

Discourse and Coreference

LING 571 — Deep Processing Methods in NLP

November 20, 2019

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Clarification

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


Alternative Resnik WSD Pseudocode

Given: *input word* w_0 and *probe words* $\{p_1, \dots, p_n\}$

```
for  $p_i$  in  $\{p_1, \dots, p_n\}$ :  
    supported_sense = null  
    most_information = 0.0  
    for  $sense_w$  in SENSES( $w_0$ ):  
        for  $sense_p$  in SENSES( $p_i$ ):  
             $lcs_{synset}$  = LOWESTCOMMONSUBSUMER( $sense_w$ ,  $sense_p$ )  
             $lcs_{info}$  = INFORMATIONCONTENT( $lcs_{synset}$ )  
            if  $lcs_{info} > most\_information$ :  
                most_information =  $lcs_{info}$   
                supported_sense =  $sense_w$   
increment support[supported_sense] by most_information
```

Ambiguity of the Week

FrameNet Data	
Giving	
Definition:	
A Donor transfers a Theme from a Donor to a Recipient . This frame includes only actions that are initiated by the Donor (the one that starts out owning the Theme). Theme . Following the transfer the Donor no longer has the Theme and the Recipient does. Barney GAVE the beer to Moe. \$300 was ENDOWED to the university to build a new performing arts building.	
FEs:	
Core:	
Donor [Donor]	The person that begins in possession of the Theme and causes it to be in the possession of the Recipient .
Recipient [Rec]	The entity that ends up in possession of the Theme .
Theme [Thm]	The object that changes ownership.
Semantic Type: Physical_object	




Did you get flowers???

Today 7:19 AM

Lol yes yes I did what about it

From who?!?!?!?

Go on





Ohhh hahaha

Amazing

FrameNet Data	
Getting	
Definition:	
A Recipient starts off without the Theme in their possession, and then comes to possess it. Although the Source from which the Theme came is logically necessary, the Recipient and its changing relationship to the Theme is profiled. I GOT you whistles from John.	
FEs:	
Core:	
Recipient [Rec]	The Recipient indicates the entity that ends up in possession of the Theme .
Theme [Thm]	The Theme is the object that changes possession.
Semantic Type: Physical_object	

Roadmap

- Introduction to Discourse
- Coreference Resolution
 - Phenomena
 - Pronominal Anaphora Resolution
 - Hobbs' Algorithm

Introduction to Discourse

What is 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*

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

Why Discourse?

- Understanding depends on context
 - Word sense — *plant*
 - Intention — *Do you have the time?*
 - Referring expressions — *it, that, the screen*
 - Domain restriction — “All of the students read the announcement.”

Why Discourse?

- Applications: Discourse in NLP
 - Question-Answering
 - Information Retrieval
 - Summarization
 - Dialogue / Conversational AI
 - Automatic Essay Grading

Reference Resolution

- Knowledge sources:
 - *Domain Knowledge*
 - *Discourse Knowledge*
 - *World Knowledge*

User: Where is *A Bug's Life* playing in *Summit*?

System: A Bug's Life is playing at the Summit Theater.

User: When is *it* playing *there*?

System: It's playing at 2PM, 5PM, and 8PM.

User: I'd like 1 *adult* and 2 *children* for *the first show*. How much would *that* cost?

From Carpenter and Chu-Carroll, [*Tutorial on Spoken Dialogue Systems*](#), ACL '99

Not All Sentences Are Created Equal

- *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 tomorrow.^[2]
- Summary:
 - *First Union President John R. Georgius is planning to announce his retirement tomorrow.*
- Inter-sentence coherence relations:
 - Second sentence: main concept (nucleus)
 - First sentence: background

Coherence Relations

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

John hid Bill's car keys. He likes spir🤔h.

- Why is this odd?
 - No obvious relation between sentences
 - Breaks our assumption as readers that information presented in discourse is relevant
- How is the first pair related?
 - statment — explanation/cause
- Assumption: utterances should have meaningful connection
 - Establish through *coherence relations*

Coherence Relations

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

John hid Bill's car keys. He likes s🤔nach.

- **Assumption**

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

Discourse: Looking Ahead

Coreference

Cohesion

Coherence

Structure / Segmentation

Coreference Resolution

Reference: Terminology

- **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.

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.

Reference: Terminology

- ***Antecedent:***
 - An expression that introduces an item to the discourse for other items to refer back to
 - Queen Elizabeth... her

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.

Reference: Terminology

- **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.*”

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 Expressions

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

Referring Expressions

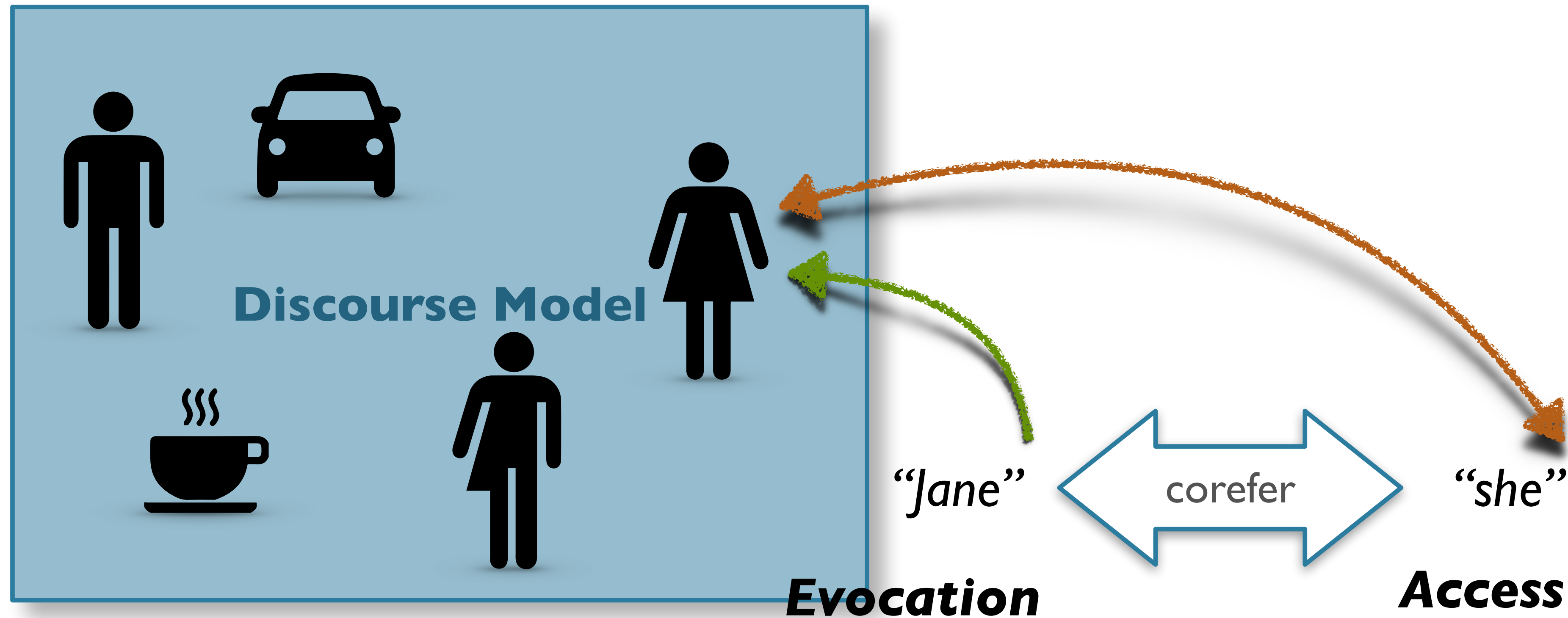
- *Queen Elizabeth* – *she/her* – *the Queen* – *HRM* – *the British Monarch*
- “Correct” form depends on discourse context
 - *she, her* presume prior mention or presence in the world
 - *the Queen* presumes an Anglocentric geopolitical discourse context generally or the UK (or British Commonwealth) specifically

(...i.e. likely a different interpretation during a RPDR viewing party.)

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
 - [“introduces a discourse referent (dref)”]
 - Subsequent mentions **access** entity *from* the model.

Reference and Model



Reference Tasks

- **Coreference resolution:**
 - Find all expressions referring to the same entity in a text.
 - A set of coreferring expressions is a *coreference chain*.
- **Pronominal anaphora resolution:**
 - Find antecedent for a single pronoun.
 - Subtask of coreference resolution

Pronominal Anaphora Resolution

Reference Phenomena

Expression Type	Examples	Constraints
Indefinite NP	“ <i>a cat</i> ”, “ <i>some geese</i> ”	Introduces new entity to context
Definite NP	“ <i>the dog</i> ”	Refers to entity identifiable by hearer in context
Pronouns	“ <i>he</i> ,” “ <i>them</i> ,” “ <i>they</i> ”	Refers to entity, must be “ <i>salient</i> ”
Demonstratives	“ <i>this</i> ,” “ <i>that</i> ”	Refers to entity, sense of distance (literal/figurative)
Names	“ <i>Dr. Woodhouse</i> ,” “ <i>IBM</i> ”	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.

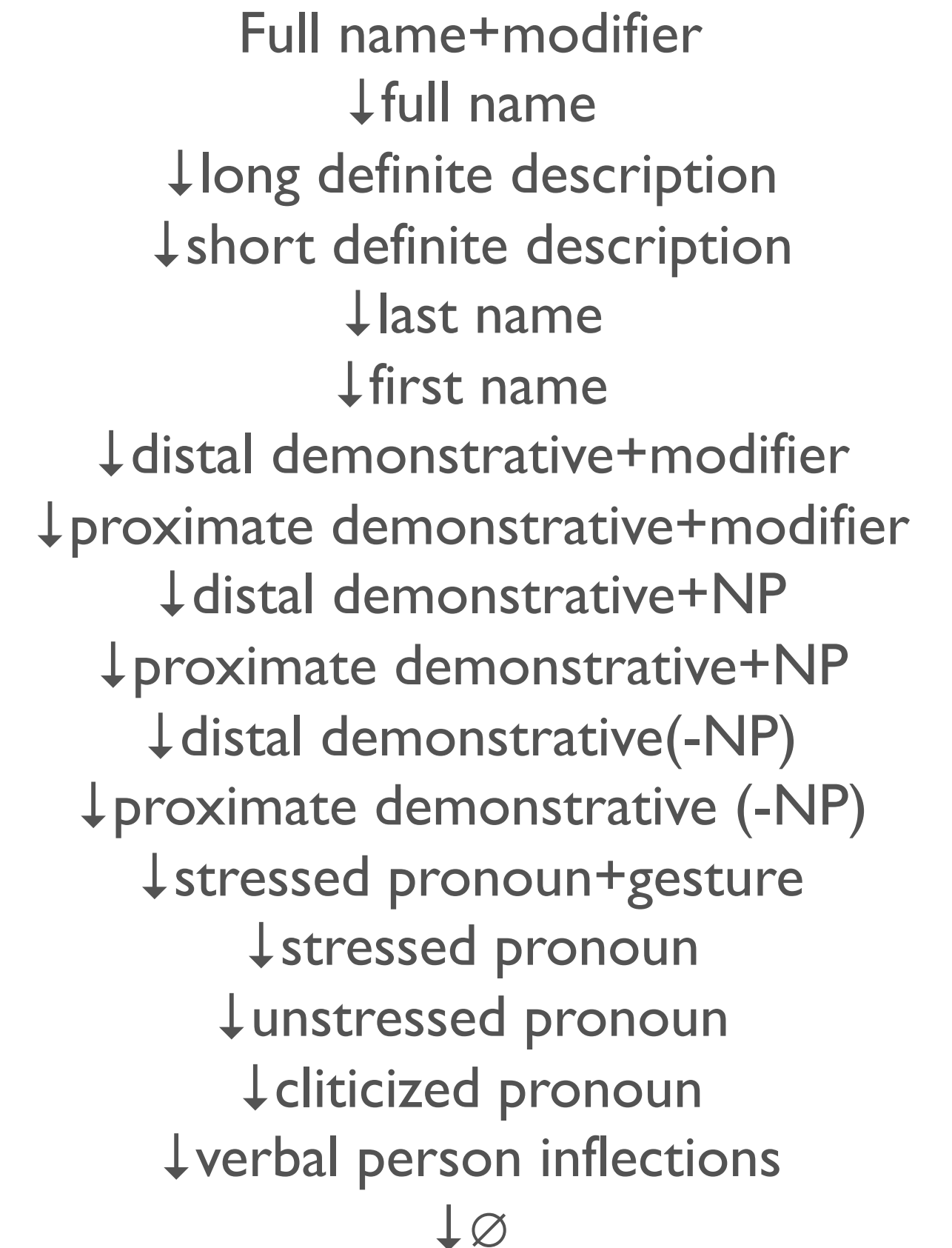
Information Status

- 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 >	uniquely identifiable >	referential >	type identifiable
<i>it</i>	<i>this</i>	<i>that N</i>	<i>the N</i>	<i>indef. this N</i>	<i>a N</i>
	<i>that</i>				
	<i>this N</i>				

Information Status

- ***Accessibility scale:*** ([Ariel, 2001](#))
 - More salient elements easier to call up, can be shorter
 - correlates with length: more accessible, shorter refexp



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)

Features for Anaphora Resolution: Constraints

- **Number:**

- *Anjali has a Corvette.* **They are red.* *It is red.*

- **Person:**

- 1st: *I, we* 2nd: *you, y'all* 3rd: *he, she, it, they*

- **Gender:**

- *Janae plays the guitar.* *She sounds great.*
- *Janae plays the guitar.* *It sounds great.*

Features for Anaphora Resolution: Constraints

- **Binding Theory**

- How to handle reflexive pronouns vs. nonreflexives

- *Aaron bought **themselves** a new car.*

- *Aaron bought **them** a new car.*

[them ≠ Aaron]

- *Jen said that Imani bought **herself** a new car.*

[herself = Imani]

- *Jen said that Imani bought **her** a new car.*

[her ≠ Imani]

- ***He**₁ said that **he**₂ bought **Willie** a new car.*

[He₁ ≠ Willie, he₂ ≠ Willie]

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

- Reflexives do corefer with subject of containing clause

Features for Anaphora Resolution: Preferences

- **Recency:**

- Prefer closer antecedents.
- *The doctor found **an old map** in the captain's chest. Jim found **an even older map** on the 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**. **He** called for a glass of rum.*
 - ***Jim Hawkins** went to the bar with **Billy Bones**. **He** called for a glass of rum.*

Features for Anaphora Resolution: Preferences

- **Repeated Mention:**

- Once entity is focused, likely to continue to be focused → more likely pronomialized.
- ***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.*

Features for Anaphora Resolution: Preferences

- Verb Semantics

- Some verbs semantically bias for one of their argument positions.

John telephoned Bill. He had lost the laptop.

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' Algorithm

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

Hobbs Algorithm

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

Hobbs Algorithm Detail (Hobbs, 1978)

1. 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.

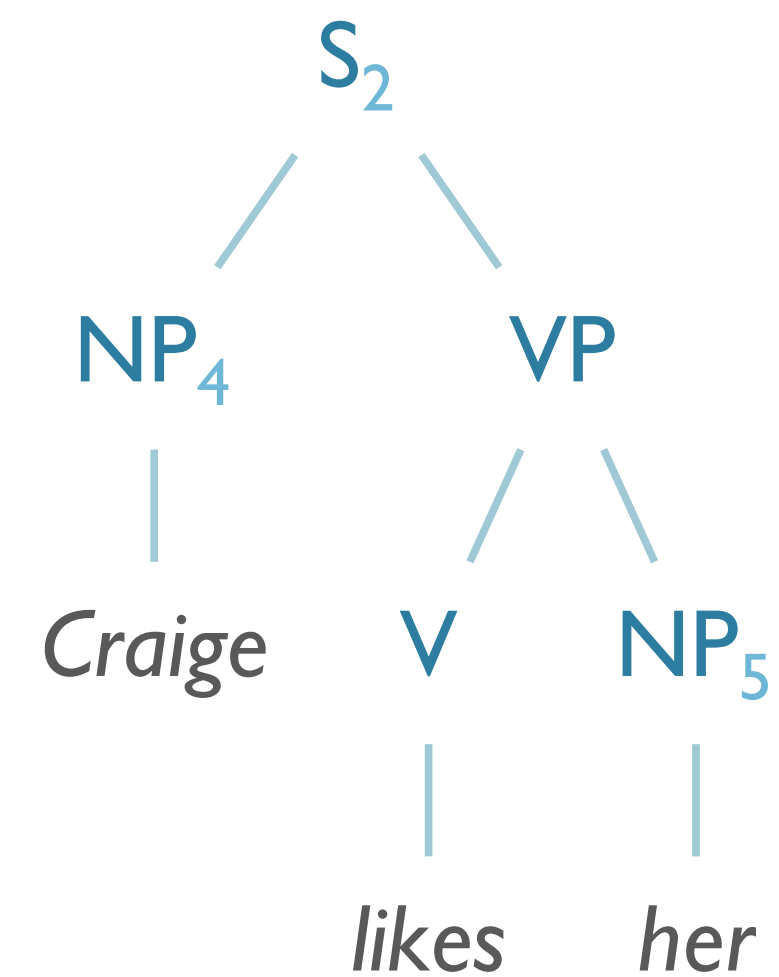
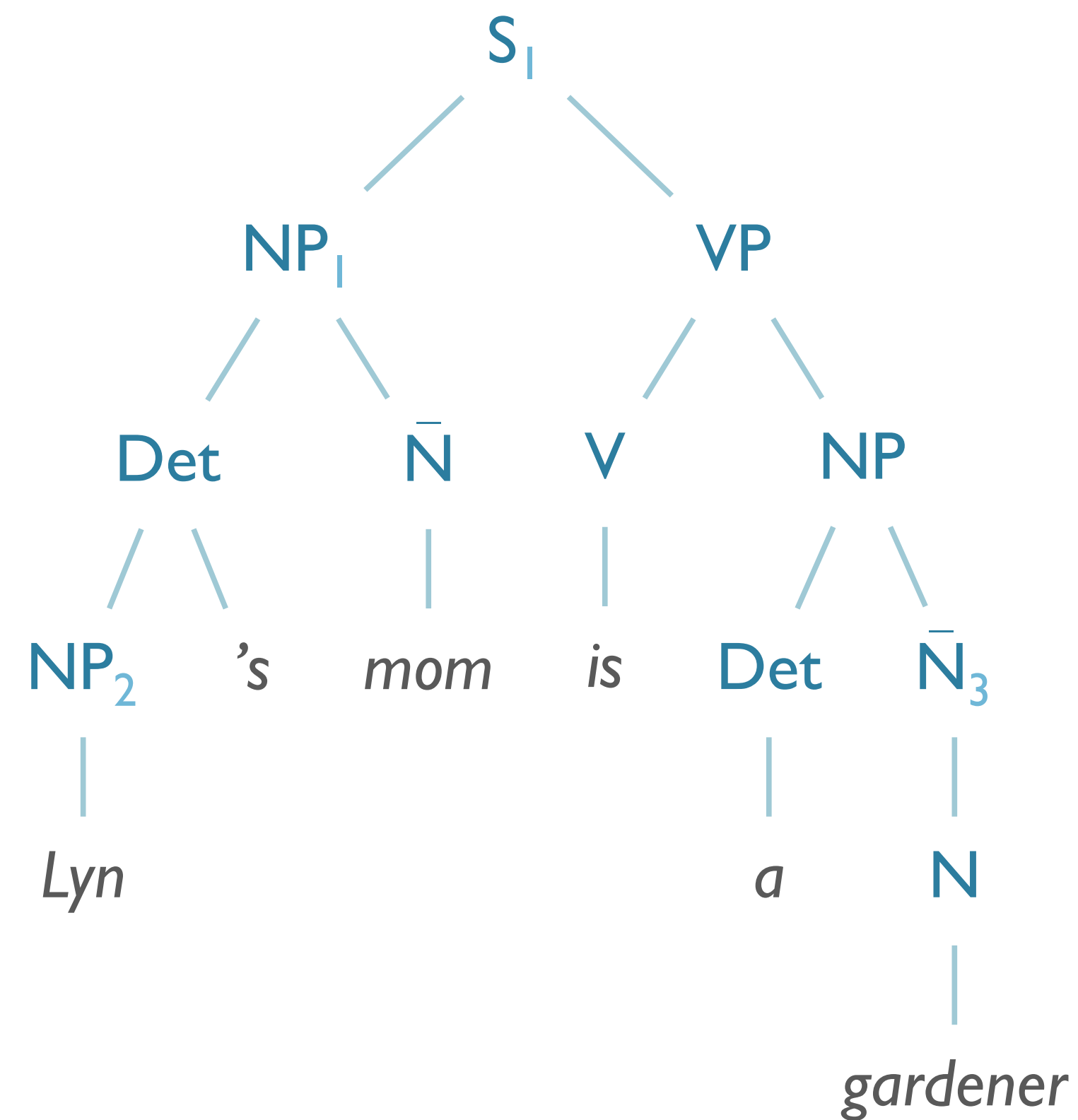
Hobbs Algorithm Detail (Hobbs, 1978)

5. From node **X**, go up the tree to the first NP or S node encountered. Call this new node **X**, and call the path traversed to reach it p .
6. If **X** is an NP node and if the path p to **X** did not pass through the Nominal node that **X** immediately dominates, propose **X** as the antecedent.
7. Traverse all branches below node **X** to the *left* of path p in a left-to-right, breadth-first manner. Propose any NP node encountered as the antecedent.
8. If **X** is an S node, traverse all branches of node **X** to the *right* of path p in a left-to-right, breadth-first manner, but do not go below any NP or S node encountered. Propose any NP node encountered as the antecedent.
9. Go to step 4.

Hobbs Example

Lyn's mom is a gardener.

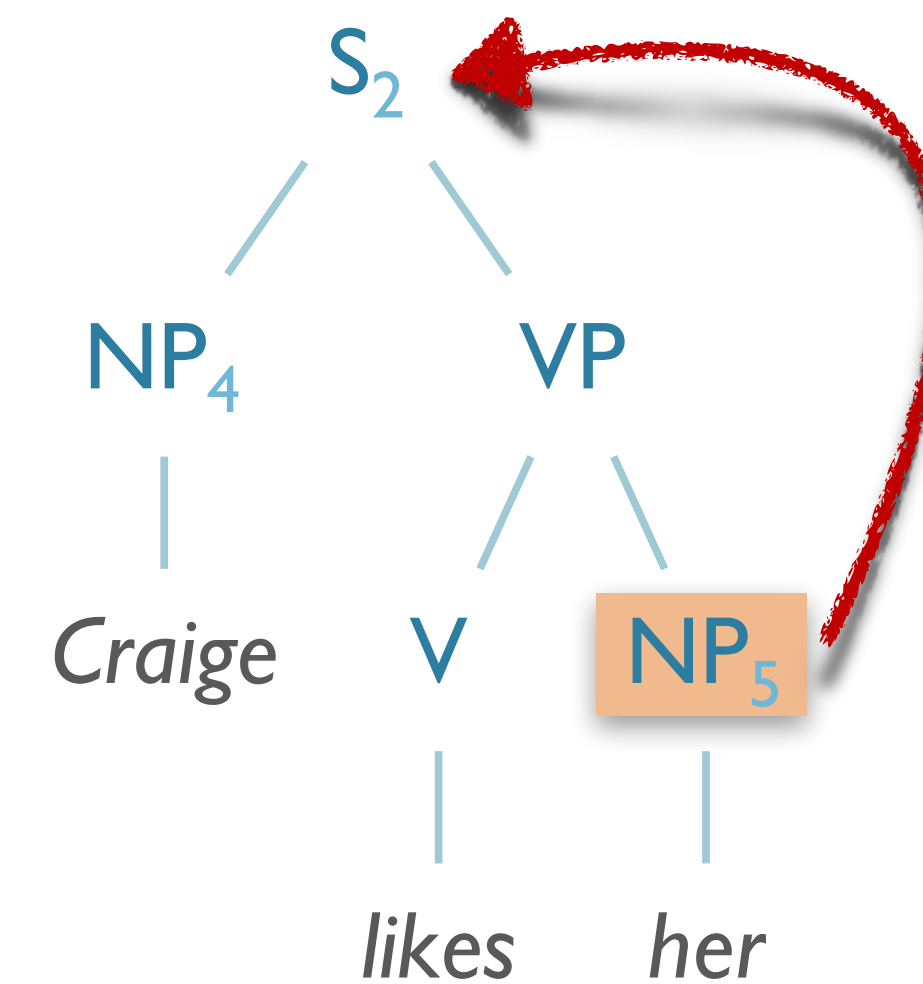
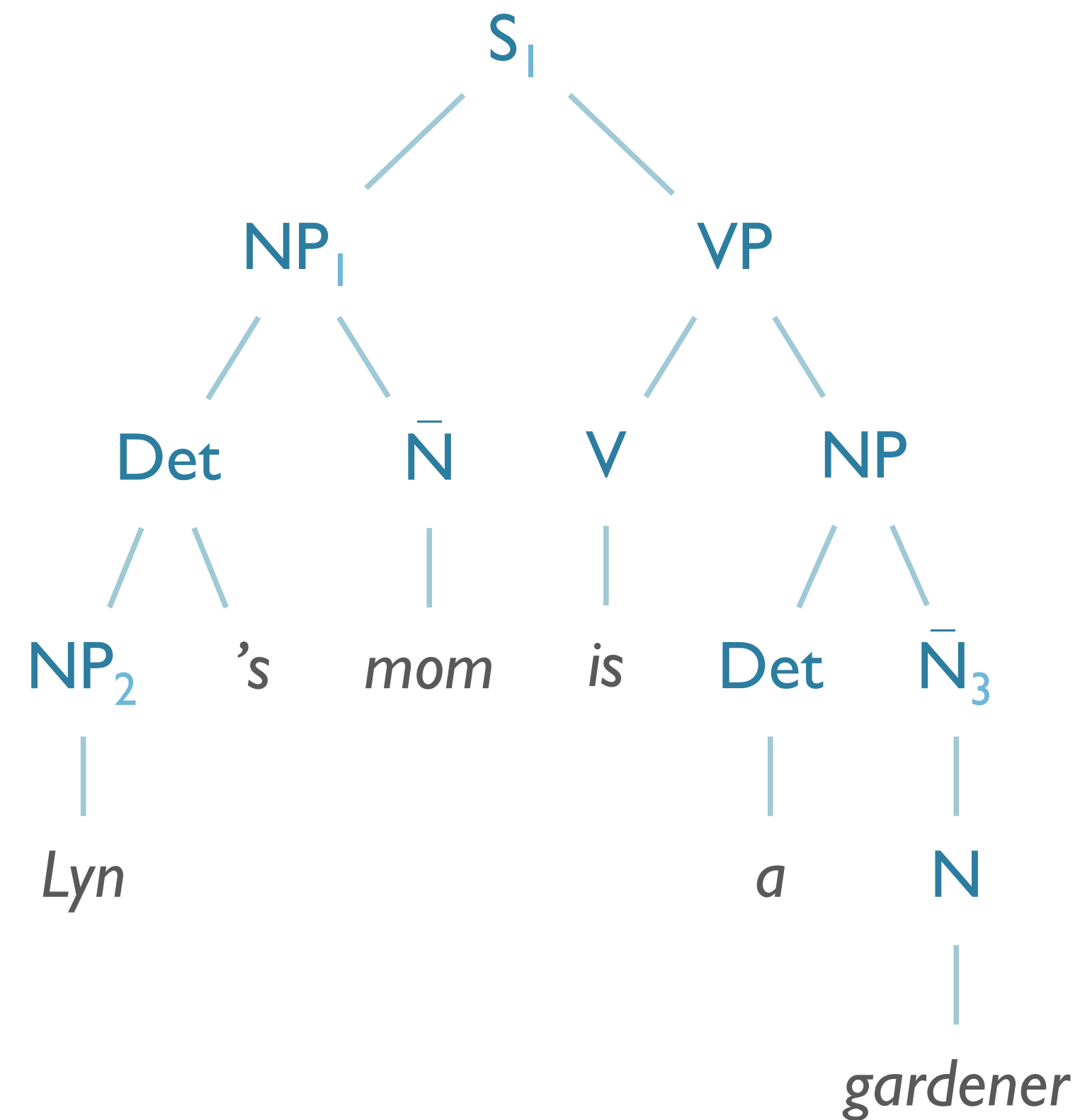
Craige likes her.



Hobbs Example

Lyn's mom is a gardener.

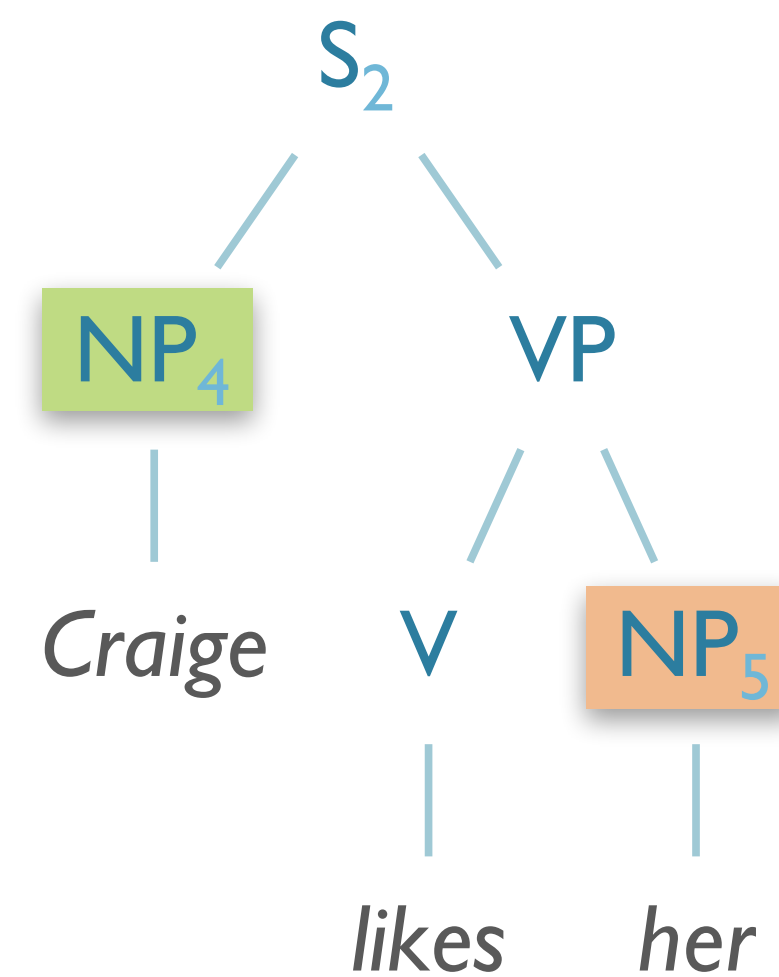
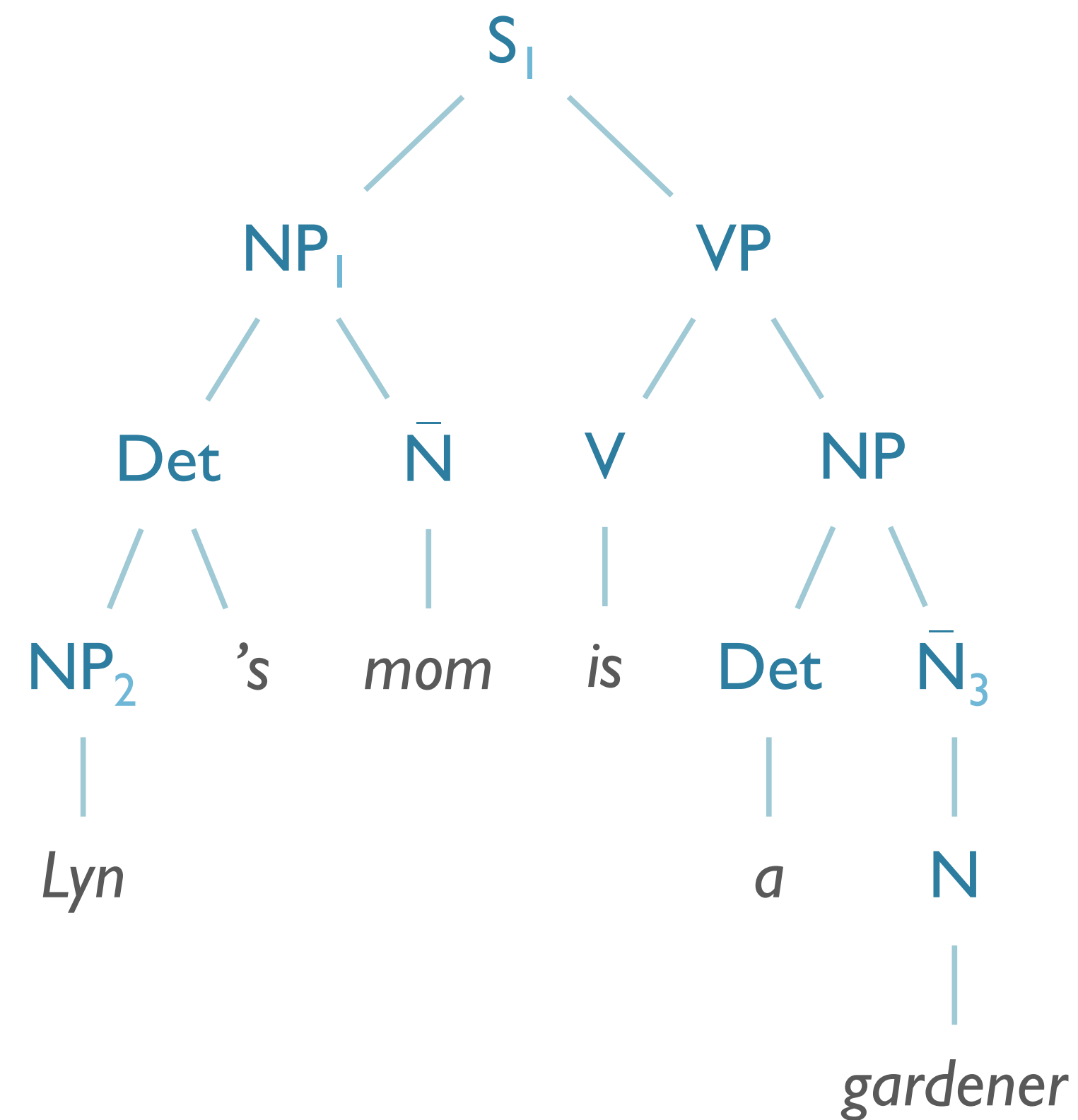
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Hobbs Example

Lyn's mom is a gardener.

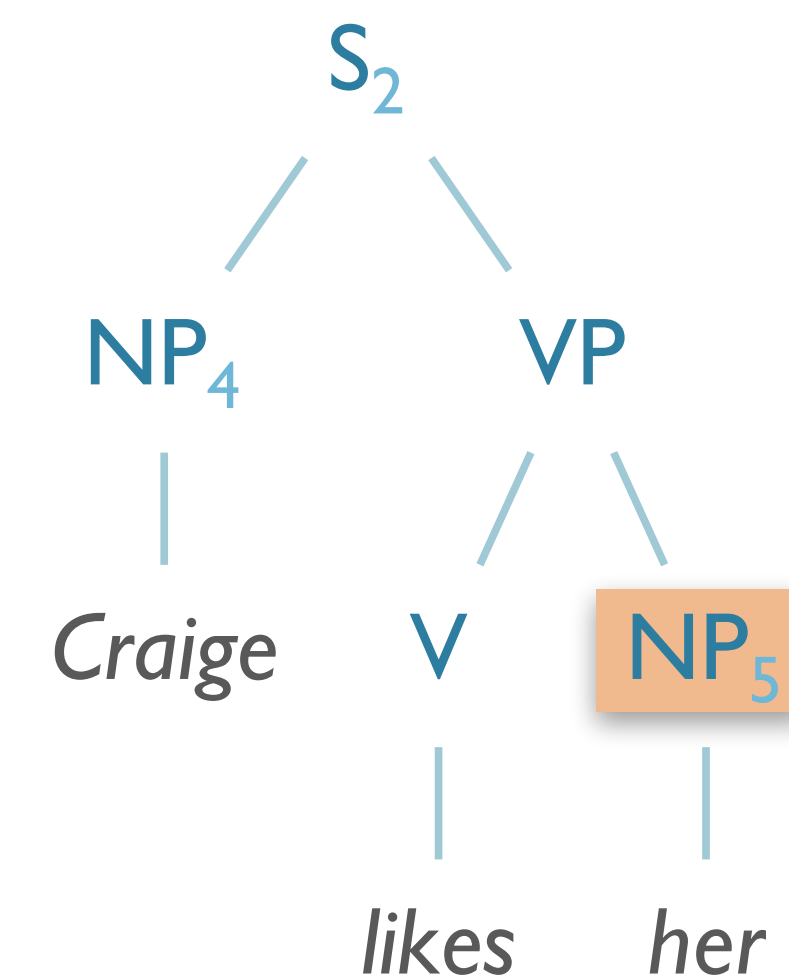
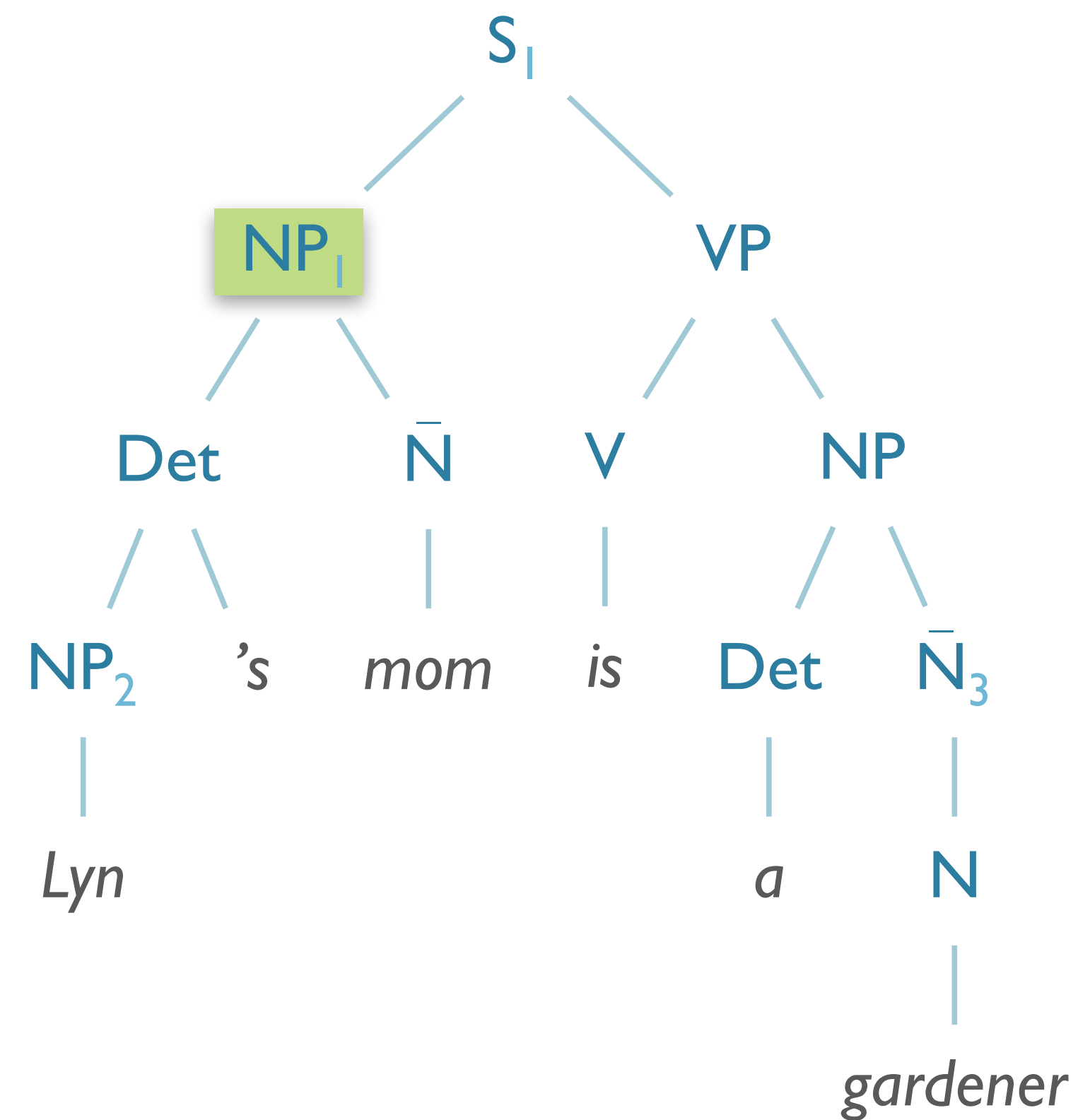
Craige likes her.



Hobbs Example

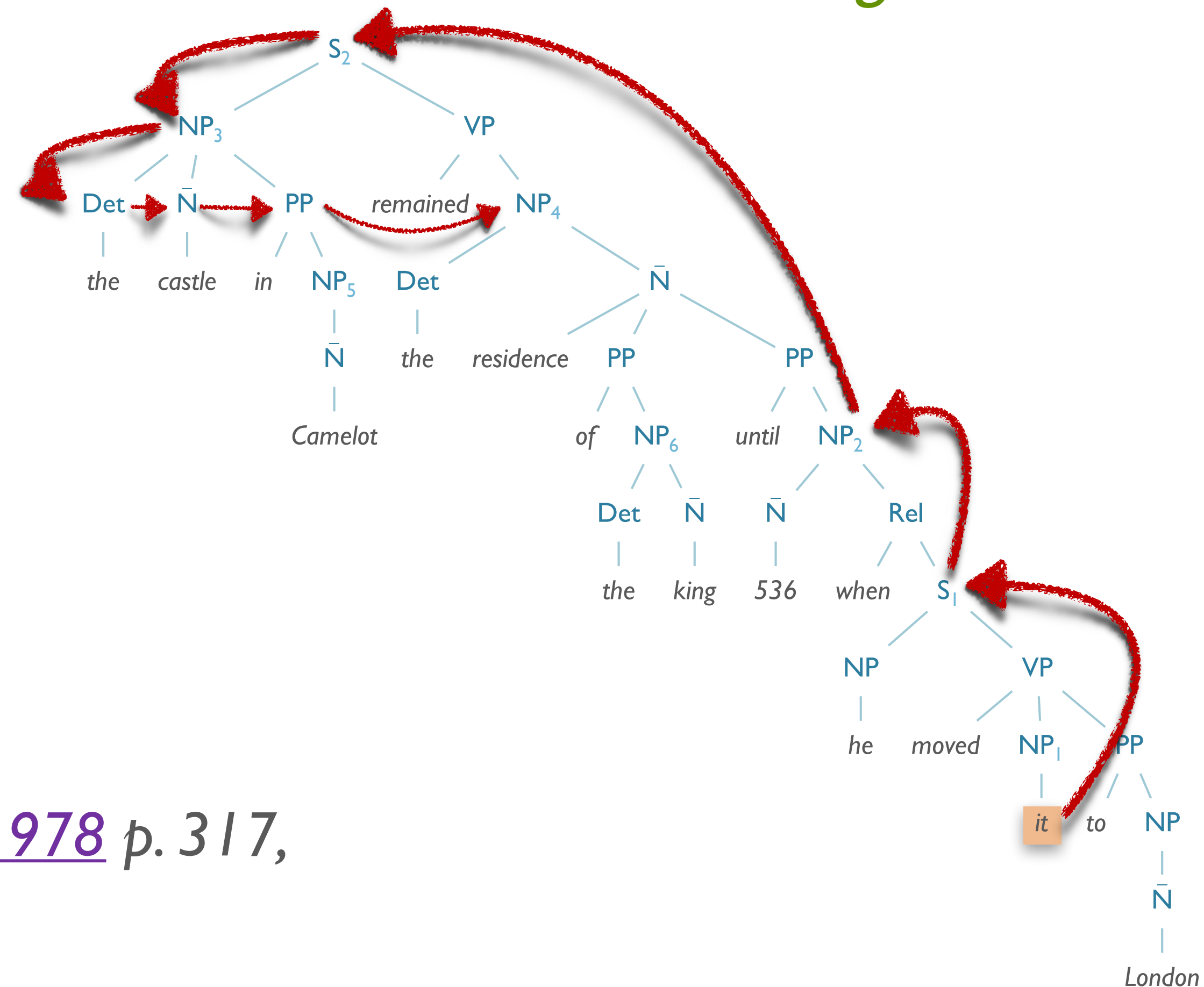
Lyn's mom is a gardener.

Craige likes her.



Another Hobbs Example

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



for full walkthrough see [Hobbs, 1978](#) p. 317,
and the end of today's slides

Hobbs Algorithm

- Results: 88% Accuracy; 90% intrasentential
 - ...on perfect, manually parsed sentences
- Useful ***baseline*** for evaluating pronominal anaphora
- Issues:
 - **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

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)*

*more on this later
- Get this reading by knowledge of city councils and permitting, and reasons why permits would be refused.

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 naïve 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.*

*“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 antecedent.” — Hobbs (1978), *Lingua*, p. 345*

Coreference and World Knowledge

- The trophy doesn't fit into the brown suitcase because it's too [small/large]. What is too [small/large]?
 - Answers: The suitcase/the trophy.
- Joan made sure to thank Susan for all the help she had [given/received]. Who had [given/received] help?
 - Answers: Susan/Joan.
- Paul tried to call George on the phone, but he wasn't [successful/available]. Who was not [successful/available]?
 - Answers: Paul/George.
- The lawyer asked the witness a question, but he was reluctant to [answer/repeat] it . Who was reluctant to [answer/repeat] the question?
 - Answers: The witness/the lawyer.

Winograd Schema Challenge

- Still hard!
- WSC

Heavily supervised

SuperGLUE GLUE Paper </> Code Tasks Leaderboard FAQ Diagnostics Submit

Leaderboard Version: 2.0

Rank	Name	Model	URL	Score	BoolQ	CB	COPA	MultiRC	ReCoRD	RTE	WiC	WSC	AX-g	AX-b
1	SuperGLUE Human Baselines	SuperGLUE Human Baselines		89.8	89.0	95.8/98.9	100.0	81.8/51.9	91.7/91.3	93.6	80.0	100.0	99.3/99.7	76.6
2	T5 Team - Google	T5		88.9	91.0	93.0/96.4	94.8	88.2/62.3	93.3/92.5	92.5	76.1	93.8	92.7/91.9	65.6
3	Facebook AI	RoBERTa		84.6	87.1	90.5/95.2	90.6	84.4/52.5	90.6/90.0	88.2	69.9	89.0	91.0/78.1	57.9
4	IBM Research AI	BERT-mlt		73.5	84.8	89.6/94.0	73.8	73.2/30.5	74.6/74.0	84.1	66.2	61.0	97.8/57.3	29.6
5	SuperGLUE Baselines	BERT++		71.5	79.0	84.8/90.4	73.8	70.0/24.1	72.0/71.3	79.0	69.6	64.4	99.4/51.4	38.0
		BERT		69.0	77.4	75.7/83.6	70.6	70.0/24.1	72.0/71.3	71.7	69.6	64.4	97.8/51.7	23.0
		Most Frequent Class		47.1	62.3	21.7/48.4	50.0	61.1/0.3	33.4/32.5	50.3	50.0	65.1	100.0/50.0	0.0
		CBoW		44.5	62.2	49.0/71.2	51.6	0.0/0.5	14.0/13.6	49.7	53.1	65.1	100.0/50.0	-0.4
		Outside Best		-	80.4	-	84.4	70.4/24.5	74.8/73.0	82.7	-	-	-	-
-	Stanford Hazy Research	Snorkel [SuperGLUE v1.9]		-	-	88.6/93.2	76.2	76.4/36.3	-	78.9	72.1	72.6	-	47.6

HW #9

Goals

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

Task

- 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.

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)

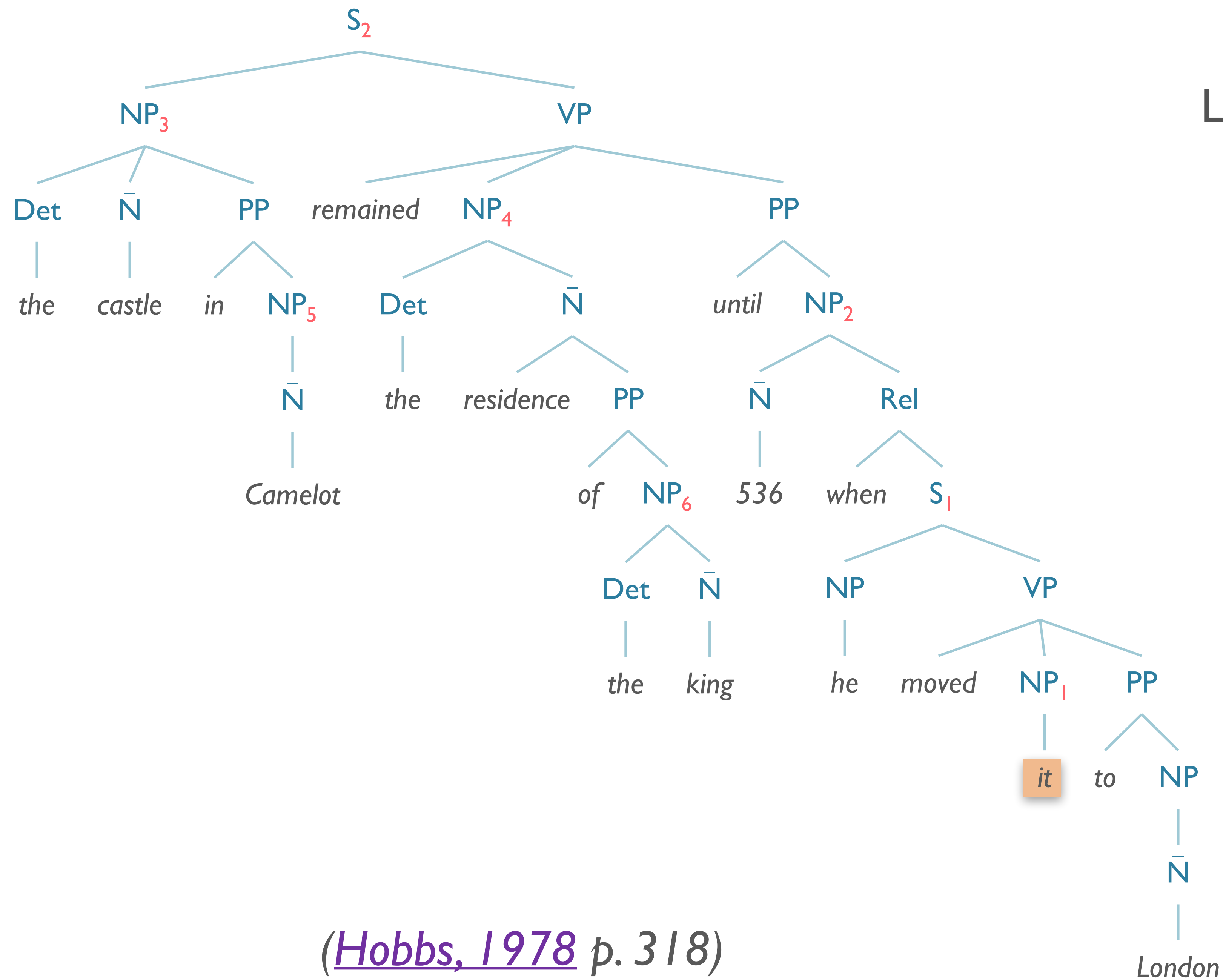
NLTK Tools

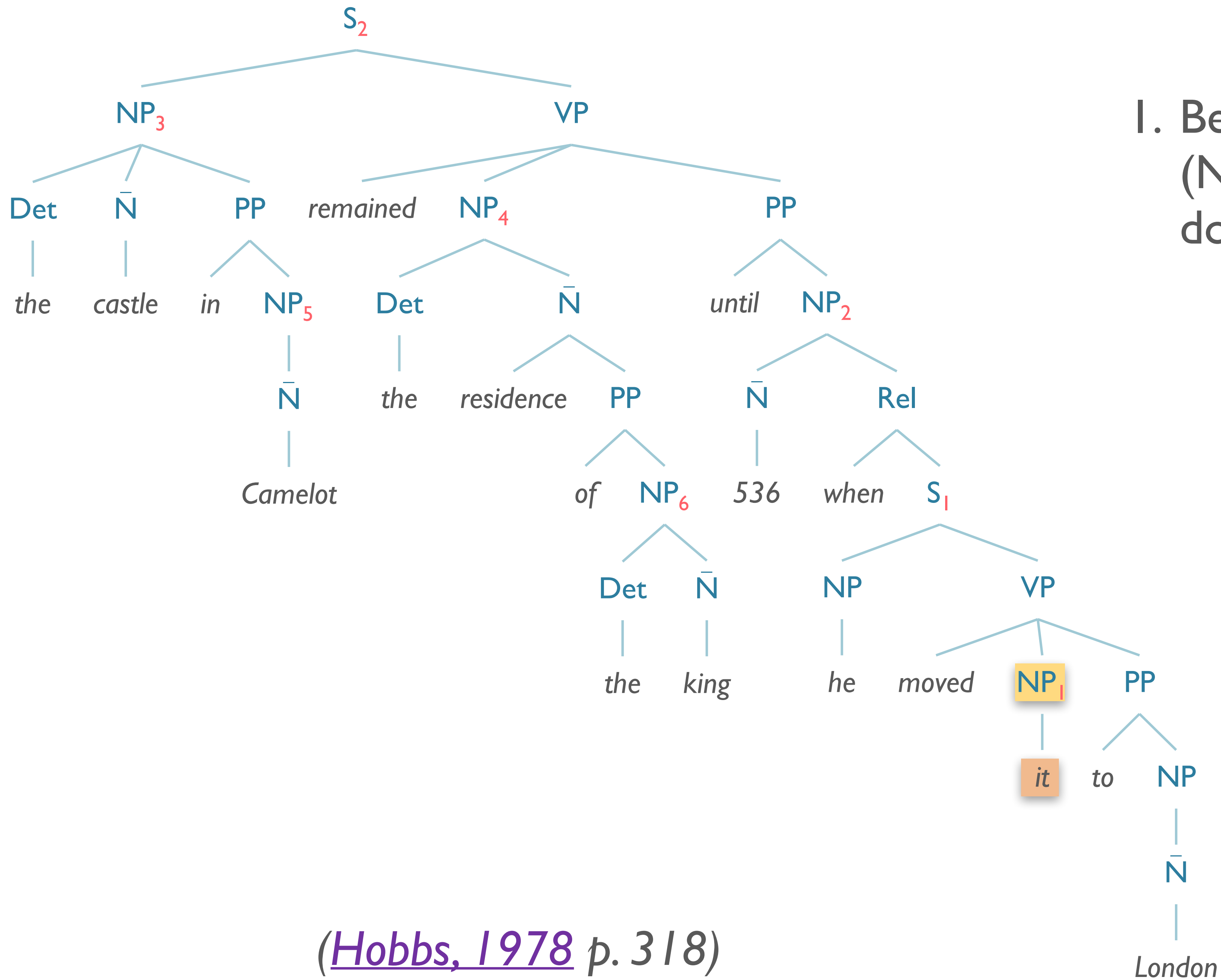
- “Climbing” parse trees:
 - NLTK ParentedTree
 - nltk.org/howto/tree.html
 - Conversion from standard tree **t**
 - `parented_tree = nltk.tree.ParentedTree.convert(t)`
- Accessing feature structures

```
fs = nltk.grammar.FeatStructNonterminal(parented_tree.label())
pronoun_agr = fs['agr']
antecedent_agr.subsumes(pronoun_agr)
```


Hobbs Algorithm Walkthrough

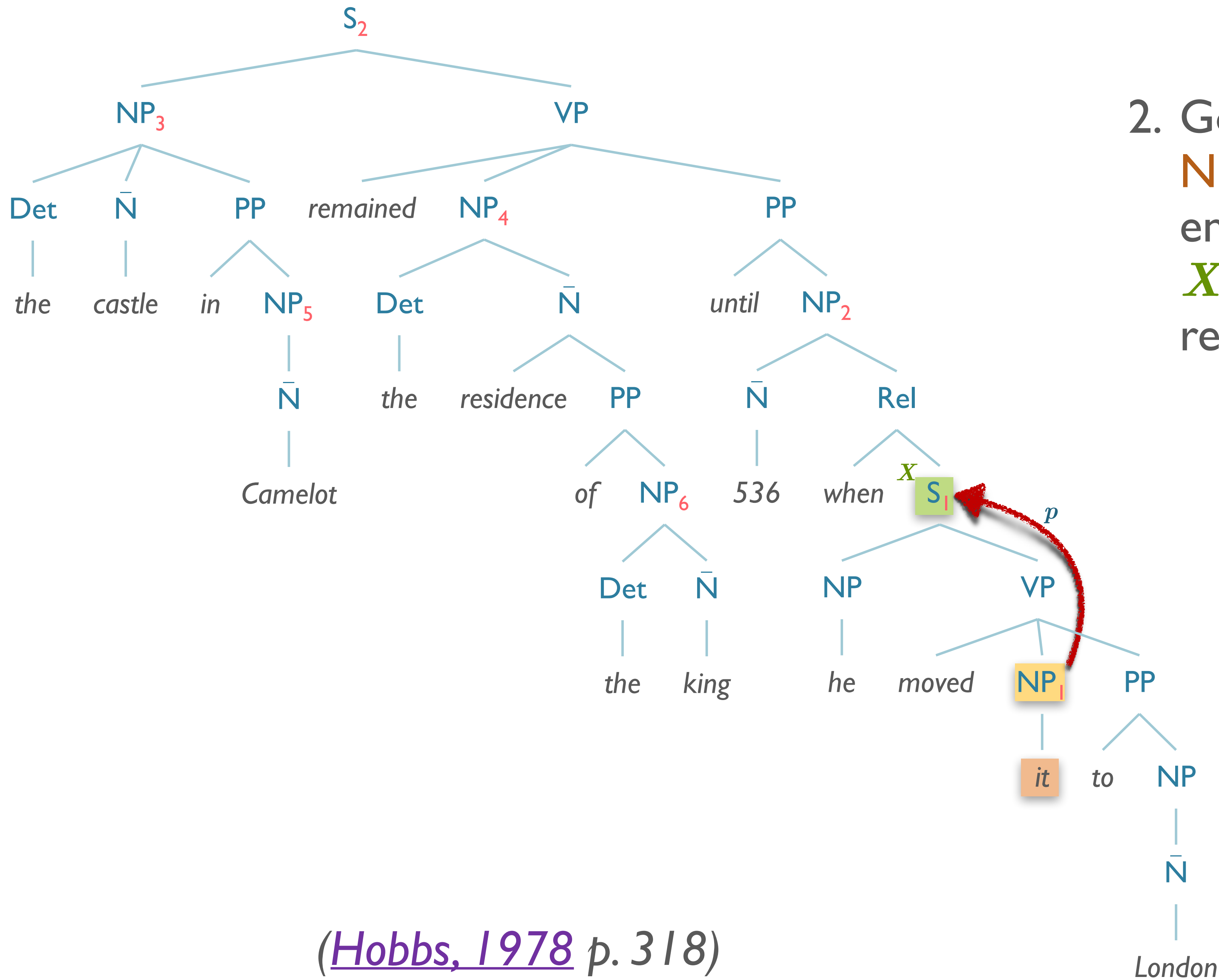
(h/t Ryan Georgi)





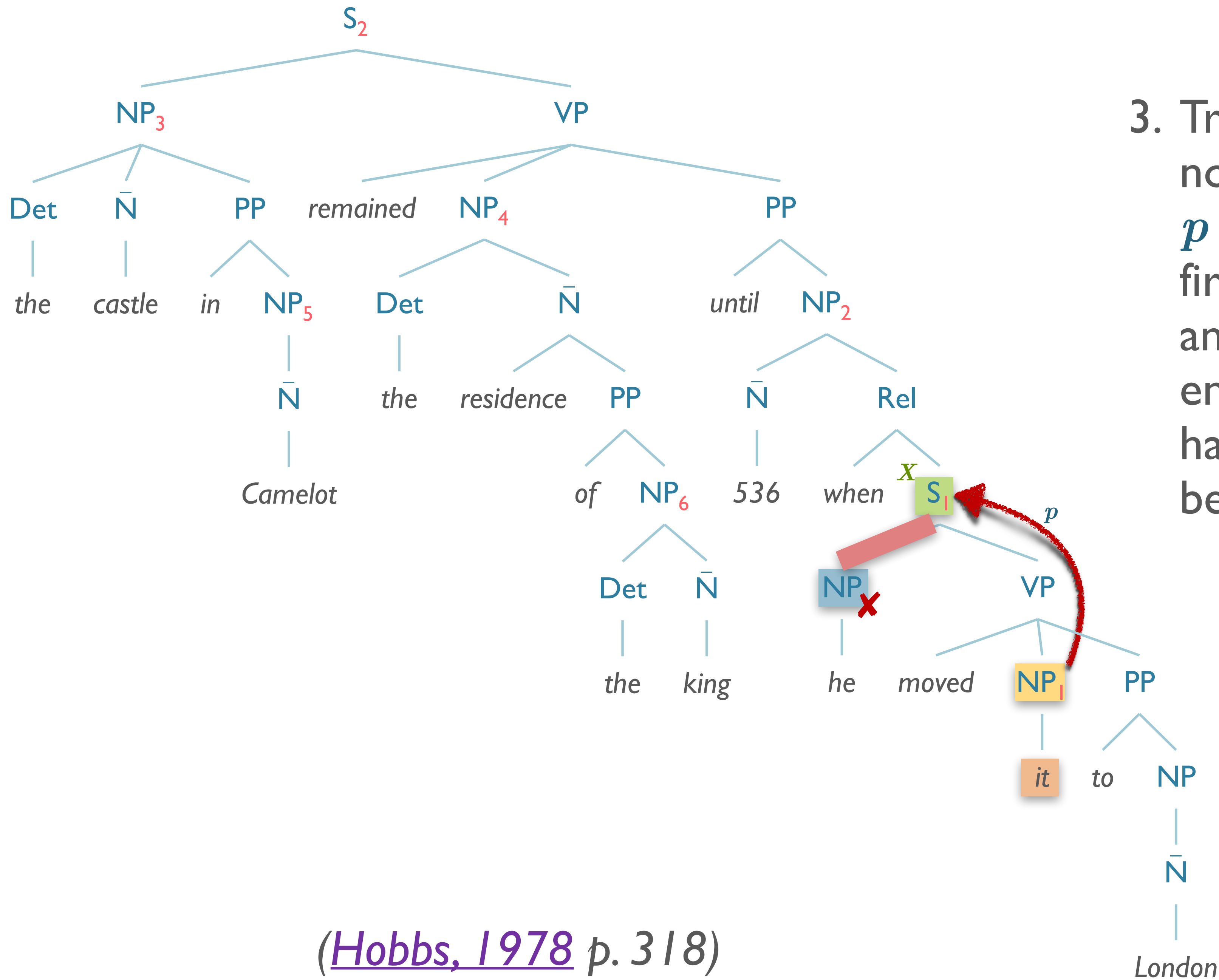
I. Begin at the noun phrase (NP) node immediately dominating the pronoun

(Hobbs, 1978 p. 318)



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*.

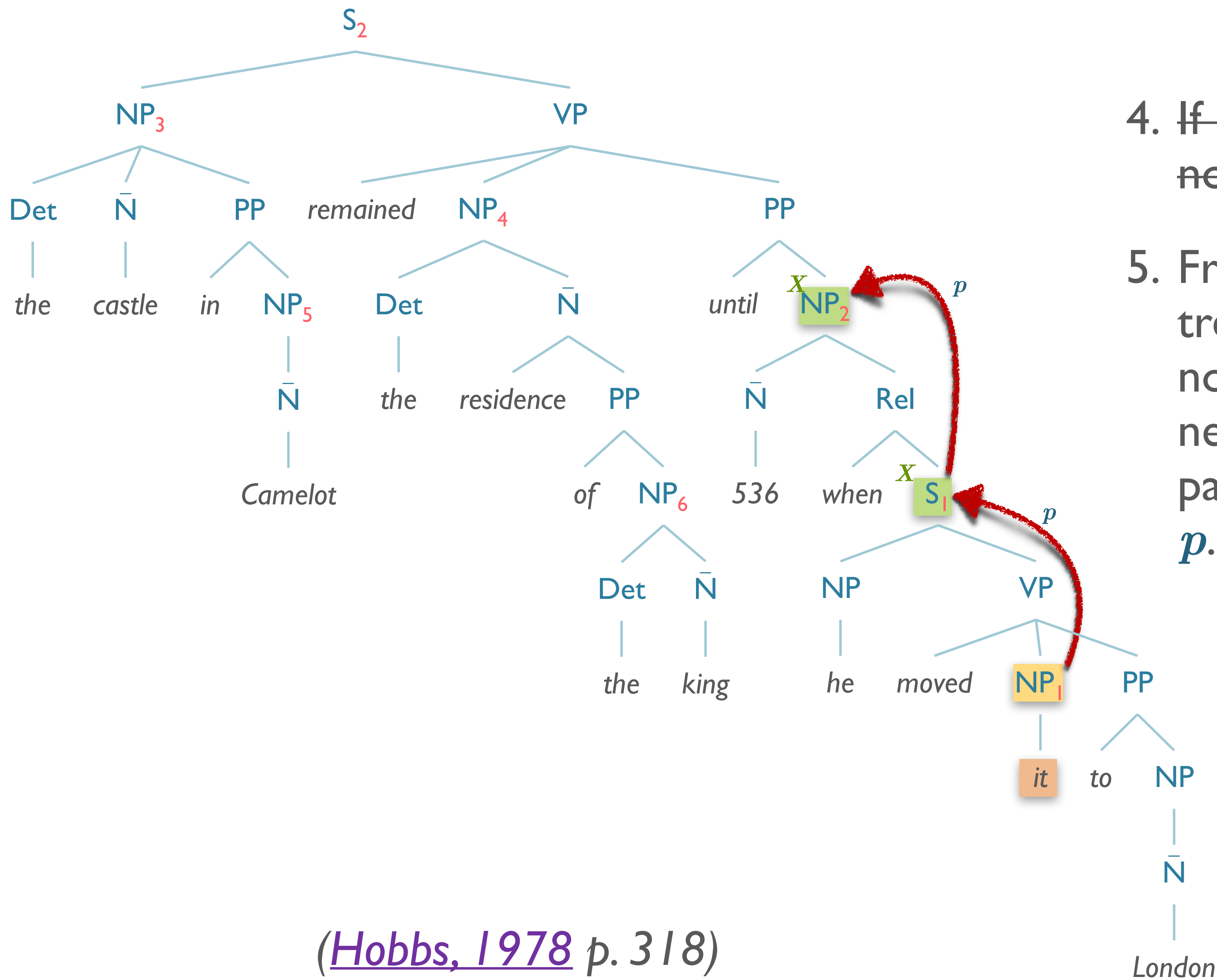
(Hobbs, 1978 p. 318)



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 .

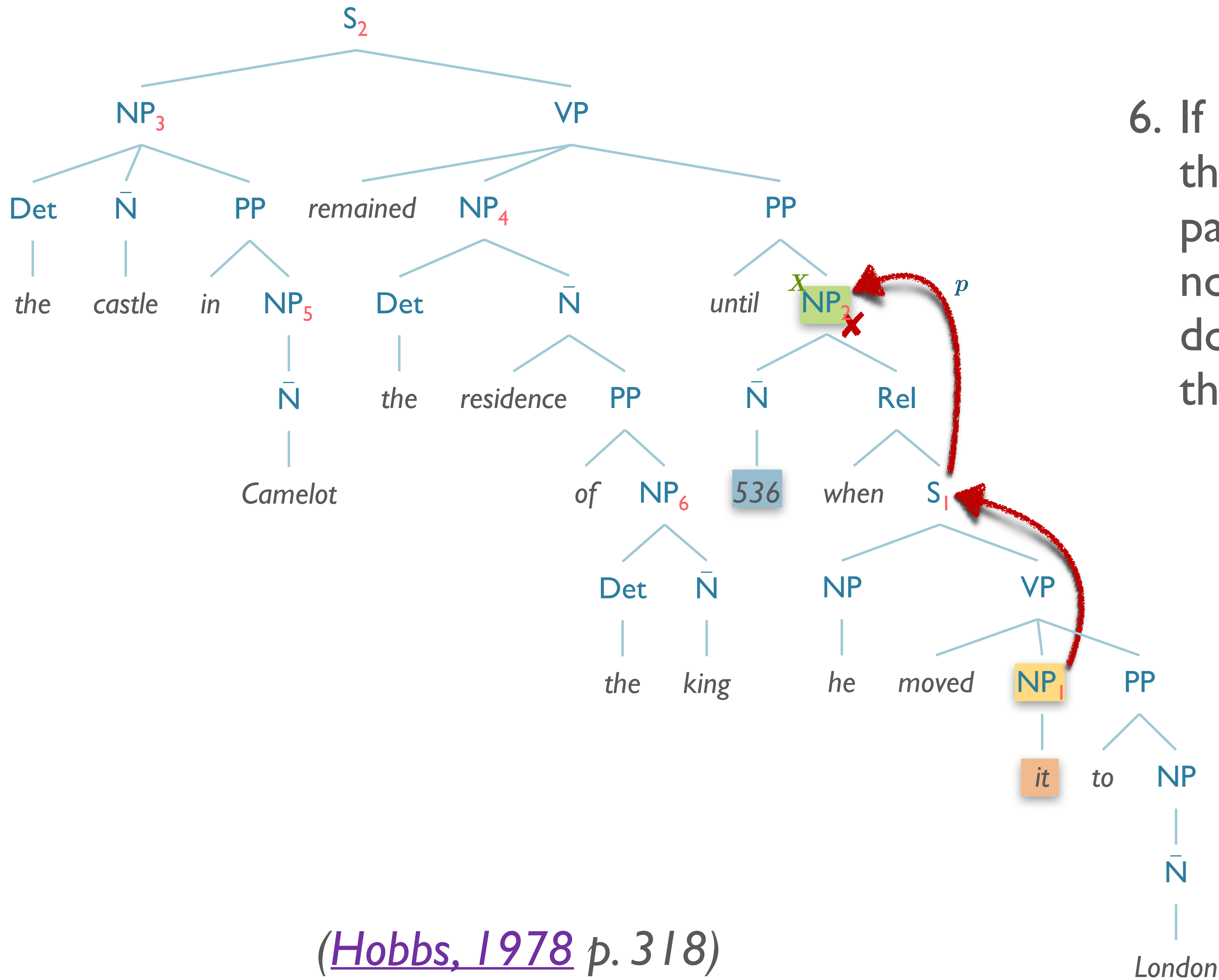
No NP or S between "he" NP and X

(Hobbs, 1978 p. 318)



(Hobbs, 1978 p. 318)

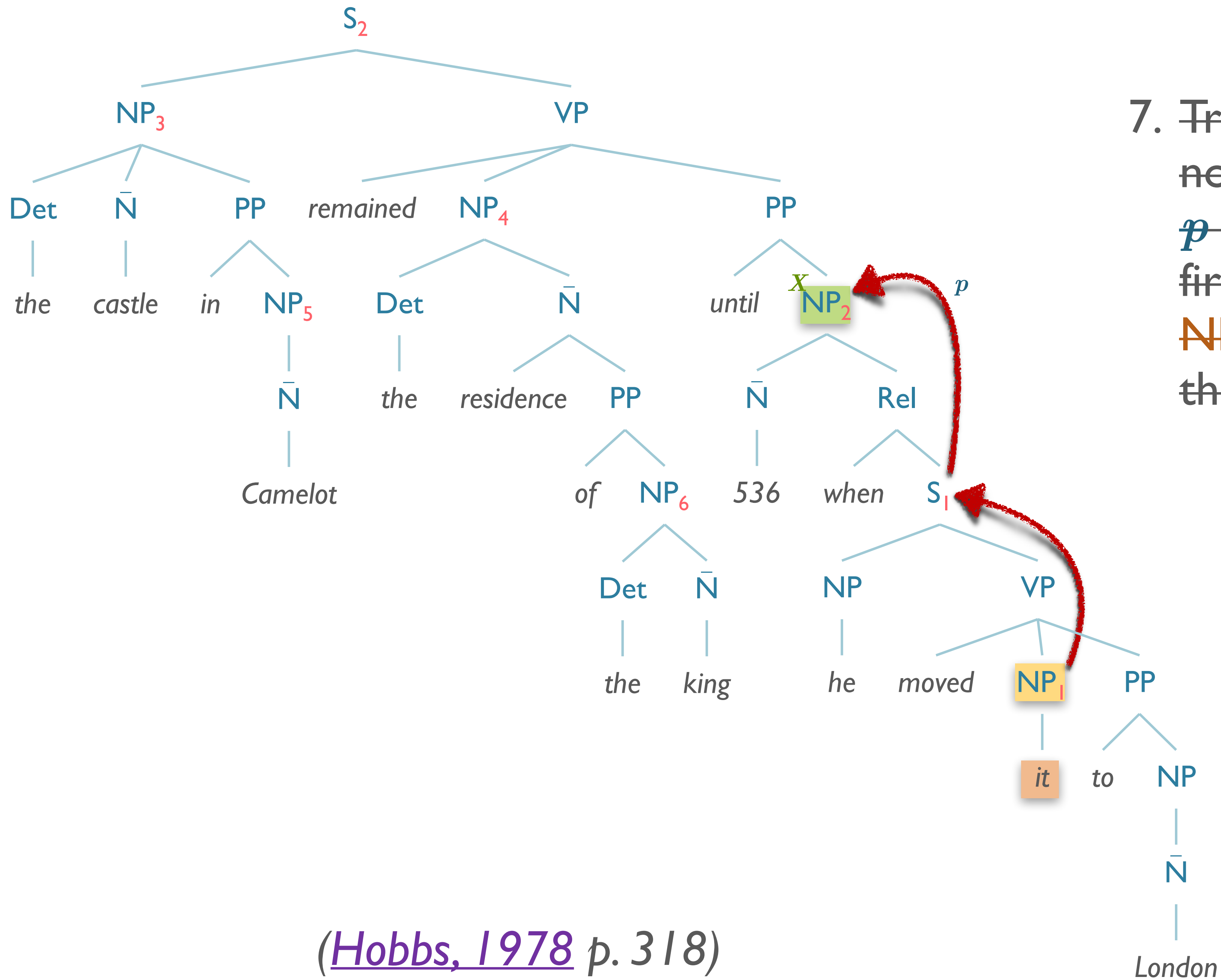
4. If node X is the highest S node in the sentence...
5. From node X , go up the tree to the first NP or S node encountered. Call this new node X , and call the path traversed to reach it p .



6. If X is an NP node and if the path p to X did not pass through the Nominal node that X immediately dominates, propose X as the antecedent.

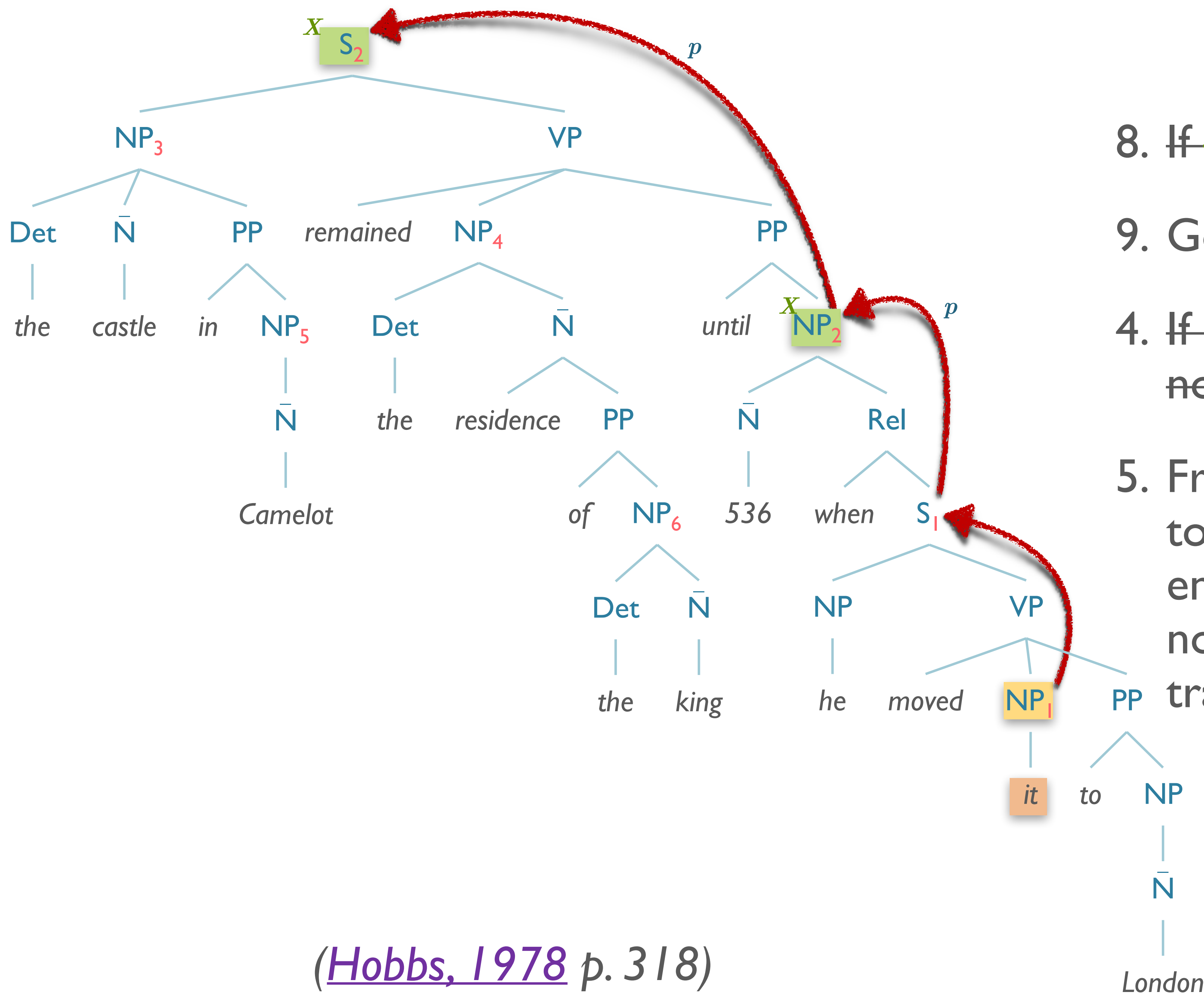
“536” can’t be “moved”!

(Hobbs, 1978 p. 318)



(Hobbs, 1978 p. 318)

7. ~~Traverse all branches below node **X** to the left of path *p* in a left-to-right, breadth-first manner. Propose any **NP** node encountered as the antecedent.~~



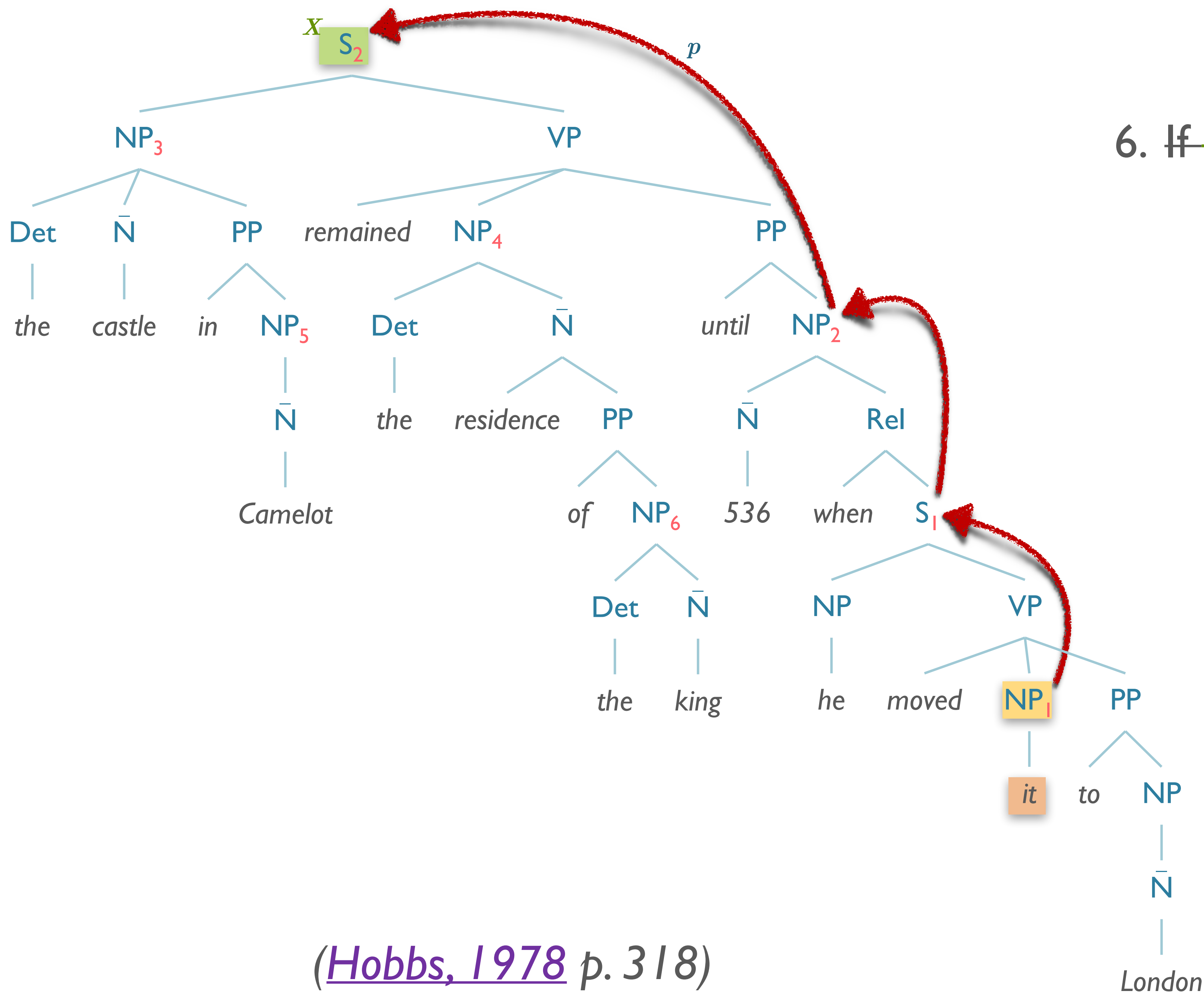
(Hobbs, 1978 p. 318)

8. If ~~X~~ is an ~~S~~ node...

9. Go to step 4.

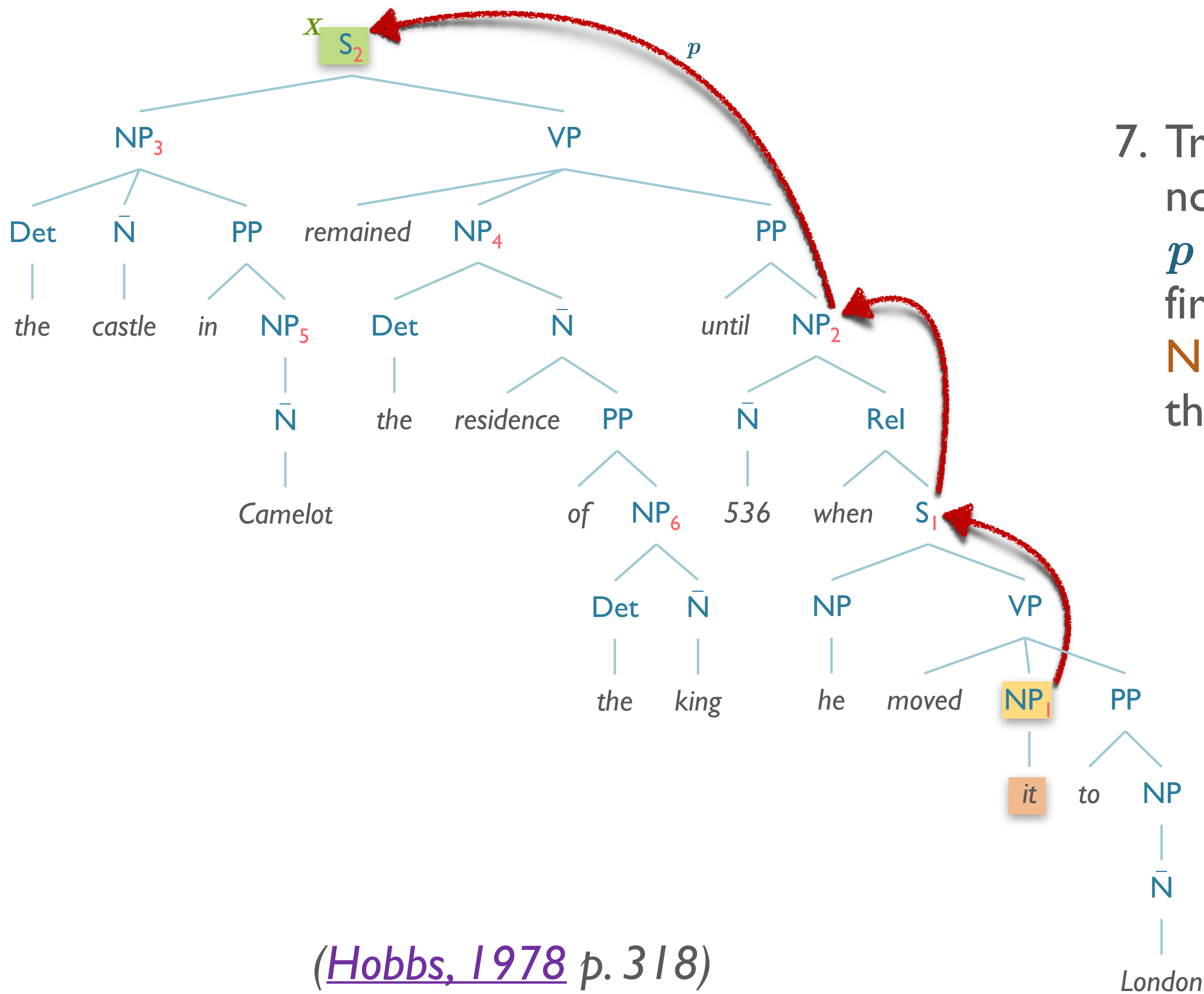
4. If node ~~X~~ is the highest ~~S~~ node in the sentence...

5. From node X , go up the tree to the first NP or S node encountered. Call this new node X , and call the path traversed to reach it p .



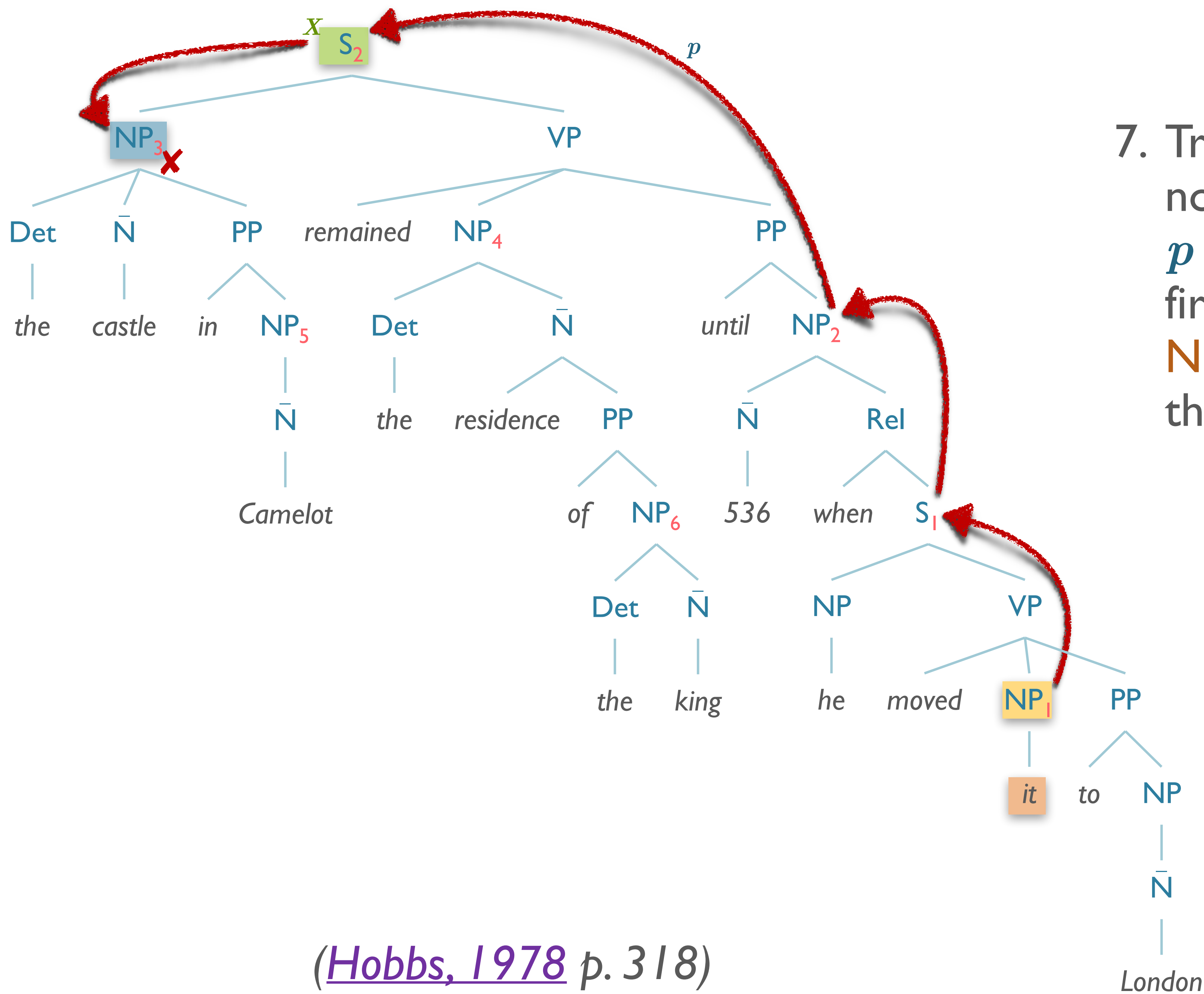
6. If ~~X~~ is an ~~NP~~ node...

(Hobbs, 1978 p. 318)



7. Traverse all branches below node X to the *left* of path p in a left-to-right, breadth-first manner. Propose any **NP** node encountered as the antecedent.

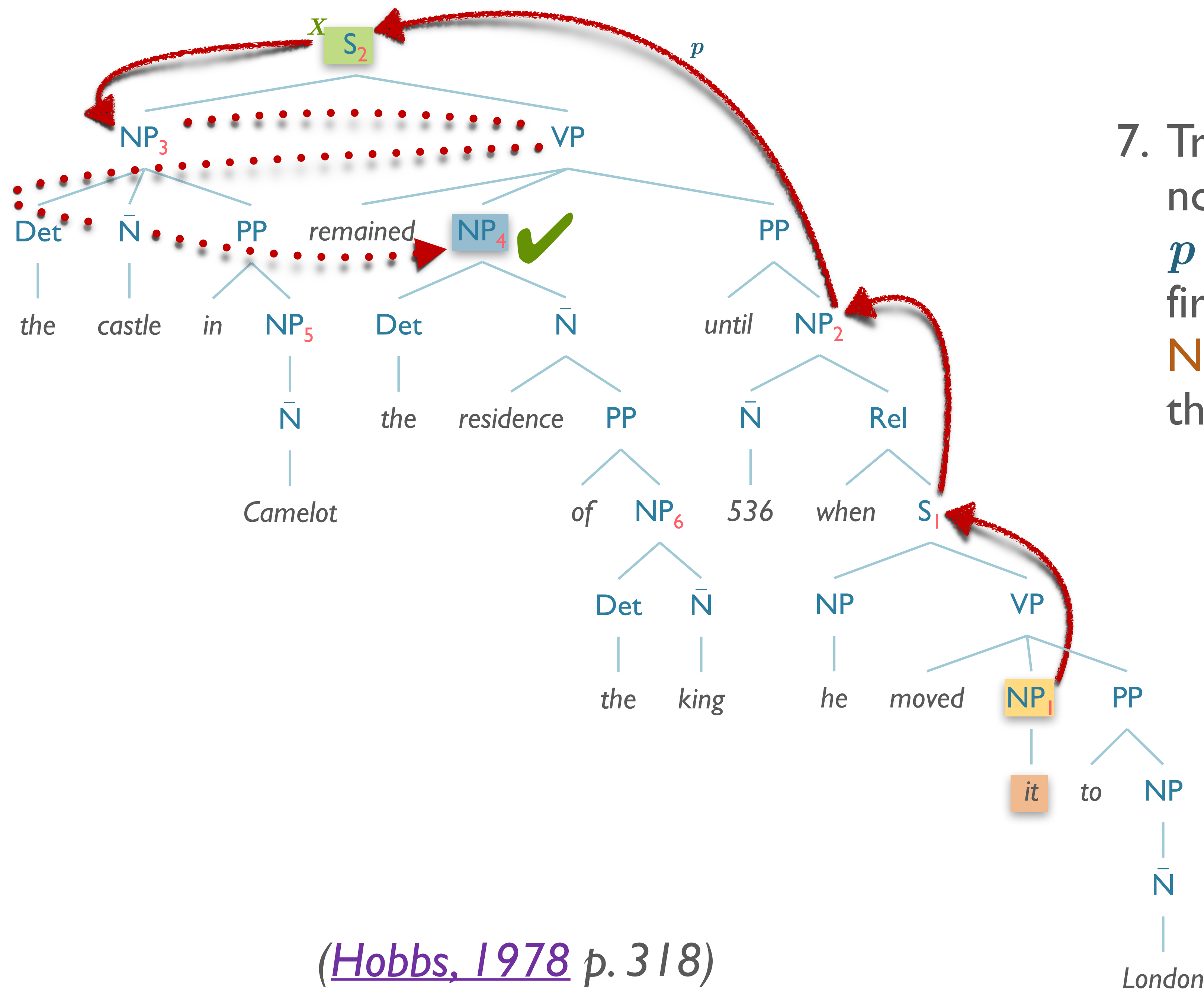
(Hobbs, 1978 p. 318)



7. Traverse all branches below node **X** to the *left* of path **p** in a left-to-right, breadth-first manner. Propose any **NP** node encountered as the antecedent.

Moving castles? 🤔

(Hobbs, 1978 p. 318)



7. Traverse all branches below node **X** to the *left* of path **p** in a left-to-right, breadth-first manner. Propose any **NP** node encountered as the antecedent.

Hobbs Algorithm Detail (Hobbs, 1978)

1. 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.

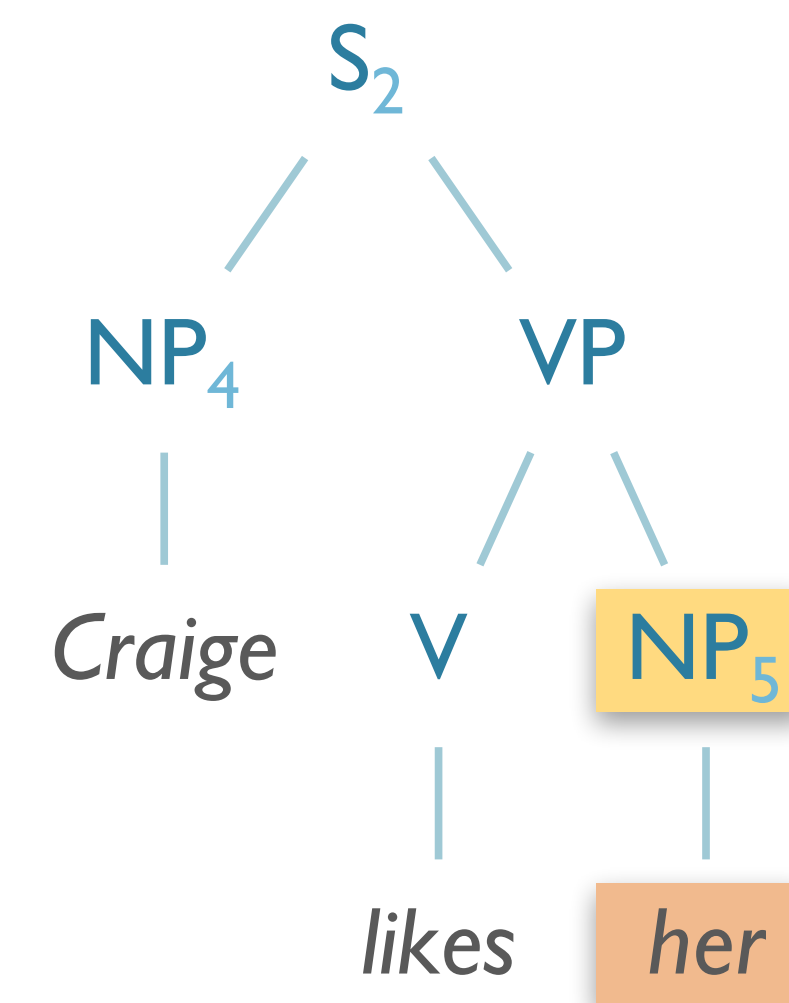
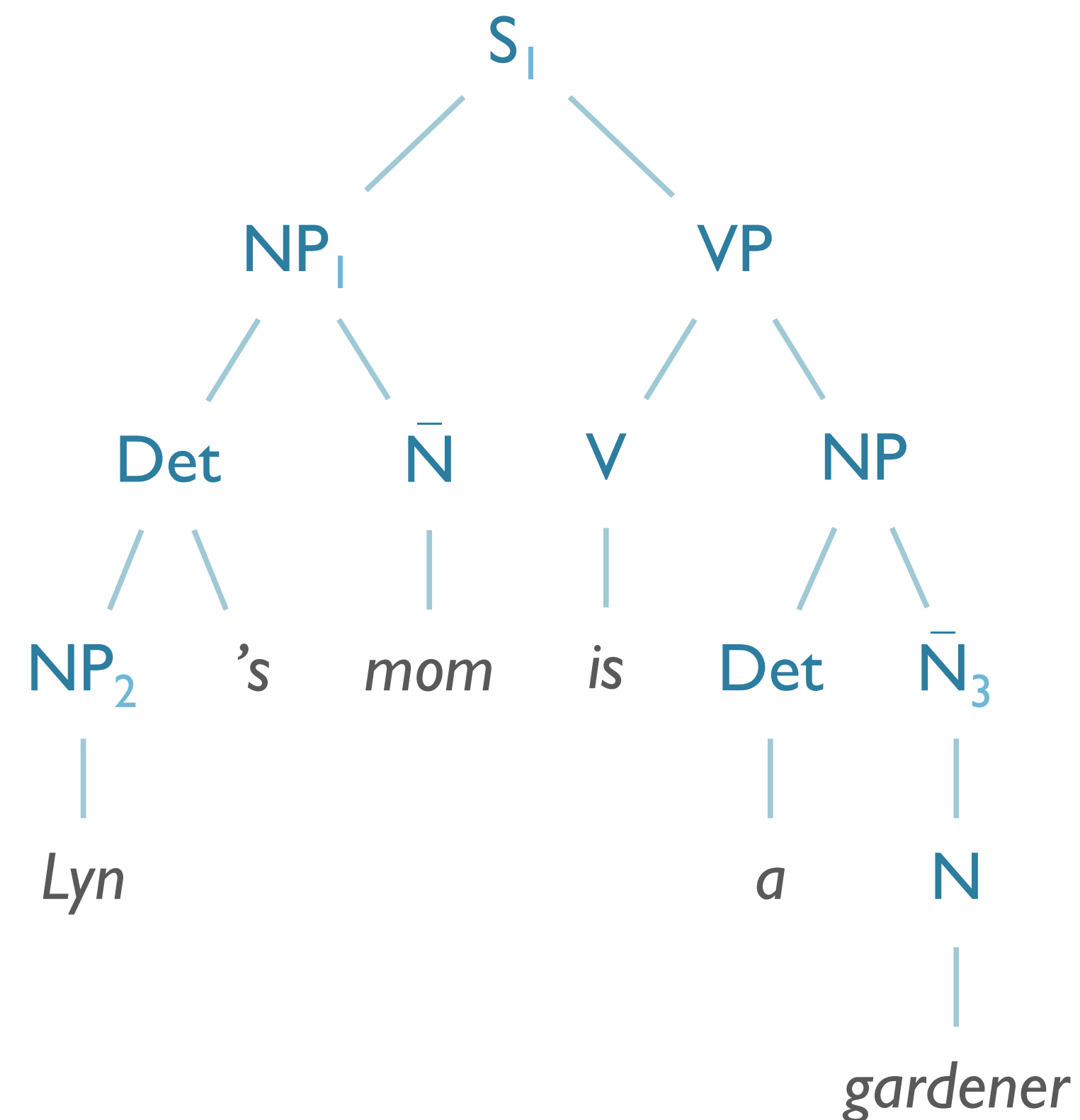
Hobbs Algorithm Detail (Hobbs, 1978)

5. From node X , go up the tree to the first NP or S node encountered. Call this new node X , and call the path traversed to reach it p .
6. If X is an NP node and if the path p to X did not pass through the Nominal node that X immediately dominates, propose X as the antecedent.
7. Traverse all branches below node X to the *left* of path p in a left-to-right, breadth-first manner. Propose any NP node encountered as the antecedent.
8. If X is an S node, traverse all branches of node X to the *right* of path p in a left-to-right, breadth-first manner, but do not go below any NP or S node encountered. Propose any NP node encountered as the antecedent.
9. Go to step 4.

Hobbs Example

Lyn's mom is a gardener.

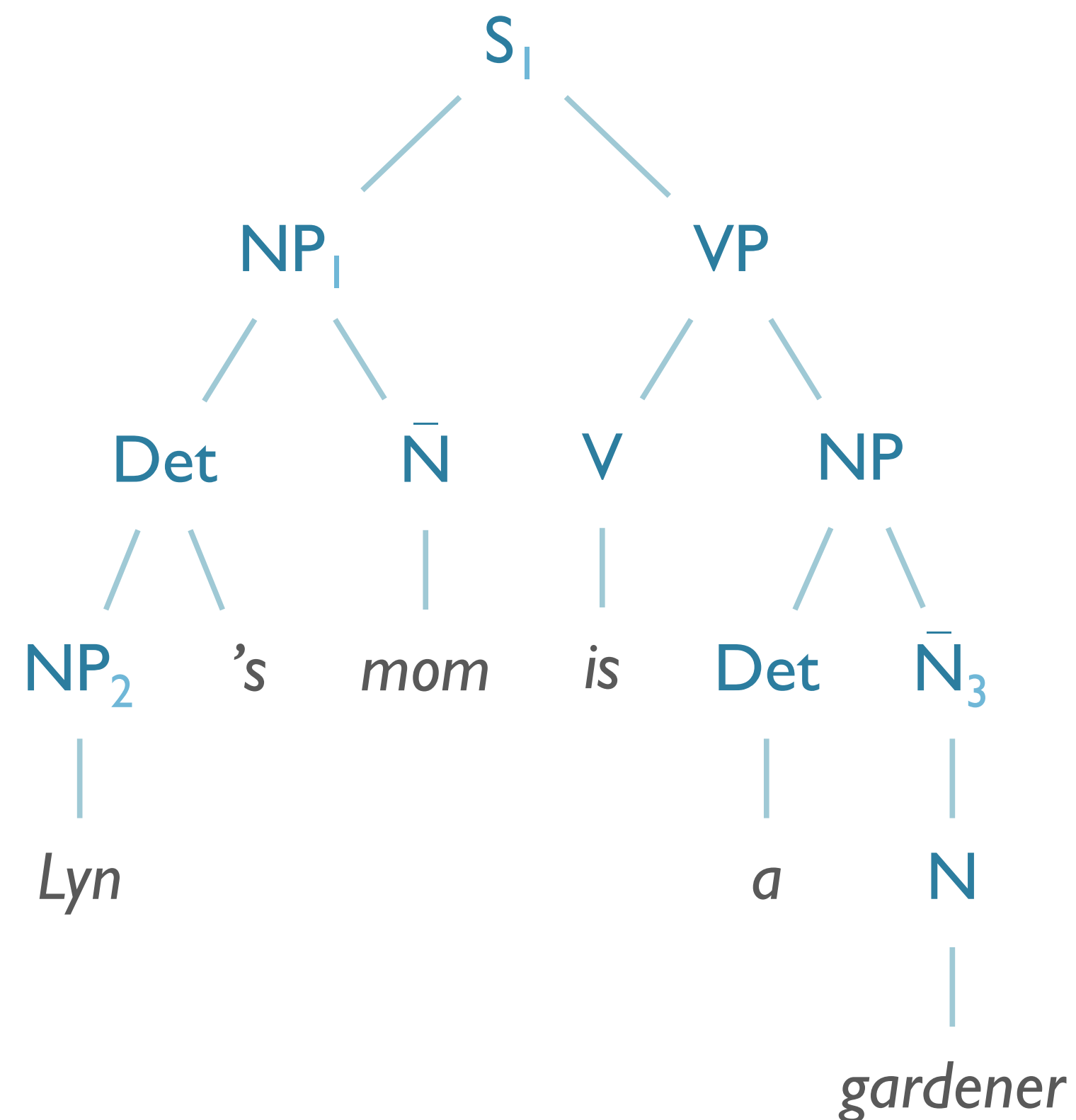
Craige likes her.



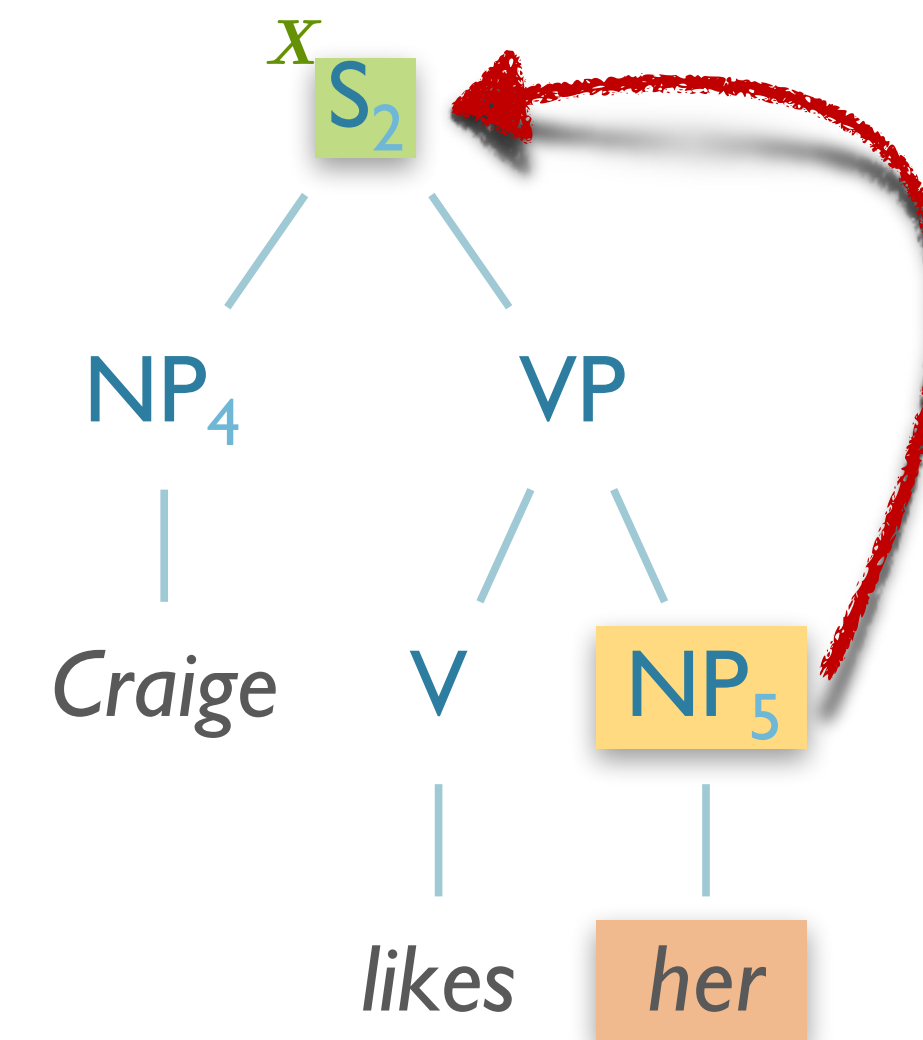
1. Begin at the noun phrase (**NP**) node immediately dominating the pronoun

Hobbs Example

Lyn's mom is a gardener.



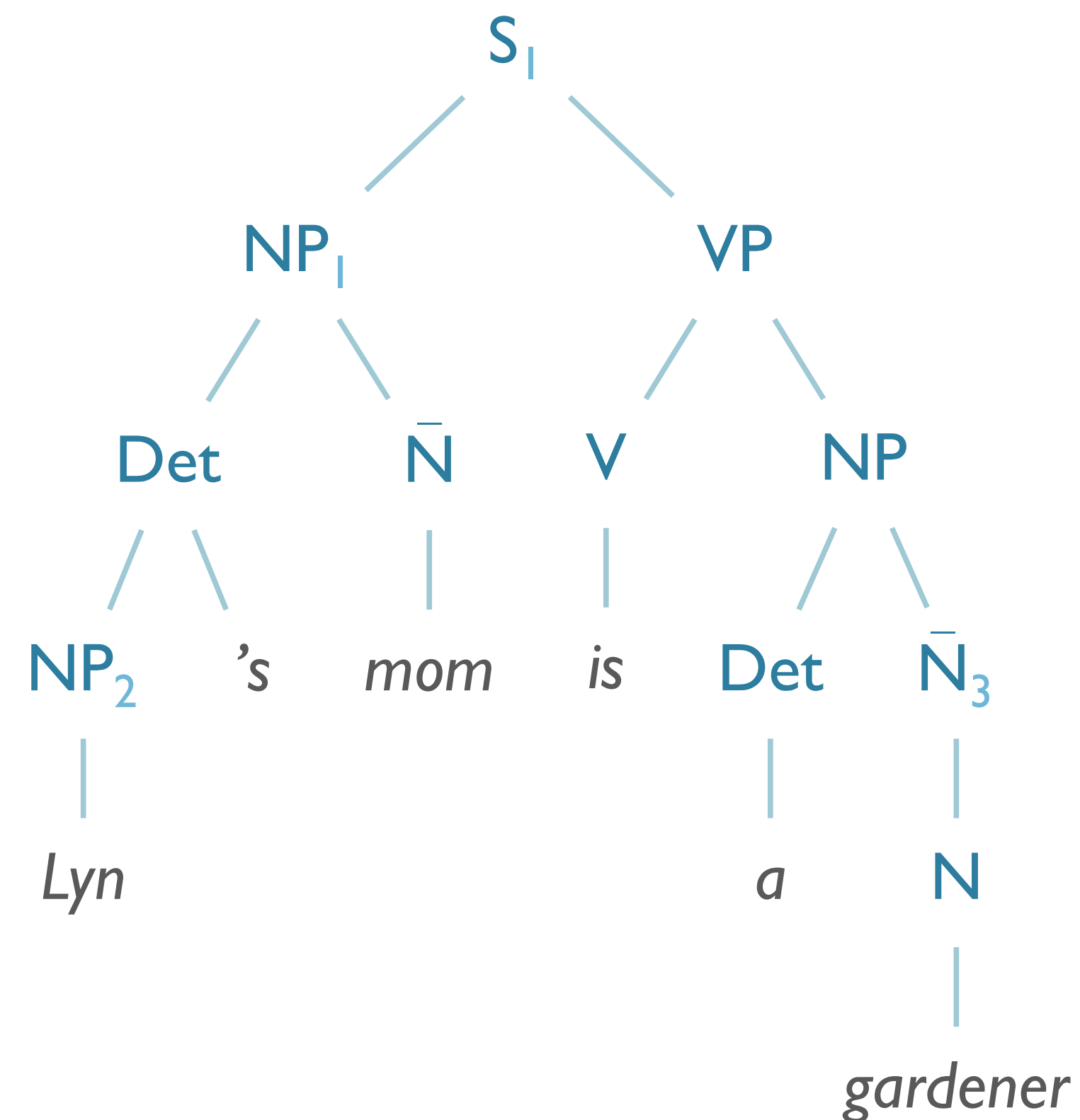
Craige likes her.



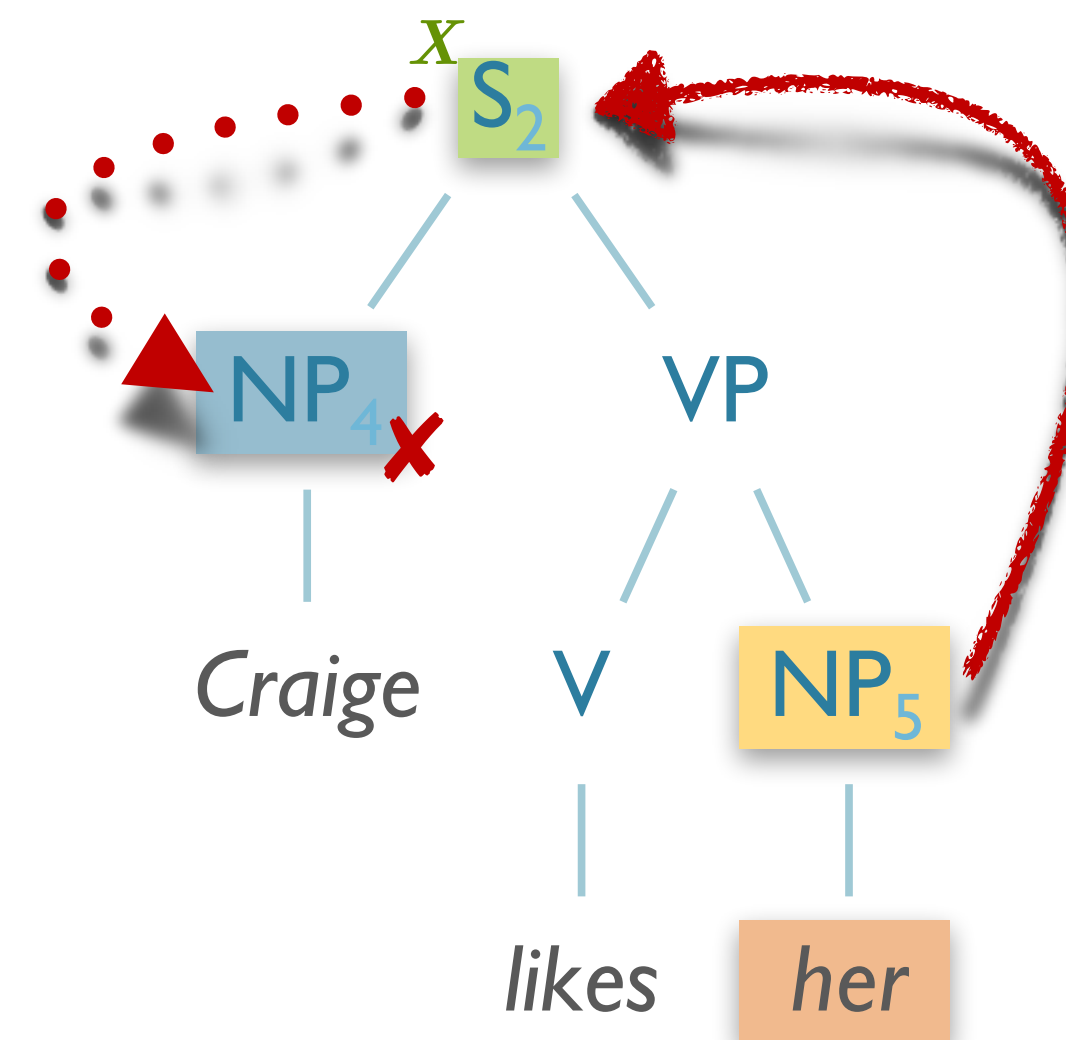
2. Go up the tree to the first **NP** or sentence (**S**) node encountered. Call this node **X**, and call the path used to reach it p .

Hobbs Example

Lyn's mom is a gardener.



Craige likes her.

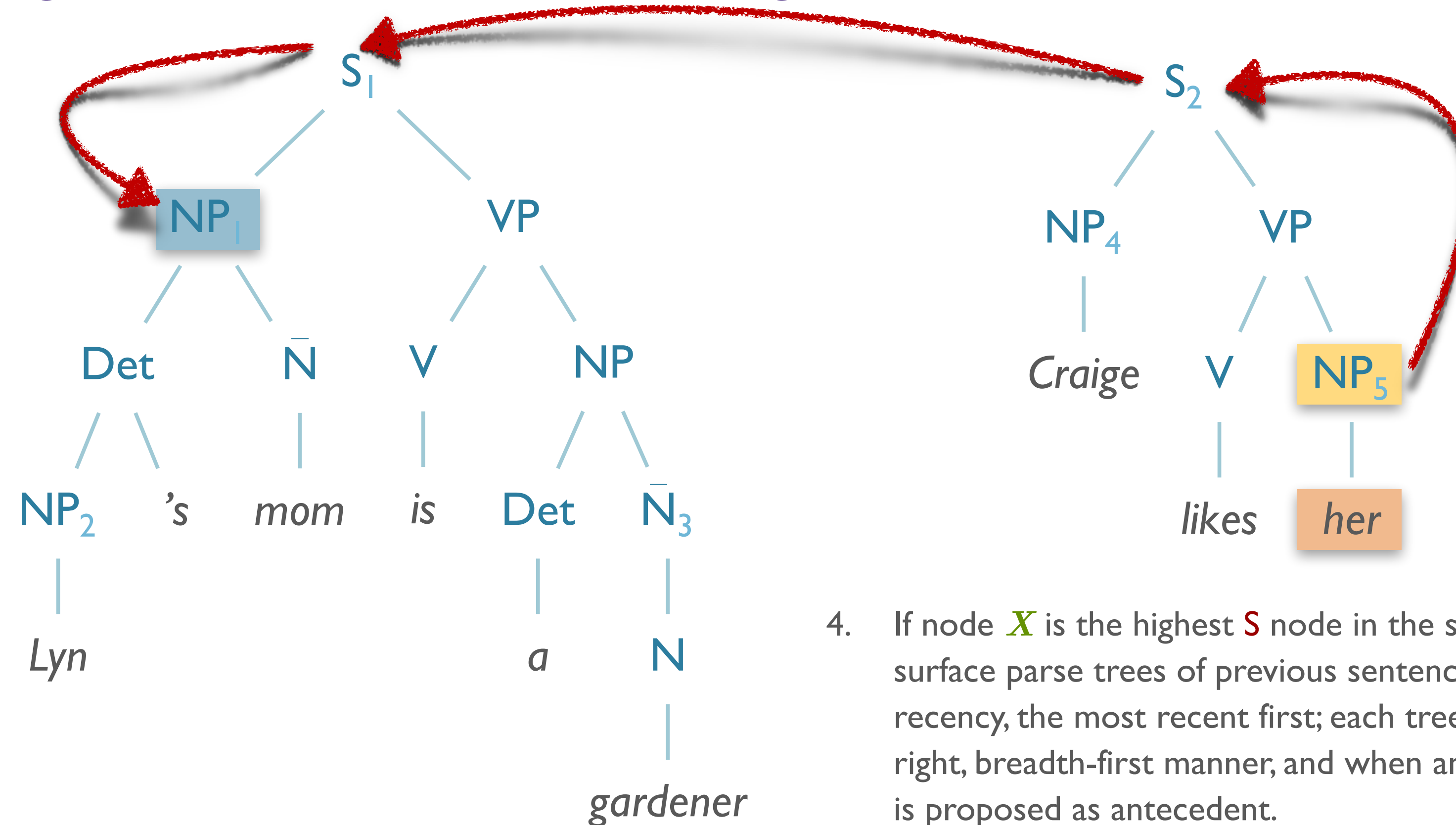


3. Traverse all branches below node X to the left of path p in a left-to-right, breadth-first fashion. Propose as the antecedent any encountered NP node that has an NP or S node between it and X .

Hobbs Example

Lyn's mom is a gardener.

Craige likes her.



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.

Hobbs Example

- What about...?
 - *Lyn's mom ~~is~~ hired a gardener.*
 - *Craige likes her.*

Coherence Relations

- **Elaboration:** Infer the same proposition P from the assertions of S_0 and S_1 .
 - *Dorothy was from Kansas. She lived in the midst of the great Kansas prairies.*
- **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 assertion of S_1 .
 - *Dorothy picked up the oil-can. She oiled the Tin Woodman's joints.*

Coherence Relation Hierarchy

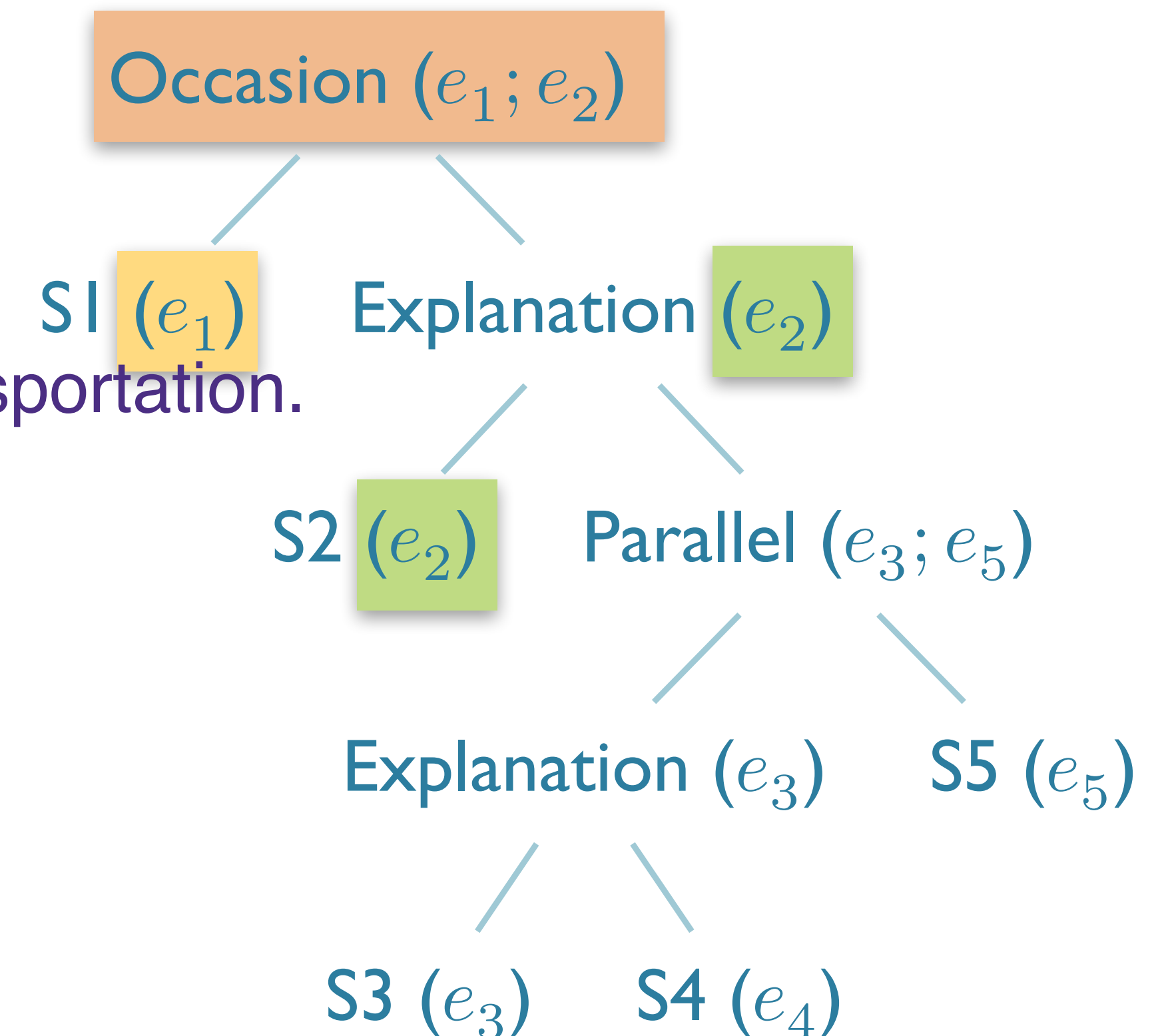
S1 – Armin went to the bank to deposit his paycheck

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.



Coherence Relation Hierarchy

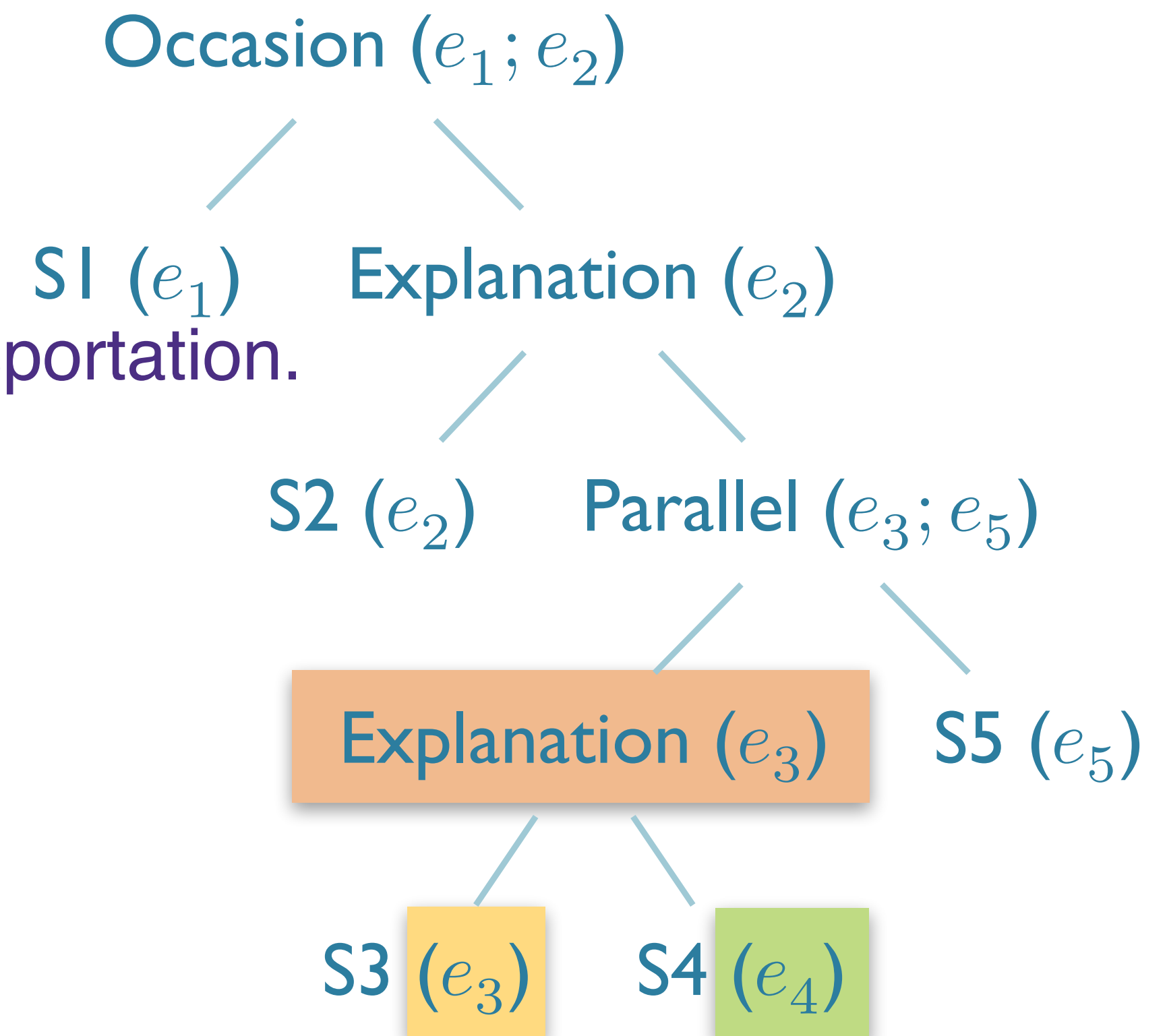
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Coherence Relation Hierarchy

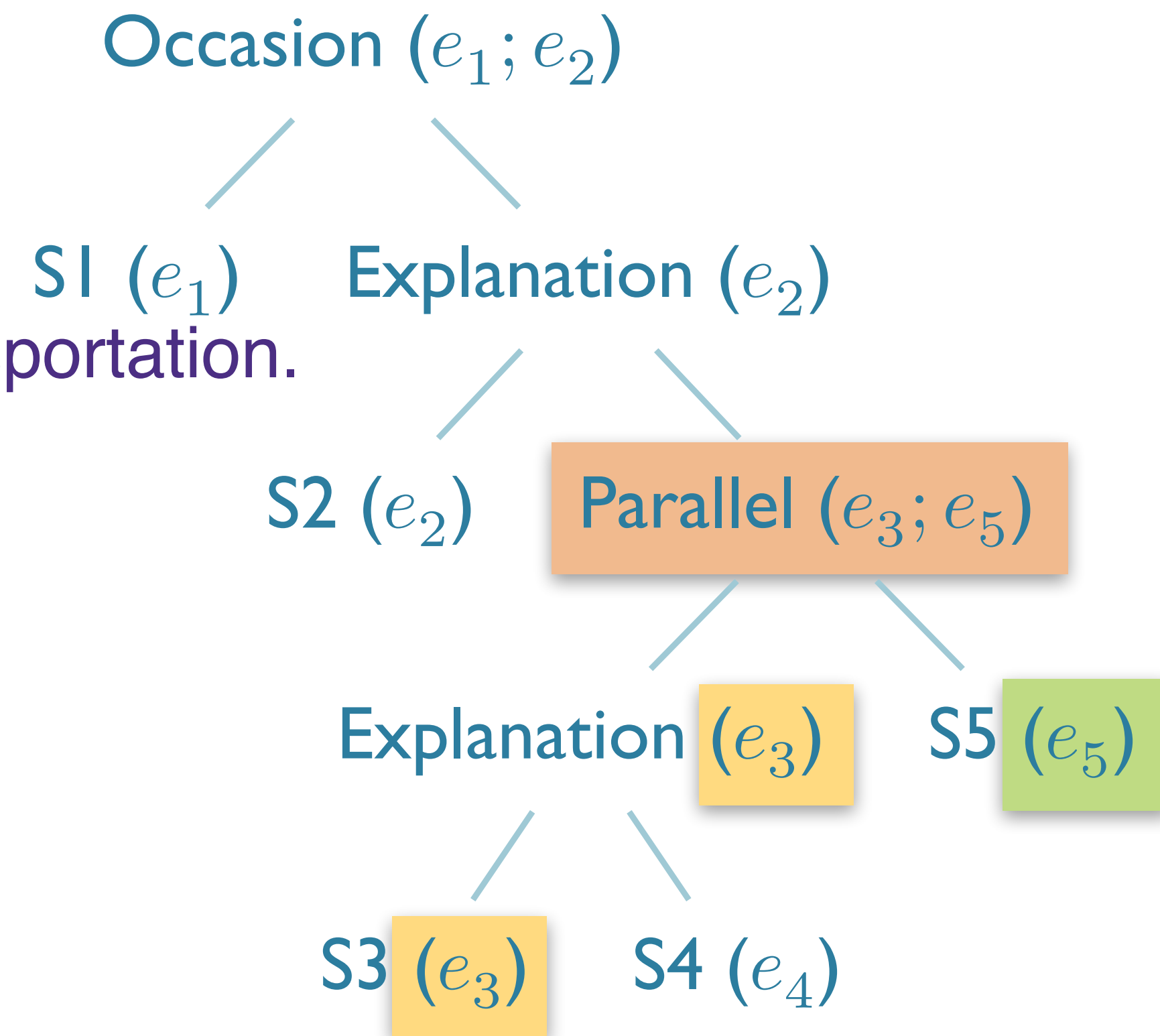
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