

CS 181: NATURAL LANGUAGE PROCESSING

Lecture 21: Pronoun Resolution Algorithms

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PRONOUNS

- 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.
 - John 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
 - 1st, 2nd, 3rd person match
- Gender agreement
 - he/she/it

ANTECEDENT GAME

- Binding theory constraints:
 - John bought himself an ice cream.
 - John bought him an ice cream
 - John said that Bill bought him an ice cream
 - John 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:
 - John 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.
- Verb 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 encountered. 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-to-right, 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

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

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_n , written $C_b(U_n)$, represents focus after U_n interpreted.
- ☼ Forward looking centers of U_n , written $C_f(U_n)$, forms ordered list of entities in U_n that can serve as $C_b(U_{n+1})$.
- ☼ $C_b(U_{n+1})$ is highest ranking elt of $C_f(U_n)$ mentioned in U_{n+1} .

CENTERS

- ☼ Order of entities in $C_f(U_n)$:
 - ☼ subject > existential predicate nominal > object > indirect object > demarcated adverbial PP
- ☼ Let $C_p(U_{n+1})$ be highest ranked forward looking center

STATE-BASED TRANSITIONS

	$C_b(U_{n+1}) = C_b(U_n)$ 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 $C_f(U_n)$ is realized by a pronoun in U_{n+1} then $C_b(U_{n+1})$ must be realized as a pronoun also.
- ☼ Rule 2: Transition states are ordered.
Continue > Retain > Smooth-Shift > Rough-Shift.

CENTERING ALGORITHM

- ☼ Generate possible $C_b - C_f$ 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.

EXAMPLE REDUX

- ☼ John saw a beautiful MGB at the dealership.
- ☼ He showed it to Bob.
- ☼ He bought it.

EXAMPLE REDUX

- ☼ John saw a beautiful MGB at the dealership.
 - ☼ $C_f(U_1) = \{\text{John, MGB, dealership}\}$ - in order
 - ☼ $C_p(U_1) = \text{John}$
 - ☼ $C_b(U_1)$: undefined (*highest ranked from prev C_f*)

- ☼ He showed it to Bob. {it = MGB?}
 - ☼ $C_f(U_2) = \{\text{John, MGB, Bob}\}$
 - ☼ $C_p(U_2) = \text{John}$
 - ☼ $C_b(U_2): \text{John}$ *highest from $C_f(U_1)$*
 - ☼ Result: continue - $C_p(U_2) = C_b(U_2)$, $C_b(U_1)$ *undefined*
- ☼ He showed it to Bob. {it = dealership?}
 - ☼ $C_f(U_2) = \{\text{John, dealership, Bob}\}$
 - ☼ $C_p(U_2) = \text{John}$
 - ☼ $C_b(U_2): \text{John}$
 - ☼ Result: continue - $C_p(U_2) = C_b(U_2)$, $C_b(U_1)$ *undefined*
- ☼ Tied, arb pick MGB since 1st in $C_f(U_1)$

- ☼ He bought it. {it = MGB, he = John?}
 - ☼ $C_f(U_3) = \{\text{John, MGB}\}$
 - ☼ $C_p(U_3) = \text{John}$
 - ☼ $C_b(U_3): \text{John}$ *highest from $C_f(U_2)$*
 - ☼ Result: continue - $C_p(U_3) = C_b(U_3) = C_b(U_2)$
- ☼ He bought it. {it = MGB, he = Bob?}
 - ☼ $C_f(U_3) = \{\text{Bob, MGB}\}$
 - ☼ $C_p(U_3) = \text{Bob}$
 - ☼ $C_b(U_3): \text{Bob}$
 - ☼ Result: Smooth-Shift - $C_p(U_3) = C_b(U_3)$, $C_b(U_3) \neq C_b(U_2)$
- ☼ 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 $C_b(U_2) = \text{Bob}$ 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. (U_1)
- ☼ He showed it to Bob. (U_2)
- ☼ He bought it. (U_3)

	He (U_2)	it (U_2)	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,...]:
$$100 * \frac{m - (s + i + d)}{m}$$
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?