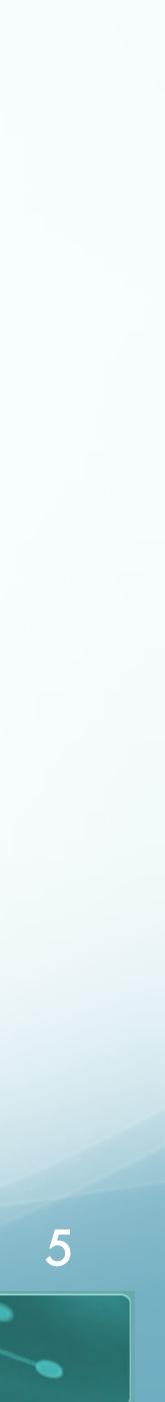
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Introduction to Discourse





Discourse is "a coherent structured group of sentences." (J&M p. 681)

- Discourse is language *in situ*
 - rather than synthetic, isolated sentences.
 - language use toward a goal



What is Discourse?





Different Parameters of Discourse

- Number of participants
 - Single author/voice → Monologue
 - Multiple participants → Dialogue
- Modality
 - Spoken vs.Written
- Goals
 - Transactional (message passing) vs. Interactional (relations, attitudes)
 - Cooperative task-oriented rational interaction



eractional (relations, attitudes) eraction







- Understanding depends on context
 - Word sense plant
 - Intention Do you have the time?
 - Referring expressions *it*, *that*, *the screen*



Why Discourse?

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- Applications: Discourse in NLP
 - Question-Answering
 - Information Retrieval
 - Summarization
 - Spoken Dialogue
 - Automatic Essay Grading



Why Discourse?





Reference Resolution

- Where is **A Bug's Life** playing in **Summit**? User:
- A Bug's Life is playing at the Summit Theater. System:
 - User: When is *it* playing *there*?
- It's playing at 2PM, 5PM, and 8PM. System:
 - I'd like I **adult** and 2 **children** for **the first show**. How much would **that** cost? User:
- Knowledge sources:
 - **Domain Knowledge**
 - **Discourse Knowledge**
 - World Knowledge

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From Carpenter and Chu-Carroll, Tutorial on Spoken Dialogue Systems, ACL '99





Not All Sentences Are Created Equal

- tomorrow.^[2]
- Summary:
- Inter-sentence coherence relations:
 - Second sentence: main concept (nucleus)
 - First sentence: background



• First Union Corp. is continuing to wrestle with severe problems.^[1] According to industry insiders at PW, their president, John R. Georgius, is planning to announce his retirement

• First Union President John R. Georgius is planning to announce his retirement tomorrow.





Coherence Relations

John hid Bill's car keys. He was drunk. John hid Bill's car keys. He likes spinach. 😐

- Why is this odd?
 - No obvious relation between sentences
- How is the first pair related?
 - statment explanation/cause
- Assumption: utterances should have meaningful connection
 - Establish through coherence relations

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Breaks our assumption as readers that information presented in discourse is relevant





Coherence Relations

John hid Bill's car keys. He was drunk. John hid Bill's car keys. He likes spinach.

Assumption

- Segments of discourse should have meaningful connection.
- Establish through coherence relations







Discourse: Looking Ahead



- Coreference
 - Cohesion
- Coherence
- Structure / Segmentation







Coreference Resolution





Queen Elizabeth set about transforming her husband, King George VI, into a viable monarch. Logue, a renowned speech therapist, was summoned to help the King overcome his speech impediment.

- referring expression: (refexp)
 - An expression that picks out entity (*referent*) in some knowledge model
 - Referring expressions used for the same entity **corefer**
 - Queen Elizabeth, her, the Queen
 - Logue, a renowned speech therapist
 - Entities in **purple** do not corefer to anything.

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Queen Elizabeth set about transforming her husband, King George VI, into a viable monarch. Logue, a renowned speech therapist, was summoned to help the King overcome his speech impediment.

• Antecedent:

- to
- Queen Elizabeth... her





• An expression that introduces an item to the discourse for other items to refer back





Reference: Terminology

Queen Elizabeth set about transforming her husband, King George VI, into a viable monarch. Logue, a renowned speech therapist, was summoned to help the King overcome his speech impediment.

Anaphora: An expression that refers back to a previously introduced entity.
cataphora: Introduction of expression before referent:
"Even before she saw it, Dorothy had been thinking about..."

*Not all anaphora is referential! e.g. "No dancer hurt their knee."







- Many forms:
 - Queen Elizabeth
 - she/her
 - the Queen
 - HRM
 - the British Monarch



Referring Expressions

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• Queen Elizabeth – she/her – the Queen – HRM – the British Monarch

- "Correct" form depends on discourse context
 - she, her presume prior mention or presence in the world
 - the Queen presumes an Anglocentric geopolitical discourse context generally or the UK (or British Commonwealth) specifically
 - (...i.e. likely a different interpretation during a RPDR viewing party.)



Referring Expressions





Discourse Model Correct interpretation of reference requires Discourse Model

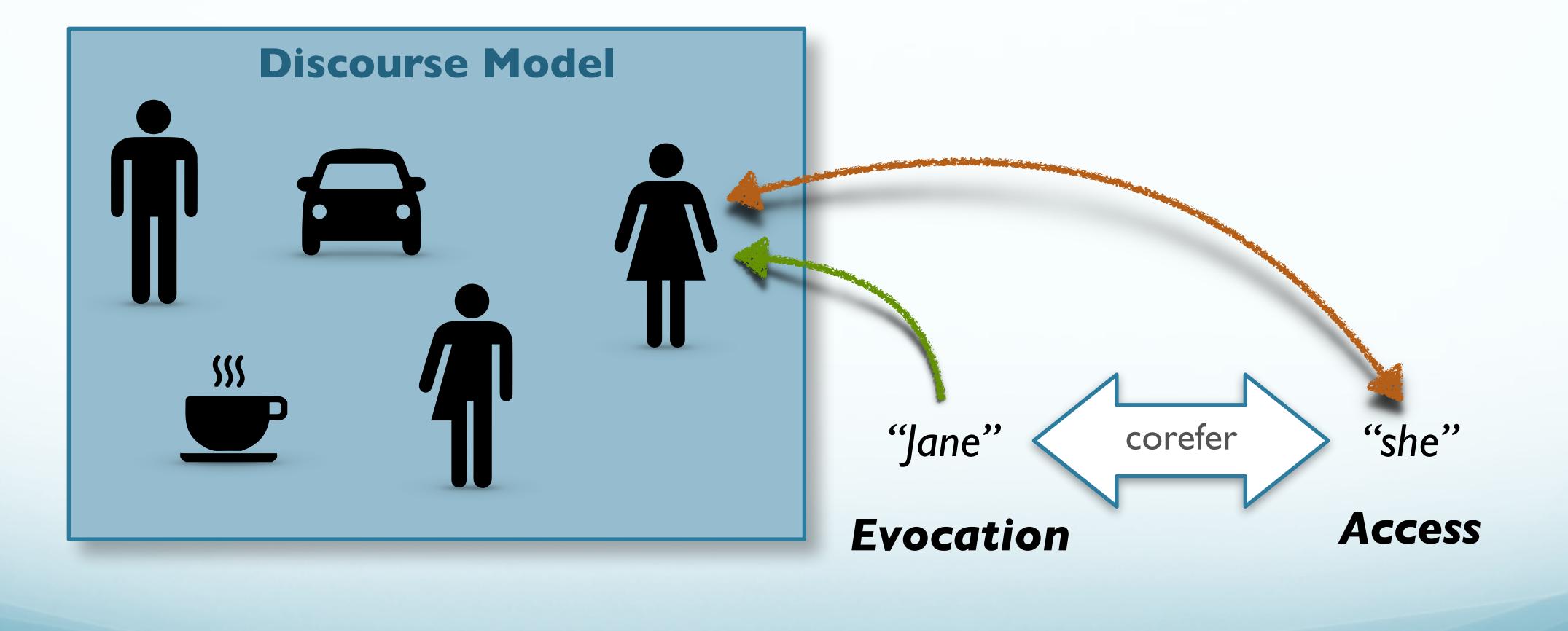
- - Entities referred to in the discourse
 - Relationships of these entities
- Need way to construct, update model
 - First mention of entity **evokes** entity *into* model
 - Subsequent mentions access entity from the model.







Reference and Model





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Reference Tasks

• Coreference resolution:

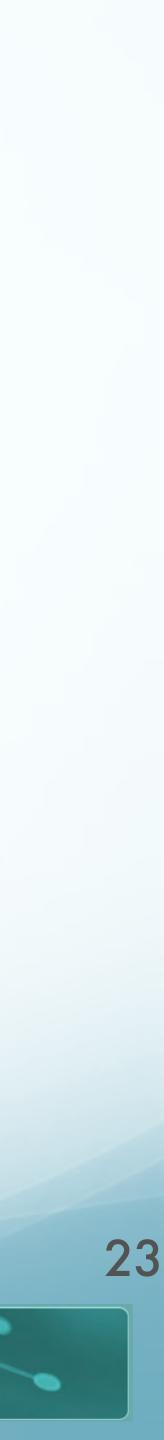
- Find all expressions referring to the same entity in a text.
- A set of coreferring expressions is a coreference chain.

• **Pronomial anaphora resolution**:

- Find antecedent for a single pronoun.
- Subtask of coreference resolution







Pronomial Anaphora Resolution







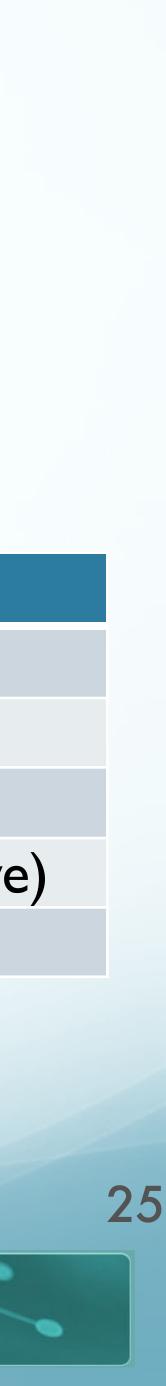
Reference Phenomena

Expression Type	Examples
Indefinite NP	"a cat", "some geese"
Definite NP	"the dog"
Pronouns	"he," "them," "zir"
Demonstratives	"this," "that"
Names	"Dr. Woodhouse," "IBM"



Constraints

Introduces new entity to context Refers to entity identifiable by hearer in context Refers to entity, must be "**salient**" Refers to entity, sense of distance (literal/figurative) New or old entities



Reference Phenomena: Activation/Salience

- a) John went to Erin's party, and parked next to a classic Ford Falcon.
- b) He went inside and talked to Erin for more than an hour.
- c) **Erin** told **him** that **she** recently got engaged.
- d) **?? She** also said that **she** bought **it** yesterday.
 - e) She also said that she bought the Falcon yesterday.
- d) is problematic because the Falcon has lost its salience.
- e) is acceptable because the definite NP has a further range for salience.







Information Status

- Some expressions introduce new information (ex: indefinite NPs)
- Other expressions refer to previous referents (ex: Pronouns)
- "Givenness hierarchy" (Gundel et al. 1993)

in focus >	activated >	familiar >
it	this	that N
	that	
	this N	



uniquely type identifiable > referential > identifiable the N indef. this N a N





Information Status

• Accessibility scale: (Ariel, 2001)

- More salient elements easier to call up, can be shorter
- correlates with length: more accessible, shorter refexp



o, can be shorter e, shorter refexp Full name+modifier ↓full name ↓long definite description ↓ short definite description \downarrow last name \downarrow first name \downarrow distal demonstrative+modifier ↓ proximate demonstrative+modifier ↓ distal demonstrative+NP ↓ proximate demonstrative+NP ↓ distal demonstrative(-NP) \downarrow proximate demonstrative (-NP) ↓ stressed pronoun+gesture ↓ stressed pronoun ↓unstressed pronoun ↓ cliticized pronoun ↓verbal person inflections ↓Ø PROFESSIONAL MASTER'S IN COMPUTATIONAL LINGUISTICS



Complicating Factors

Inferrables

- refexp refers to inferentially related entity:
- I bought **a car** today, but **a door** had a dent, and **the engine** was noisy.
 - a door, the engine \in a car

• Generics:

- I want to buy **a Jaguar**. **They** are very stylish.
- General group evoked by instance.
- Non-referential cases:
 - It's raining. (Pleonasm)
 - It was good that Frodo carried the ring. (Extraposition)

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Features for Anaphora Resolution: Constraints

Number:

• Anjali has a Corvette. *They are red.

Person:

• | st: *I*, we 2nd: you, y'all 3rd: he, she, it, they

Gender:

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- Janae plays the guitar.
- Janae plays the guitar.

It is red.

She sounds great.

It sounds great.





Features for Anaphora Resolution: Constraints

Binding Theory

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- How to handle reflexive pronouns vs. nonreflexives
 - Aaron bought themself a new car.
 - Aaron bought them a new car.
 - Jen said that Imani had bought her a new car.
 - Jen said that Imani had bought herself a new car.
 - He_1 said that he_2 had bought Willie a new car.

Pronoun/Def. NP: can't corefer with subject of clause

[them \neq Aaron] [her \neq]en] [herself = Imani] $[He_1 \neq Willie, he_2 \neq Willie]$





Features for Anaphora Resolution: Preferences

• Recency:

- Prefer closer antecedents.
- The doctor found an old map in the capa shelf. It described an island.

• Grammatical role:

- Saliency hierarchy of roles
- e.g. Subj > Object > Ind. Object > Oblique > AdvP
 - **Billy Bones** went to the bar with **Jim Hawkins**.
 - Jim Hawkins went to the bar with Billy Bones. He called for a glass of rum.



The doctor found an old map in the captain's chest. Jim found an even older map on the

ique > AdvP Hawkins. illy Bones.

He called for a glass of rum.He called for a glass of rum.

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Features for Anaphora Resolution: Preferences

Repeated Mention:

- - **Billy Bones** had been thinking of a glass of rum. **He** hobbled over to the bar. **Jim Hawkins** went with him. **He** called for a glass of rum.

Parallelism:

- Prefer entity in same role.
- **Silver** went with **Jim** to the bar. **Billy Bones** went with **him** to the inn.



Once entity is focused, likely to continue to be focused \rightarrow more likely pronomialized.





Features for Anaphora Resolution: Preferences

Verb Semantics

• Some verbs semantically bias for one of their argument positions. He had lost the laptop. John telephoned Bill. John criticized Bill. He had lost the laptop.

Selectional Restrictions

- Other kinds of semantic knowledge
 - John parked his car in the garage after driving it around for hours.
 - Understood that a car has the ability to **drive** whereas garage does not.







Reference Resolution Approaches

- Common features:
 - Use of a "Discourse Model"
 - Referents evoked in discourse, available for reference
 - Structure indicating relative salience
 - Syntactic & Semantic Constraints
 - Syntactic & Semantic Preferences
- Differences:
 - Which constraints/preferences? How to combine? Rank?







Hobbs' Resolution Algorithm

• **Requires**:

- Syntactic parser
- Gender & number checker

Input:

- Pronoun
- Parse of current and previous sentences

Captures:

- Preferences: Recency, grammatical role
- Constraints: binding theory, gender, person, number

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- Summary:
 - English-centric, rule-based algorithm.
 - Exploits English features of:
 - Agreement
 - Right-branching
 - SOV order
 - Inter-sententially, exploits notions of recency.



Hobbs Algorithm

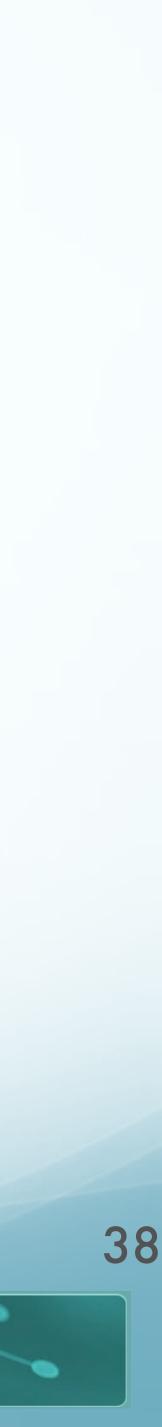


Hobbs Algorithm Detail (Hobbs, 1978)

- I. Begin at the noun phrase (NP) node immediately dominating the pronoun
- 2. Go up the tree to the first NP or sentence (S) node encountered. Call this node X, and call the path used to reach it p.
- 3. Traverse all branches below node X to the left of path p in a left-to-right, breadth-first fashion. Propose as the antecedent any encountered NP node that has an NP or S node between it and X.
- 4. If node X is the highest S node in the sentence, traverse the surface parse trees of previous sentences in the text in order of recency, the most recent first; each tree is traversed in a left-to-right, breadth-first manner, and when an NP node is encountered, it is proposed as antecedent. If X is not the highest S node in the sentence, continue to step 5.

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Hobbs Algorithm Detail (Hobbs, 1978)

- node X, and call the path traversed to reach it p.
- X immediately dominates, propose X as the antecedent.
- manner. Propose any NP node encountered as the antecedent.
- Propose any NP node encountered as the antecedent.
- 9. Go to step 4.

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5. From node X, go up the tree to the first NP or S node encounteed. Call this new

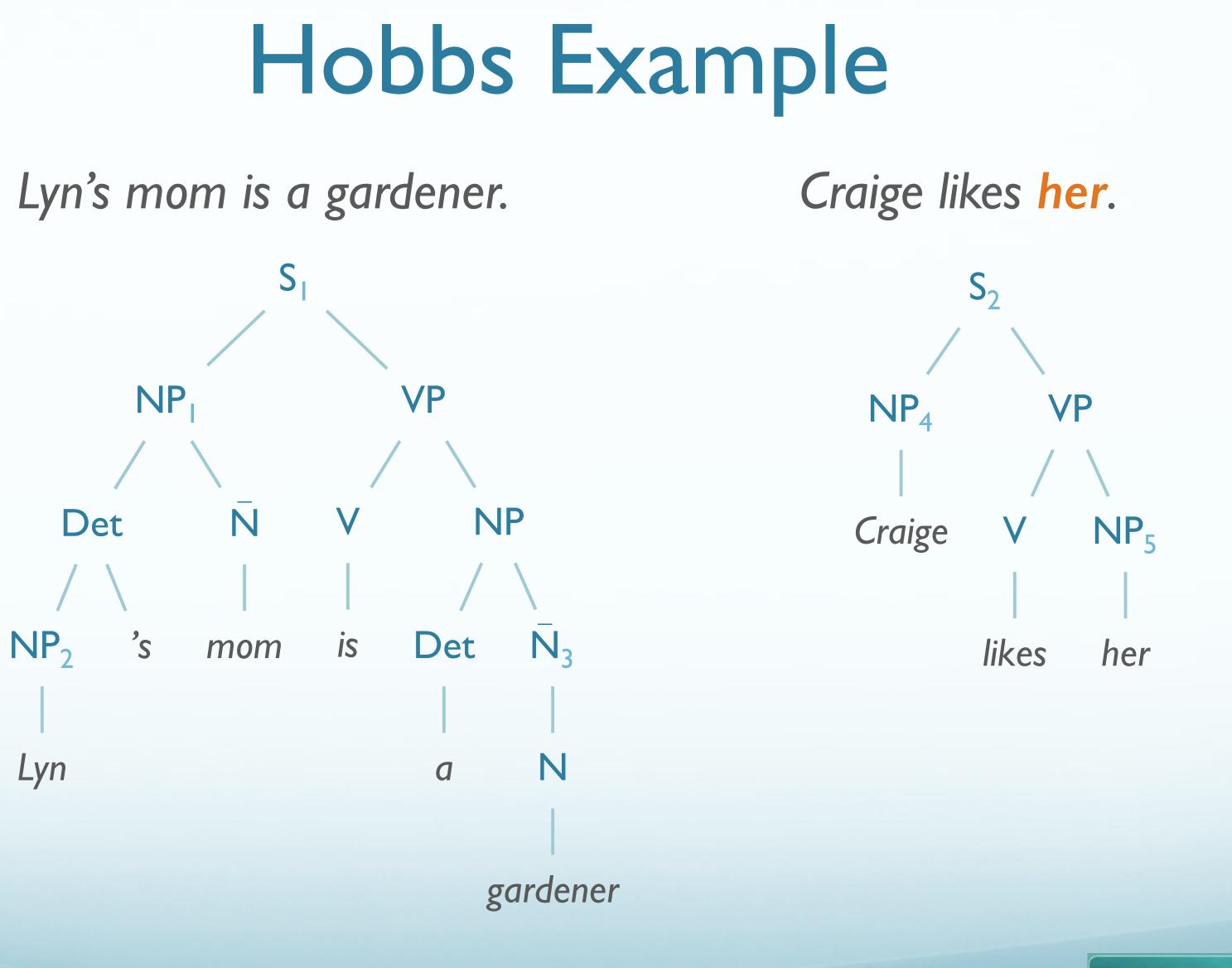
6. If X is an NP node and if the path p to X did not pass through the Nominal node that

7. Traverse all branches below node X to the left of path p in a left-to-right, breadth-first

8. If X is an S node, traverse all branches of node X to the right of path p in a left-toright, breadth-first manner, but do not go below any NP or S node encountered.

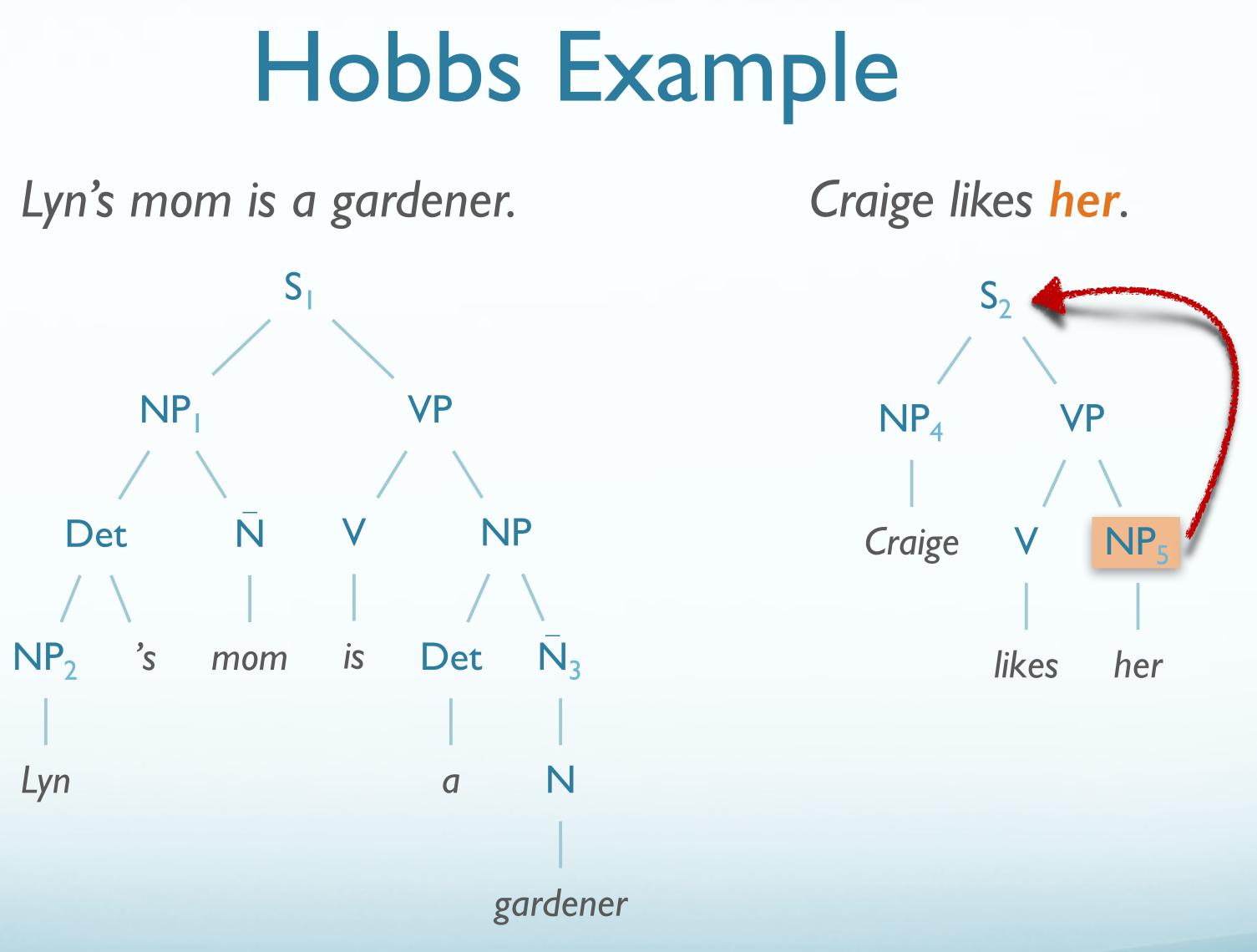






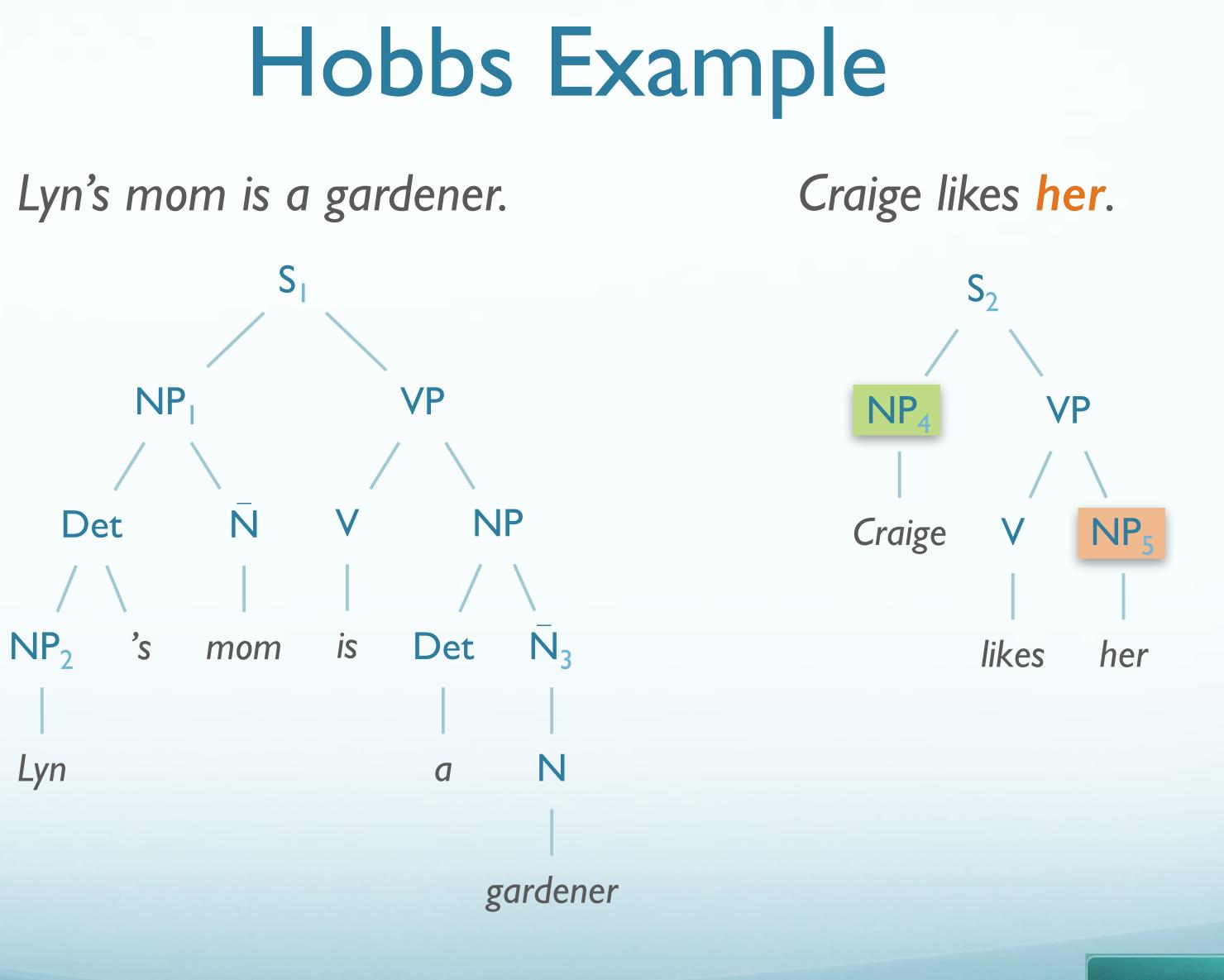






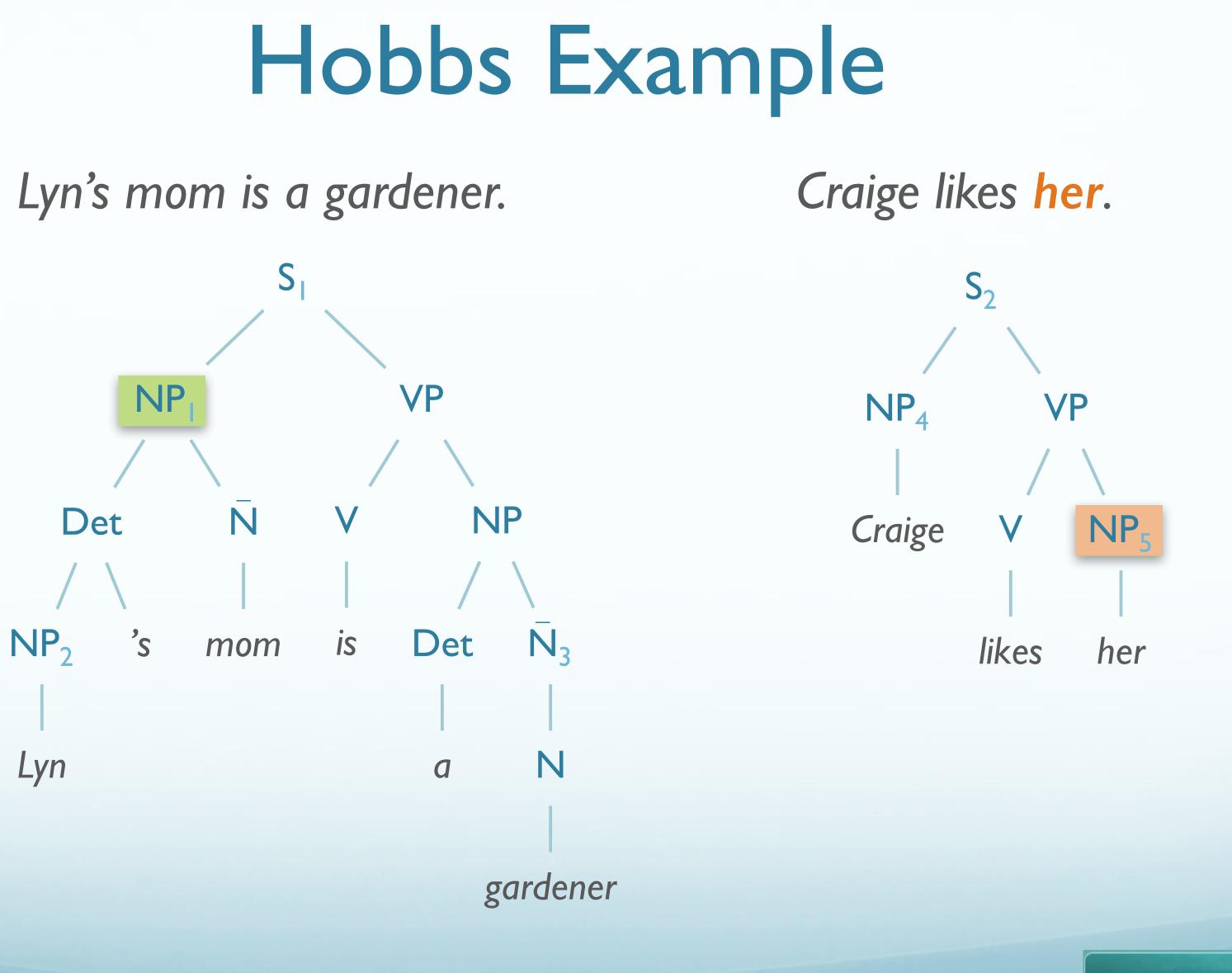






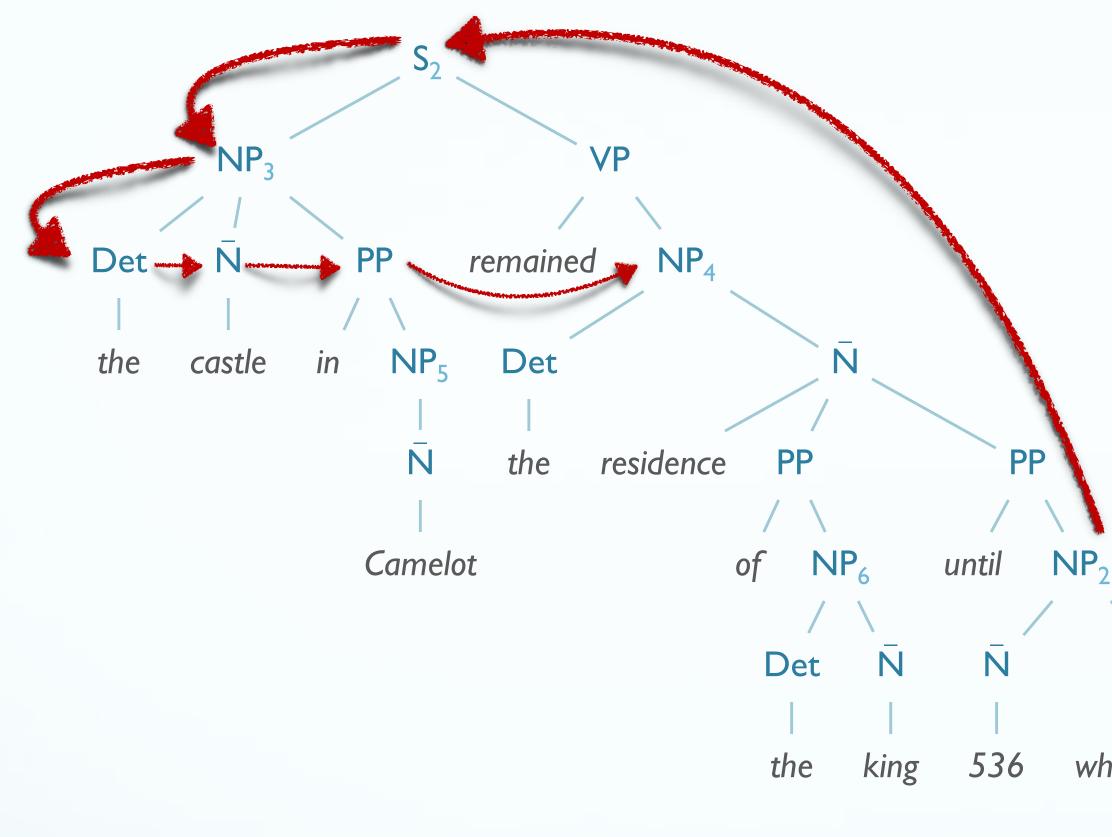








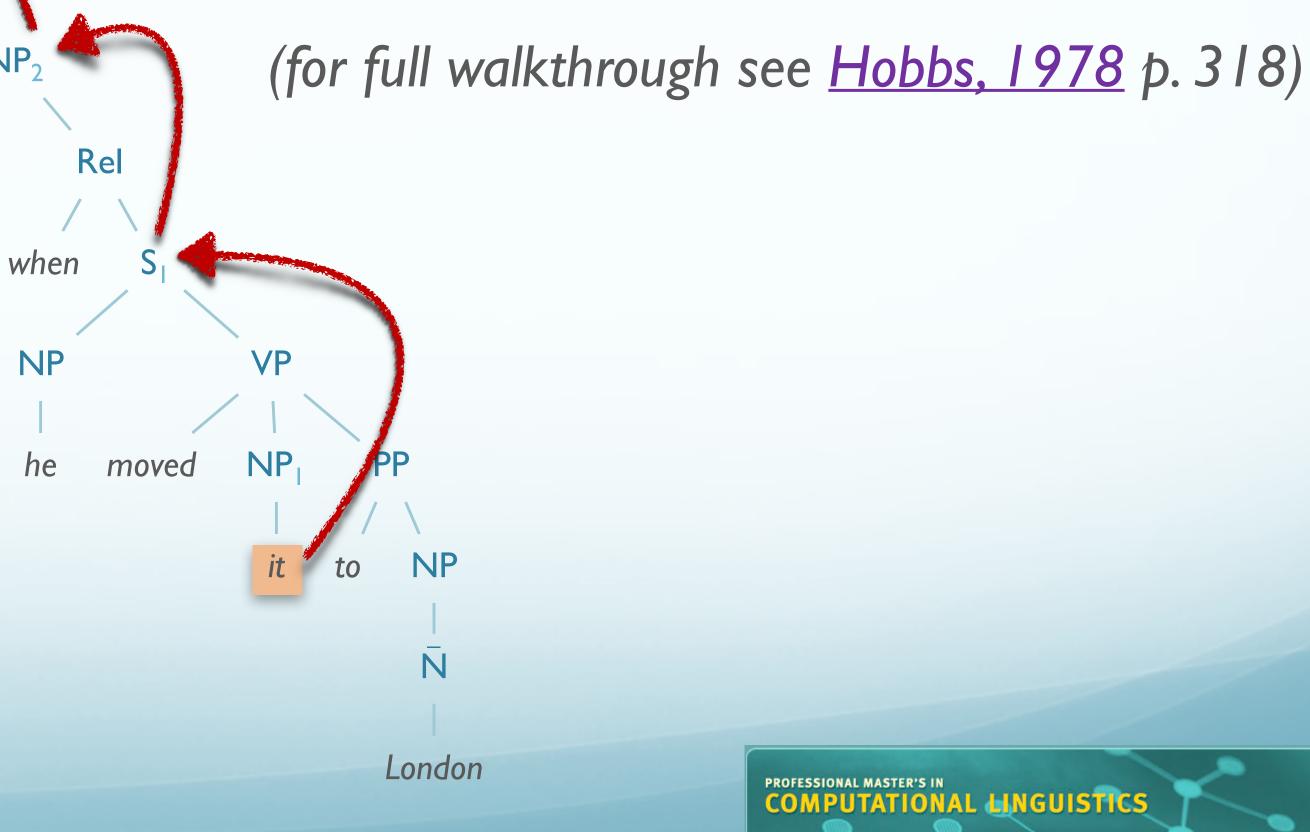






Another Hobbs Example

... the castle in Camelot remained the residence of the king until 536 when he moved it to London.







- Results: 88% Accuracy; 90% intrasentential
 - ... on perfect, manually parsed sentences
- Useful **baseline** for evaluating pronomial anaphora
- ssues:
 - Parsing:
 - Not all languages have parsers
 - Parsers not always accurate
 - **Constraints/Preferences:**
 - Captures: Binding theory, grammatical role, recency
 - But not: parallelism, repetition, verb semantics, selection

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Hobbs Algorithm





- Other issue: does not implement world knowledge
 - The city council refused the women a permit because they feared violence.
 - The city council refused the women a permit because they advocated violence. (Winograd, 1972)
- permits would be refused.



Hobbs Algorithm

• Get this reading by knowledge of city councils and permitting, and reasons why





Hobbs Algorithm: A Parable

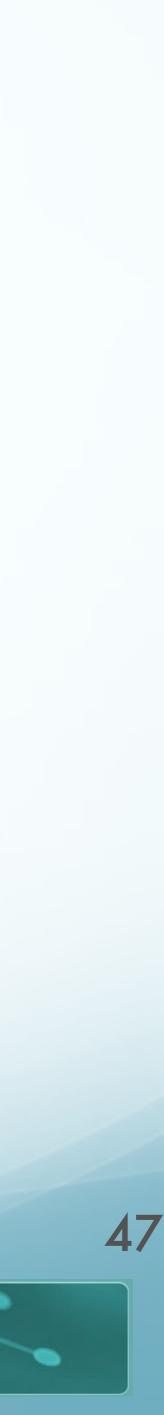
- Was actually one of the first instances in NLP where a researcher tried an informed, if "naïve" baseline
 - ...found that (in 1972) no system he could build could beat it!
- "the naive approach is quite good. Computationally speaking, it will be a long time before a semantically based algorithm is sophisticated enough to perform as well, and these results set a very high standard for any other approach to aim for.

antecedent." — Hobbs (1978), Lingua, p. 345

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"Yet there is every reason to pursue a semantically based approach. The naïve algorithm does not work. Any one can think of examples where it fails. In these cases it not only fails; it gives no indication that it has failed and offers no help in finding the real













Goals

- Explore the task of pronomial anaphora resolution
- Gain familiarity with syntax-based resolution techniques
- Analyze the effectiveness of the Hobbs algorithm by applying it to pairs of parsed sentences.







- Given pairs of sentences (S_0, S_1) as context
 - Resolve the pronoun(s) in S_1 using the Hobbs algorithm. • J&M p. 704-705

Subtasks:

- **Parsing Sentences** Automatic (CKY, Earley, etc)
- Hobbs Algorithm May be done either:
 - Manually manually mark up the output parse tree
 - **Coded** implement Hobbs algorithm will require feature grammar or similar for finding agreement, etc.



Task





Notes

- For implementation
 - May use any NLTK tools for parse tree manipulation
 - ... as long as it doesn't directly implement the Hobbs algorithm!
 - May create lookup table/dictionary for agreement
- Two results files:
 - One for all parsed output
 - One for remaining manual steps
 - (Based on a copy of the first)







- "Climbing" parse trees:
 - NLTK ParentedTree
 - <u>nltk.org/howto/tree.html</u>
 - Conversion from standard tree t
 - oupport = nltk.tree.ParentedTree.convert(t)
- Accessing feature structures pronoun agr = fs['agr'] antecedent agr.subsumes(pronoun agr)



NLTK Tools

fs = nltk.grammar.FeatStructNonterminal(parented tree.label())







More on Coherence





Coherence Relations

John hid Bill's car keys. He was drunk. **??** John hid Bill's car keys. He likes spinach.

- Why is this odd?
 - No obvious relation between sentences
 - Readers often try to construct relations
- How are the first two related?
 - Explanation/cause
- Utterances should have meaningful connection
 - Establish through coherence relations

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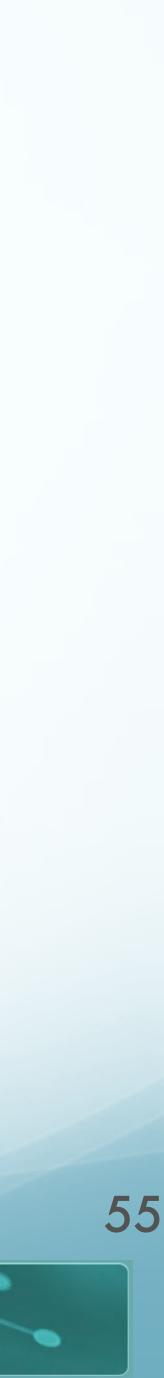
Coherence Relations

- **Result**: Infer that the state or event asserted by S_0 causes, or could cause the state asserted by S_1 .
 - The Tin Woodman was caught in the rain. His joints rusted.
- **Explanation:** Infer that the state or event asserted by S_1 causes or could cause the state or event asserted by S_0 .
 - John hid Bill's car keys. He was drunk.

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- **Parallel:** Infer $p(a_1, a_2, ...)$ from the assertion of S_0 and $p(b_1, b_2, ...)$ from the assertion of S_1 , where a_i and b_i are similar, for all i.
 - The Scarecrow wanted some brains. The Tin Woodman wanted a heart.





Coherence Relations

- - **Dorothy was from Kansas.** She lived in the midst of the great Kansas prairies.
- assertion of S_1 .
 - Dorothy picked up the oil-can. She oiled the Tin Woodman's joints.



• **Elaboration**: Infer the same proposition P from the assertions of S_0 and S_1 .

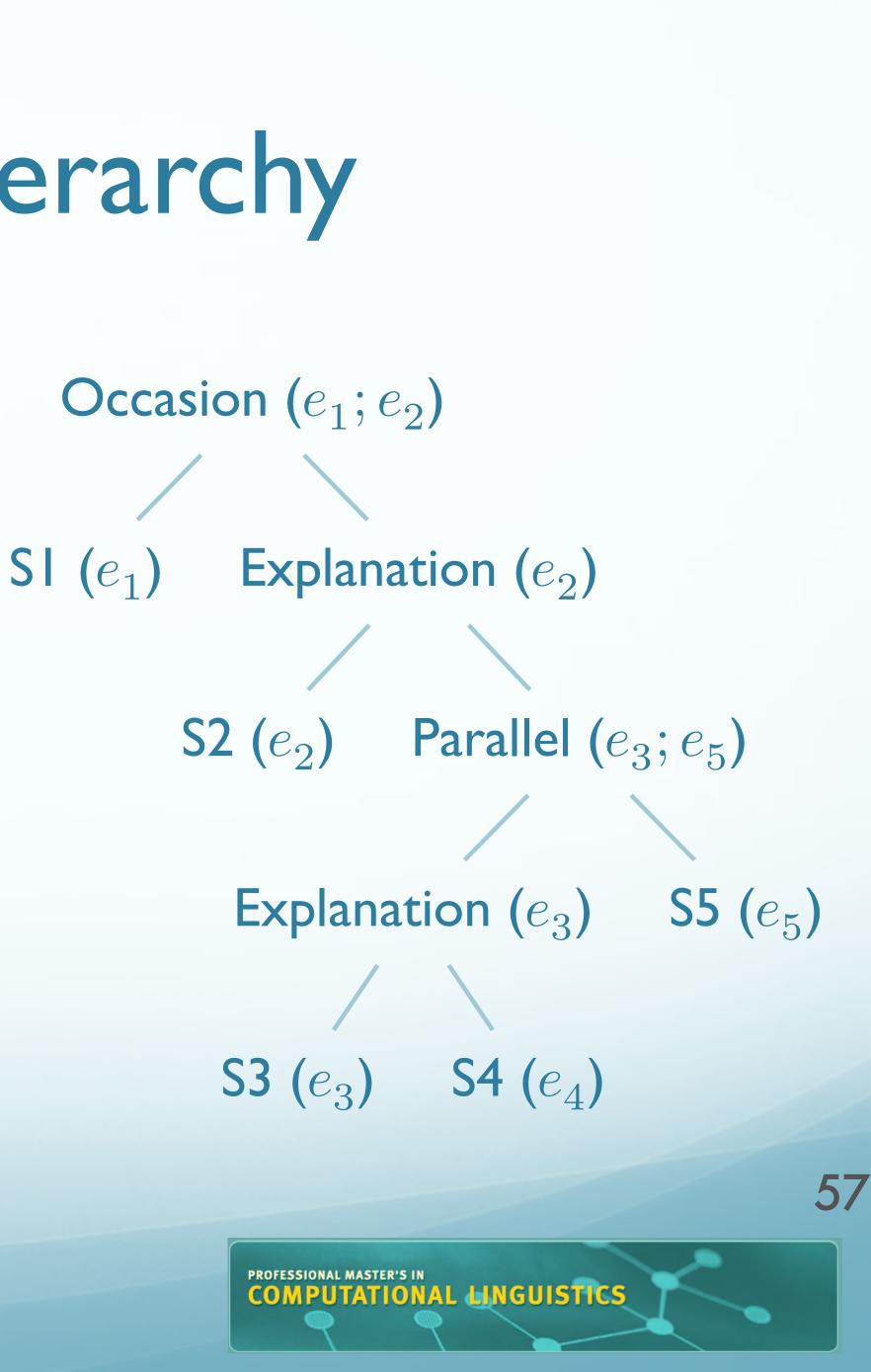
• Occasion: A change of state can be inferred from the assertion of S_0 whose final state can be inferred from S_1 , or a change of state can be inferred from the





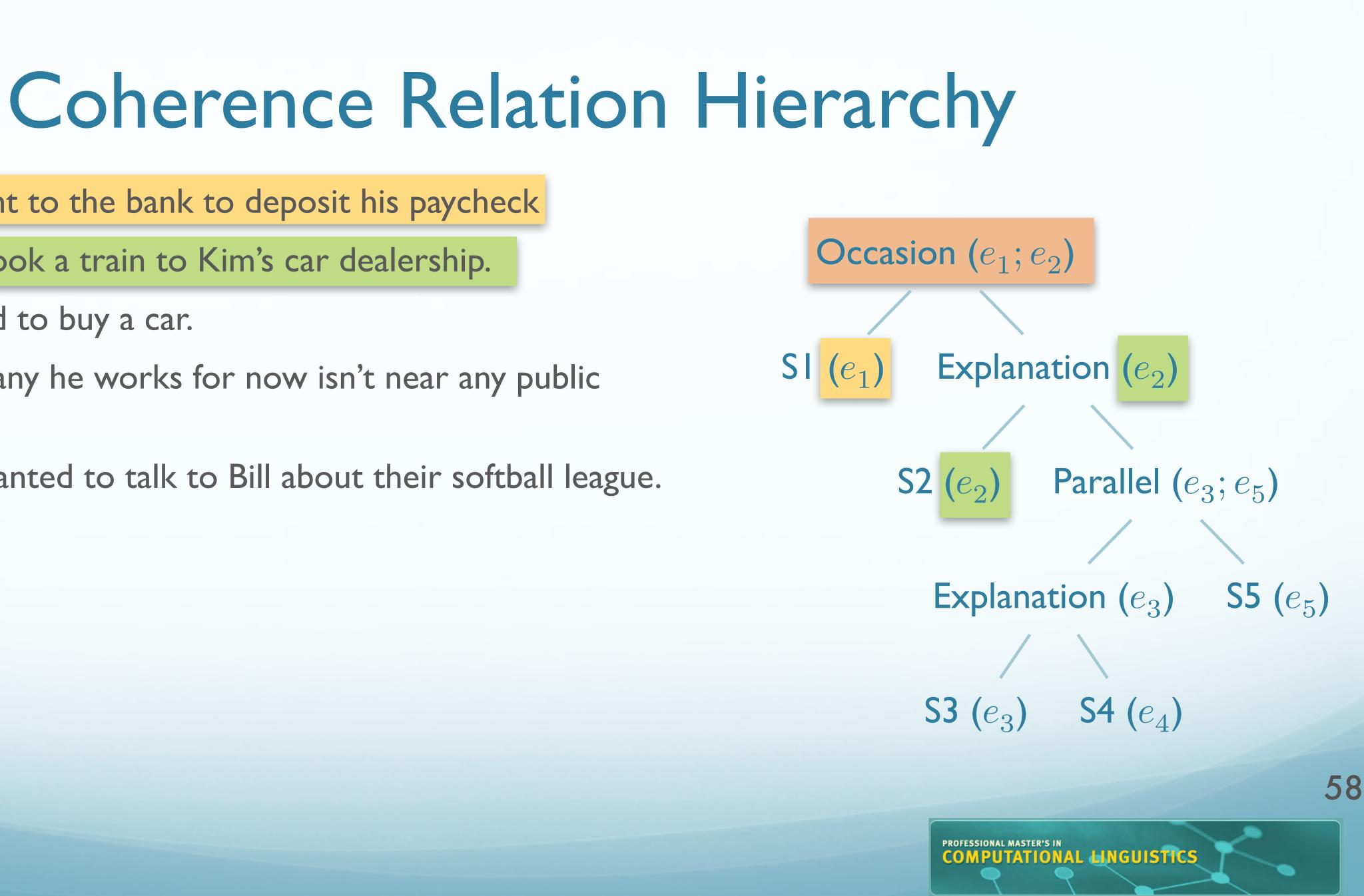
- SI Armin went to the bank to deposit his paycheck
- S2 He then took a train to Kim's car dealership.
- S3 He needed to buy a car.
- S4 The company he works for now isn't near any public transportation.
- S5 He also wanted to talk to Bill about their softball league.





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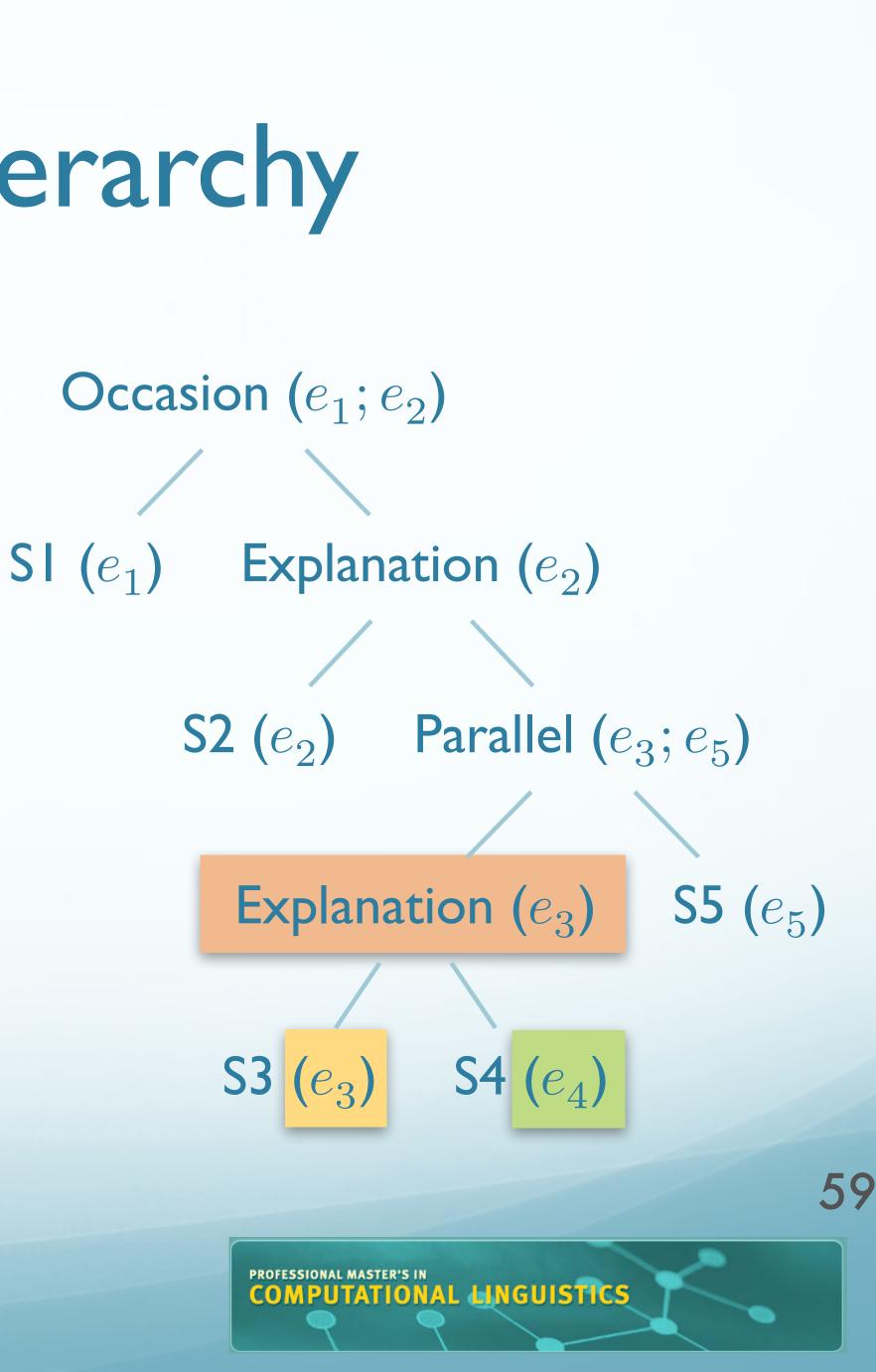


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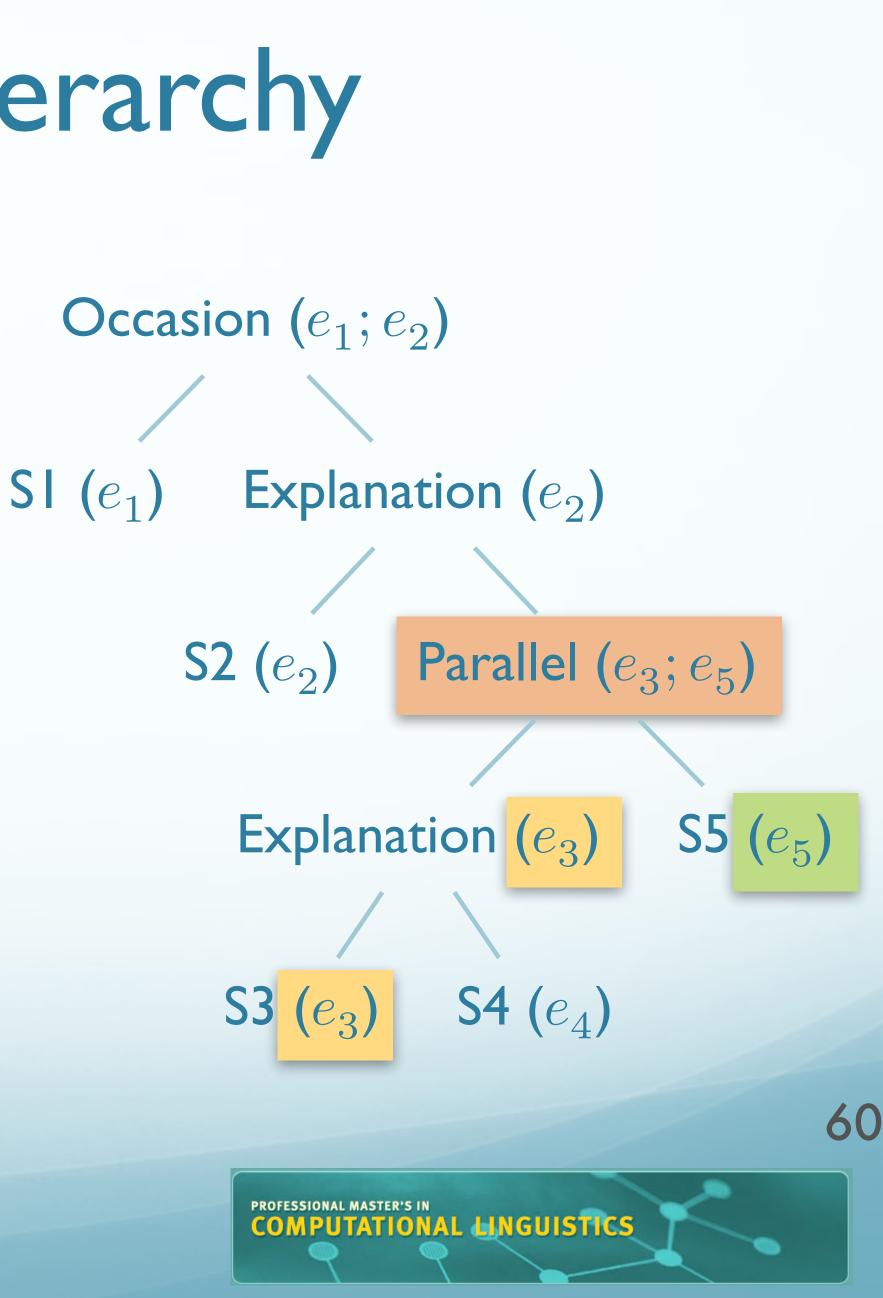




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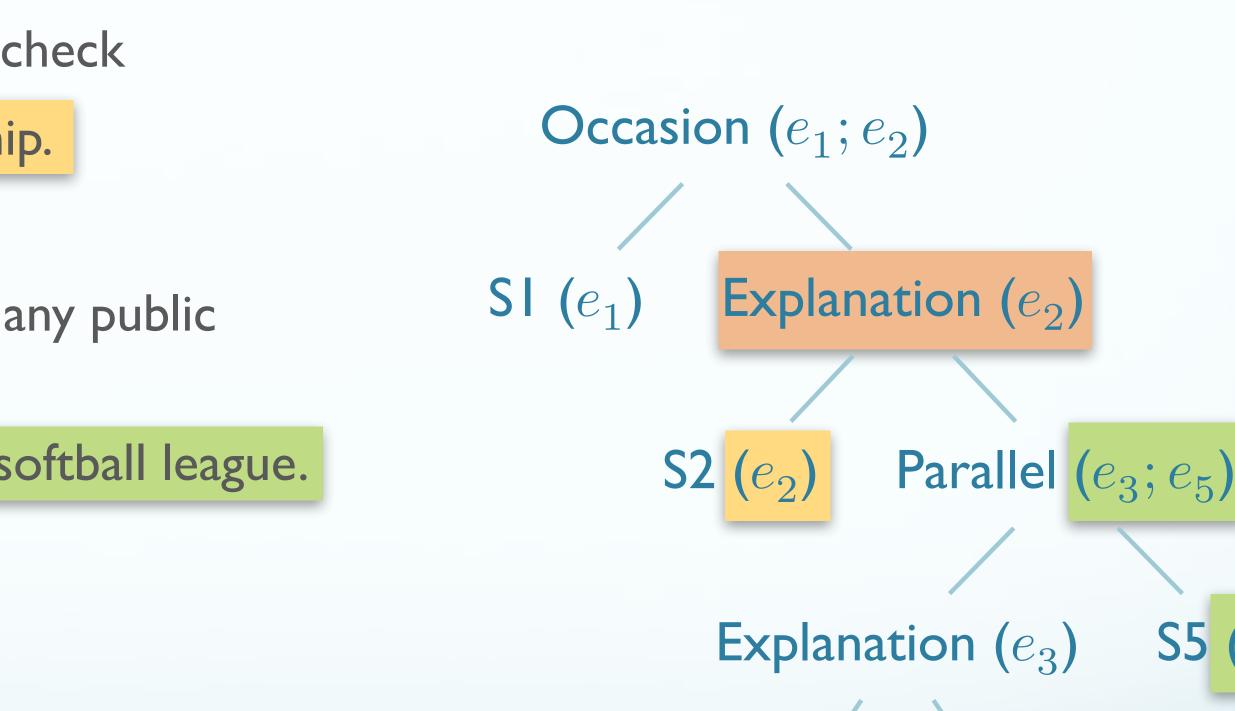




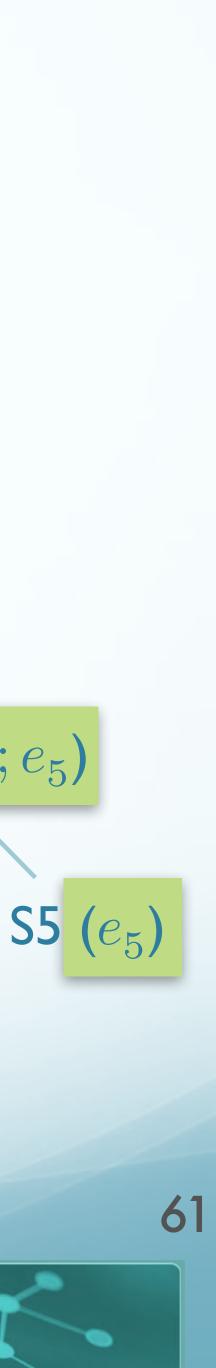
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S4 (e_{A})



Entity-Based Coherence

John went to his favorite music store to buy a piano. He had frequented the store for many years. He was excited that he could finally buy a piano.

• Versus:

John went to his favorite music store to buy a piano. It was a store John had frequented for many years. He was excited that he could finally buy a piano. It was closing just as John arrived.

- Which is better? Why?
 - First focuses on a single entity

• Second interleaves entities John and the music store

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