

NATURAL LANGUAGE PROCESSING FOR COMMUNICATION

Sections 23.1 – 23.3
(not covering 23.2.1-2)

Please set your mobile devices to
silent.

Last Time

Introduction to Learning

- Supervised Learning - Induction from observations
- Trading model fit for simplicity
- Algorithms
 - KNN
 - Naïve Bayes
 - Decision Trees / Information Gain

Outline

- Formal Grammar
- Parsing: Syntactic Analysis
- Augmented Grammars
- The larger context
 - Communication as Action
 - Semantic Interpretation
 - Ambiguity and Disambiguation
 - Discourse Understanding

Language

- Formal language: A (possibly infinite) set of strings
- Grammar: A finite set of rules that specifies a language
- Rewrite rules
 - Non-terminal symbols [not observed] (S, NP, etc.)
 - Terminal symbols [observed] (“he”)
 - $S \rightarrow NP VP$
 - $NP \rightarrow \text{Pronoun}$
 - $\text{Pronoun} \rightarrow \text{“he”}$

Convention:
Uppercase are
for non-
terminals,
lowercase for
terminals

Generative Capacity

Noam Chomsky described four grammatical formalisms:

- Recursively enumerable grammars
 - Unrestricted rules: both sides of the rewrite rules can have any number of terminal and non-terminal symbols; full Turing machines
 $ABd \rightarrow CaE$
- Context-sensitive grammars
 - The RHS must contain at least as many symbols as the LHS
 $ASB \rightarrow AXB$
- Context-free grammars (CFG)
 - LHS is a single non-terminal symbol
 $S \rightarrow XYa$
- Regular grammars
 - LHS is single non-terminal; RHS a terminal plus optional non-terminal
 $X \rightarrow a$ $X \rightarrow aY$

Formal Grammar

- The lexicon for ε_0 :

Noun → stench | breeze | glitter | wumpus | pit | pits | gold | ...

Verb → is | see | smell | shoot | stinks | go | grab | turn | ...

Adjective → right | left | east | dead | back | smelly | ...

Adverb → here | there | nearby | ahead | right | left | east | ...

Pronoun → me | you | I | it | ...

Name → John | Mary | Boston | Aristotle | ...

Article → the | a | an | ...

Preposition → to | in | on | near | ...

Conjunction → and | or | but | ...

Digit → 0 | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9

Formal Grammar

- The grammar for ϵ_0 :

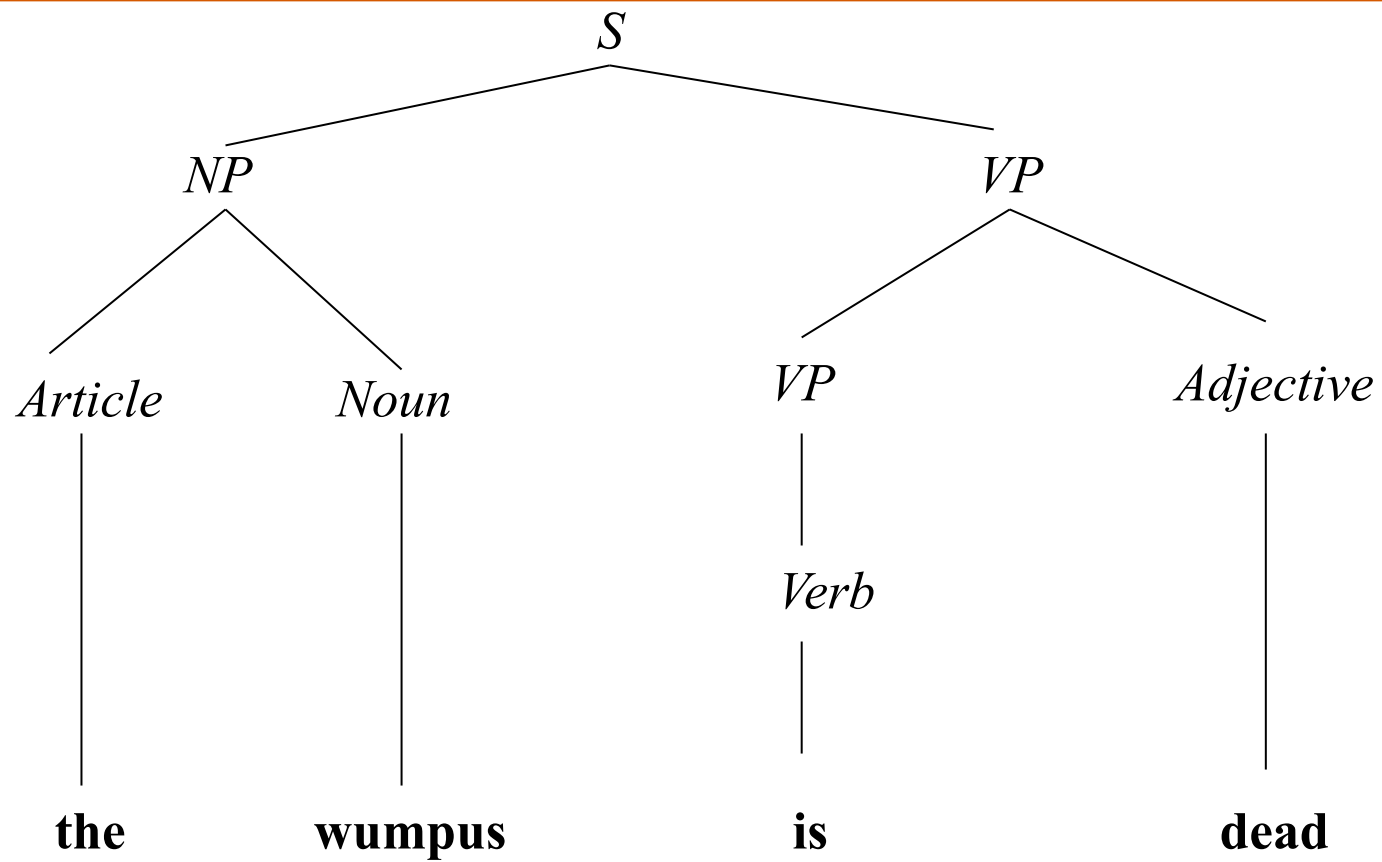
S	→	NP VP	I + feel a breeze
		S Conjunction S	I feel a breeze + and + I smell a wumpus
NP	→	Pronoun	I
		Name	John
		Noun	pits
		Article Noun	the + wumpus
		Digit Digit	3 4
		NP PP	the wumpus + to the east
		NP RelClause	the wumpus + that is smelly

← We'll deal with probabilistic rules later

Formal Grammar

- Parts of speech
 - Open class: noun, verb, adjective, adverb
 - Closed class: pronoun, article, preposition, conjunction, ...
- Shortcomings of our grammar
 - Overgenerate: “Me go Boston”
 - Undergenerate: “I think the wumpus is smelly”

Parse Tree



Syntactic Analysis (Parsing)

- Parsing: The process of finding a parse tree for a given input string
- Top-down parsing
 - Start with the S symbol and search for a tree that has the words as its leaves
- Bottom-up parsing
 - Start with the words and search for a tree with root S

Trace of Bottom-up Parsing

List of nodes	Subsequence	Rule
the wumpus is dead	the	Article → the
<i>Article</i> wumpus is dead	wumpus	Noun → wumpus
<i>Article Noun</i> is dead	<i>Article Noun</i>	NP → Article Noun
<i>NP</i> is dead	is	Verb → is
<i>NP Verb</i> dead	dead	Adjective → dead
<i>NP Verb Adjective</i>	<i>Verb</i>	VP → Verb
<i>NP VP Adjective</i>	<i>VP Adjective</i>	VP → VP Adjective
<i>NP VP</i>	<i>NP VP</i>	S → NP VP
S		

- Left to Right processing per token

Intrasentence ambiguity

“Have the students of CS 3243

Artificial Intelligence ...

Probabilistic Grammars

- Probabilistic lexicon

Noun → stench [.05] | breeze [.1] | wumpus [.15] | pits [.05] | ...

Verb → is [.1] | feel [.1] | stinks [.05] | ...

Digit → 0 [.1] | 1 [.1] | 2 [.1] | ...

- Probabilistic Grammar

VP→	Verb	[.4] stinks
	VP NP	[.35] feel + a breeze
	VP Adjective	[.05] is + smelly
	VP PP	[.1] turn + to the east
	VP Adverb	[.1] go + ahead

Where each category (e.g., VP) rule has probabilities that sum to one.

CYK Chart Parsing

- Uses dynamic programming to memoize intermediate results, saving to a **chart**.
- Bottom Up Iterative Processing
- Converts context free grammar into a special form: Chomsky Normal Form
 - $X \rightarrow "a"$
 - $X \rightarrow YZ$
- Uses space $O(n^2m) \cong O(n^3)$, despite $O(2^n)$ possible parses.
- Suitable for probabilistic CFGs (PCFGs).

CYK Parsing

function CYK-PARSE(*words*, *grammar*) **returns** P , a table of probabilities

$N \leftarrow \text{LENGTH}(\textit{words})$

$M \leftarrow$ the number of nonterminal symbols in *grammar*

$P \leftarrow$ an array of size $[M, N, N]$, initially all 0

/ Insert lexical rules for each word */*

for $i = 1$ **to** N **do**

for each rule of form $(X \rightarrow \textit{words}_i [p])$ **do**

$P[X, i, 1] \leftarrow p$

/ Combine first and second parts of right-hand sides of rules, from short to long */*

for $\textit{length} = 2$ **to** N **do**

for $\textit{start} = 1$ **to** $N - \textit{length} + 1$ **do**

for $\textit{len1} = 1$ **to** $N - 1$ **do**

$\textit{len2} \leftarrow \textit{length} - \textit{len1}$

for each rule of the form $(X \rightarrow Y Z [p])$ **do**

$P[X, \textit{start}, \textit{length}] \leftarrow \text{MAX}(P[X, \textit{start}, \textit{length}]$,

$P[Y, \textit{start}, \textit{len1}] \times P[Z, \textit{start} + \textit{len1}, \textit{len2}] \times p)$

return P

An Ambiguous Example

N,A	N,V	N,V	P	A	N
British	left	waffles	on	Falkland	Islands

- $S \rightarrow NP VP$
- $S \rightarrow NP VP PP$
- $PP \rightarrow P NP$
- $NP \rightarrow A NP$
- $NP \rightarrow NP PP$
- $VP \rightarrow V PP$
- $VP \rightarrow V NP$
- $VP \rightarrow V$
- $NP \rightarrow N$

See whether you understand what each of the "S"s stand for

Dealing with Probabilities

S[3.6864E-4], S[4.608E-4], S[5.37E-3]					
	S[.00576], VP[.00384]				
		?			
S[.0144]			PP [.16]		
S[.048], NP[.16]	?			NP[.16]	
N,A	N,V	N,V	P	A	N
British	left	waffles	on	Falkland	Islands

- S → NP VP [.3]
- S → NP VP PP [.7]
- PP → P NP [1]
- NP → A NP [.4]
- NP → NP PP [.2]
- NP → N [.4]
- VP → V PP [.4]
- VP → V NP [.3]
- VP → V [.4]

We ignored the probabilistic lexicon for this example which should also be used.

Subjective & Objective Cases

- Overgeneration:
 - $S \rightarrow NP VP \rightarrow NP VP NP \rightarrow NP Verb NP$
 - Pronoun Verb NP \rightarrow Pronoun Verb Pronoun

She loves him

*her loves he

She ran towards him

*She ran towards he

Handling Subjective & Objective Cases

S → NPs VP | ...
NP_s → Pronoun_s | Name | Noun | ...
NP_o → Pronoun_o | Name | Noun | ...
VP → VP NP_o | ...
PP → Preposition NP_o
Pronoun_s → **I | you | he | she | it** | ...
Pronoun_o → **me | you | him | her | it** | ...

- Disadvantage: Grammar size grows exponentially

Augmented Grammars

- Handling case, agreement, etc
- Augment grammar rules to allow parameters on nonterminal categories
 - NP(Subjective)
 - NP(Objective)
 - NP(case)

Definite Clause Grammar (DCG)

- The grammar for ϵ_1 :

S	→	NP(Subjective) VP ...
NP(case)	→	Pronoun(case) Name Noun ...
VP	→	VP NP(Objective) ...
PP	→	Preposition NP(Objective)
Pronoun(Subjective)	→	I you he she it ...
Pronoun(Objective)	→	me you him her it ...

Definite Clauses,
where did we
hear that before?

Definite Clause Grammar (DCG)

- Each grammar rule is a definite clause in logic:
 - $S \rightarrow NP VP$
 - $NP(s1) \wedge VP(s2) \Rightarrow S(s1 + s2)$
 - $NP(case) \rightarrow Pronoun(case)$
 - $Pronoun(case, s1) \Rightarrow NP(case, s1)$
- DCG enables parsing as logical inference:
 - Top-down parsing is backward chaining
 - Bottom-up parsing is forward chaining

Prolog as a suitable language for deterministic NLP

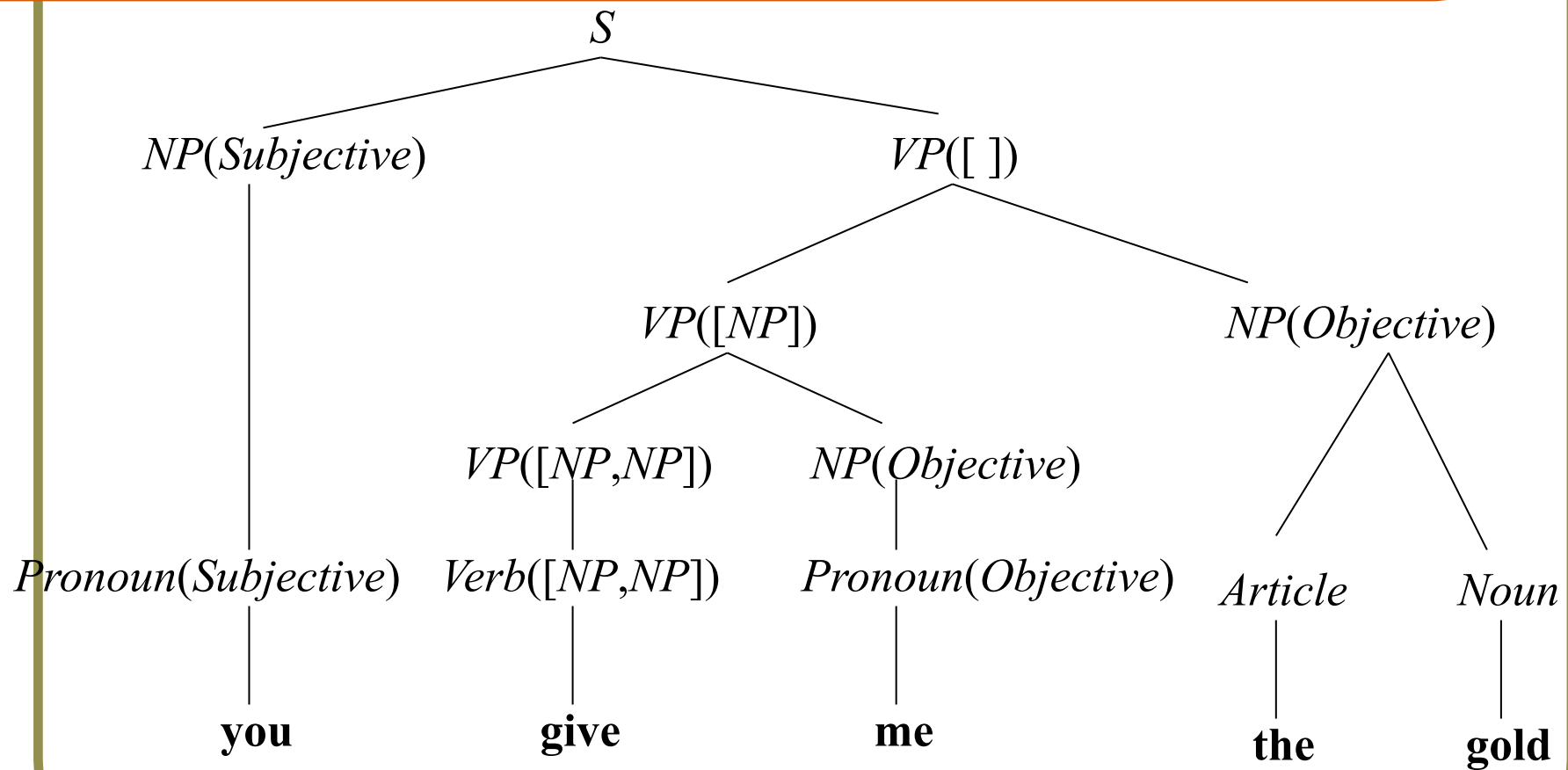
Verb Subcategorization

Verb	Subcats	Example Verb Phrase
give	[<i>NP,PP</i>]	give the gold to me
	[<i>NP,NP</i>]	give me the gold
smell	[<i>NP</i>]	smell a wumpus
	[<i>Adjective</i>]	smell awful
	[<i>PP</i>]	smell like a wumpus
is	[<i>Adjective</i>]	is smelly
	[<i>PP</i>]	is in 2 2
	[<i>NP</i>]	is a pit
died	[]	died
believe	[<i>S</i>]	believe the wumpus is dead

Verb Subcategorization

S	→	NP(Subjective) VP([])
VP(subcat)	→	Verb(subcat)
		VP(subcat + [NP]) NP(Objective)
		VP(subcat + [Adjective]) Adjective
		VP(subcat + [PP]) PP
VP(subcat)	→	VP(subcat) PP
		VP(subcat) Adverb
Verb([NP,NP])	→	give hand ...

Parsing Using Verb Subcategorization



The larger context – communication

- Communication
 - **Intentional** exchange of information brought about by the production and perception of signs drawn from a shared system of conventional signs
- Humans use **language** to communicate most of what is known about the world
- The Turing test is based on language

Communication as Action

- **Speech act**
 - Language production viewed as an action
 - Speaker, hearer, utterance
- **Examples:**
 - Query: “Have you smelled the wumpus anywhere?”
 - Inform: “There’s a breeze here in 3 4.”
 - Request: “Please help me carry the gold.” “I could use some help carrying this.”
 - Acknowledge: “OK”
 - Promise: “I’ll shoot the wumpus.”

Component Steps of Communication

SPEAKER:

- Intention

Know(H, \neg Alive(Wumpus, S_3))

- Generation

“The wumpus is dead”

- Synthesis

[thaxwahmpaxsihzdehd]

Component Steps of Communication

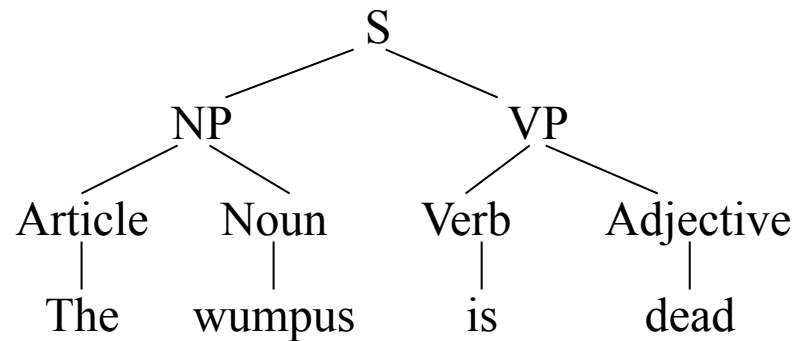
HEARER:

- Perception:

“The wumpus is dead”

- Analysis

(Parsing):



(Semantic Interpretation): \neg Alive(Wumpus, Now)

Tired(Wumpus, Now)

(Pragmatic Interpretation): \neg Alive(Wumpus₁, S₃)

Tired(Wumpus₁, S₃)

Component Steps of Communication

HEARER:

- Disambiguation:

$\neg \text{Alive}(\text{Wumpus}_1, S_3)$

- Incorporation:

$\text{TELL}(\text{KB}, \neg \text{Alive}(\text{Wumpus}_1, S_3))$

Semantic Interpretation

- Semantics: meaning of utterances
- First-order logic as the representation language
- Compositional semantics: meaning of a phrase is composed of meaning of the constituent parts of the phrase

Semantic Interpretation

Exp(x) \rightarrow Exp(x₁) Operator(op) Exp(x₂)
 $\{ x = \text{Apply}(\text{op}, x_1, x_2) \}$

Exp(x) \rightarrow (Exp(x))

Exp(x) \rightarrow Number(x)

Number(x) \rightarrow Digit(x)

Number(x) \rightarrow Number(x₁) Digit(x₂) $\{ x = 10 \times x_1 + x_2 \}$

Digit(x) \rightarrow x $\{ 0 \leq x \leq 9 \}$

Operator(x) \rightarrow x $\{ x \in \{ +, -, \times, \div \} \}$

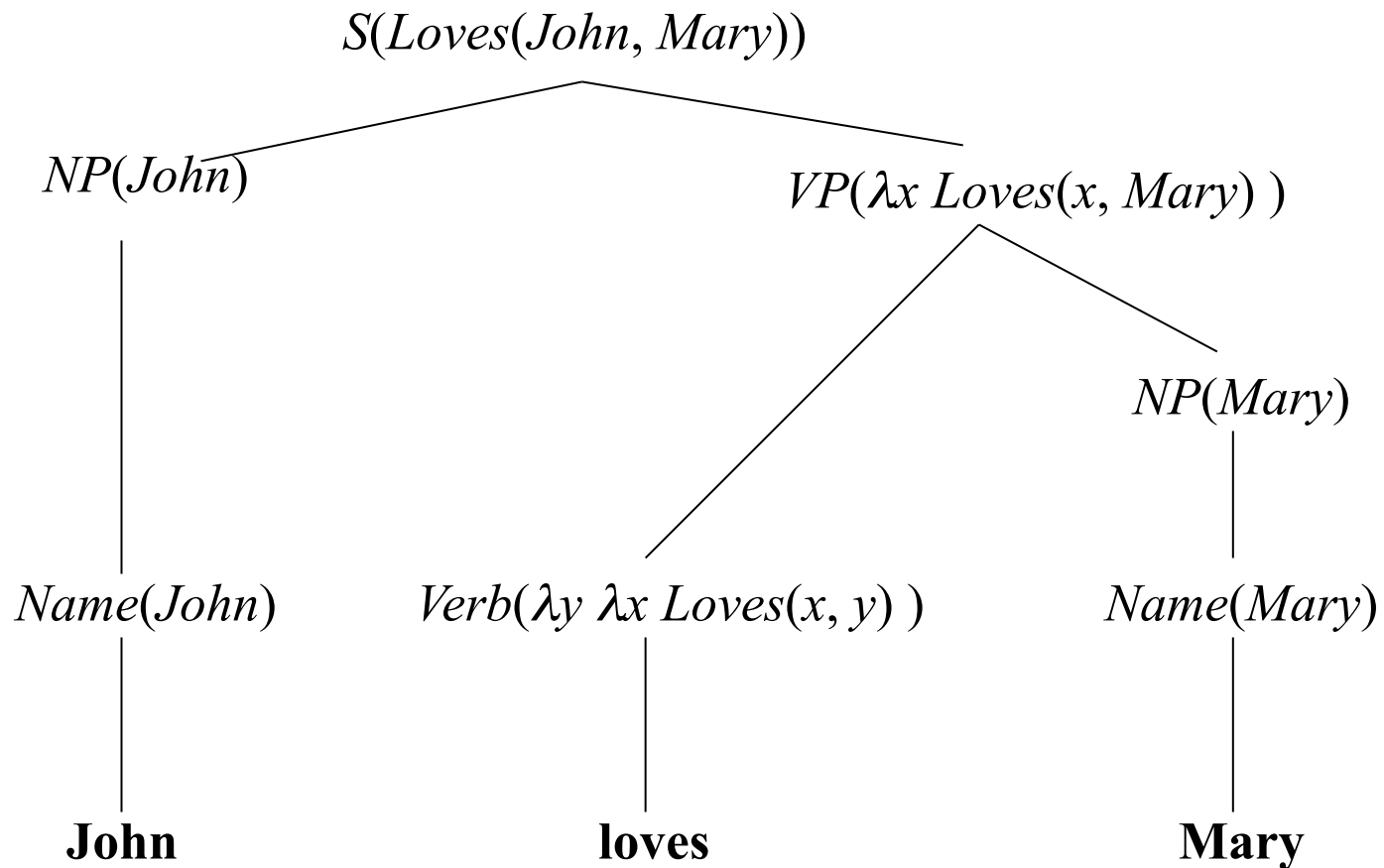
Semantic Interpretation

John loves Mary
Loves(John, Mary)

$(\lambda y \lambda x \text{Loves}(x,y)) (\text{Mary}) \equiv \lambda x \text{Loves}(x, \text{Mary})$
 $(\lambda x \text{Loves}(x, \text{Mary})) (\text{John}) \equiv \text{Loves}(\text{John}, \text{Mary})$

S(rel(obj))	→	NP(obj) VP(rel)
VP(rel(obj))	→	Verb(rel) NP(obj)
NP(obj)	→	Name(obj)
Name(John)	→	John
Name(Mary)	→	Mary
Verb($\lambda y \lambda x \text{Loves}(x,y)$)	→	loves

Semantic Interpretation



Pragmatic Interpretation

- Adding context-dependent information about the current situation to each candidate semantic interpretation
- Indexicals: phrases that refer directly to the current situation
 - “I am in Boston today”
(“I” refers to speaker and “today” refers to now)

Language Generation

The same DCG can be used for parsing and generation

- Parsing:

- Given: $S(\text{sem}, [\text{John}, \text{loves}, \text{Mary}])$
- Return: $\text{sem} = \text{Loves}(\text{John}, \text{Mary})$

- Generation:

- Given: $S(\text{Loves}(\text{John}, \text{Mary}), \text{words})$
- Return: $\text{words} = [\text{John}, \text{loves}, \text{Mary}]$

Ambiguity

- Lexical ambiguity
 - “the **back** of the room” vs. “**back** up your files”
 - “In the **interest** of stimulating the economy, the government lowered the **interest** rate.”
- Syntactic ambiguity (structural ambiguity)
 - “I smelled a wumpus in 2,2”
- Semantic ambiguity
 - “the IBM lecture”
- Pragmatic ambiguity
 - “I’ll meet you next Friday”

Metonymy

Denotes a concept by naming some other concept closely *related* to it

- Examples:

- Company for company's spokesperson ("IBM announced a new model")
- Author for author's works ("I read Shakespeare")
- Producer for producer's product ("I drive a Honda")

Metonymy

- Representation of “IBM announced”

$\exists m, x, e \ x = IBM \wedge e \in Announce(m) \wedge Af\ tek(Now, e) \wedge Metonymy(m, x)$

$\forall m, x \ (m = x) \Rightarrow Metonymy(m, x)$

$\forall m, x \ x \in Organizations \wedge Spokesperson(m, x) \Rightarrow Metonymy(m, x)$

Metaphor

Refer to concepts using words whose meanings are appropriate to other completely different kinds of concepts

- Example: corporation-as-person metaphor:
 - Speak of a corporation as if it is a person and can experience emotions, has a mind, etc.
 - “That doesn’t **scare** Digital, which has grown to be the world’s second-largest computer maker.”
 - “But if the company changed its **mind**, however, it would do so for investment reasons, the filing said.”

Disambiguation

$$\arg \max_{intent} \text{Likelihood}(intent \mid words, situation)$$

- Disambiguation is like diagnosis
- The speaker's intent to communicate is an unobserved cause of the words in the utterance
- The hearer's job is to work backwards from the words and from knowledge of the situation to recover the most likely intent of the speaker

Discourse Understanding

- Discourse: multiple sentences
- Reference resolution: The interpretation of a pronoun or a definite noun phrase that refers to an object in the world
- “John flagged down the waiter. He ordered a ham sandwich.”
 - “He” refers to “John”
- “After John proposed to Mary, they found a preacher and got married. For the honeymoon, they went to Hawaii.”
 - “they”? “the honeymoon”?

Discourse Understanding

- Structure of coherent discourse: Sentences are joined by coherence relations
- Examples of coherence relations between S1 and S2:
 - Enable or cause: S1 brings about a change of state that causes or enables S2
 - “I went outside. I drove to school.”
 - Explanation: the reverse of enablement, S2 causes or enables S1 and is an explanation for S1
 - “I was late for school. I overslept.”
 - Exemplification: S2 is an example of the general principle in S1
 - “This algorithm reverses a list. The input [A,B,C] is mapped to [C,B,A].”
 - Etc.

Summary

- NLP is full of ambiguity
 - Natural languages a testament to human intelligence, creativity
- NLP largely processes one utterance at a time
- Deterministic methods can follow context free grammars, (DCGs, hence Prolog)
- PCFGs add probabilistic interpretation
- Both parsable using DP (CYK algorithm)