

# Discourse: Reference

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Deep Processing Techniques for NLP

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# What is a Discourse?

- Discourse is:
  - Extended span of text
  - Spoken or Written
  - One or more participants
  - Language in Use
  - Goals of participants
    - Processes to produce and interpret

# Why Discourse?

- Understanding depends on context
  - Referring expressions: it, that, the screen
  - Word sense: plant
  - Intention: Do you have the time?
- Applications: Discourse in NLP
  - Question-Answering
  - Information Retrieval
  - Summarization
  - Spoken Dialogue
  - Automatic Essay Grading

# Reference Resolution

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

S: A Bug's Life is playing at the Summit theater.

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

S: It's playing at 2pm, 5pm, and 8pm.

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

- Knowledge sources:

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  - Domain knowledge

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

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  - **Discourse knowledge**
  - **World knowledge**

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# Coherence

- *First Union Corp. is continuing to wrestle with severe problems. According to industry insiders at PW, their president, John R. Georgius, is planning to announce his retirement tomorrow.*
- 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: subsidiary, background



# Different Parameters of Discourse

- Number of participants
  - Multiple participants -> Dialogue
- Modality
  - Spoken vs Written
- Goals
  - Transactional (message passing) vs Interactional (relations, attitudes)
  - Cooperative task-oriented rational interaction

# Spoken vs Written Discourse

- Speech
  - Paralinguistic effects
    - Intonation, gaze, gesture
  - Transitory
  - Real-time, on-line
  - Less “structured”
    - Fragments
    - Simple, Active, Declarative
    - Topic-Comment
    - Non-verbal referents
    - Disfluencies
      - Self-repairs
      - False Starts
      - Pauses
- Written text
  - No paralinguistic effects
  - “Permanent”
  - Off-line. Edited, Crafted
  - More “structured”
    - Full sentences
    - Complex sentences
    - Subject-Predicate
    - Complex modification
    - More structural markers
    - No disfluencies

# Spoken vs Written: Representation

- Spoken “text” “same” if:
  - Recorded (Audio/Video Tape)
  - Transcribed faithfully
    - Always some interpretation
  - Text (normalized) transcription
    - Map paralinguistic features
    - e.g. pause = .,+,++
    - Notate accenting, pitch
- Written text “same” if:
  - Same words
  - Same order
  - Same punctuation (headings)
  - Same lineation

# Agenda

- Coherence: Holding discourse together
  - Coherence types and relations
- Reference resolution
  - Referring expressions
  - Information status and structure
  - Features and Preferences for resolution
  - Knowledge-rich, deep analysis approaches
    - Lappin&Leass,
    - Hobbs

# Coherence Relations

- John hid Bill's car keys. He was drunk.
- ?? John hid Bill's car keys. He likes spinach.
- Why odd?
  - No obvious relation between sentences
    - Readers often try to construct relations
- How are first two related?
  - Explanation/cause
- Utterances should have meaningful connection
  - Establish through **coherence relations**

# 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.*
- VS
  - *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?
  - 'about' one entity vs two, focuses on it for coherence

# Reference Resolution

- Match referring expressions to referents
- Syntactic & semantic constraints
- Syntactic & semantic preferences
- Reference resolution algorithms

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# Reference Resolution: Global Focus/ Task

- (From Grosz “Typescripts of Task-oriented Dialogues”)
- E: Assemble the air compressor.
- .
- .
- ... 30 minutes later...
- E: Plug it in / See if it works
- (From Grosz)
- E: Bolt the pump to the base plate
- A: What do I use?
- ....
- A: What is a ratchet wrench?
- E: Show me the table. The ratchet wrench is [...]. Show it to me.
- A: It is bolted. What do I do now?

# Relation Recognition: Intention

- A: You seem very quiet today; is there a problem?
  - B: I have a headache.
  - **Answer**
- A: Would you be interested in going to dinner tonight?
  - B: I have a headache.
  - **Reject**

# Reference

- 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)

Linguistic form that picks out entity in some model

That entity is the “referent”

When introduces entity, “evokes” it

Set up later reference, “antecedent”

2 refexps with same referent “co-refer”

# Reference (terminology)

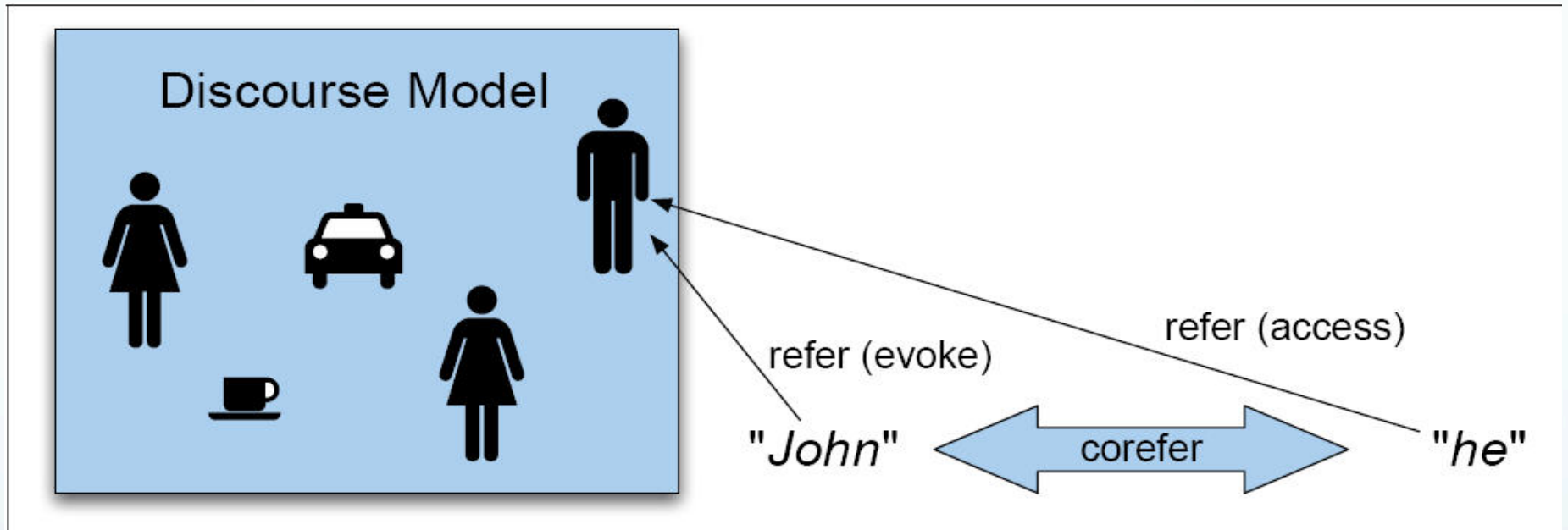
- Queen Elizabeth set about transforming **her** **husband**, **King George VI**, into a viable monarch. Logue, a renowned speech therapist, was summoned to help **the King** overcome **his** speech impediment...

- Anaphor:
  - Abbreviated linguistic form interpreted in context
    - Her, his, the King
  - Refers to previously introduced item (“accesses”)
    - Referring expression is then anaphoric

# Referring Expressions

- Many alternatives:
  - Queen Elizabeth, she, her, the Queen, etc
  - Possible correct forms depend on discourse context
    - E.g. she, her presume prior mention, or presence in world
- Interpretation (and generation) requires:
  - Discourse Model with representations of:
    - Entities referred to in the discourse
    - Relationships of these entities
  - Need way to construct, update model
  - Need way to map refexp to hearer's beliefs

# Reference and Model



# Reference 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...

Coreference resolution:

Find all expressions referring to same entity, 'corefer'

Colors indicate coreferent sets

Pronominal anaphora resolution:

Find antecedent for given pronoun

# Referring Expressions

- Indefinite noun phrases (NPs): e.g. “a cat”
  - Introduces new item to discourse context
- Definite NPs: e.g. “the cat”
  - Refers to item identifiable by hearer in context
    - By verbal, pointing, or environment availability; implicit
- Pronouns: e.g. “he”, “she”, “it”
  - Refers to item, must be “salient”
- Demonstratives: e.g. “this”, “that”
  - Refers to item, sense of distance (literal/figurative)
- Names: e.g. “Miss Woodhouse”, “IBM”
  - New or old entities



# Information Status

- Some expressions (e.g. indef NPs) introduce **new** info
- Others refer to old referents (e.g. pronouns)
- Theories link form of refexp to given/new status

## The givenness hierarchy:

|          |   |                                                                                                   |   |          |   |                          |   |                        |   |                      |
|----------|---|---------------------------------------------------------------------------------------------------|---|----------|---|--------------------------|---|------------------------|---|----------------------|
| in focus | > | activated                                                                                         | > | familiar | > | uniquely<br>identifiable | > | referential            | > | type<br>identifiable |
| {it}     |   | $\left\{ \begin{array}{l} \textit{that} \\ \textit{this} \\ \textit{this N} \end{array} \right\}$ |   | {that N} |   | {the N}                  |   | {indef. <i>this</i> N} |   | {a N}                |

- Accessibility:
  - 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 the door had a dent, and the engine was noisy.*
  - E.g. car -> door, engine
- Generics:
  - *I want to buy a Mac. They are very stylish.*
  - General group evoked by instance.
- Non-referential cases:
  - *It's raining.*

# Syntactic Constraints for Reference Resolution

- Some fairly rigid rules constrain possible referents
- Agreement:
  - Number: Singular/Plural
  - Person: 1st: I,we; 2nd: you; 3rd: he, she, it, they
  - Gender: he vs she vs it

# Syntactic & Semantic Constraints

- Binding constraints:
  - Reflexive (x-self): corefers with subject of clause
  - Pronoun/Def. NP: can't corefer with subject of clause
- “Selectional restrictions”:
  - “animate”: The cows eat grass.
  - “human”: The author wrote the book.
  - More general: drive: John drives a car....

# Syntactic & Semantic Preferences

- Recency: Closer entities are more salient
  - The doctor found an old map in the chest. Jim found an even older map on the shelf. It described an island.
- Grammatical role: Saliency hierarchy of roles
  - e.g. Subj > Object > I. Obj. > Oblique > AdvP
    - Billy Bones went to the bar with Jim Hawkins. He called for a glass of rum. [he = Billy]
    - Jim Hawkins went to the bar with Billy Bones. He called for a glass of rum. [he = Jim]

# Syntactic & Semantic Preferences

- Repeated reference: Pronouns more salient
  - Once focused, likely to continue to be focused
    - 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. [he=Billy]
- Parallelism: Prefer entity in same role
  - Silver went with Jim to the bar. Billy Bones went with him to the inn. [him = Jim]
  - Overrides grammatical role
- Verb roles: “implicit causality”, thematic role match,...
  - John telephoned Bill. He lost the laptop.
  - John criticized Bill. He lost the laptop.

# Reference Resolution Approaches

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

# A Resolution Algorithm (Lappin & Leass)

- Discourse model update:
  - Evoked entities:
    - Equivalence classes: Coreferent referring expressions
  - Salience value update:
    - Weighted sum of salience values:
      - Based on syntactic preferences
- Pronoun resolution:
  - Exclude referents that violate syntactic constraints
  - Select referent with highest salience value



# Salience Factors (Lappin & Leass 1994)

- Weights empirically derived from corpus
  - Recency: 100
  - Subject: 80
  - Existential: 70
  - Object: 50
  - Indirect Object/Oblique: 40
  - Non-adverb PP: 50
  - Head noun: 80
  - Parallelism: 35, Cataphora: -175
- Divide by 50% for each sentence distance

# Example

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

# Example

- John saw a beautiful Acura Integra in the dealership.

| Referent   | Phrases                        | Value |
|------------|--------------------------------|-------|
| John       | {John}                         | 310   |
| Integra    | {a beautiful<br>Acura Integra} | 280   |
| Dealership | {the dealership}               | 230   |

# Example

- He showed it to Bob.

| Referent   | Phrases                        | Value |
|------------|--------------------------------|-------|
| John       | {John, he1}                    | 465   |
| Integra    | {a beautiful<br>Acura Integra} | 140   |
| Dealership | {the dealership}               | 115   |

| Referent   | Phrases                        | Value |
|------------|--------------------------------|-------|
| John       | {John, he1}                    | 465   |
| Integra    | {a beautiful<br>Acura Integra} | 420   |
| Dealership | {the dealership}               | 115   |

# Example

- He showed it to Bob.

| Referent   | Phrases                        | Value |
|------------|--------------------------------|-------|
| John       | {John, he1}                    | 465   |
| Integra    | {a beautiful<br>Acura Integra} | 140   |
| Bob        | {Bob}                          | 270   |
| Dealership | {the dealership}               | 115   |

# Example

| Referent   | Phrases                        | Value |
|------------|--------------------------------|-------|
| John       | {John, he1}                    | 232.5 |
| Integra    | {a beautiful<br>Acura Integra} | 210   |
| Bob        | {Bob}                          | 135   |
| Dealership | {the dealership}               | 57.5  |

- He bought it.

| Referent   | Phrases                        | Value |
|------------|--------------------------------|-------|
| John       | {John, he1}                    | 542.5 |
| Integra    | {a beautiful<br>Acura Integra} | 490   |
| Bob        | {Bob}                          | 135   |
| Dealership | {the dealership}               | 57.5  |

# Hobbs' Resolution Algorithm

- Requires:
  - Syntactic parser
  - Gender and number checker
- Input:
  - Pronoun
  - Parse of current and previous sentences
- Captures:
  - Preferences: Recency, grammatical role
  - Constraints: binding theory, gender, person, number

# Hobbs Algorithm

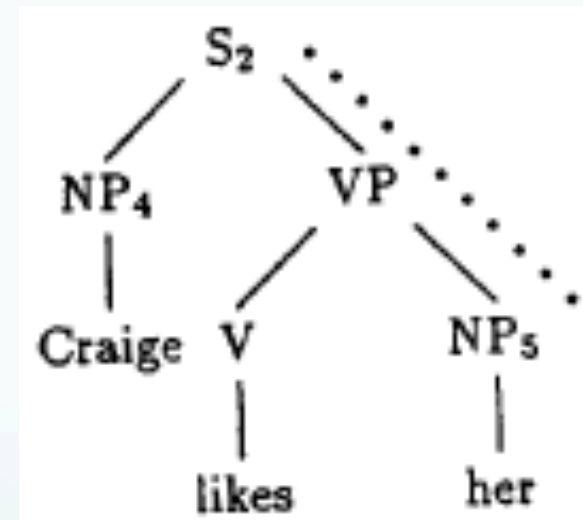
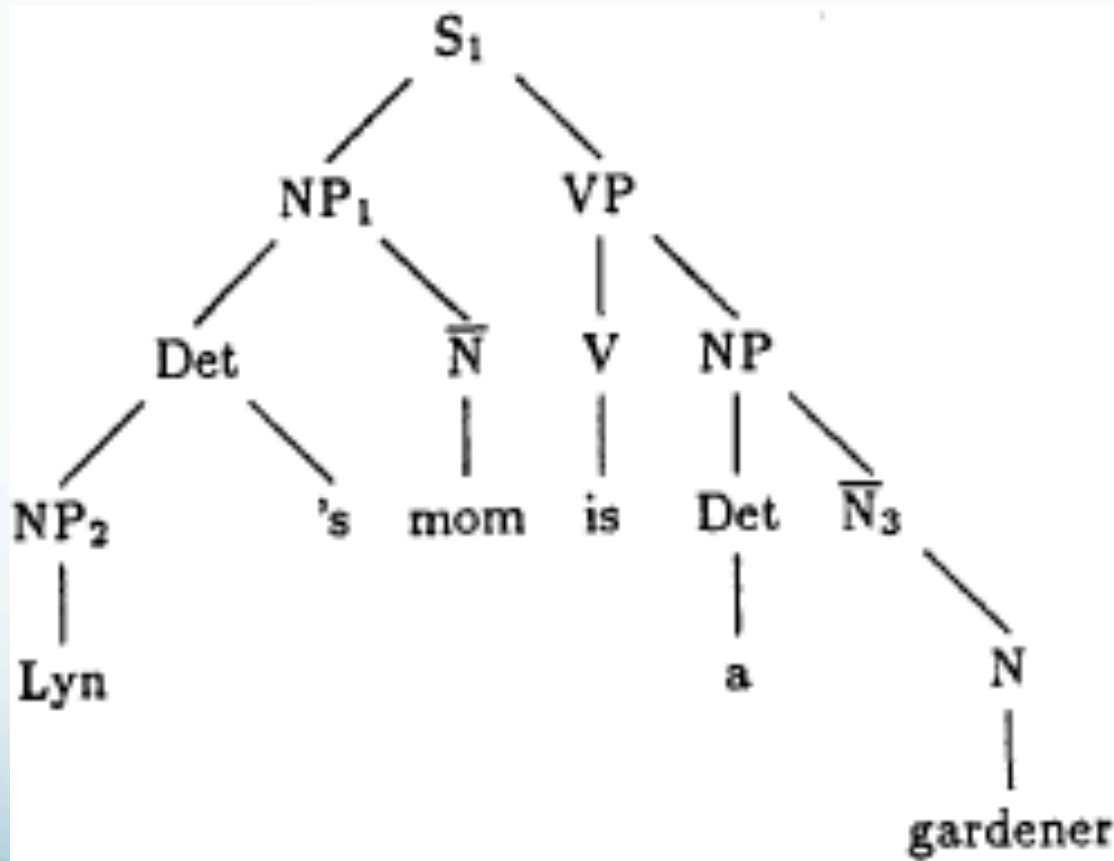
- Intuition:
  - Start with target pronoun
  - Climb parse tree to S root
  - For each NP or S
    - Do breadth-first, left-to-right search of children
      - Restricted to left of target
    - For each NP, check agreement with target
  - Repeat on earlier sentences until matching NP found



# Hobbs Algorithm Detail

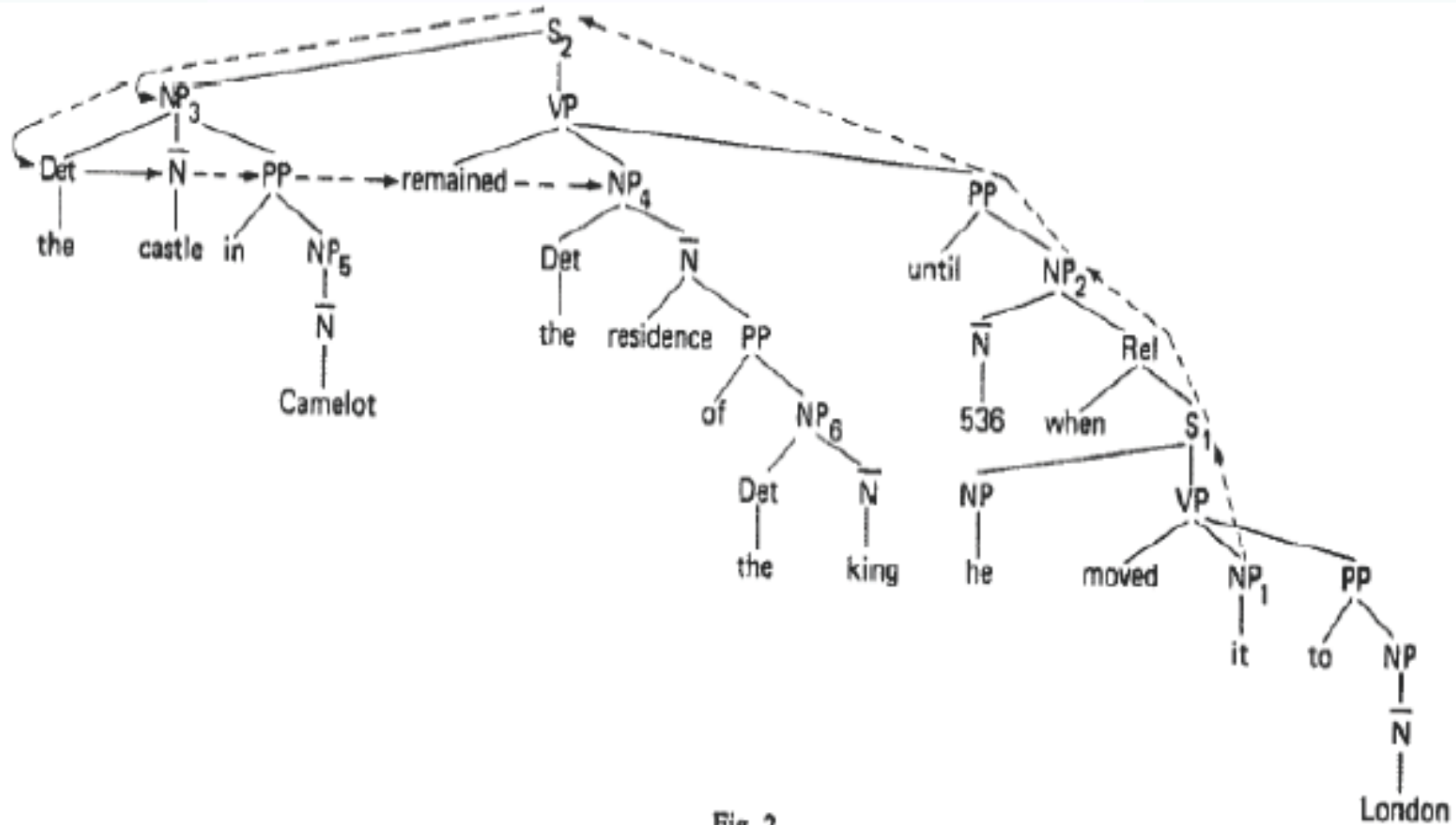
- Begin at NP immediately dominating pronoun
- Climb tree to NP or S: X=node, p = path
- Traverse branches below X, and left of p
  - Breadth-first, Left-to-Right
  - If find NP, propose as antecedent
    - If separated from X by NP or S
- Loop: If X highest S in sentence, try previous sentences.
- If X not highest S, climb to next NP or S: X = node
- If X is NP, and p not through X's nominal, propose X
- Traverse branches below X, left of p: BF,LR
  - Propose any NP
- If X is S, traverse branches of X, right of p: BF, LR
  - Do not traverse NP or S; Propose any NP
  - Go to Loop

# Hobbs Example



Lyn's mom is a gardener. Craig likes her.

# Another Hobbs Example



# 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 are not always accurate
  - Constraints/Preferences:
    - Captures: Binding theory, grammatical role, recency
    - But not: parallelism, repetition, verb semantics, selection

# Reference Resolution: Agreements

- Knowledge-based
  - Deep analysis: full parsing, semantic analysis
  - Enforce syntactic/semantic constraints
  - Preferences:
    - Recency
    - Grammatical Role Parallelism (ex. Hobbs)
    - Role ranking
    - Frequency of mention
- Local reference resolution
- Little/No world knowledge
- Similar levels of effectiveness

# Questions

- 80% on (clean) text. What about...
  - Conversational speech?
    - Ill-formed, disfluent
  - Dialogue?
    - Multiple speakers introduce referents
  - Multimodal communication?
    - How else can entities be evoked?
    - Are all equally salient?

# More Questions

- 80% on (clean) (English) text: What about..
  - Other languages?
    - Salience hierarchies the same
      - Other factors
    - Syntactic constraints?
      - E.g. reflexives in Chinese, Korean,...
    - Zero anaphora?
      - How do you resolve a pronoun if you can't find it?

# Reference Resolution Algorithms

- Many other alternative strategies:
  - Linguistically informed, saliency hierarchy
    - Centering Theory
- Machine learning approaches:
  - Supervised: Maxent
  - Unsupervised: Clustering
- Heuristic, high precision:
  - Cogniac



# Reference Resolution: Extensions

- Cross-document co-reference
  - (Baldwin & Bagga 1998)
  - Break “the document boundary”
  - Question: “John Smith” in A = “John Smith” in B?
  - Approach:
    - Integrate:
      - Within-document co-reference
    - with
      - Vector Space Model similarity

# Cross-document Co-reference

- Run within-document co-reference (CAMP)
  - Produce chains of all terms used to refer to entity
- Extract all sentences with reference to entity
  - Pseudo per-entity summary for each document
- Use Vector Space Model (VSM) distance to compute similarity between summaries

# Cross-document Co-reference

- Experiments:
  - 197 NYT articles referring to “John Smith”
    - 35 different people, 24: 1 article each
  - With CAMP: Precision 92%; Recall 78%
  - Without CAMP: Precision 90%; Recall 76%
  - Pure Named Entity: Precision 23%; Recall 100%

# Conclusions

- Co-reference establishes coherence
- Reference resolution depends on coherence
- Variety of approaches:
  - Syntactic constraints, Recency, Frequency, Role
- Similar effectiveness - different requirements
- Co-reference can enable summarization within and across documents (and languages!)