Syntax: Constituency and Dependency Structure

CS 490A, Fall 2021 https://people.cs.umass.edu/~brenocon/cs490a_f21/

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Announcements

- HW3 coming out shortly (show)
 - Due Friday, Oct 29
 - Work with POS tags, NER, dependency parsing
- Project proposals due tomorrow!
 - Question: how will they be graded?
- Discuss project ideas after class?
 - (outdoors please)

- Syntax: how do words structurally combine to form sentences and meaning?
- Representations
 - Constituents
 - [the big dogs] chase cats
 - [colorless green clouds] chase cats
 - Dependencies
 - The **dog** ← **chased** the cat.
 - My dog, who's getting old, chased the cat.
- Idea of a grammar (G): global template for how sentences / utterances / phrases w are formed, via latent syntactic structure y
 - Linguistics: what do G and P(w,y | G) look like?
 - Generation: score with, or sample from, P(w, y | G)
 - Parsing: predict P(y | w, G)

Syntax for NLP

- If we could predict syntactic structure from raw text, that could help with...
 - Language understanding: meaning formed from structure
 - Grammar checking
 - Preprocessing: Extract phrases and semantic relationships between words for features, viewing, etc.
- Provides a connection between the theory of generative linguistics and computational modeling of language

Is language context-free?

- Regular language: repetition of repeated structures
 - e.g. "base noun phrases": (Noun | Adj)* Noun
 - subset of the JK pattern
- Context-free: hierarchical recursion
- Center-embedding: classic theoretical argument for CFG vs. regular languages
 - (10.1) The cat is fat.
 - (10.2) The cat that the dog chased is fat.
 - (10.3) *The cat that the dog is fat.
 - (10.4) The cat that the dog that the monkey kissed chased is fat.
 - (10.5) *The cat that the dog that the monkey chased is fat.
- Competence vs. Performance?

Hierarchical view of syntax

 "a Sentence made of Noun Phrase followed by a Verb Phrase"



[From Phillips (2003)]

Context-free grammars (CFG)

• A CFG is a 4-tuple:

- a set of non-terminals N
- Σ a set of terminals (distinct from *N*)
- a set of productions, each of the form $A \rightarrow \beta$, R_{-} where $A \in N$ and $\beta \in (\Sigma \cup N)^*$

Example: see handout!

- a designated start symbol S
- Derivation: a sequence of rewrite steps from S to a string (sequence of terminals, i.e. words)
- Yield: the final string (sentence)
- The parse tree or constituency tree corresponds to the rewrite steps that were used to derive the string

- A CFG is a "boolean language model"
- A probabilistic CFG is a probabilistic language model:
 - Every production rule has a probability; defines prob dist. over strings.

• Example: derivation from worksheet's grammar





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[Examples from Eisenstein (2017)]

NINIC

Is language context-free?

- CFGs nicely explain nesting and agreement (if you stuff grammatical features into the nonterminals)
 - The processor <u>has</u> 10 million times fewer transistors on it than todays typical microprocessors, <u>runs</u> much more slowly, and <u>operates</u> at five times the voltage...

• $S \rightarrow NN VP$ $VP \rightarrow VP3S | VPN3S | \dots$ $VP3S \rightarrow VP3S, VP3S, and VP3S | VBZ | VBZ NP | \dots$

• Ambiguities in syntax

Attachment ambiguity we eat sushi with chopsticks, I shot an elephant in my pajamas.
Modifier scope southern food store
Particle versus preposition The puppy tore up the staircase.
Complement structure The tourists objected to the guide that they couldn't hear.
Coordination scope "I see," said the blind man, as he picked up the hammer and saw.
Multiple gap constructions The chicken is ready to eat

Probabilistic CFGs

- $S \rightarrow NP VP$ [.80] $Det \rightarrow that [.10] \mid a [.30] \mid the [.60]$ Noun \rightarrow book [.10] | flight [.30] $S \rightarrow Aux NP VP$ [.15] *meal* [.15] | *money* [.05] [.05] $S \rightarrow VP$ *flights* [.40] | *dinner* [.10] $NP \rightarrow Pronoun$ [.35] $Verb \rightarrow book [.30] \mid include [.30]$ $NP \rightarrow Proper-Noun$ [.30] *prefer*;[.40] $NP \rightarrow Det Nominal$ [.20] *Pronoun* $\rightarrow I[.40]$ *she* [.05] $NP \rightarrow Nominal$ [.15] [.75] *me* [.15] | *you* [.40] *Nominal* \rightarrow *Noun Nominal* \rightarrow *Nominal Noun* .20] *Proper-Noun* \rightarrow *Houston* [.60] Nominal \rightarrow Nominal PP [.05] *TWA* [.40] $Aux \rightarrow does [.60] \mid can [40]$ $VP \rightarrow Verb$ [.35] *Preposition* \rightarrow *from* [.30] | *to* [.30] $VP \rightarrow Verb NP$ [.20] $VP \rightarrow Verb NP PP$ [.10] on [.20] | near [.15] $VP \rightarrow Verb PP$ [.15] through [.05] $VP \rightarrow Verb NP NP$ [.05] $VP \rightarrow VP PP$ [.15] [1.0] $PP \rightarrow Preposition NP$
- Defines a probabilistic generative process for words in a sentence
 - Can parse with a modified form of CKY
- How to learn? Fully supervised with a treebank... unsupervised learning possible too, but doesn't give great results...
 []&M textbook]

```
( (S
                       (NP-SBJ (NNP General) (NNP Electric) (NNP Co.) )
                       (VP (VBD said)
                         (SBAR (-NONE- 0)
                           (S
                             (NP-SBJ (PRP it) )
                             (VP (VBD signed)
                               (NP
                                 (NP (DT a) (NN contract) )
                                 (PP (-NONE- *ICH*-3)))
                               (PP (IN with)
                                 (NP
                                   (NP (DT the) (NNS developers) )
                                   (PP (IN of)
                                     (NP (DT the) (NNP Ocean) (NNP State) (NNP Power) (NN project) )))
                               (PP-3 (IN for)
                                 (NP
                                   (NP (DT the) (JJ second) (NN phase) )
                                   (PP (IN of)
                                     (NP
Treebank
                                       (NP (DT an) (JJ independent)
                                         (ADJP
                                           (QP ($ $) (CD 400) (CD million) )
                                           (-NONE- *U*) )
                                         (NN power) (NN plant) )
                                       (, ,)
                                       (SBAR
                                         (WHNP-2 (WDT which))
                                         (S
                                           (NP-SBJ-1 (-NONE- *T*-2))
                                           (VP (VBZ is)
                                             (VP (VBG being)
                                               (VP (VBN built)
                                                 (NP (-NONE- *-1))
                                                 (PP-LOC (IN in)
                                                   (NP
                                                     (NP (NNP Burrillville) )
                                                    (, ,)
```

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Penn

- A **dependency parse** is a tree or directed graph among the words in the sentence
- Directly encodes word-to-word grammatical relationships. Each edge is between:
 - Head (or governor), to the
 - **Dependent** (or **child**, or modifier)
- [Example]

- There isn't really a generative grammar view of dependencies
 - They're more of a *descriptive* formalism
- Dependency parsers are often convenient to use for downstream applications
 - Subtree = a phrase
 - "Dependency bigrams": (parent, child) pairs in tree
 - Do these correspond to phrases?
- Dependency treebanks are available for many languages (<u>https://universaldependencies.org/</u>), and therefore parsers are widely available.

From Constituents to Dependencies

 Theories of grammar postulate that every phrase has a **head word**, which contains or typifies the grammatical content of the phrase

From Constituents to Dependencies



Figure 11.1: Dependency grammar is closely linked to lexicalized context free grammars: each lexical head has a dependency path to every other word in the constituent. (This example is based on the lexicalization rules from § 10.5.2, which make the preposition the head of a prepositional phrase. In the more contemporary Universal Dependencies annotations, the head of *with claws* would be *claws*, so there would be an edge *scratch* \rightarrow *claws*.)

• stopped here 10/19





Head rules

- Idea: Every phrase has a head word
- Head rules: for every nonterminal in tree, choose one of its children to be its "head". This will define head words.
- Every nonterminal type has a different head rule; e.g. from Collins (1997):
- If parent is NP,
 - Search from right-to-left for first child that's NN, NNP, NNPS, NNS, NX, JJR
 - Else: search left-to-right for first child which is NP
 - Heads are super useful if you want just a single token to stand-in for a phrase!
 - Not just dep parsing. Entities, etc.

 Dependencies tend to be less specific than constituent structure



[<u>Eisenstein (2017)</u>]