

Speech and Language Processing

Discourse: Anaphora Resolution

Dan Jurafsky

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Reference Resolution

- Two examples:
 - John went to Bill's car dealership to check out an Acura Integra. He looked at it for half an hour
 - I'd like to get from Boston to San Francisco, on either December 5th or December 6th. It's ok if it stops in another city along they way
- What is the target of "it"?
 - First example: two possible targets
 - Second example: where is the target?

Why reference resolution?

- **Conversational Agents:** Airline reservation system needs to know what “it” refers to in order to book correct flight
- **Information Extraction:** **First Union Corp.** is continuing to wrestle with severe problems unleashed by a botched merger and a troubled business strategy. According to industry insiders at **Paine Webber**, **their** president, John R. Georgius, is planning to retire by the end of the year.

Some terminology

- **John** went to Bill's car dealership to check out **an Acura Integra**. **He** looked at **it** for half an hour
- **Reference**: process by which speakers use words **John** and **he** to denote a particular person
 - **Referring expression**: John, he
 - **Referent**: the actual entity (but as a shorthand we might call "John" the referent).
 - John and he "corefer"
 - **Antecedent**: John
 - **Anaphor**: he
- **Cataphora**: pronoun before the referent
 - Before he bought **it**, John checked over the **Integra** very carefully

Many types of reference

- (after Webber, '91)
- According to John, Bob bought Sue an Integra, and Sue bought Fred a Legend
 - But **that** turned out to be a lie (a speech act)
 - But **that** was false (proposition)
 - **That** struck me as a funny way to describe the situation (manner of description)
 - **That** caused Sue to become rather poor (event)
- But we focus on references to entities

Reference Phenomena

- Indefinite noun phrases: new to hearer
 - I saw **an Acura** Integra today
 - **Some Acura** Integras were being unloaded...
- Definite noun phrases: identifiable to hearer because
 - **Mentioned**: I saw an Acura Integra today. **The Integra** was white
 - **Identifiable from beliefs (common knowledge)**: **The Indianapolis 500**
 - **Inherently unique**: **The fastest car** in Indianapolis 500...

Reference Phenomena: Pronouns

- Simple example:
 - I saw an **Acura Integra** today. **It** was white
- Referent **salience**, in case of discourse:
 1. John went to Bob's party, and parked next to a beautiful Acura Integra
 2. He went inside and talked to Bob for more than an hour.
 3. Bob told him that he recently got engaged.
 4. a) He also said that he bought **it** yesterday. **“it”... what?**
b) He also said that he bought **the Acura** yesterday.

Reference phenomena

- **Demonstratives**

- I bought **an Integra yesterday**. It's similar to **the one I bought five years ago**. **That one** was really nice, but I like **this one** even better

- **A non-pronominal anaphora**

- I saw no less than **6 Acura Integra** today. I want **one**
 - ... **one (of them)**

Reference phenomena

- **Inferrable**
 - I almost bought an Acura Integra today, but **the engine** seemed noisy.
- **Pronominal**
 - I saw no less than 6 **Acura Integras** today. **They** are the coolest cars.

Pronominal Reference Resolution

- **Given a pronoun, find the reference**
- **Constraints to leverage**
 - **Hard constraints on reference**
 - **Soft constraints on reference**
- **Algorithms which use/don't use these constraints**

Hard constraints: syntax

- **Number agreement**
 - *John has an **Acura**. **They???** are red
 - John has an **Acura**. **It** is red
- **Person and case agreement**
 - *John and Mary have **Acuras**. We love **them???**
(**who/what???**)
 - John and I have **Acuras**. We love **them**.
- **Gender agreement**
 - **John** has an **Acura**. **He / it** is attractive.
- **Syntactic constraints**
 - **John** bought **himself** a new Acura (himself == John)
 - John bought **him** a new Acura (him != John)

Soft constraints: Pronoun Interpretation Preferences

- **Selectional Restrictions**
 - John parked **his Acura** in **the garage**. He had driven **it** around for hours.
- **Recency**
 - John has **an Integra**. Bill has a **Legend**. Mary likes to drive **it**.

Soft constraints: Pronoun Interpretation Preferences

- **Grammatical Role: Subject preference**
 - **John** went to the Acura dealership with **Bill**.
He bought an Integra.
 - **John** and **Bill** went to the Acura dealership. **He**
bought an Integra

Soft constraints: Repeated Mention preference

- **John** needed a car to get to his new job. **He** decided that he wanted something sporty. **Bill** went to the Acura dealership with **him**. **He** bought an Integra.

Soft constraints: Parallelism Preference

- **Same structure**
 - **Mary** went with **Sue** to the Acura dealership.
Sally went with **her** to the Mazda dealership.
- **But... with similar structure**
 - **Mary** went with **Sue** to the Acura dealership.
Sally told **her** not to buy anything.

Soft constraints: Verb Semantics Preferences

- **John** telephoned **Bill**. **He** lost the pamphlet on Acuras.
- **John** criticized **Bill**. **He** lost the pamphlet on Acuras.
- **Implicit causality**
 - Implicit cause of criticizing is object.
 - Implicit cause of telephoning is subject.

Algorithms for pronoun anaphora resolution

- **Knowledge-rich approach**
 - Syntactic-based: Hobbs' algorithm
 - Discourse-based: Centering Theory
 - Hybrid approaches: Lappin and Leas
 - Corpus-based: Charniak, Hale, and Ge
- **Knowledge-poor approach**
 - Machine Learning

Hobbs

- Hobbs (1978) proposes an algorithm that searches parse trees for antecedents of a pronoun.
- starting at the NP node *immediately dominating the pronoun*
- in a *specified search order*
- looking for the *first match of the correct gender and number*
- Idea: discourse and other preferences will be approximated by search order.

Hobbs

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- Extending Hobbs:
 - Add “simple” selectional restrictions, e.g. (for the verb “to move”):
 - dates can't move
 - places can't move
 - large fixed objects can't move
 - For “they”, in addition to accepting plural NPs, collects selectionally compatible entities (somehow), e.g., conjoined NPs.
 - Assume some process that recovers elided constituents and inserts them in the tree.

Example:

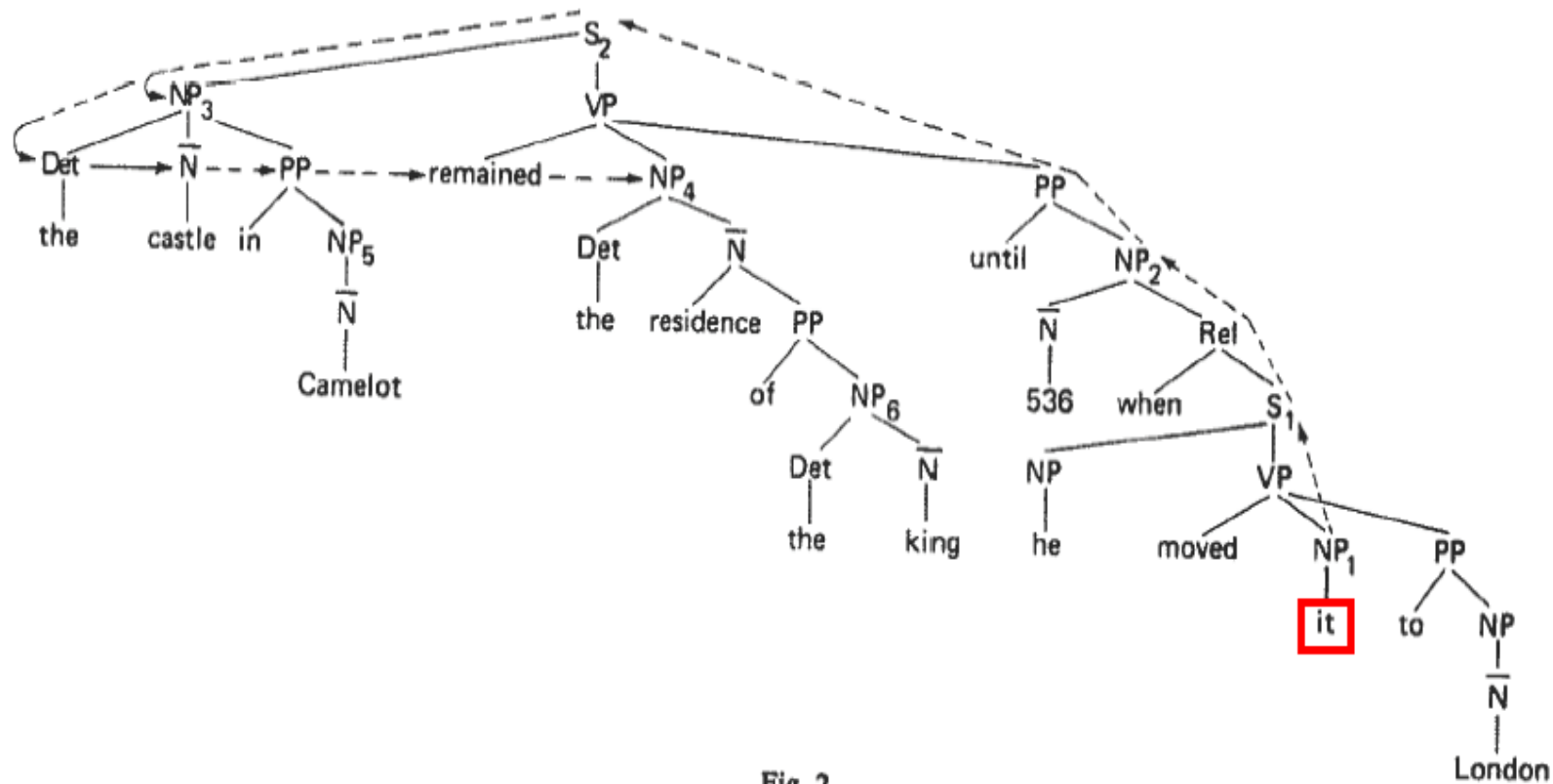


Fig. 2.

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

Hobbs' Algorithm

- 1. Begin at the NP immediately dominating the pronoun.**
- 2. Go up tree to first NP or S encountered.**
 - a. Call node X, and path to it, p.**
 - b. Search left-to-right below X and to left of p, proposing any NP node which has an NP or S between it and X.**
- 3. If X is highest S node in sentence,**
 - Search previous trees, in order of recency, left-to-right, breadth-first, proposing NPs encountered.**
- 4. Otherwise, from X, go up to first NP or S node encountered,**
 - Call this X, and path to it p.**
- 5. If X is an NP, and p does not pass through an N-bar that X immediately dominates, propose X.**
- 6. Search below X, to left of p, left-to-right, breadth-first, proposing NP encountered.**
- 7. If X is an S, search below X to right of p, left-to-right, breadth-first, but not going through any NP or S, proposing NP encountered.**
- 8. Go to 2.**

Example:

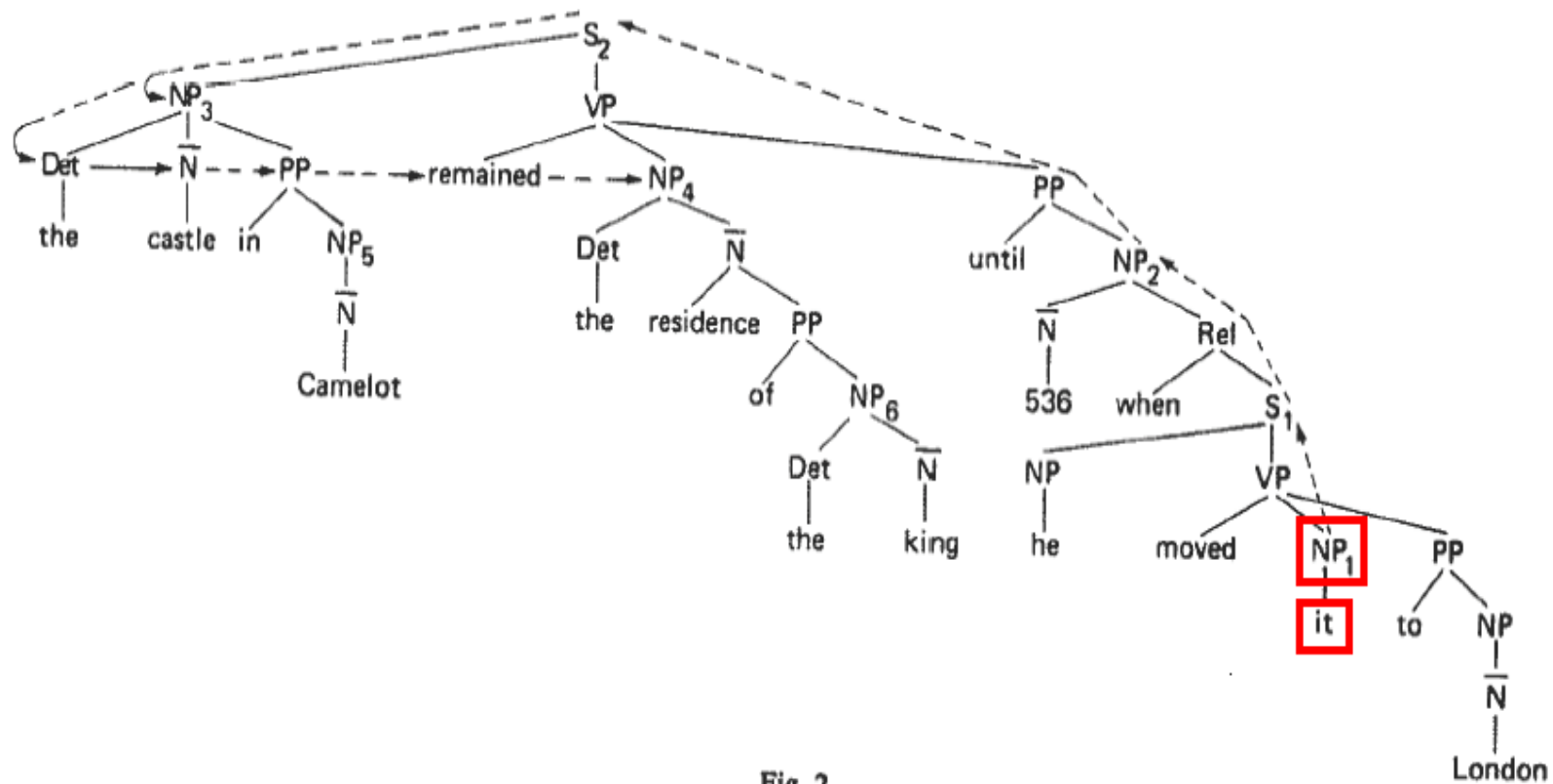


Fig. 2.

1. Begin at the NP immediately dominating the pronoun.

Hobbs' Algorithm

1. Begin at the NP immediately dominating the pronoun.
2. Go up tree to first NP or S encountered.
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Example:

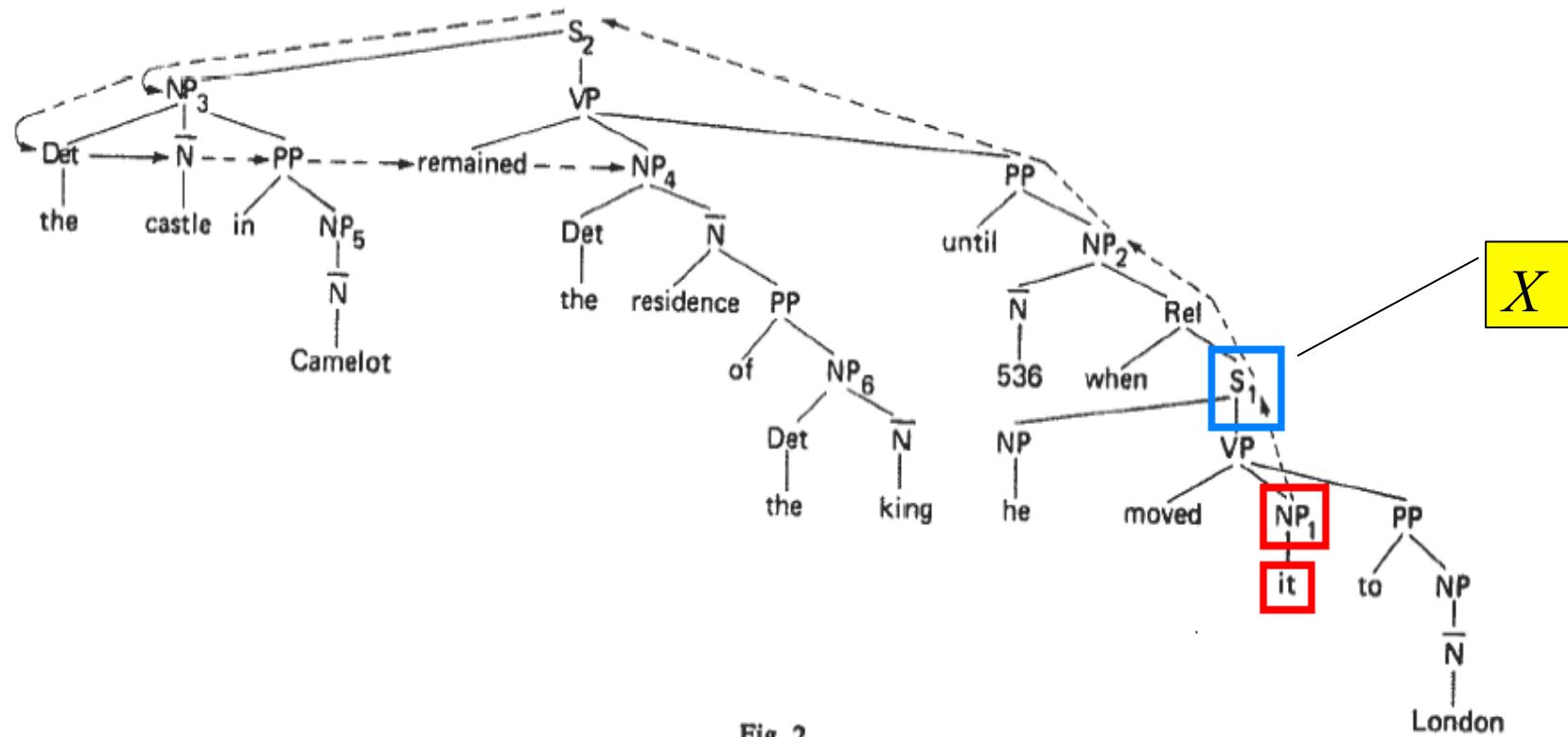


Fig. 2.

2. Go up tree to first NP or S encountered
 - a. Call node X, and path to it, p.

Example:

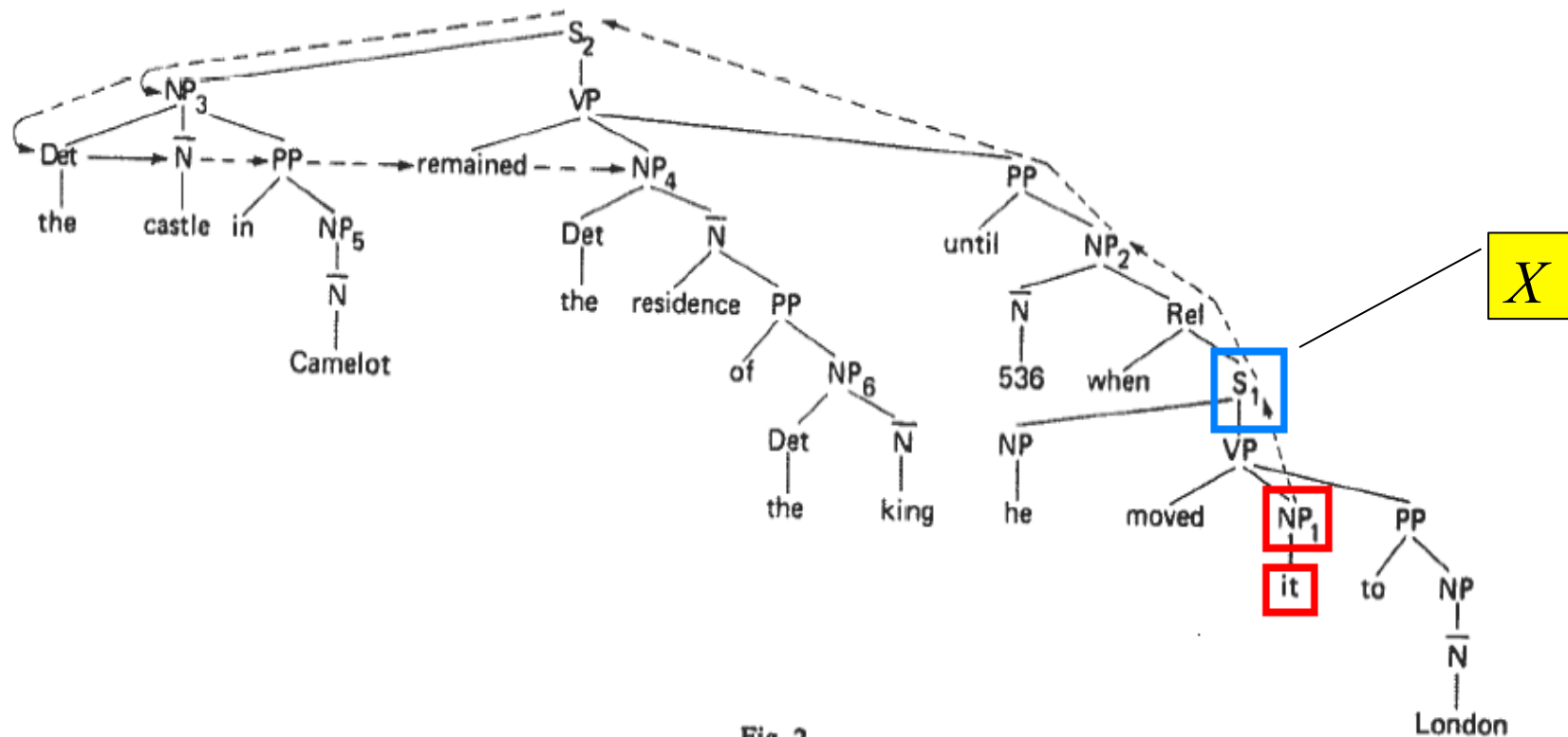


Fig. 2.

b. Search left-to-right below X and to left of p , proposing any NP node which has an NP or S between it and X

→ search yields no candidate. Go to next step of the algorithm.

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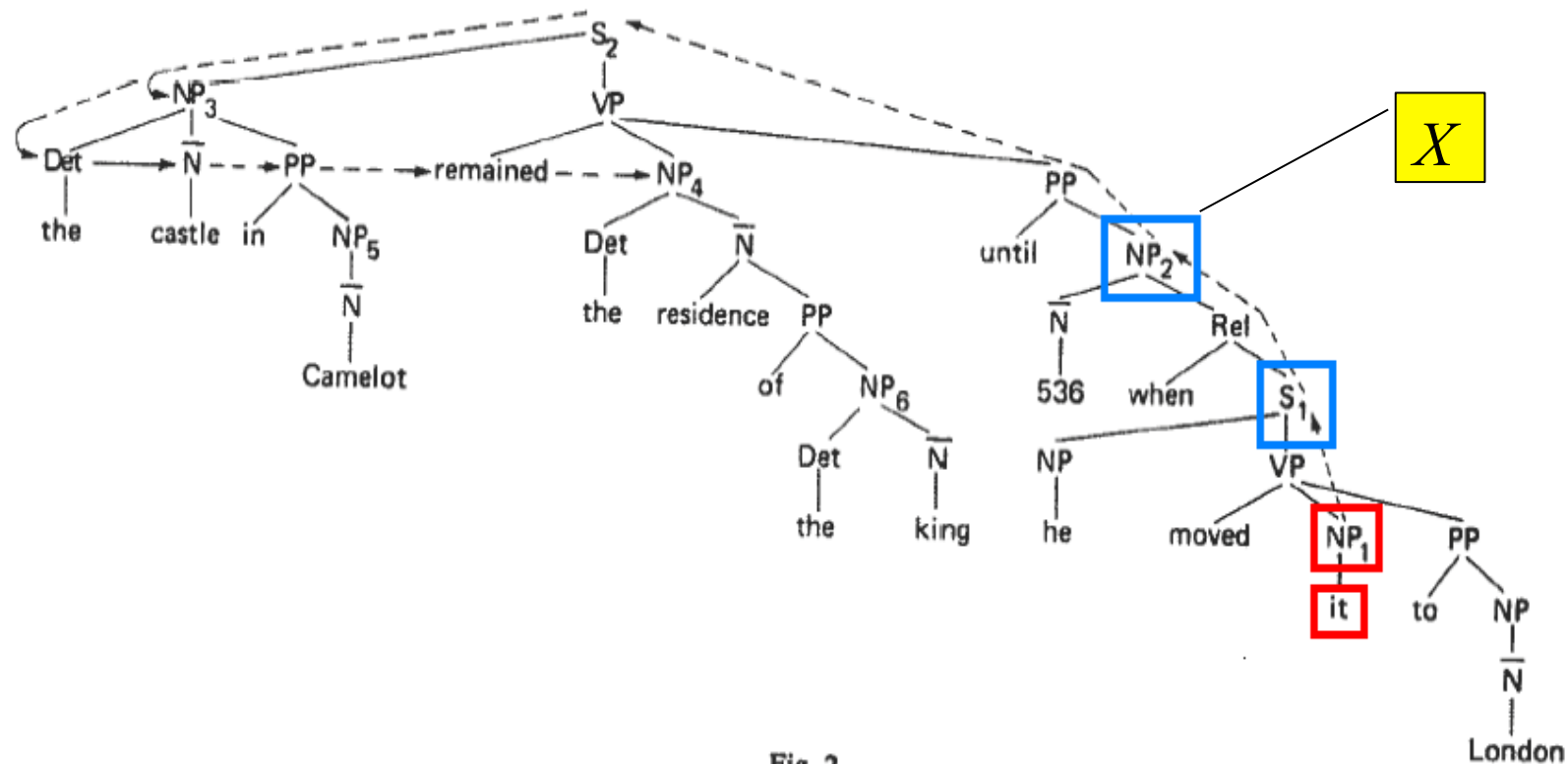


Fig. 2.

3. false

4. Otherwise, from X go up to first NP or S node encountered.

a. Call this X , and path to it p .

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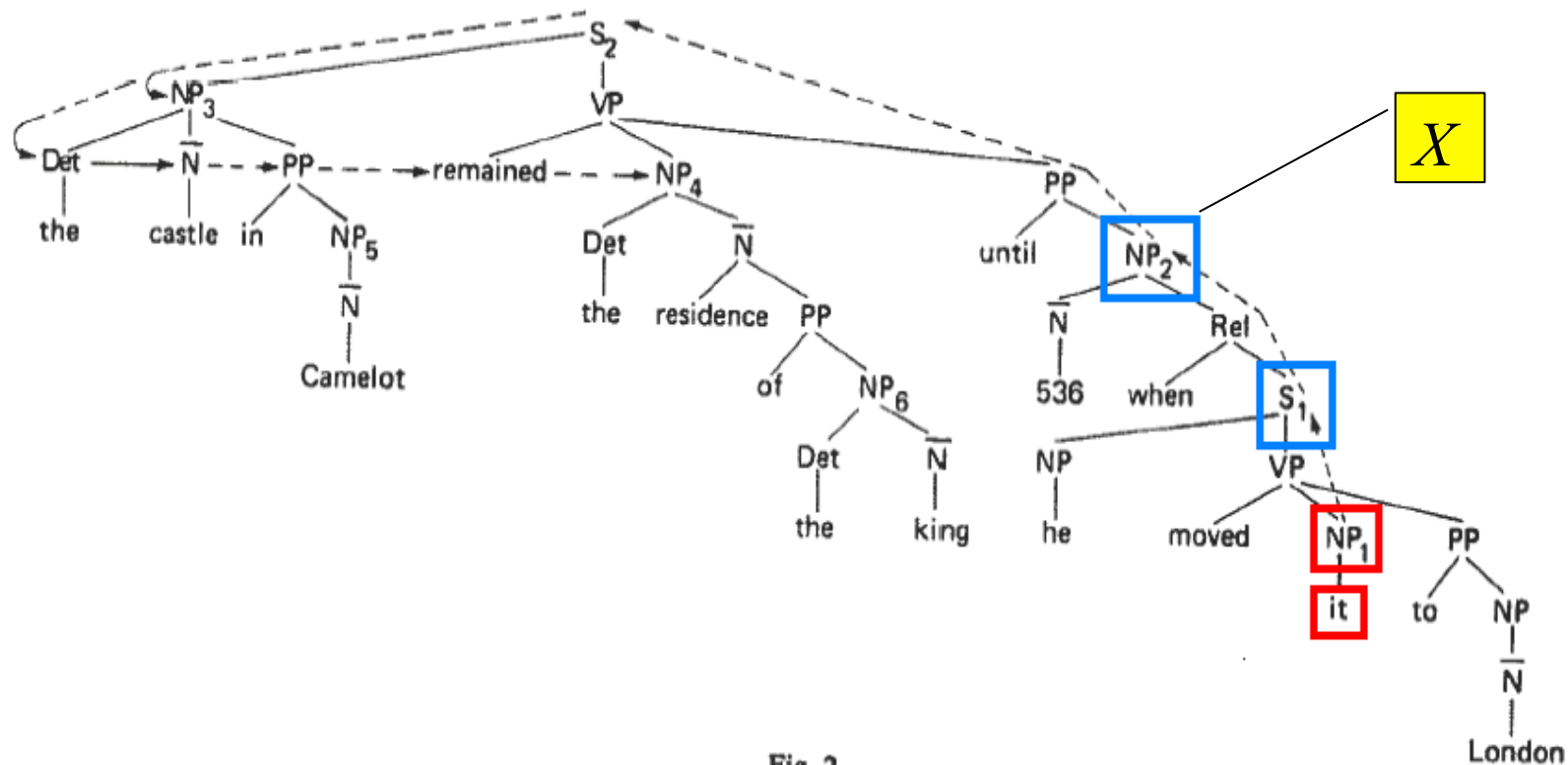


Fig. 2.

5. If X is an NP, and p does not pass through an N-bar that X immediately dominates, propose X

→ NP_2 is proposed. Rejected by selectional restrictions (dates can't move).

Hobbs' Algorithm

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8. **Go to 2.**

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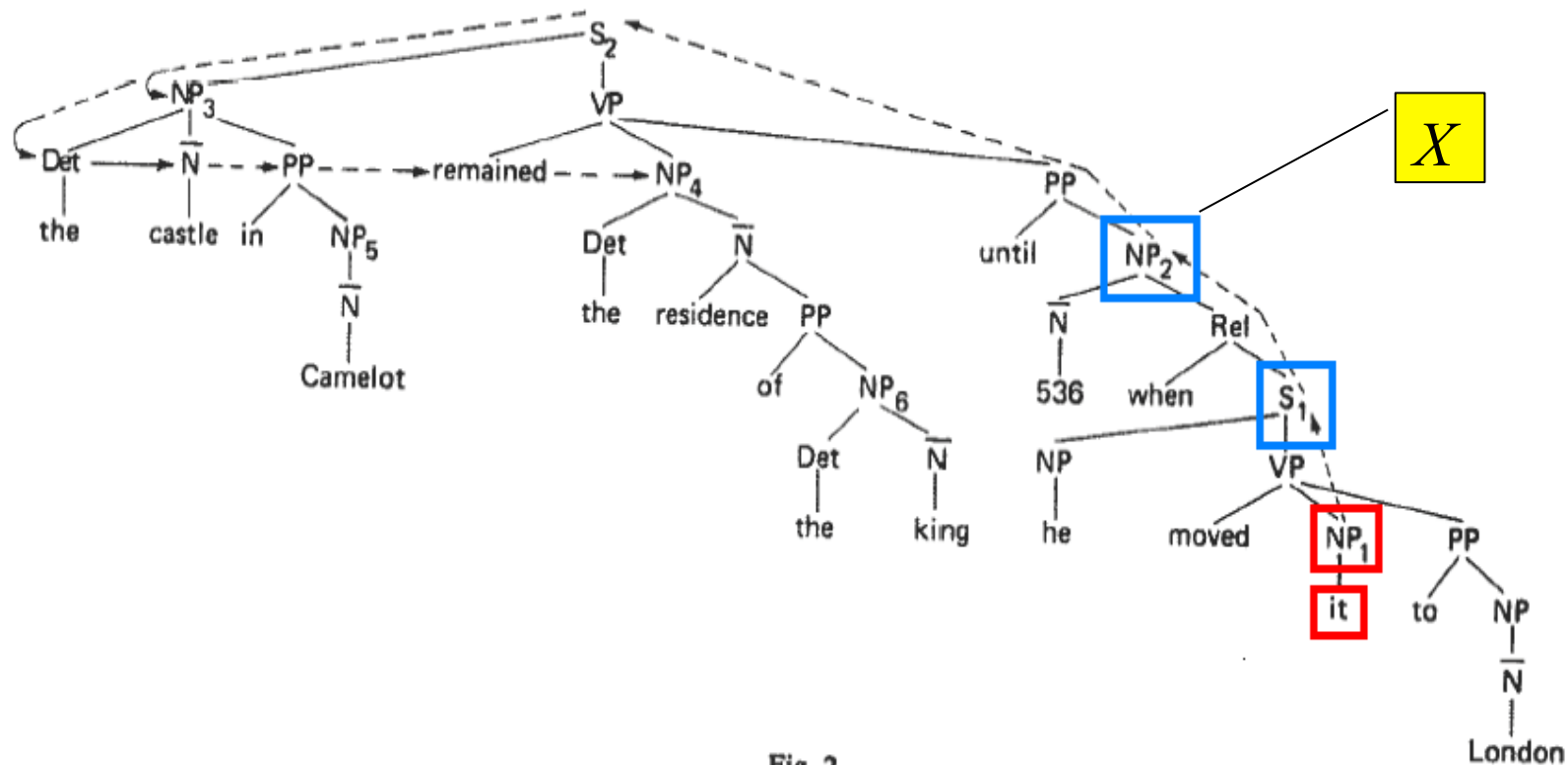


Fig. 2.

6. Search below X , to left of p , left-to-right, breadth-first, proposing NP encountered.

→ search yields no candidate. Go to next step of the algorithm.

Hobbs' Algorithm

1. **Begin at the NP immediately dominating the pronoun.**
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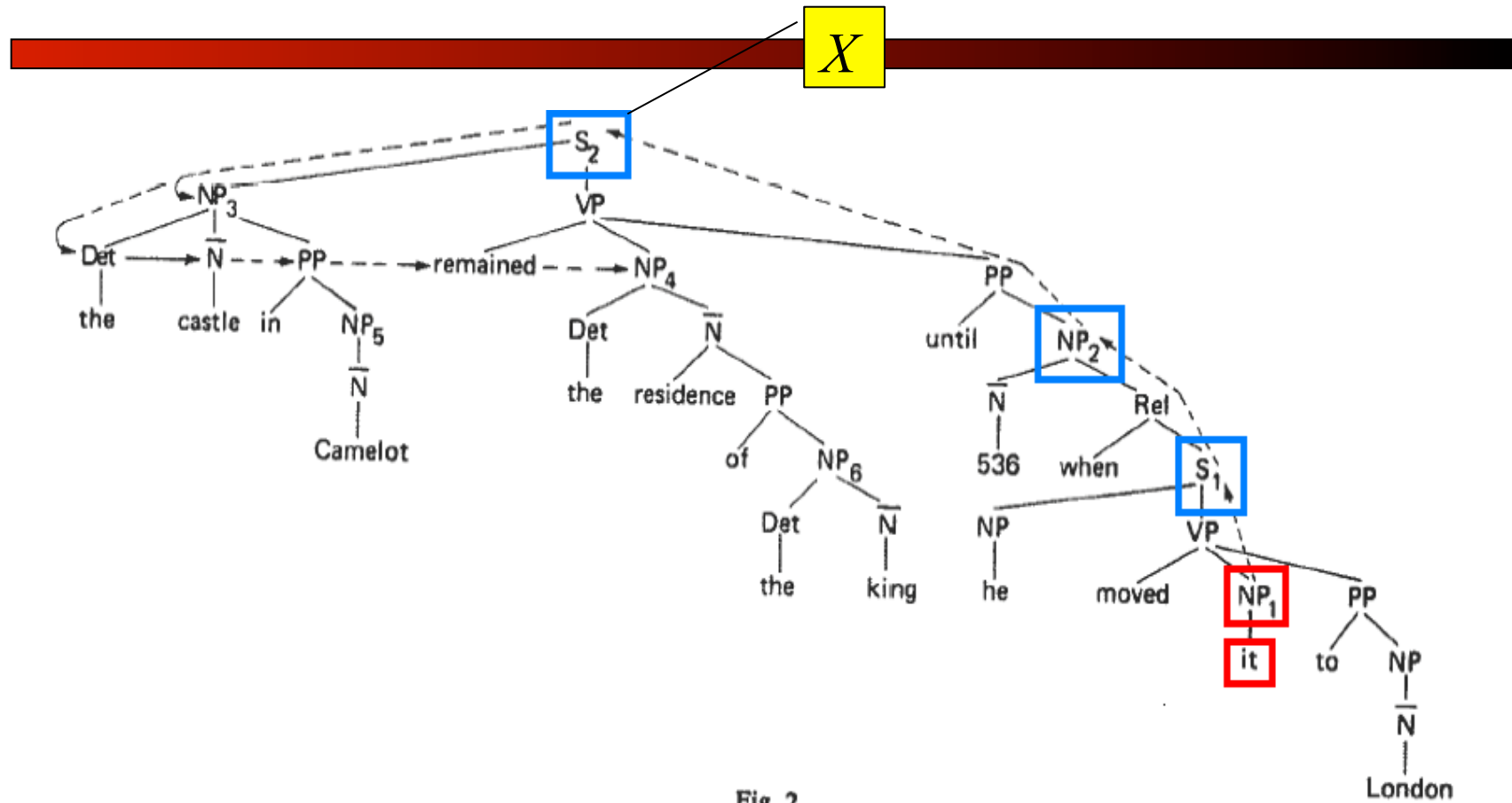


Fig. 2.

7. false
2. Go up tree to first NP or S encountered
 - a. Call node X , and path to it, p

Example:

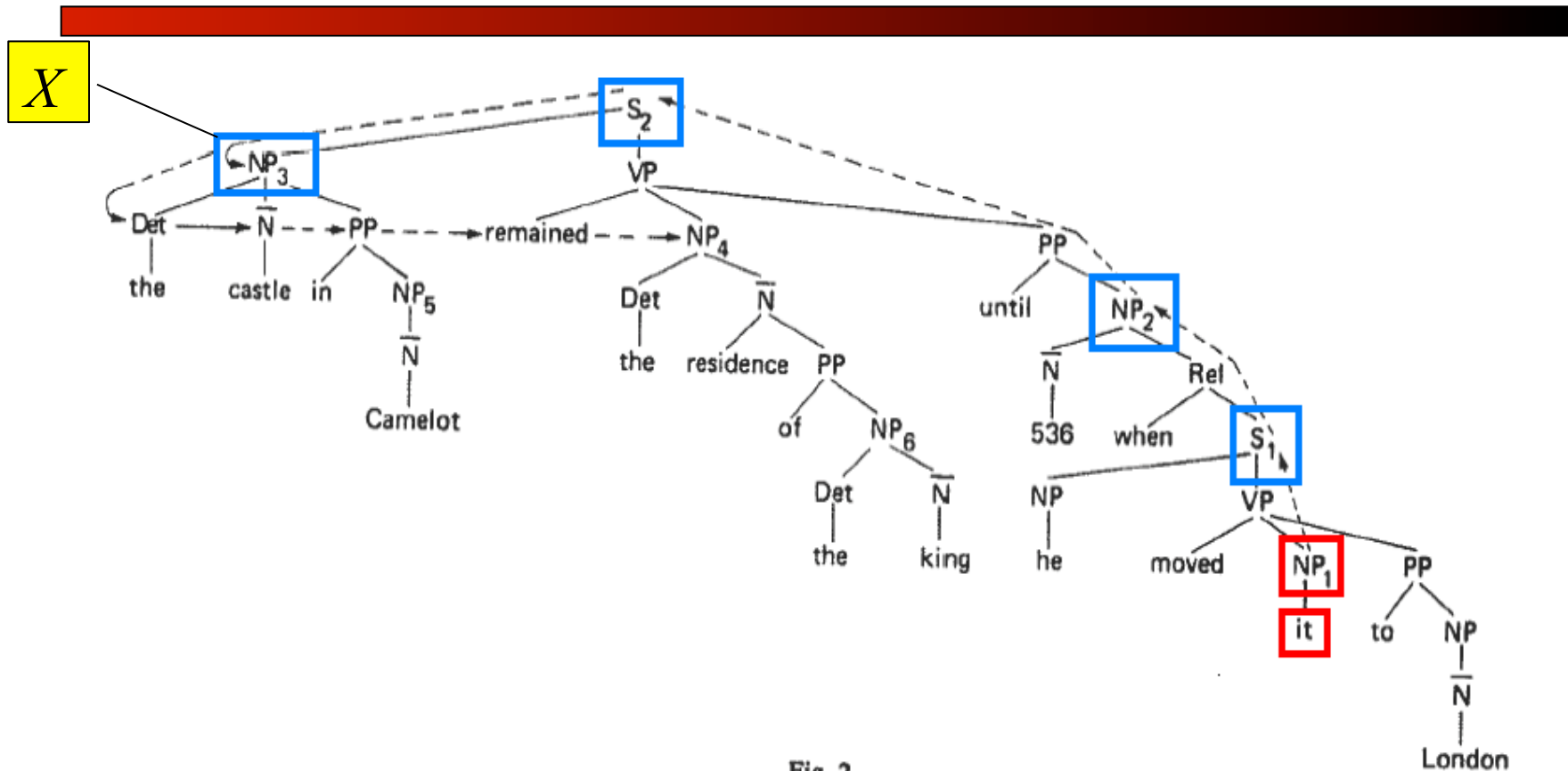


Fig. 2.

b. Search left-to-right below X and to left of p, proposing any NP node which has an NP or S between it and X.

NP₃ proposed → Rejected by rejected by selectional restrictions (can't move large fixed objects.)

Example:

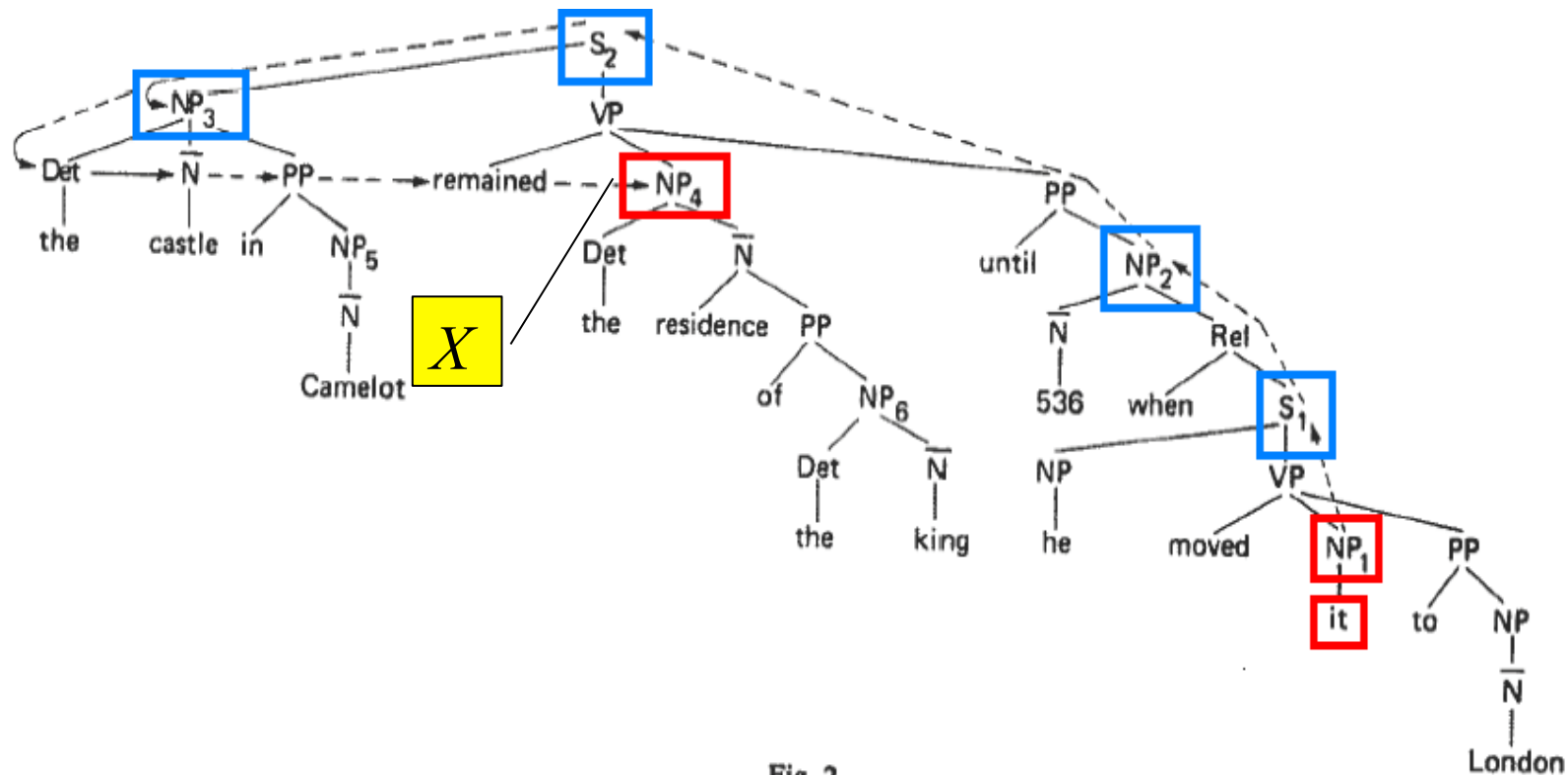


Fig. 2.

b. Search left-to-right below X and to left of p , proposing any NP node which has an NP or S between it and X .

\rightarrow NP₄ proposed. Accepted.

Lappin and Leass

- Lappin and Leass (1994): Given **he/she/it**, assign antecedent.
- Implements only recency and syntactic preferences
- Two steps
 - Discourse model update
 - When a new noun phrase is encountered, add a representation to discourse model with a salience value
 - Modify saliences.
 - Pronoun resolution
 - Choose the most salient antecedent

Saliency Factors and Weights

- Saliency given to NP

| | |
|--|------------|
| 1 Recency | 100 |
| 2 Subject emphasis | 80 |
| 3 Existential emphasis | 70 |
| 4 Accusative (direct object) emphasis | 50 |
| 5 Ind. Obj and oblique emphasis | 40 |
| 6 Non-adverbial emphasis | 50 |
| 7 Head noun emphasis | 80 |

Information leveraged

- *NP saliencies*
 - according to recency of the NP (1)
 - according to syntactical role (2-7)
 - cut in half after each sentence is processed

Syntactic roles

- Salience factors 2-6: Syntactic role preference
 - Subject > existential predicate nominal > object > indirect object > demarcated adverbial PP
- Examples for 2-5
 - An Acura Integra is parked in the lot (subject)
 - There is an Acura Integra parked in the lot (ex. pred nominal)
 - John parked an Acura Integra in the lot (object)
 - John gave his Acura Integra a bath (indirect obj)
 - In his Acura Integra, John showed Susan his new CD player (demarcated adverbial PP)
- Add salience if 6 holds (not part of demarcated adverbial PP):
 - Inside his Acura Integra, John showed Susan his new CD player (here, it is part of demarcated adverbial PP → no salience)
- Add salience if 7 holds (part of head noun):
 - The owner's manual for an Acura Integra is on John's desk

Lappin and Leass Algorithm

- Collect the potential referents (up to 4 sentences back)
- Remove potential referents that do not agree in number or gender with the pronoun
- Remove potential references that do not pass syntactic coreference constraints
- Compute total salience value of referent from all factors, including, if applicable:
 - role parallelism (+35)
 - or cataphora (-175).
- Select referent with highest salience value. In case of tie, select closest.

Example

- John saw a beautiful Acura Integra at the dealership. He showed it to Bob. He bought it.

Sentence 1:

| Referent | 1 recency | 2 Subj | 3 Exist | 4 Obj | 5 Ind- obj | 6 Non- adv | 7 Head N | Total |
|------------|--------------|-----------|------------|----------|---------------|------------------|-------------|-------|
| John | 100 | 80 | | | | 50 | 80 | 310 |
| Integra | 100 | | | 50 | | 50 | 80 | 280 |
| dealership | 100 | | | | | 50 | 80 | 230 |

After sentence 1

- Cut all values in half

| Referent | Phrases | Value |
|------------|-----------------------------|-------|
| John | {John} | 155 |
| Integra | {a beautiful Acura Integra} | 140 |
| dealership | {the dealership} | 115 |

Sentence 2:

He showed it to Bob

- **He** specifies male gender
- So Step 2 reduces set of referents to only **John**.
- Now update discourse model:
 - **He** in current sentence (recency=100), subject position (=80), not adverbial (=50) not embedded (=80), so add 310:

| Referent | Phrases | Value |
|------------|-----------------------------|---------|
| John | {John, he ₁ } | 155+310 |
| Integra | {a beautiful Acura Integra} | 140 |
| dealership | {the dealership} | 115 |

Sentence 2: *He showed it to Bob*

- Need to add “it”, which can be “Integra” or “dealership”.
- Need to add weights:
 - Parallelism: “it” and “Integra” are objects (“dealership” is not), so +35 for “Integra”
 - Integra: 175, dealership: 115
→ pick Integra
- Update discourse model:
 - “it” is object, gets $100+50+50+80=280$:

Sentence 2: *He showed it to Bob*

| Referent | Phrases | Value |
|------------|--|---------|
| John | {John, he ₁ } | 465 |
| Integra | {a beautiful Acura Integra, it ₁ } | 140+280 |
| dealership | {the dealership} | 115 |

Sentence 2: *He showed it to Bob*

- Bob is a new referent
- Update discourse model:
 - Bob is oblique argument, weight is $100+40+50+80=270$

| Referent | Phrases | Value |
|------------|---|-------|
| John | {John, he ₁ } | 465 |
| Integra | {a beautiful Acura Integra, it ₁ } | 420 |
| Bob | {Bob} | 270 |
| dealership | {the dealership} | 115 |

Sentence 3:

He bought it

- Drop weights in half:

| Referent | Phrases | Value |
|------------|---|-------|
| John | {John, he ₁ } | 232.5 |
| Integra | {a beautiful Acura Integra, it ₁ } | 210 |
| Bob | {Bob} | 135 |
| dealership | {the dealership} | 57.5 |

He₂ will be resolved to John, and it₂ to Integra

Evaluation

- Referential Rate (Byron, 2001)
- $RR = C / (T+E)$

C: # pronouns correctly resolved

T: all pronouns in the test set

E: all excluded referential pronouns



REFERENCES

References

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`www.cs.cornell.edu/courses/cs674/2005sp/projects/tejaswini-deoskar.doc`

- **Anaphora Resolution**

`https://people.ucsc.edu/~abrsvn/anaphora%20resolution.ppt`