

# Statistical NLP Spring 2008



## Lecture 21: Compositional Semantics

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Includes examples from Johnson, Jurafsky and Gildea, Luo, Palmer

## Semantic Role Labeling (SRL)

- Characterize clauses as *relations with roles*:

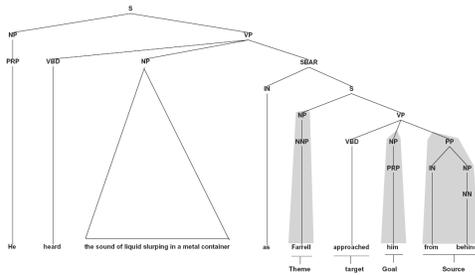
[*Judge* She ] *blames* [*Evaluate* the Government ] [*Reason* for failing to do enough to help ] .

Holman would characterise this as **blaming** [*Evaluate* the poor ] .

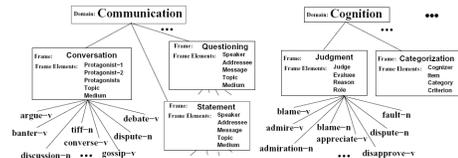
The letter quotes Black as saying that [*Judge* white and Navajo ranchers ] misrepresent their livestock losses and *blame* [*Reason* everything ] [*Evaluate* on coyotes ] .

- Want to more than which NP is the subject (but not much more):
- Relations like *subject* are syntactic, relations like *agent* or *message* are semantic
- Typical pipeline:
  - Parse, then label roles
  - Almost all errors locked in by parser
  - Really, SRL is quite a lot easier than parsing

## SRL Example



## PropBank / FrameNet



- FrameNet: roles shared between verbs
- PropBank: each verb has its own roles
- PropBank more used, because it's layered over the treebank (and so has greater coverage, plus parses)
- Note: some linguistic theories postulate even fewer roles than FrameNet (e.g. 5-20 total: agent, patient, instrument, etc.)

## PropBank Example

**fall.01** sense: move downward  
 roles: Arg1: thing falling  
 Arg2: extent, distance fallen  
 Arg3: start point  
 Arg4: end point

Sales fell to \$251.2 million from \$278.7 million.  
 arg1: Sales  
 rel: fell  
 arg4: to \$251.2 million  
 arg3: from \$278.7 million

## PropBank Example

**rotate.02** sense: shift from one thing to another  
 roles: Arg0: causer of shift  
 Arg1: thing being changed  
 Arg2: old thing  
 Arg3: new thing

Many of Wednesday's winners were losers yesterday as investors quickly took profits and rotated their buying to other issues, traders said. (wsj\_1723)  
 arg0: investors  
 rel: rotated  
 arg1: their buying  
 arg3: to other issues

## PropBank Example

**aim.01** sense: intend, plan  
 roles: Arg0: aimer, planner  
 Arg1: plan, intent

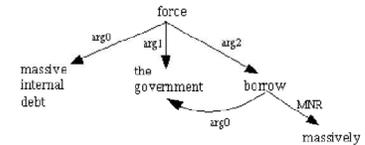
The Central Council of Church Bell Ringers aims \*trace\* to improve relations with vicars (wsj\_0089)  
 arg0: The Central Council of Church Bell Ringers  
 rel: aims  
 arg1: \*trace\* to improve relations with vicars

**aim.02** sense: point (weapon) at  
 roles: Arg0: aimer  
 Arg1: weapon, etc.  
 Arg2: target

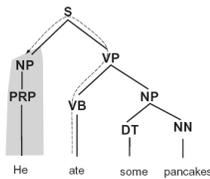
Banks have been aiming packages at the elderly.  
 arg0: Banks  
 rel: aiming  
 arg1: packages  
 arg2: at the elderly

## Shared Arguments

(NP-SBJ (JJ massive) (JJ internal) (NN debt) )  
 (VP (VBZ has)  
 (VP (VBN forced)  
 (S  
 (NP-SBJ-1 (DT the) (NN government) )  
 (VP  
 (VP (TO to)  
 (VP (VB borrow)  
 (ADVP-MNR (RB massively) )...)



## Path Features



Path	Description
VB VP PP	PP argument/adjunct
VB VP S NP	subject
VB VP NP	object
VB VP VP S NP	subject (embedded VP)
VB VP ADVP	adverbial adjunct
NN NP NP PP	prepositional complement of noun

## Results

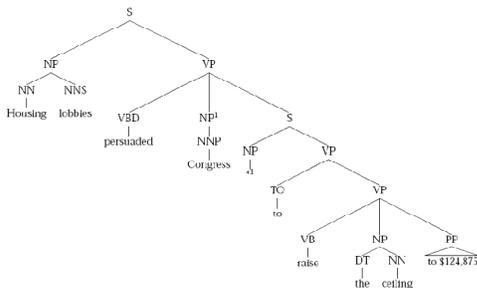
- Features:
  - Path from target to filler
  - Filler's syntactic type, headword, case
  - Target's identity
  - Sentence voice, etc.
  - Lots of other second-order features
- Gold vs parsed source trees
  - SRL is fairly easy on gold trees
  - Harder on automatic parses

CORE		ARGM	
F1	Acc.	F1	Acc.
92.2	89.7	89.9	71.8

CORE		ARGM	
F1	Acc.	F1	Acc.
84.1	66.3	81.4	55.6

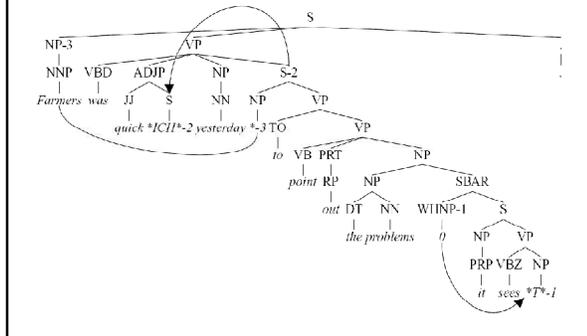
## Interaction with Empty Elements



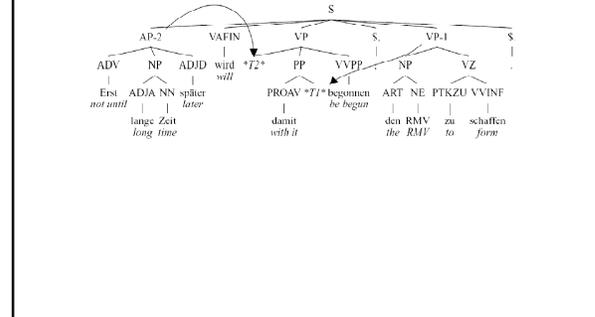
## Empty Elements

- In the PTB, three kinds of empty elements:
  - Null items (usually complementizers)
  - Dislocation (WH-traces, topicalization, relative clause and heavy NP extraposition)
  - Control (raising, passives, control, shared argumentation)
- Need to reconstruct these (and resolve any indexation)

## Example: English



## Example: German

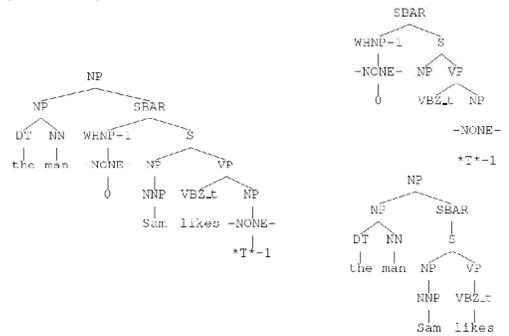


## Types of Empties

Antecedent	POS	Label	Count	Description
NP	NP	*	18,334	NP trace (e.g., <i>Sam was seen</i> *)
NP	NP	*	9,812	NP PRO (e.g., <i>* to sleep is nice</i> )
WHNP	NP	*T*	8,620	WH trace (e.g., <i>the woman <u>wh</u> you saw</i> *T*)
		*U*	7,478	Empty units (e.g., <i>\$25 *U*</i> )
		0	5,635	Empty complementizers (e.g., <i>Sam said 0 Sasha snores</i> )
S	S	*T*	4,062	Moved clauses (e.g., <i>Sam had to go, Sasha explained</i> *T*)
WHADV	ADVP	*T*	2,492	WH-trace (e.g., <i>Sam explained <u>how</u> to leave</i> *T*)
	SBAR		2,033	Empty clauses (e.g., <i>Sam had to go, Sasha explained</i> (SBAR))
	WHNP	0	1,759	Empty relative pronouns (e.g., <i>the woman 0 we saw</i> )
	WHADV	0	575	Empty relative pronouns (e.g., <i>no reason 0 to leave</i> )

## A Pattern-Matching Approach

▪ [Johnson 02]



## Pattern-Matching Details

- Something like transformation-based learning
- Extract patterns
  - Details: transitive verb marking, auxiliaries
  - Details: legal subtrees
- Rank patterns
  - Pruning ranking: by correct / match rate
  - Application priority: by depth
- Pre-order traversal
- Greedy match

## Top Patterns Extracted

Count	Match	Pattern
5816	6223	(S (NP (-NONE- *))) VP)
3605	7895	(SBAR (-NONE- 0) S)
5312	5338	(SBAR WHNP-1 (S (NP (-NONE- *T*-1)) VP))
4434	5217	(NP QP (-NONE- *U*))
1682	1682	(NP \$ CD (-NONE- *U*))
1327	1593	(VP VBNLE (NP (-NONE- *))) PP)
700	700	(ADJP QP (-NONE- *U*))
667	1219	(SBAR (WHNP-1 (-NONE- 0)) (S (NP (-NONE- *T*-1)) VP))
618	635	(G \$ 1, NP (VP VDD (SBAR (NONE 0) (G (NONE *T* 1)))) .)
499	512	(SINV `` S-1, `` (VP VBZ (S (-NONE- *T*-1))) NP .)
361	369	(SINV `` S-1, `` (VP VBD (S (-NONE- *T*-1))) NP .)
352	320	(S NP-1 (VP VBZ (S (NP (-NONE- *-1)) VP)))
346	273	(G NP 1 (VP AUX (VP VBNLE (NP (NONE * 1)) PP)))
322	467	(VP VRD.t (NP (-NONE- *))) PP)
269	275	(G `` \$ 1, `` NP (VP VDD (S (NONE *T* 1))) .)

## Results

Empty node		Section 23			Parser output		
POS	Label	P	R	f	P	R	f
(Overall)		0.93	0.83	0.88	0.85	0.74	0.79
NP	*	0.95	0.87	0.91	0.86	0.79	0.82
NP	*T*	0.93	0.88	0.91	0.85	0.77	0.81
	0	0.94	0.99	0.96	0.86	0.89	0.88
	*U*	0.92	0.98	0.95	0.87	0.96	0.92
S	*T*	0.98	0.83	0.90	0.97	0.81	0.88
ADVP	*T*	0.91	0.52	0.66	0.84	0.42	0.56
SBAR		0.90	0.63	0.74	0.88	0.58	0.70
WHNP	0	0.75	0.79	0.77	0.48	0.46	0.47

## A Machine-Learning Approach

- [Levy and Manning 04]
- Build two classifiers:
  - First one predicts where empties go
  - Second one predicts if/where they are bound
- Use syntactic features similar to SRL (paths, categories, heads, etc)

	Performance on gold trees						Performance on parsed trees						
	ID		Rel		Combo		ID		Rel		Combo		
	P	R	F1	Acc	P	R	F1	P	R	F1	P	R	F1
WSJ(full)	92.0	82.9	87.2	95.0	89.6	80.1	84.6	34.5	47.6	40.0	17.8	24.3	20.5
WSJ(sm)	92.3	79.5	85.5	93.3	90.4	77.2	83.2	38.0	47.3	42.1	19.7	24.3	21.7
NEGRA	73.9	64.6	69.0	85.1	63.3	55.4	59.1	48.3	39.7	43.6	20.9	17.2	18.9

## Semantic Interpretation

- Back to meaning!
  - A very basic approach to computational semantics
  - Truth-theoretic notion of semantics (Tarskian)
  - Assign a "meaning" to each word
  - Word meanings combine according to the parse structure
  - People can and do spend entire courses on this topic
  - We'll spend about an hour!
- What's NLP and what isn't?
  - Designing meaning representations?
  - Computing those representations?
  - Reasoning with them?
- Supplemental reading will be on the web page.

## Meaning

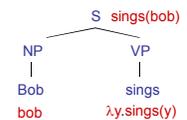
- "Meaning"
  - What is meaning?
    - "The computer in the corner."
    - "Bob likes Alice."
    - "I think I am a gummy bear."
  - Knowing whether a statement is true?
  - Knowing the conditions under which it's true?
  - Being able to react appropriately to it?
    - "Who does Bob like?"
    - "Close the door."
- A distinction:
  - Linguistic (semantic) meaning
    - "The door is open."
  - Speaker (pragmatic) meaning
- Today: assembling the semantic meaning of sentence from its parts

## Entailment and Presupposition

- Some notions worth knowing:
  - Entailment:
    - A entails B if A being true necessarily implies B is true
    - ? "Twitchy is a big mouse" → "Twitchy is a mouse"
    - ? "Twitchy is a big mouse" → "Twitchy is big"
    - ? "Twitchy is a big mouse" → "Twitchy is furry"
  - Presupposition:
    - A presupposes B if A is only well-defined if B is true
    - "The computer in the corner is broken" presupposes that there is a (salient) computer in the corner

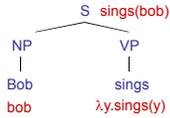
## Truth-Conditional Semantics

- Linguistic expressions:
  - "Bob sings"
- Logical translations:
  - sings(bob)
  - Could be `p_1218(e_397)`
- Denotation:
  - [[bob]] = some specific person (in some context)
  - [[sings(bob)]] = ???
- Types on translations:
  - bob : e (for entity)
  - sings(bob) : t (for truth-value)



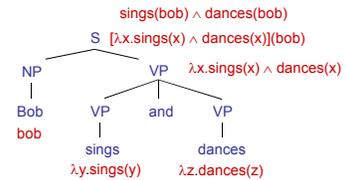
## Truth-Conditional Semantics

- Proper names:
  - Refer directly to some entity in the world
  - Bob : bob      $[[\text{bob}]]^w \rightarrow ???$
- Sentences:
  - Are either true or false (given how the world actually is)
  - Bob sings :  $\text{sings}(\text{bob})$
- So what about verbs (and verb phrases)?
  - sings must combine with bob to produce sings(bob)
  - The  $\lambda$ -calculus is a notation for functions whose arguments are not yet filled.
  - sings :  $\lambda x.\text{sings}(x)$
  - This is *predicate* – a function which takes an entity (type e) and produces a truth value (type t). We can write its type as  $e \rightarrow t$ .
  - Adjectives?



## Compositional Semantics

- So now we have meanings for the words
- How do we know how to combine words?
- Associate a combination rule with each grammar rule:
  - $S : \beta(\alpha) \rightarrow NP : \alpha \quad VP : \beta$  (function application)
  - $VP : \lambda x . \alpha(x) \wedge \beta(x) \rightarrow VP : \alpha$  and  $VP : \beta$  (intersection)
- Example:

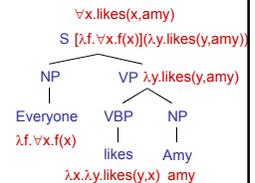


## Denotation

- What do we do with logical translations?
  - Translation language (logical form) has fewer ambiguities
  - Can check truth value against a database
    - Denotation ("evaluation") calculated using the database
  - More usefully: assert truth and modify a database
  - Questions: check whether a statement in a corpus entails the (question, answer) pair:
    - "Bob sings and dances"  $\rightarrow$  "Who sings?" + "Bob"
- Chain together facts and use them for comprehension

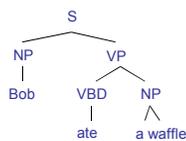
## Other Cases

- Transitive verbs:
  - likes :  $\lambda x.\lambda y.\text{likes}(y,x)$
  - Two-place predicates of type  $e \rightarrow (e \rightarrow t)$ .
  - likes Amy :  $\lambda y.\text{likes}(y,\text{Amy})$  is just like a one-place predicate.
- Quantifiers:
  - What does "Everyone" mean here?
  - Everyone :  $\lambda f.\forall x.f(x)$
  - Mostly works, but some problems
    - Have to change our NP/VP rule.
    - Won't work for "Amy likes everyone."
  - "Everyone likes someone."
  - This gets tricky quickly!



## Indefinites

- First try
  - "Bob ate a waffle" :  $\text{ate}(\text{bob},\text{waffle})$
  - "Amy ate a waffle" :  $\text{ate}(\text{amy},\text{waffle})$
- Can't be right!
  - $\exists x : \text{waffle}(x) \wedge \text{ate}(\text{bob},x)$
  - What does the translation of "a" have to be?
  - What about "the"?
  - What about "every"?



## Grounding

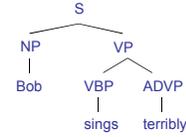
- Grounding
  - So why does the translation  $\text{likes} : \lambda x.\lambda y.\text{likes}(y,x)$  have anything to do with actual liking?
  - It doesn't (unless the denotation model says so)
  - Sometimes that's enough: wire up *bought* to the appropriate entry in a database
- Meaning postulates
  - Insist, e.g.  $\forall x,y.\text{likes}(y,x) \rightarrow \text{knows}(y,x)$
  - This gets into lexical semantics issues
- Statistical version?

## Tense and Events

- In general, you don't get far with verbs as predicates
- Better to have event variables  $e$ 
  - "Alice danced" :  $\text{danced}(\text{alice})$
  - $\exists e : \text{dance}(e) \wedge \text{agent}(e, \text{alice}) \wedge (\text{time}(e) < \text{now})$
- Event variables let you talk about non-trivial tense / aspect structures
  - "Alice had been dancing when Bob sneezed"
  - $\exists e, e' : \text{dance}(e) \wedge \text{agent}(e, \text{alice}) \wedge \text{sneeze}(e') \wedge \text{agent}(e', \text{bob}) \wedge (\text{start}(e) < \text{start}(e') \wedge \text{end}(e) = \text{end}(e')) \wedge (\text{time}(e') < \text{now})$

## Adverbs

- What about adverbs?
  - "Bob sings terribly"
  - $\text{terribly}(\text{sings}(\text{bob}))?$
  - $(\text{terribly}(\text{sings}))(\text{bob})?$
  - $\exists e \text{ present}(e) \wedge \text{type}(e, \text{singing}) \wedge \text{agent}(e, \text{bob}) \wedge \text{manner}(e, \text{terrible})?$
  - It's really not this simple..



## Propositional Attitudes

- "Bob thinks that I am a gummi bear"
  - $\text{thinks}(\text{bob}, \text{gummi}(\text{me}))?$
  - $\text{thinks}(\text{bob}, \text{"I am a gummi bear"})?$
  - $\text{thinks}(\text{bob}, \wedge \text{gummi}(\text{me}))?$
- Usual solution involves intensions ( $\wedge X$ ) which are, roughly, the set of possible worlds (or conditions) in which  $X$  is true
- Hard to deal with computationally
  - Modeling other agents models, etc
  - Can come up in simple dialog scenarios, e.g., if you want to talk about what your bill claims you bought vs. what you actually bought

## Trickier Stuff

- Non-Intersective Adjectives
  - green ball :  $\lambda x. [\text{green}(x) \wedge \text{ball}(x)]$
  - fake diamond :  $\lambda x. [\text{fake}(x) \wedge \text{diamond}(x)]?$   $\rightarrow \lambda x. [\text{fake}(\text{diamond}(x))]$
- Generalized Quantifiers
  - the :  $\lambda f. [\text{unique-member}(f)]$
  - all :  $\lambda f. \lambda g. [\forall x. f(x) \rightarrow g(x)]$
  - most?
  - Could do with more general second order predicates, too (why worse?)
    - $\text{the}(\text{cat}, \text{meows}), \text{all}(\text{cat}, \text{meows})$
- Generics
  - "Cats like naps"
  - "The players scored a goal"
- Pronouns (and bound anaphora)
  - "If you have a dime, put it in the meter."
- ... the list goes on and on!

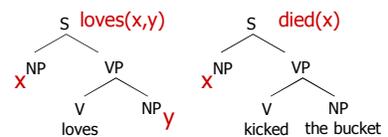
## Multiple Quantifiers

- Quantifier scope
  - Groucho Marx celebrates quantifier order ambiguity: "In this country a woman gives birth every 15 min. Our job is to find that woman and stop her."
- Deciding between readings
  - "Bob bought a pumpkin every Halloween"
  - "Bob put a warning in every window"
  - Multiple ways to work this out
    - Make it syntactic (movement)
    - Make it lexical (type-shifting)

## Implementation, TAG, Idioms

- Add a "sem" feature to each context-free rule
  - $S \rightarrow \text{NP loves NP}$
  - $S[\text{sem}=\text{loves}(x,y)] \rightarrow \text{NP}[\text{sem}=x] \text{ loves } \text{NP}[\text{sem}=y]$
  - Meaning of S depends on meaning of NPs

TAG version:



- Template filling:  $S[\text{sem}=\text{showflights}(x,y)] \rightarrow$   
I want a flight from  $\text{NP}[\text{sem}=x]$  to  $\text{NP}[\text{sem}=y]$

## Modeling Uncertainty

- Gaping hole warning!
- Big difference between statistical disambiguation and statistical reasoning.

*The scout saw the enemy soldiers with night goggles.*

- With probabilistic parsers, can say things like "72% belief that the PP attaches to the NP."
  - That means that *probably* the enemy has night vision goggles.
  - However, you can't throw a logical assertion into a theorem prover with 72% confidence.
  - Not clear humans really extract and process logical statements symbolically anyway.
  - Use this to decide the expected utility of calling reinforcements?
- In short, we need probabilistic reasoning, not just probabilistic disambiguation followed by symbolic reasoning!

## CCG Parsing

### Combinatory Categorial Grammar

- Fully (mono-) lexicalized grammar
- Categories encode argument sequences
- Very closely related to the lambda calculus
- Can have spurious ambiguities (why?)

$John \vdash NP : john'$   
 $shares \vdash NP : shares'$   
 $buys \vdash (S \backslash NP) / NP : \lambda x. \lambda y. buys' x y$   
 $sleeps \vdash S \backslash NP : \lambda x. sleeps' x$   
 $well \vdash (S \backslash NP) \backslash (S \backslash NP) : \lambda f. \lambda x. well''(fx)$

