

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*:

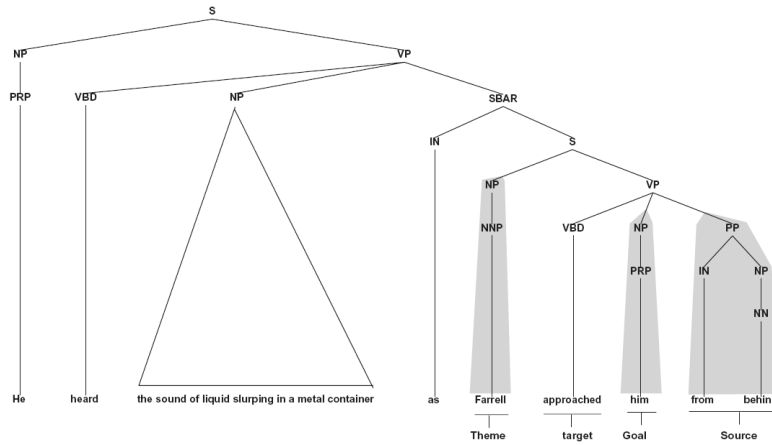
[*Judge* She] **blames** [*Evaluate* the Government] [*Reason* for failing to do enough to help] .

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

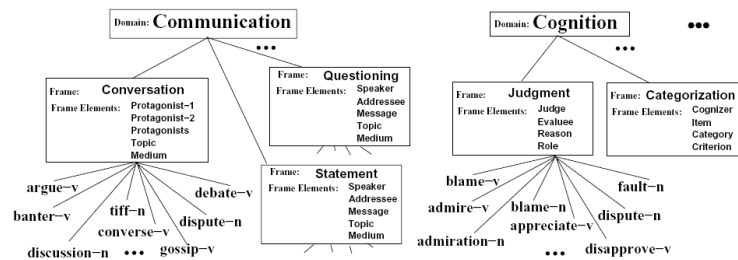
The letter quotes Black as saying that [*Judge* white and Navajo ranchers] misrepresent their livestock losses and **blame** [*Reason* everything] [*Evaluate* on coyotes] .

- 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

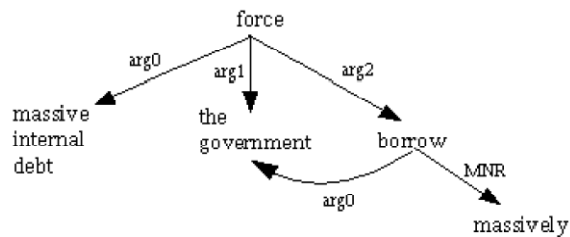
The Central Council of Church Bell Ringers aims *trace* to improve relations with vicars. (wsj_0089)
 arg0: The Central Council of Church Bell Ringers
 rel: aims
 arg1: *trace* to improve relations with vicars

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

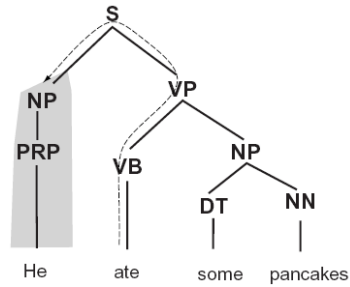
Banks have been aiming packages at the elderly.
 arg0: Banks
 rel: aiming
 arg1: packages
 arg2: at the elderly

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



<i>Path</i>	<i>Description</i>
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

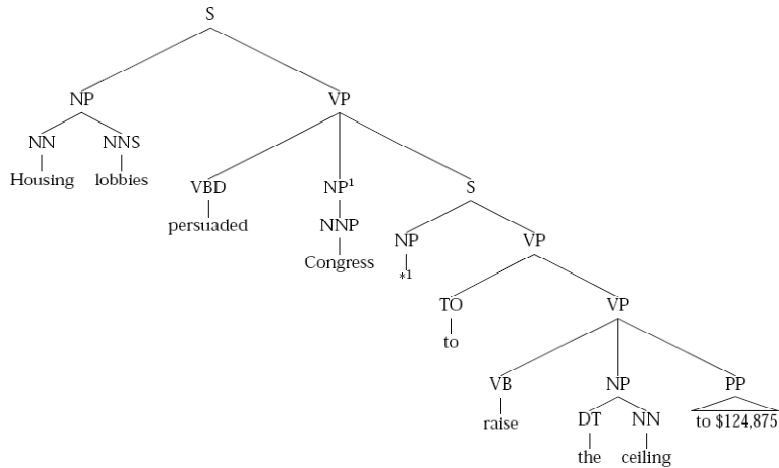
- **Gold vs parsed source trees**

- SRL is fairly easy on gold trees
- Harder on automatic parses

CORE		ARGM	
F1	Acc.	F1	Acc.
92.2	80.7	89.9	71.8

CORE		ARGM	
F1	Acc.	F1	Acc.
84.1	66.5	81.4	55.6

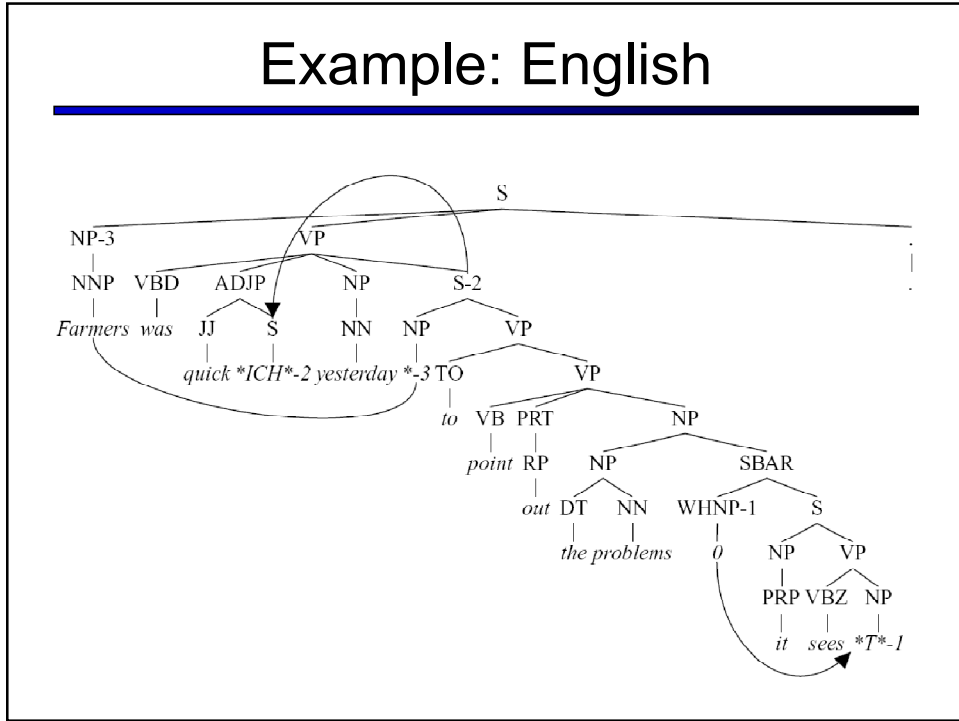
Interaction with Empty Elements



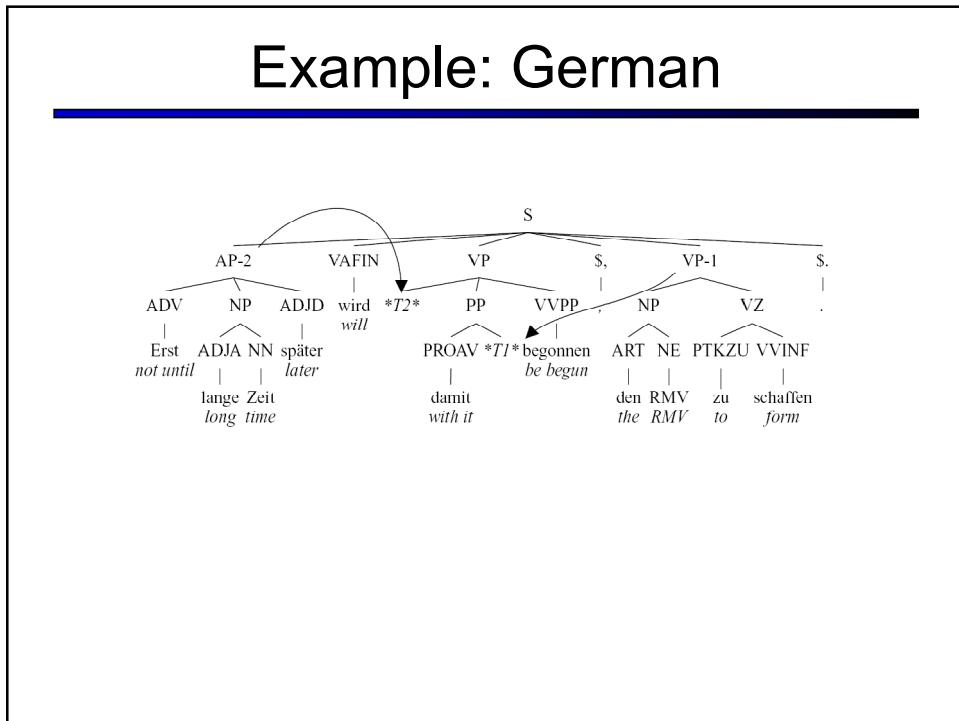
Empty Elements

- 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

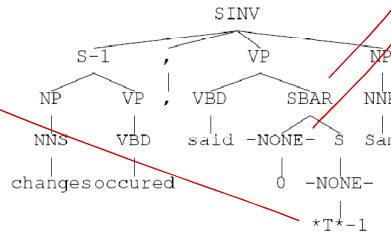
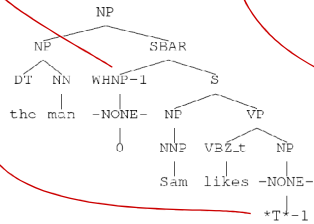


Example: German



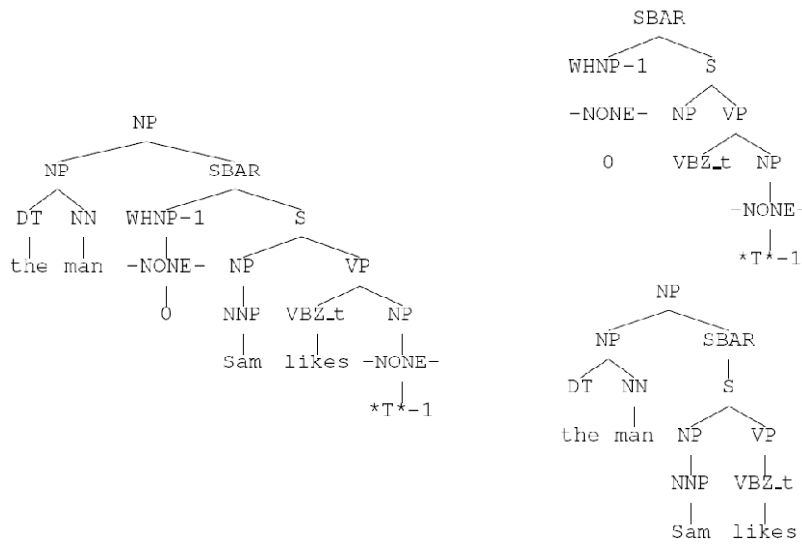
Types of Empties

Antecedent	POS	Label	Count	Description
NP	NP	*	18,334	NP trace (e.g., <i>Sam was seen</i> *)
	NP	*	9,812	NP PRO (e.g., <i>* to sleep is nice</i>)
WHNP	NP	*T*	8,620	WH trace (e.g., <i>the woman who you saw</i> *T*)
		U	7,478	Empty units (e.g., <i>\$ 25</i> *U*)
		0	5,635	Empty complementizers (e.g., <i>Sam said 0 Sasha snores</i>)
S	S	*T*	4,063	Moved clauses (e.g., <i>Sam had to go, Sasha explained</i> *T*)
WHADVP	ADVP	*T*	2,492	WH-trace (e.g., <i>Sam explained how 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>)
	WHADVP	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)
5605	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 VBN_t (NP (-NONE- *)) PP)
700	700	(ADJP QP (-NONE- *U*))
662	1219	(SBAR (WHNP-1 (-NONE- 0)) (S (NP (-NONE- *T*-1)) VP))
618	635	(S S-1 , NP (VP VBD (SBAR (-NONE- 0) (S (-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	(S NP-1 (VP AUX (VP VBN_t (NP (-NONE- *-1)) PP)))
322	467	(VP VBD_t (NP (-NONE- *)) PP)
269	275	(S `` S-1 , '' NP (VP VBD (S (-NONE- *T*-1))) .)

Results

Empty node		Section 23			Parser output		
POS	Label	<i>P</i>	<i>R</i>	<i>f</i>	<i>P</i>	<i>R</i>	<i>f</i>
(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						
	P	ID	Rel	Acc	Combo	F1	P	ID	Rel	Acc	Combo	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

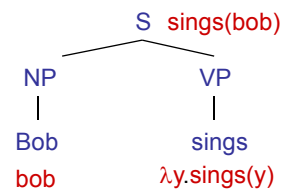
- **“Meaning”**
 - What is meaning?
 - “The computer in the corner.”
 - “Bob likes Alice.”
 - “I think I am a gummi bear.”
 - Knowing whether a statement is true?
 - Knowing the conditions under which it’s true?
 - Being able to react appropriately to it?
 - “Who does Bob like?”
 - “Close the door.”
- **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

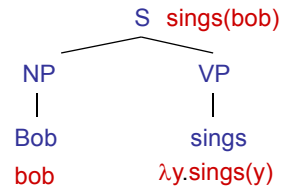
Truth-Conditional Semantics

- 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)



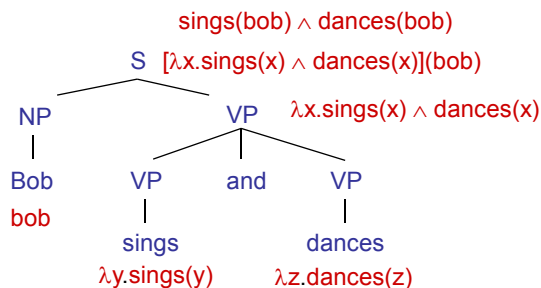
Truth-Conditional Semantics

- 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 : sings(bob)
- So what about verbs (and verb phrases)?
 - sings must combine with bob to produce sings(bob)
 - The λ -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)
 - $VP : \lambda x . \alpha(x) \wedge \beta(x) \rightarrow VP : \alpha \quad \text{and} : \emptyset \quad VP : \beta$ (intersection)
- Example:

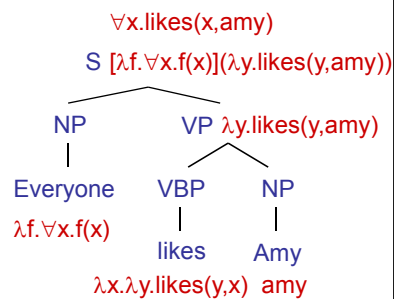


Denotation

- 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” → “Who sings?” + “Bob”
 - Chain together facts and use them for comprehension

Other Cases

- Transitive verbs:
 - likes : $\lambda x.\lambda y.likes(y,x)$
 - Two-place predicates of type $e \rightarrow (e \rightarrow t)$.
 - likes Amy : $\lambda y.likes(y,Amy)$ 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!



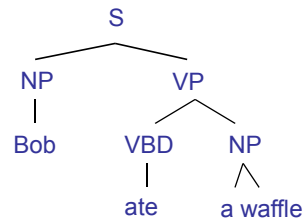
Indefinites

- 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

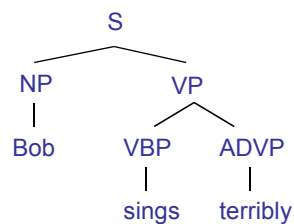
- Statistical version?

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

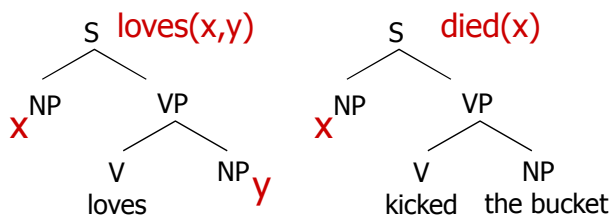
- Non-Intersective Adjectives
 - green ball : $\lambda x. [\text{green}(x) \wedge \text{ball}(x)]$
 - fake diamond : $\lambda x. [\text{fake}(x) \wedge \text{diamond}(x)] ? \longrightarrow \lambda x. [\text{fake}(\text{diamond}(x))]$
- Generalized Quantifiers
 - the : $\lambda f. [\text{unique-member}(f)]$
 - 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
 - $S \rightarrow NP \text{ loves } NP$
 - $S[\text{sem}=\text{loves}(x,y)] \rightarrow NP[\text{sem}=x] \text{ loves } NP[\text{sem}=y]$
 - 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 $NP[\text{sem}=x]$ to $NP[\text{sem}=y]$

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?)

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

