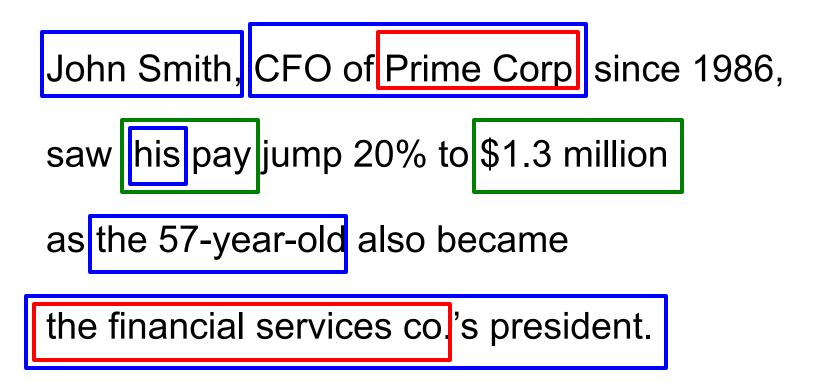
Coreference resolution

Reference Resolution

 Noun phrases refer to entities in the world, many pairs of noun phrases co-refer:



Applications

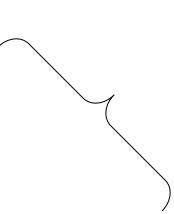
- detecting more named entity mentions
- entity-specific sentiment
- topic modeling / document clustering
- document summarization
- question answering

Kinds of Reference

- Referring expressions
 - John Smith
 - President Smith
 - the president
 - the company's new executive
- Free variables
 - Smith saw his pay increase
- Bound variables
 - The dancer hurt herself.

More common in newswire, generally harder in practice

More interesting grammatical constraints, more linguistic theory, easier in practice



Not all NPs are referring!

- Every dancer twisted her knee.
- (No dancer twisted her knee.)
- There are three NPs in each of these sentences; because the first one is non-referential, the other two aren't either.

Features for Pronominal Anaphora Resolution

Constraints:

- Number agreement
 - Singular pronouns (it/he/she/his/her/him) refer to singular entities and plural pronouns (we/they/us/them) refer to plural entities
- Person agreement
 - He/she/they etc. must refer to a third person entity
- Gender agreement
 - He → John; she → Mary; it → car
 - Jack gave Mary a gift. She was excited.
- Certain syntactic constraints
 - John bought himself a new car. [himself → John]
 - John bought him a new car. [him can not be John]

Features for Pronominal Anaphora Resolution

Preferences:

- Recency: More recently mentioned entities are more likely to be referred to
 - John went to a movie. Jack went as well. He was not busy.
- Grammatical Role: Entities in the subject position is more likely to be referred to than entities in the object position
 - John went to a movie with Jack. He was not busy.
- Parallelism:
 - John went with Jack to a movie. Joe went with him to a bar.

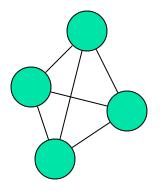
Features for Pronominal Anaphora Resolution

Preferences:

- Verb Semantics: Certain verbs seem to bias whether the subsequent pronouns should be referring to their subjects or objects
 - John telephoned Bill. He lost the laptop.
 - John criticized Bill. He lost the laptop.
- Selectional Restrictions: Restrictions because of semantics
 - John parked his car in the garage after driving it around for hours.
- Encode all these and maybe more as features

Model

- Mention Pair models
 - Treat coreference chains as a collection of pairwise links
 - Make independent pairwise decisions and reconcile them in some way (e.g. clustering or greedy partitioning)



Mention Pair Models

- Most common machine learning approach
- Build classifiers over pairs of NPs
 For each NP, pick a preceding NP or NEW
 Or, for each NP, choose link or no-link
- Clean up non transitivity with clustering or graph partitioning algorithmsE.g.: [Soon et al. 01], [Ng and Cardie 02]

 - Some work has done the classification and clustering jointly [McCallum and Wellner 03]

Pairwise Features

- strict gender [true or false]. True if there is a strict match in gender (e.g. male pronoun Pro_i with male antecedent NP_i).
- 2. **compatible gender [true** or **false]**. True if Pro_i and NP_j are merely compatible (e.g. male pronoun Pro_i with antecedent NP_j of unknown gender).
- strict number [true or false] True if there is a strict match in number (e.g. singular pronoun with singular antecedent)
- 4. **compatible number [true** or **false**]. True if Pro_i and NP_j are merely compatible (e.g. singular pronoun Pro_i with antecedent NP_j of unknown number).
- sentence distance [0, 1, 2, 3,...]. The number of sentences between pronoun and potential antecedent.
- Hobbs distance [0, 1, 2, 3,...]. The number of noun groups that the Hobbs algorithm has to skip, starting backwards from the pronoun Proi, before the potential antecedent NP_j is found.
- grammatical role [subject, object, PP]. Whether the potential antecedent is a syntactic subject, direct object, or is embedded in a PP.
- linguistic form [proper, definite, indefinite, pronoun]. Whether the potential
 antecedent NP_j is a proper name, definite description, indefinite NP, or a pronoun.

Pairwise Features

Category	Features	Remark			
Lexical	exact_strm	1 if two mentions have the same spelling; 0 otherwise			
	left_subsm	1 if one mention is a left substring of the other; 0 otherwise			
	right_subsm	1 if one mention is a right substring of the other; 0 otherwise			
	acronym	1 if one mention is an acronym of the other; 0 otherwise			
	edit_dist	quantized editing distance between two mention strings			
	spell	pair of actual mention strings			
	ncd	number of different capitalized words in two mentions			
Distance	token_dist	how many tokens two mentions are apart (quantized)			
	sent_dist	how many sentences two mentions are apart (quantized)			
	gap_dist	how many mentions in between the two mentions in question (quantized)			
Syntax	POS_pair	POS-pair of two mention heads			
	apposition	1 if two mentions are appositive; 0 otherwise			
Count	count	pair of (quantized) numbers, each counting how many times a mention string is seen			
Pronoun	gender	pair of attributes of {female, male, neutral, unknown }			
	number	pair of attributes of {singular, plural, unknown}			
	possessive	1 if a pronoun is possessive; 0 otherwise			
	reflexive	1 if a pronoun is reflexive; 0 otherwise			

	Devtest		Feb02		Sep02	
Model	ACE-val(%)	ECM-F(%)	ACE-val(%)	ECM-F(%)	ACE-val(%)	ECM-F(%)
MP	89.8	$73.2 (\pm 2.9)$	90.0	$73.1 (\pm 4.0)$	88.0	$73.1 (\pm 6.8)$
EM	89.9	$71.7(\pm 2.4)$	88.2	$70.8 (\pm 3.9)$	87.6	$72.4 (\pm 6.2)$

Coherence relations

- Adjacent sentence pairs come in multiple flavors:
- John hid Bill's car keys.
 - He was drunk. [He=Bill?]
 - He was concerned. [He=John?]