#### PRONOUNS

# CS 181:

# NATURAL LANGUAGE PROCESSING

Lecture 21: Pronoun Resolution Algorithms

KIM BRUCE POMONA COLLEGE SPRING 2008

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# Reference to an entity already introduced called *anaphora*.

- Pronoun is *licensed* by previous mention of an *antecedent*.
- Pronoun resolution subset of general reference resolution.

### **ANTECEDENT GAME**

#### Constraints on antecedents:

- Number agreement.
  - Iohn his a ball. He threw them far.
  - but:
    - Microsoft released a new version of Windows today. *They* hope it will be more successful than Vista.
- Person agreement
- Ist, 2nd, 3rd person match
- Gender agreement
   he/she/it

#### **ANTECEDENT GAME**

- Binding theory constraints:
  - Sohn bought himself an ice cream.
  - Sohn bought him an ice cream
  - Sohn said that Bill bought him an ice cream
  - Sohn said that Bill bought himself an ice cream
  - \* He said that he bought Bill an ice cream
  - Constraints on meaning of him, himself, he.

## **ANTECEDENT GAME**

- Selectional restrictions:
  - Sohn ate his sandwich in his office.
    - It was made with roast beef.
    - It was quieter than eating in the snack bar.
- Recency:
  - Lee met Mary for lunch. They saw Sue at the restaurant. She gave Lee a hug.
- Grammatical role: Subject > object
  - Jane saw Sally at the market. She went over to say hello.

## **ANTECEDENT GAME**

- Repeated mention:
  - John had a long day. He had not gotten much sleep the night before. He and Fred went to the movies that night. He had a hard time staying awake.
- Parallelism
  - Jane helped Mary with her Physics homework. Ellen helped her with her English.
- Werb Semantics:
  - Jane gave Mary the letter.
    - She was excited to receive it.
    - She had received it yesterday.

# Algorithms for Pronominal Anaphora Resolution

## HOBBS 1978

- Works on parse trees of sentence containing pronoun and of all previous sentences.
- Approximates binding theory, recency, and grammatical role preferences.
- Uses info on gender, person, and number constraints as a final check.

#### HOBBS

- 1. Begin at NP immediately dominating the pronoun
- $2.\ Go up tree to first NP or S node enountered. Call it X and path to it p.$
- 3. Traverse all branches below X to left of path p in a left-to-right, breadth-first fashion. Propose as antecedent any NP node encountered which has an NP or S node between it and X.
- 4. If X is highest S node in sentence, traverse parse trees of previous sentences in order of recency, each in a left-to-right, breadth-first manner, and when an NP is encountered, propose as antecedent. If X not highest, go to 5

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

# EXAMPLES FINAL CHECK Parsers generally return number and person info, but usually not gender. Check hyponyms in WordNet of head noun. Person, living thing indicate animate noun female indicates female gender, ... Cues in titles: Mr., Ms.

## **CENTERING ALGORITHM**

- Claim: There is single entity being "centered" on at any point in the discourse.
- Let  $U_n$ ,  $U_{n+1}$  be 2 consecutive utterances.
- Backward looking center of U<sub>n</sub>, written C<sub>b</sub>(U<sub>n</sub>), represents focus after U<sub>n</sub> interpreted.
- Forward looking centers of U<sub>n</sub>, written C<sub>f</sub>(U<sub>n</sub>), forms ordered list of entities in U<sub>n</sub> that can serve as C<sub>b</sub>(U<sub>n+1</sub>).
- $C_b(U_{n+1})$  is highest ranking elt of  $C_f(U_n)$ mentioned in  $U_{n+1}$ .

#### CENTERS

- Order of entities in C<sub>f</sub>(U<sub>n</sub>):
  - subject > existential predicate nominal > object > indirect object > demarcated adverbial PP
- Let C<sub>p</sub>(U<sub>n+1</sub>) be highest ranked forward looking center



|                                  | $C_b(U_{n+1}) = C_b(U_n)$<br>or undefined $C_b(U_n)$ | $C_b(U_{n+1}) \neq C_b(U_n)$ |
|----------------------------------|--|------------------------------|
| $C_b(U_{n+1}) = C_p(U_{n+1})$    | Continue   | Smooth-Shift                 |
| $C_b(U_{n+1}) \neq C_p(U_{n+1})$ | Retain   | Rough-Shift                  |

- Rule 1: If any elt of Cf(Un) is realized by a pronoun in Un+1 then Cb(Un+1) must be realized as a pronoun also.
- Rule 2: Transition states are ordered. Continue > Retain > Smooth-Shift > Rough-Shift.

# **CENTERING ALGORITHM**

- Generate possible C<sub>b</sub> C<sub>f</sub> combinations for each possible set of reference assignments.
- Filter by constraints (syntactic coreference constraints, selectional, centering rules and constraints).
- Rank by transition orderings
- Assign referents based on Rule 2, if Rule 1 and other constraints not violated.



- He showed it to Bob. {it = MGB?}
  - $C_f(U_2) = {John, MGB, Bob}$
  - $Oldsymbol{ } C_p(U_2) = John$
  - $C_{b}(U_{2})$ : John *highest from C<sub>f</sub>(U<sub>1</sub>)*
  - \* Result: continue  $C_{\rho}(U_2) = C_b(U_2), C_b(U_1)$  undefined
- # He showed it to Bob. {it = dealership?}

  - $C_p(U_2) = John$
  - $Oldsymbol{ } C_b(U_2)$ : John
  - \* Result: continue  $C_{\rho}(U_2) = C_b(U_2), C_b(U_1)$  undefined
- Tied, arb pick MGB since 1st in  $C_f(U_1)$

- We bought it. {it = MGB, he = John?}
   C<sub>f</sub>(U<sub>3</sub>) = {John, MGB}
   C<sub>p</sub>(U<sub>3</sub>) = John
   C<sub>b</sub>(U<sub>3</sub>): John *bighest from C<sub>f</sub>(U<sub>2</sub>)* Result: continue C<sub>p</sub>(U<sub>3</sub>) = C<sub>b</sub>(U<sub>3</sub>) = C<sub>b</sub>(U<sub>2</sub>)
   We bought it. {it = MGB, he = Bob?}
   C<sub>f</sub>(U<sub>3</sub>) = {Bob, MGB}
   C<sub>f</sub>(U<sub>4</sub>) = D<sub>2</sub>
  - $C_p(U_3) = Bob$
  - $C_b(U_3)$ : Bob
  - Result: Smooth-Shift  $C_p(U_3) = C_b(U_3), C_b(U_3) != C_b(U_2)$
- Pick John as continue > Smooth-shift

## CENTERING

- Implicitly incorporates grammatical role, recency, and repeated mention.
- Can get confused.
  - Bob opened a new bike shop last week. John took a look at the road bikes in his shop. He ended up buying one.
  - Incorrectly assigns he to "Bob" because C<sub>b</sub>(U<sub>2</sub>)
     Bob so get continue, while "John" gets smooth-shift.

# MACHINE LEARNING

- Train classifier: Log-linear (we skipped) or Naive Bayes.
- Rely on hand-labeled corpus where each pronoun linked to antecedent.
- Present positive and negative results for training.
- \* Extract features for training.

# FEATURES

- Commonly used for anaphora resolution:
  - strict gender [boolean]
  - \* compatible gender [boolean]
  - strict number [boolean]
  - compatible number [boolean]
  - sentence distance [0,1,2,...] from pronoun
  - Hobbs distance [0,1,2,...] # Hobbs NP skipped
  - Grammatical role [subject, object, PP]
  - Linguistic form [proper, definite, indefinite, pronoun]

# EXAMPLE

- ✤ John saw an MGB at the dealership. (U1)
- He showed it to Bob. (U<sub>3</sub>)
- # He bought it. (U3)

|                   | He $(U_2)$ | it (U <sub>2</sub> ) | Bob $(U_2)$ | John $(U_1)$ |
|-------------------|------------|----------------------|-------------|--------------|
| strict number     | 1          | 1                    | 1           | 1            |
| compatible number | 1          | 1                    | 1           | 1            |
| strict gender     | 1          | 0                    | 1           | 1            |
| compatible gender | 1          | 0                    | 1           | 1            |
| sentence distance | 1          | 1                    | 1           | 2            |
| Hobbs distance    | 2          | 1                    | 0           | 3            |
| grammatical role  | subject    | object               | PP          | subject      |
| linguistic form   | pronoun    | pronoun              | proper      | proper       |

## TRAINING

- Train on vectors.
  - Filter out pleonastic "it" as in "it is raining"
  - Results in weights for each of the features and combinations of features.

# CO-REFERENCE RESOLUTION

#### COREFERENCES

- Extract coreference chains
  - Secretary of State Colin Powell, he, Mr. Powell, Powell.
  - Condoleeza Rice, she, Rice
  - President Bush, Bush
- Can use machine learning classifier as before
  - Process from left to right.
  - For each NP, search backwards for match using classifier

#### **NEED MORE FEATURES**

- Need to recognize that Microsoft is company to make sense of:
  - Microsoft announced record profits today. The company ...
- # Jane .... The 30 year old mother of two ...

# **COMMON FEATURES**

Anaphor edit distance [0,1,2,...]:

$$100 * \frac{m - (s + i + d)}{m - (s + i + d)}$$

where m = size of antecedent.

Antecedent edit distance [0,1,2,...]  $100 * \frac{n - (s + i + d)}{n}$ 

where n = size of anaphor

## **COMMON FEATURES**

- alias [true or false]: names equivalent or acronyms.
- appositive [true or false]: Mary, the new student, ...
- linguistic form [proper, definite, indef, pronoun] type of anaphor

## PSYCHOLOGICAL JUSTIFICATION

- Reading time experiments
  - Clark & Sengal found reading time faster when referent for pronoun in most recent clause, rather than 2 or 3 back (for which speeds same)
  - Crawley found subjects identified antecedent of pronoun if subject more often than if object.
  - Smyth found strong impact of parallel placement.
  - Matthews & Chodorow found slower comprehension when pronoun antecedent occupied early syntactically deep position

# ANY QUESTIONS?