

# Inferring the Antecedent: Resolving Anaphora off the Deep End

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AMPRA Panel on Computational and Experimental Approaches to  
Reference and Anaphoric Inference  
UCLA, October 18, 2014

# Background: Shallow vs. Deep Approaches to Anaphora Resolution

- ▶ Many approaches to resolving anaphora are more 'shallow,' avoiding parsing and semantic analysis
- ▶ Why? Because 'deep' approaches
  - ▶ Require large, sophisticated annotations (grammars, etc.)
  - ▶ Make categorical distinctions that lead to a built-in precision bias
  - ▶ Produce systems that are complicated to maintain
- ▶ But what if we didn't care about the first point because the necessary annotations were available?

# This Talk

- ▶ I'll discuss a deep NLP technique for resolving anaphora in discourse I'm developing, in part jointly with Kathy Dahlgren
- ▶ The approach generalizes a well-known algorithm from the literature
- ▶ Most importantly, it leverages large-scale knowledge bases to get *bridging* inferences

# Outline

Anaphora Resolution: The Problem

Resolving Anaphora with Machines

Bridging with a Knowledge Base of Entailments and Relationships

Conclusion

# What Makes It So Hard

- ▶ The problem is sometimes cast as the task of rewriting an anaphor using a previously-mentioned named entity, as in
  - (1) Who is [the President of the U.S.]<sub>*i*</sub>? How old is he<sub>*i*</sub>? (Both slots rewritten with *Barack Obama*.)

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- ▶ But this obviously doesn't generalize:
  - (2) Find an Italian restaurant<sub>*i*</sub>, and tell me how late it<sub>*i*</sub>'s open. (Where's the named entity?)

## 'Bridging' the Gap between Shallow and Deep

- ▶ One of the apparently more difficult complicating factors, inference, may be a candidate for moving toward the deep end
- (3) Find my reservation<sub>*i*</sub> at Bierhaus<sub>*j*</sub>.
    - a. What time is it<sub>*i*</sub> for?
    - b. What's on the menu<sub>*i*</sub>?

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- ▶ Getting the anaphora in (3a) requires world knowledge—reservations are the kinds of things that have times, and menus aren't, so the reservation should be more salient than the menu
- ▶ Getting (3b), an instance of *bridging*, is even tougher: we have to know that
  1. Bierhaus is an instance of a restaurant,
  2. Restaurants, unlike reservations, are the kinds of things that have menus, and
  3. *the menu* refers to the menu at Bierhaus

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# Discourse Representation Theory

DRT can be thought of as an extension of first-order logic where

- ▶ Indefinites (*a, some, etc.*) introduce *discourse referents*, rather than existential quantification
- ▶ There is a notion of semantic scope, giving rise to a relation of *accessibility* for anaphoric antecedents
- ▶ An entire discourse's meaning can be captured, rather than just a single utterance

# Benefits of DRT

- ▶ It can model everything that single-utterance first-order approaches to semantics can model
- ▶ There are well-known translations from DRT to first-order logic and back\*
- ▶ There is a fairly good algorithm for resolving anaphora over DRT representations (more on that in a minute)
- ▶ Its notation is pretty intuitive, and a fairly large community of researchers exists around it

# Simple DRT Example

- (4) An engineer walked in. Then a scientist walked in, and he ordered a beer.

$x$ $y$ $z$
engineer( $x$ )
walk-in( $x$ )
scientist( $y$ )
walk-in( $y$ )
beer( $z$ )
order( $y, z$ )

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- ▶ The referents  $x$ ,  $y$ , and  $z$  are left around so that more information can be added about them later

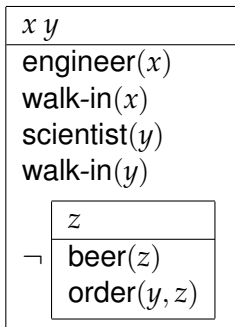
## Slightly More Complicated DRT Example

- (4) An engineer walked in. Then a scientist walked in, but he didn't order a beer.

$x$ $y$	
engineer( $x$ )	
walk-in( $x$ )	
scientist( $y$ )	
walk-in( $y$ )	
$\neg$	$z$
	beer( $z$ )
	order( $y, z$ )

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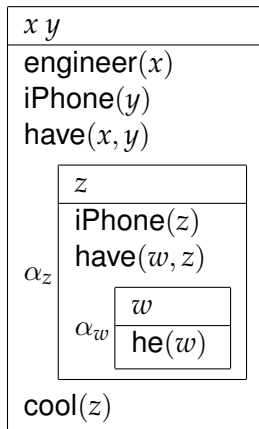
- ▶ In this version, the beer referent  $z$  is 'trapped' in the scope of negation and is not accessible as an antecedent for later anaphora



# Van der Sandt-Bos Algorithm I

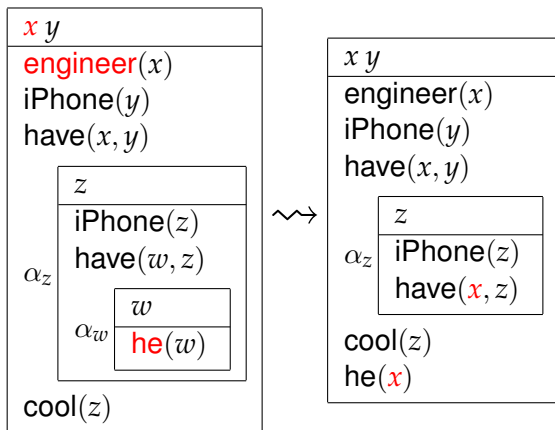
(5) An engineer<sub>*i*</sub> has an iPhone<sub>*j*</sub>. His<sub>*i*</sub> iPhone<sub>*j*</sub> is cool.

- ▶ This example gets the initial representation below:



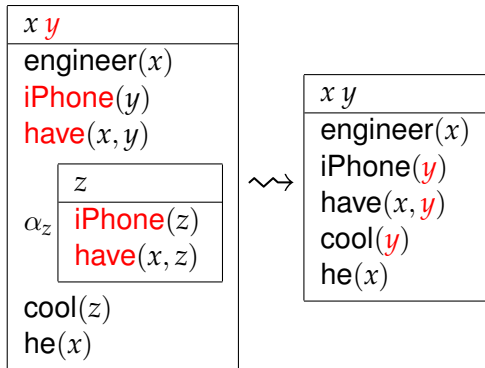
# Van der Sandt-Bos Algorithm II

- Then the most deeply embedded anaphor is resolved to  $x$ , because *engineer* is consistent with *he*



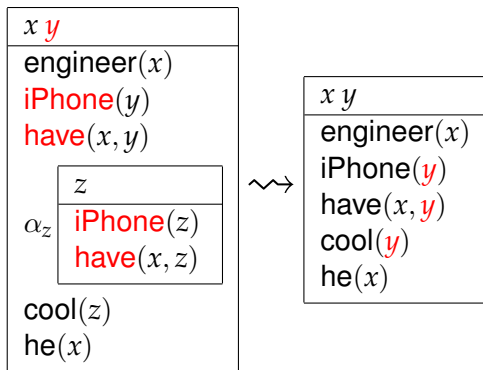
# Van der Sandt-Bos Algorithm III

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- ▶ This gives the desired interpretation: there is some male engineer  $x$  that has a cool iPhone  $y$

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- ▶ But a lot has to go on behind the scenes to decide whether an antecedent is suitable
  - Pronouns** Is the antecedent's descriptive content consistent with the pronoun's?
  - Definites** Does the antecedent's descriptive content entail the definite's?
- ▶ In case an anaphor can't be resolved, we have to decide whether to fail or *accommodate* (for which the van der Sandt-Bos algorithm has a special second phase)

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- ▶ For bridging inferences, we additionally have to ask about aspects related to the antecedent, i.e., if the antecedent is a restaurant, that it has a menu

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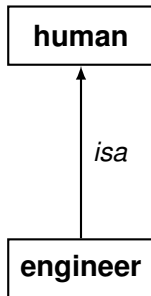


# Making the Algorithm 'Smarter'

- ▶ Using DRT representations by itself only tells us what *can't* be possible antecedents (accessibility)
- ▶ Comparing descriptive contents for consistency and entailment, and doing inferences to find unmentioned antecedents, is beyond its scope
- ▶ Fortunately, there are many resources that are up to the task: ontologies like Cognition, Cyc, Freebase, SUMO, ...

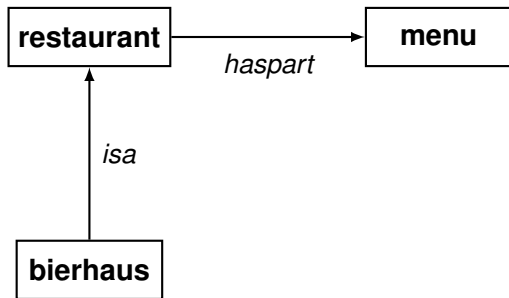
# Simple Entailments

- ▶ In the simplest case, an ontology can help out with basic entailments needed for consistency
- ▶ For example, if the ontology has an *isa* hierarchy (or one can be deduced), we can get that an engineer is a human
- ▶ This is useful because humans can be referred to as *he/him* or *she/her*



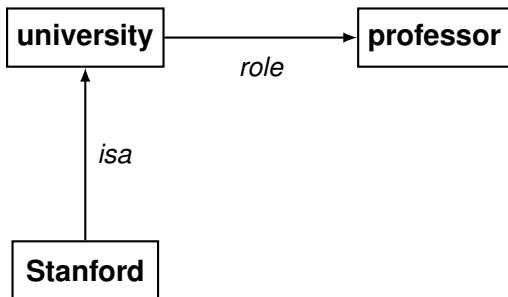
# Broadening the Coverage: Bridging Inferences I

- ▶ An ontology with *haspart* relations between concepts or classes lets us get bridging inferences
- ▶ Only a single graph traversal is needed in addition to what was already needed for simple entailments



## Broadening the Coverage: Bridging Inferences II

- ▶ The capabilities for bridging are limited only by the kinds of relations that the ontology can countenance
- ▶ For instance, if the ontology has a notion of *role* in addition to *isa* subsumption and *haspart*, we can get e.g. *Kim attended Stanford. She liked the professors.*



# Generalizing the Algorithm

Assuming a suitable ontology is available, an extension of the van der Sandt-Bos algorithm for bridging goes like this:

1. *Assume*:  $R$  is an unresolved DRT representation
2. For each anaphoric condition  $a$  in  $R$ , starting with the most deeply embedded:
  - 2.1 Traverse the accessibility hierarchy (ordered by relative salience), and for each candidate antecedent  $c$ , if  $a$  is related to  $c$ :
    - 2.1.1 If the relation is subsumption, resolve  $a$  to  $c$
    - 2.1.2 Otherwise, introduce a new, bridged discourse referent  $b$  in the DRS where  $c$  occurs that bears the relevant descriptive content
3. Fail unless  $R$  is fully resolved

# Generalizing the Algorithm: Bridging Examples I

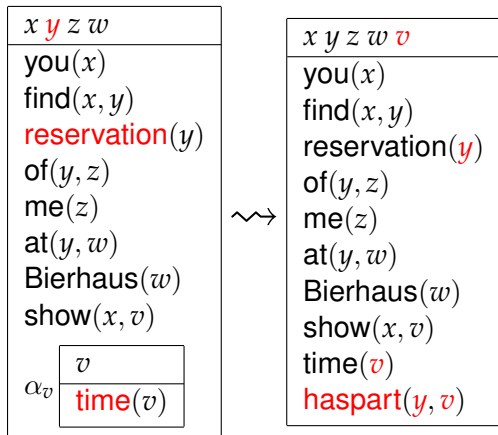
- ▶ This variant algorithm is best understood via some examples. As background, consider our earlier example
  - (3) Find my reservation at Bierhaus.
- ▶ We start by assuming the following representation of (3):

$x$ $y$ $z$ $w$
you( $x$ )
find( $x, y$ )
reservation( $y$ )
of( $y, z$ )
me( $z$ )
at( $y, w$ )
Bierhaus( $w$ )

# Generalizing the Algorithm: Bridging Examples I

- Then, recalling that reservations have times, following up (3) with (3a') Show the time.

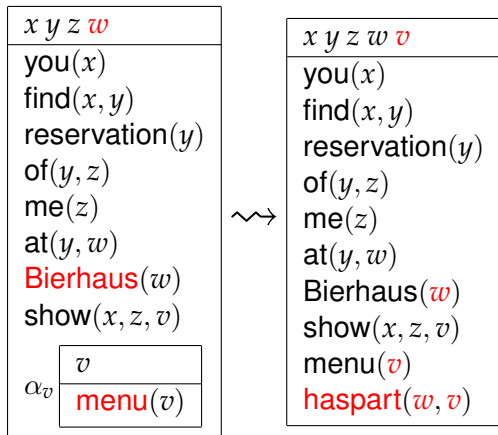
gives:



# Generalizing the Algorithm: Bridging Examples II

- On the other hand, following up (3) with (3b') Show the menu.

gets a different representation, since restaurants have menus:





# Taking Stock

- ▶ This extended algorithm allows us to improve on shallow methods, because shallow methods would be hard pressed to tell the difference between the *time* and *menu* variants
- ▶ But some caveats:
  - ▶ These are only simple cases—comparing descriptive content with an antecedent for suitability, in the general case, requires (the equivalent of) first-order theorem proving
  - ▶ Many ontologies are quite dense, so how do you know when to quit traversing?
  - ▶ When you're constantly asking about covert antecedents, does that slow down an implementation too much?
  - ▶ Already, without bridging, the cost of resolving a DRT structure is not exactly cheap

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  - ▶ Usually, just a single additional hop in the ontology traversal is needed
- ▶ The process of adding bridged antecedents can be seen as a highly constrained replacement for the *accommodation* phase of the van der Sandt-Bos algorithm

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- ▶ But also, for evaluation we can now use:
  - ▶ The Groningen Meaning Bank, a Penn Tree Bank sized corpus of texts with accompanying DRT representations
  - ▶ ISNotes, an extension to the OntoNotes corpus that has labels for anaphora resolution, including ~650 instances of bridging