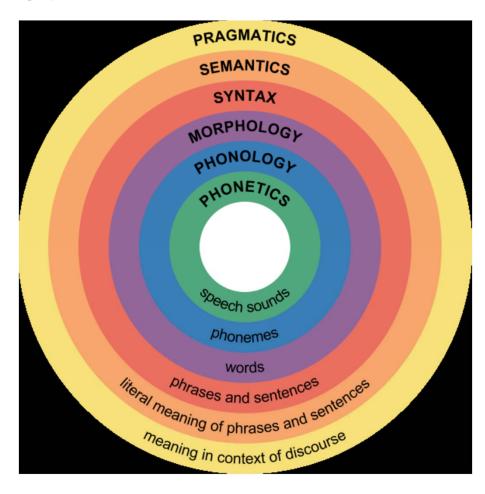
# LAP 3: COMPUTATIONAL SYNTAX INTRODUCTION

Ruben Urizar

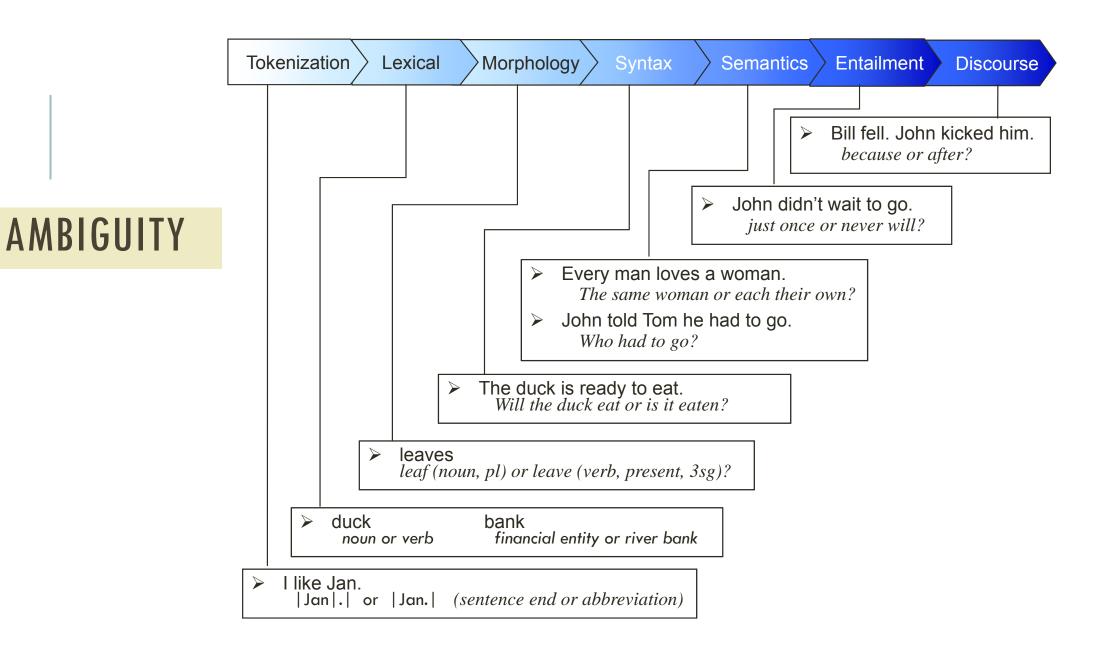
#### SYNTAX IN CONTEXT



#### WHO CARES?

- Grammar checkers
- Question answering
- Information extraction
- Machine Translation
- Generation

• • •



#### SYNTACTIC AMBIGUITY (i)

There are many types of syntactic ambiguity, just to mention some:

- 1. **PoS ambiguity** occurs when at least two words can belong to two or more parts of speech.
  - ✓ They can fish.
  - ✓ I saw her duck.

This is **quite rare** despite the large numbers of words that can be both nouns and verbs in English.

#### SYNTACTIC AMBIGUITY (ii)

- 2. **PP** (prepositional phrase) **attachment**. PPs can modify VPs as well as NPs:
  - ✓ She saw a man with a telescope
    - -[She] [saw] [a man] [with a telescope].
    - -[She] [saw] [a man with a telescope].
  - ✓ Peter waved Mary from the school with a flag (...)
    - Peter waved [from the school with a flag] => (school with a flag)
    - -Peter waved [from the school] [with a flag] => (Peter waved with a flag)

# SYNTACTIC AMBIGUITY (iii)

#### 3. Coreference:

- ✓ John told Tom he had to go
  - -he = John => John had to go
  - -he = Tom => Tom had to go.

## AMBIGUITY RESOLUTION (i)

• Sometimes ambiguity cannot be resolved; we cannot know what the producer of the sentence meant:

```
→ I saw her duck:
```

```
1. I saw [NP her duck].
```

```
2. I saw [NP her] [VP duck].
```

#### → They can fish:

```
1. They [VP can fish].
```

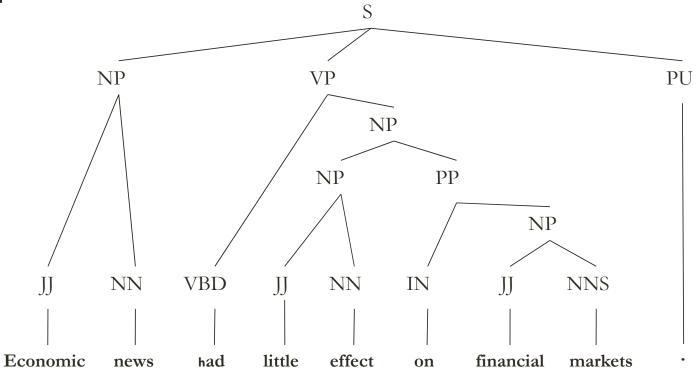
```
2. They [VP can] [NP fish].
```

## AMBIGUITY RESOLUTION (ii)

- Often, the context helps to identify the correct interpretation:
  - → [Peter waved from the school with a flag.] He kept waving the flag all the time to Mary's house. ▷ 'man with a flag' interpretation
- Sometimes, world knowledge (or 'commonsense knowledge') is needed:
  - → Peter waved from the school with a balcony... ▷ 'school with a balcony' interpretation (NOT 'man with a balcony')
- Many ambiguity cases can be resolved using linguistic context:
  - → I saw a duck in the pond ⇒ 'duck' noun
  - → What should you do when she leaves you? ⇒ leaves verb

#### TWO VIEWS OF LINGUISTIC STRUCTURE

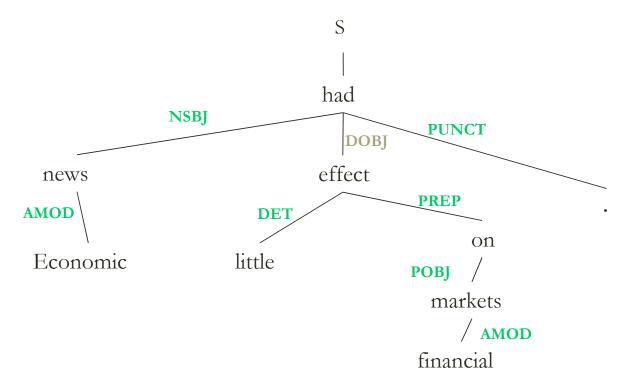
1. Phrase structure organizes words into nested constituents or phrase structures.



#### TWO VIEWS OF LINGUISTIC STRUCTURE

2. Dependency structure shows which words depend on (modify or are arguments of) which other words.

Economic news had little effect on financial markets



#### TWO VIEWS OF LINGUISTIC STRUCTURE

#### **Dependency structure (2)**

Economic news had little effect on financial markets

- root ( ROOT-0 , had-3 )
- nsubj (had-3, news-2)
- amod ( news-2 , Economic-1 )
- dobj (had-3, effect-5)
- amod (effect-5, little-4)
- prep ( effect-5 , on-6 )
- amod (markets-8, financial-7)

#### APPROACHES TO SYNTAX (i)

Both tasks (giving all the possible interpretations of a sentence and choosing one out of them) can be tackled with different approaches:

- The knowledge-based way
  - Using linguistic knowledge to eliminate nonsense analyses and/or choose coherent interpretations.
- The automatic way
  - Using statistical methods based on empirical evidence (corpora).
- A combination of both

#### APPROACHES TO SYNTAX (ii)

- 1. Based on linguistic knowledge
  - Context-free grammars (CFG)
  - Unification-based grammars
    - LFG (Lexical Functional Grammar)
    - HPSG (Head-driven Phrase Structure Grammar)
    - PATR-II
    - **>** ...
  - Finite-state mechanisms
    - CG (Constraint Grammar)
    - XFST (Xerox Finite State Tool)
- 2. Probabilistic
- 3. Combined

# Context Free Grammars CFG

#### **CONTEXT-FREE GRAMMARS**

- Formulated by Chomsky (1956) and Backus (1959)
- Capture constituents and ordering
  - Need something else for grammatical relations and dependency relations
- Consists of
  - > A set of rules
  - > A lexicon

#### **CONTEXT-FREE GRAMMARS**

#### Consist of:

- A lexicon
- A set of rules (productions) expressing the way symbols of the language can be grouped together.

#### **CONTEXT-FREE GRAMMAR: lexicon**

```
Noun 
ightarrow flights | breeze | trip | morning | ...
Verb 
ightarrow is | prefer | like | need | want | fly
Adjective 
ightarrow cheapest | non - stop | first | latest | other | direct | ...
Pronoun 
ightarrow me | I | you | it | ...
Proper-Noun 
ightarrow Alaska | Baltimore | Los Angeles | Chicago | United | American | ...
Determiner 
ightarrow the | a | an | this | these | that | ...
Preposition 
ightarrow from | to | on | near | ...
Conjunction 
ightarrow and | or | but | ...
```

#### CONTEXT-FREE GRAMMAR: set of rules

```
S 
ightarrow NP VP I + want a morning flight

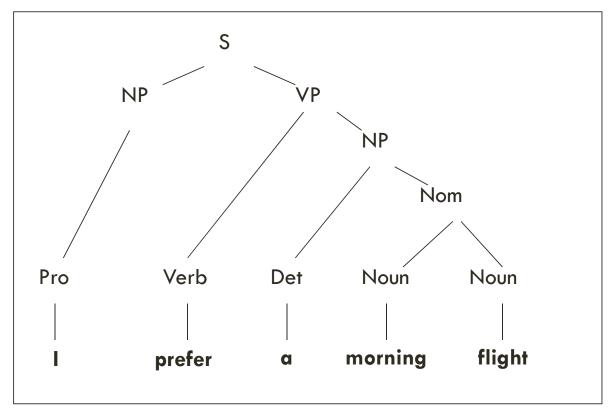
NP 
ightarrow Pronoun I
| Proper-Noun | Los Angeles | Det Nominal | a + flight |
Nominal <math>
ightarrow Noun Nominal | morning + flight |
| Noun | flights

<math>VP 
ightarrow Verb | do |
| Verb NP | want + a flight |
| Verb NP PP | leave + Boston + in the morning |
| Verb PP | leaving + on Thursday

<math>PP 
ightarrow Preposition NP | from + Los Angeles
```

#### **CONTEXT-FREE GRAMMAR:** derivations and trees

GRAMMAR rules			LEXIC	ON entries
S	$\rightarrow$	NP VP	Pro	$\rightarrow$
VP	$\rightarrow$	V NP	V	→ prefer
NP	$\rightarrow$	Pro	Det	$\rightarrow$ a
NP	$\rightarrow$	Det Nom	Noun	→ morning
Nom	$\rightarrow$	Noun Noun	Noun	→ flight



# Unification-based grammars

#### FEATURE STRUCTURES (i)

- Context-free grammars do not deal with issues like agreement
- Unification-based grammars use features such as 'number', 'person', 'gender'...
- A feature structure allows us to state properties e.g. about a noun phrase

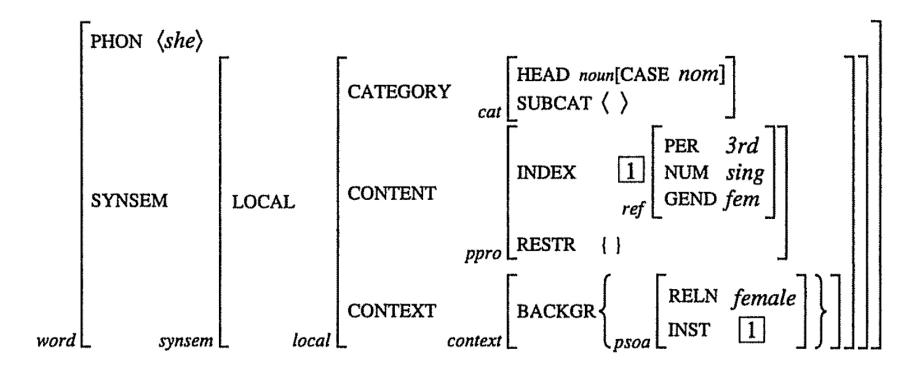
```
[she NP
number sg
person 3
gender fem ]
```

#### or a verb phrase

```
[walks VP
number sg
person 3
tense present ]
```

#### FEATURE STRUCTURES (ii)

- Each feature (e.g., 'number') is paired with a value (e.g., 'sg')
- A bundle of feature-value pairs can be put into an attribute-value matrix (AVM)

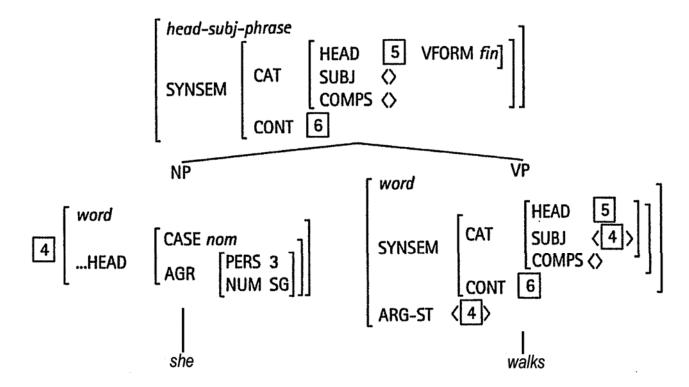


## UNIFICATION (i)

- Unification (U) = a basic operation to merge two feature structures into a resultant feature structure (FS)
- The two-feature structures must be compatible, i.e., have no values that conflict
  - Identical FSs:
    - [number sg] U [number sg] = [number sg]
  - Conflicting FSs:
    - [number sg] U [number pl] = Fail
  - Merging with an unspecified FS:
    - [number sg] U [number []] = [number sg]

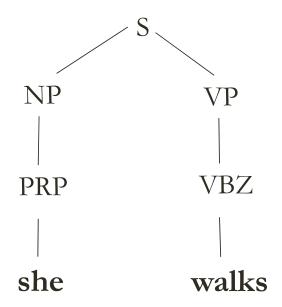
#### UNIFICATION (ii)

A feature structure matrix of type *phrase* in HPSG states that the VP must agree with its subject in number and person (3 sg). In this case, the feature structures of *she* and *walks* are compatible so unification takes place.



# UNIFICATION (iii)

The resulting parse tree would be the following:



# Probabilistic Parsing and Treebanks

## STATISTICAL PARSING (i)

- The basic idea is:
  - Start with a treebank
    - <u>treebank</u>: a corpus with syntactic annotation, i.e., already-parsed sentences (e.g. the Penn Treebank)
  - Examine which parse trees occur frequently
  - Extract grammar rules corresponding to those parse trees, estimating the probability of the grammar rule based on its frequency
- That is, we'll have a context-free grammar (CFG) augmented with probabilities

## STATISTICAL PARSING (ii)

#### Computing the probabilities of a particular parse tree

We have the following rules and probabilities:

• Being P(T) = the probability of a particular parse tree,

```
P(T) = P(S \rightarrow VP) * P(VP \rightarrow V NP) * ... * P(N \rightarrow flight)
= .05*.40*.20*.30*.05*.25 = .000015, or 1.5 x 10-5
```

#### USING PROBABILITIES

- So, the probability for that parse is 0.000015. What's the big deal?
  - Probabilities are useful for comparing with other probabilities
- Whereas we couldn't decide between two parses using a regular CFG, we now can.
- For example, "TWA flights" is ambiguous between being two separate NPs (as in I gave [NP John] [NP money]) or one NP:
  - > A: [book [TWA] [flights]]
  - > B: [book [TWA flights]]
- Comparing probabilities (previous slide) allows us to choose option B

#### **TREEBANKS**

The rise of annotated data: The Penn Treebank

```
( (S
    (NP-SBJ (DT The) (NN move))
    (VP (VBD followed)
      (NP
        (NP (DT a) (NN round))
        (PP (IN of)
          (NP
            (NP (JJ similar) (NNS increases))
            (PP (IN by)
              (NP (JJ other) (NNS lenders)))
            (PP (IN against)
              (NP (NNP Arizona) (JJ real) (NN estate) (NNS loans))))))
      (,,)
      (S-ADV
        (NP-SBJ (-NONE- *))
        (VP (VBG reflecting)
          (NP
            (NP (DT a) (VBG continuing) (NN decline))
            (PP-LOC (IN in)
              (NP (DT that) (NN market))))))
    (. .)))
```

#### **TREEBANKS**

- Building a treebank may seem a lot slower and less useful than building a grammar... but
- A treebank gives us many things
  - Reusability of the labor
  - Broad coverage
  - Frequencies and distributional information
  - A way to evaluate systems

# Parts of Speech PoS

## PARTS OF SPEECH (PoS)

... or 'word classes'

Here are some sentences extracted from different conversations in a café.

- Our friends are sitting in the corner, look.
- I have an important conference at work tomorrow, so I am rather busy
- Would you like to come to our party on Saturday, Jessica?
- This coffee is really good.
- And it's cheap here.
- 1. What different parts of speech can you distinguish?
- 2. Classify all the words in their corresponding part of speech

#### PARTS OF SPEECH (PoS)

#### 1. Verb

- Lexical: have, am, is, like, come, sitting, look
- Auxiliary: would, are

#### 2. Noun

- Common: conference, work, coffee, party, Saturday, friends, corner
- Proper: Jessica
- 3. Adjective: important, busy, good cheap
- 4. Adverb: really, tomorrow, rather, here
- 5. Preposition: at, to, on, in
- **6. Determiner:** an, this, our, the
- 7. Pronoun: I, it, you
- 8. Conjunction: because, and

# PENN TREEBANK TAGSET (i)

VERB	Lexical and Auxiliary (not modal)	VB	Verb, base form	
		VBD	Verb, past tense	
		VBG	Verb, gerund or present participle	
		VBN	Verb, past participle	
		VBP	Verb, non-3rd person singular present	
		VBZ	Verb, 3rd person singular present	
	Modal MD		Modal	
NOUN	Common	NN	Noun, singular or mass	
		NNS	Noun, plural	
	Proper	NNP	Proper noun, singular	
		NNPS	Proper noun, plural	
		JJ	Adjective	
ADJECTIVE		JJR	Adj. comparative	
		JJS	Adjective, superlative	

# PENN TREEBANK TAGSET (ii)

	RB	Adverb	
ADVERB	RBR	Adverb, comparative	
ADVERD	RBS	Adverb, superlative	
	WRB	Wh-adverb	
PREPOSITION	IN	Preposition (or subordinating conjunction)	
	DT	Determiner	
DETERMINER	WDT	Wh-determiner	
	PDT	Predeterminer	
	PRP	Personal pronoun	
PRONOUN	PRP\$	Possessive pronoun	
PRONOUN	WP	Wh-pronoun	
	WP\$	Possessive wh-pronoun	
CONJUNCTION	CC	Coordinating conjunction	
CONJUNCTION	IN	Subordinating conjunction (or preposition)	

# PENN TREEBANK TAGSET (iii)

	EX	Existential there	
	CD	Cardinal number	
	FW	Foreign word	
	LS	List item marker	
OTHER	POS	Possessive ending	
	RP	Particle	
	SYM	Symbol	
	ТО	to	
	UH	Interjection	

## VISLCG TAGSET

	Part of Speech Tags
ADJ	adjective
ADV	adverb
ART	article, cp. <art> (*)</art>
DET	determiner pronoun (inflecting)
IN	interjection
INDP	independent pronoun (non-inflecting)
INFM	infinitive marker
KC	coordinating conjunction
KS	subordinating conjunction
N	noun
PROP	proper noun, name
PRP	preposition
PERS	personal pronoun
V	verb> PR, IMPF, IMP, INF, PCP[12]

Morphological information Tags		
1P	1. person plural (PERS +, V ¤)	
15	1. person singular (PERS +, V ¤)	
2P	2. person plural (PERS +, V ¤)	
25	2. person singular (PERS +, V ¤)	
3P	3. person plural (PERS +, V ¤)	
35	3. person singular (PERS +, V ¤)	
ACC	accusative case (N, PERS)	
COM	comparative degree (ADJ, ADV)	
CONJ	conjunction, cp. KS and KC	
GEN	genitive case	
IMP	imperative (V)	
IMPF	past tense (V), also PAST (*)	
INF	infinitive (V)	
NOM	nominative case	
NUM	numeral> <card>, <ord></ord></card>	
P	plural	
PCP	past participle, cp. PCP2, PED (*)	
PCP1	present participle, cp. GER, PING (*)	
PCP2	past participle, cp. PCP, PED (*)	
PED	-ed participle, cp. PCP, PCP2 (*)	
PR	present tense> AKT, PAS	
S	singular, cp. SG (*)	
SUBJ	subjunctive, cp. CONJ (*)	
SUP	superlative degree (ADJ, ADV)	
()		

# **EXERCISES**

Exercise 01 in eGela

#### **DEPENDENCY TREE**

Global emissions of CO2 will rise for the first time in four years.

#### **PARSERS**

- Stanford: <a href="mailto:nlp.stanford.edu:8080/corenlp/">nlp.stanford.edu:8080/corenlp/</a>
- Freeling: <a href="nlp.lsi.upc.edu/freeling/demo/demo.php">nlp.lsi.upc.edu/freeling/demo/demo.php</a>

#### **EXERCISES**

Exercise 02 in eGela

Exercise 03 in eGela