

Dependency Parsing

CS 485, Fall 2023

Applications of Natural Language Processing

https://people.cs.umass.edu/~brenocon/cs485_f23/

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CFG issue

- Substitutability is too strong (e.g. “she” as subject vs object)

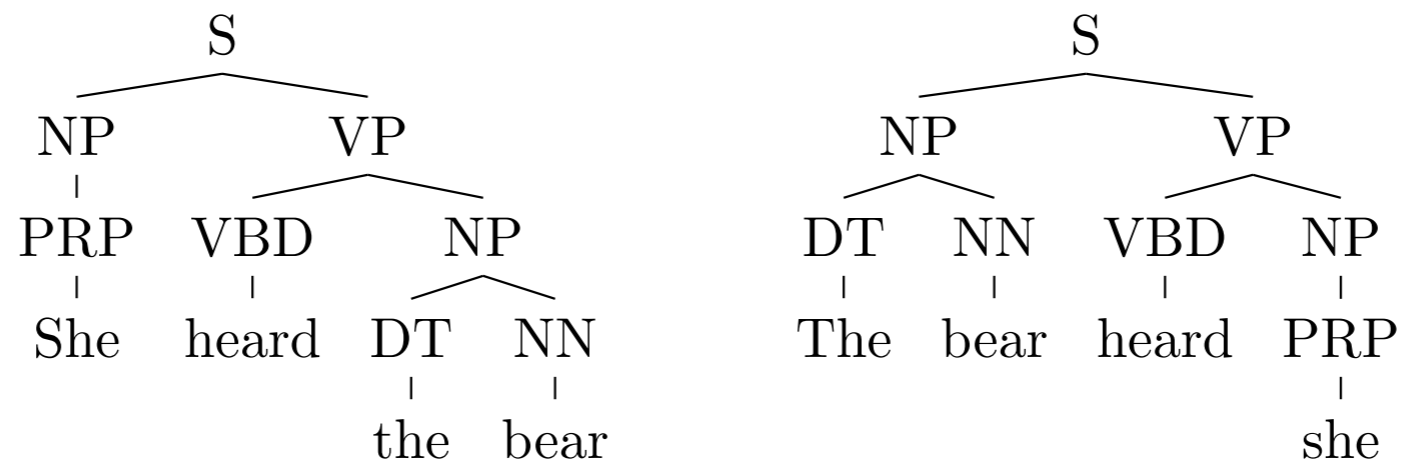
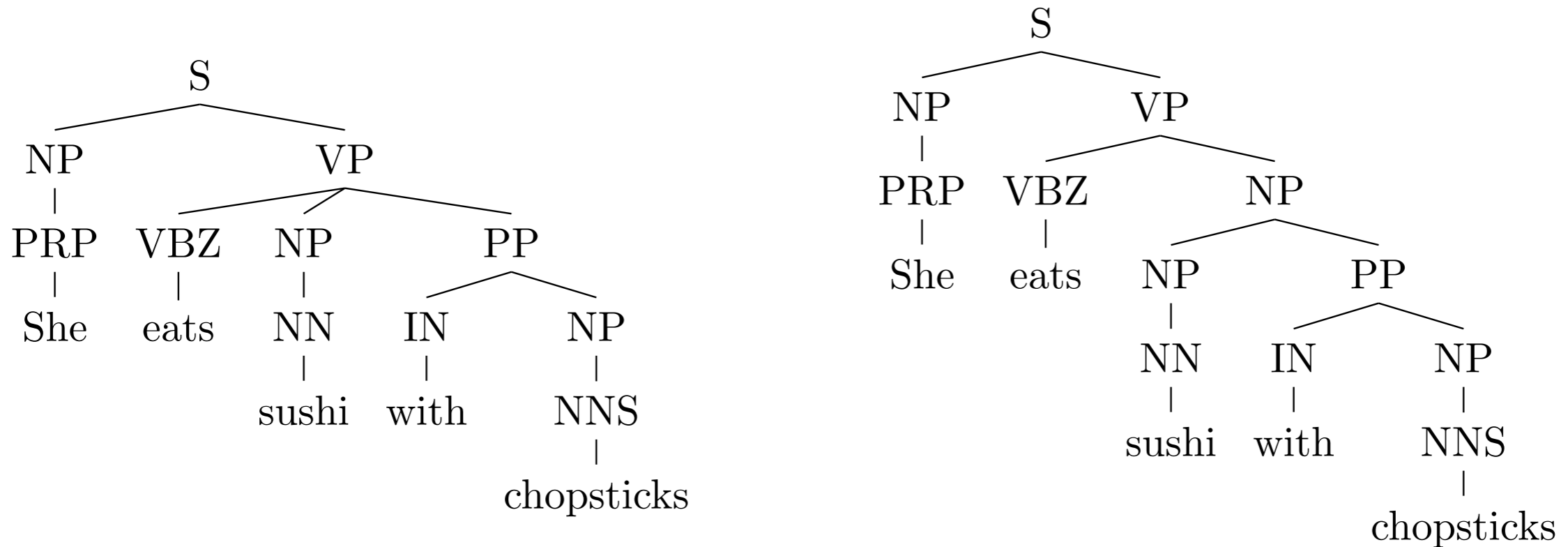


Figure 11.5: A grammar that allows *she* to take the object position wastes probability mass on ungrammatical sentences.

CFG issue

- Substitutability is too strong (PP attachment ambiguity)

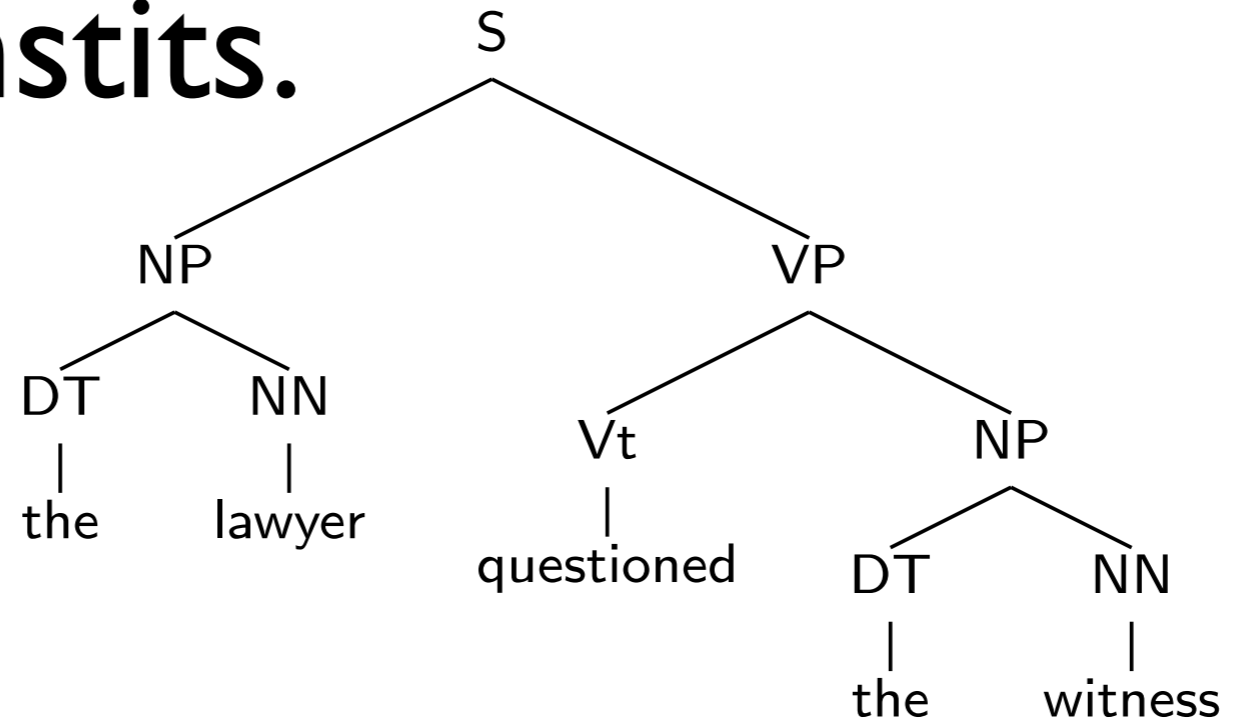


Head rules

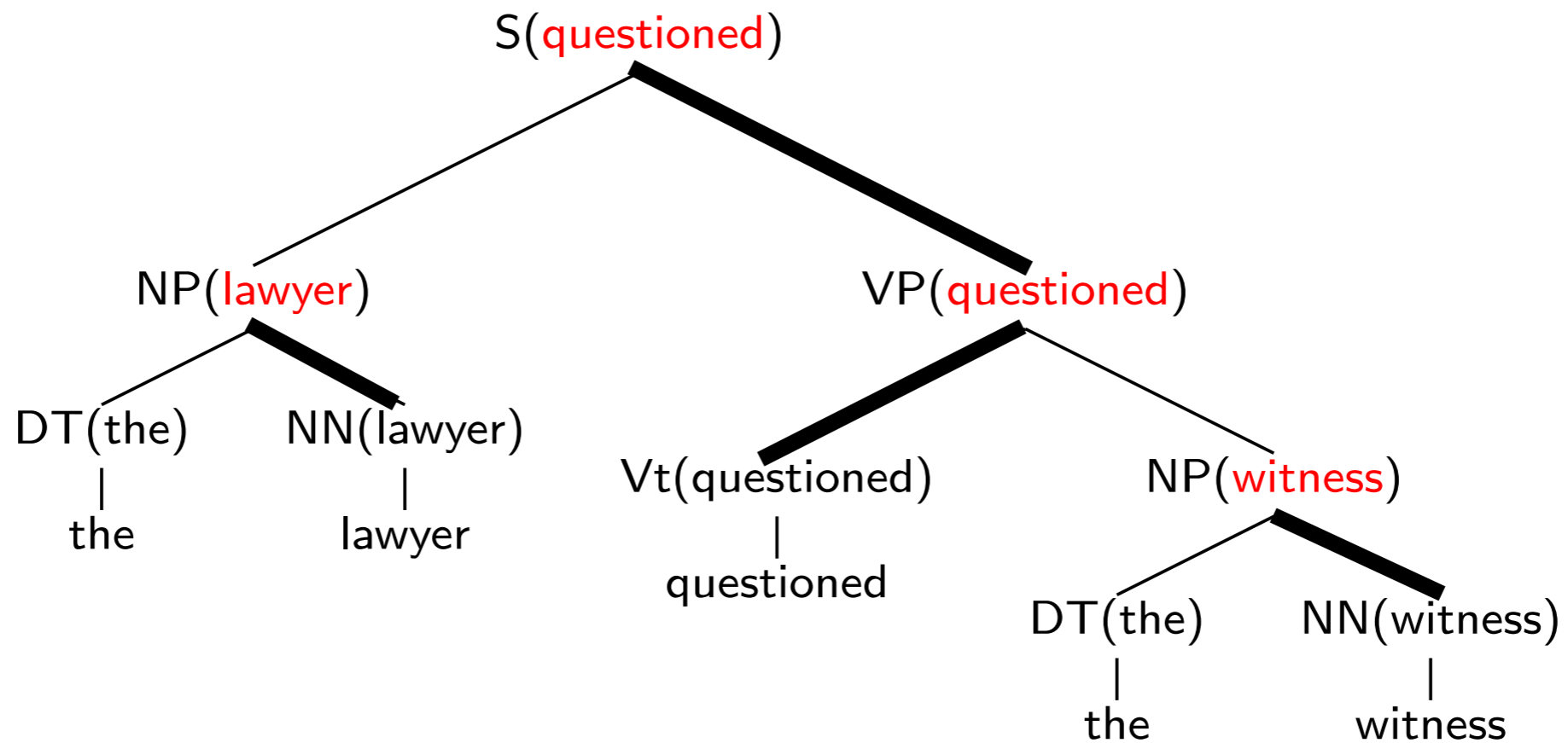
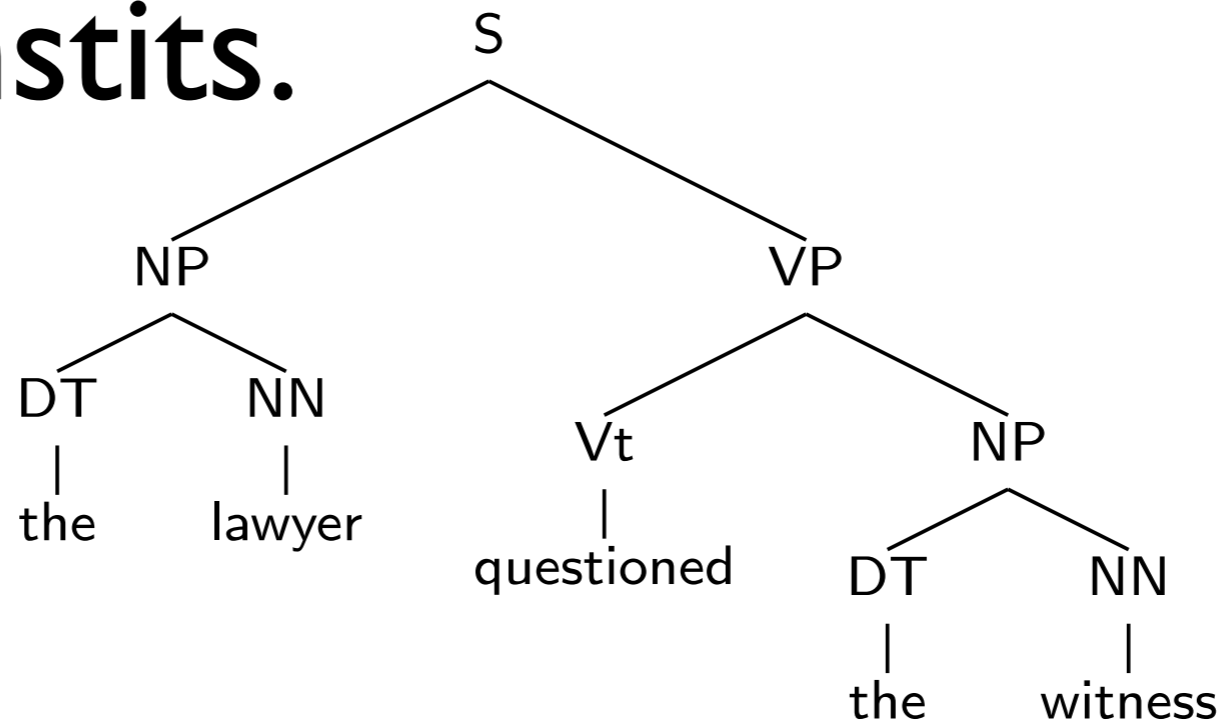
- Idea: Every phrase has a *head word*, that is the "core" or "nucleus" determining its syntactic role
- Head rules: for every nonterminal in tree, choose one of its children to be its "head". This will define head words.
- Every nonterminal type has a different head rule; e.g. from Collins (1997):

- 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 in constits.



Heads in constits.



Lexicalized CFGs

Non-terminal	Direction	Priority
S	right	VP SBAR ADJP UCP NP
VP	left	VBD VBN MD VBZ TO VB VP VBG VBP ADJP NP
NP	right	N* EX \$ CD QP PRP ...
PP	left	IN TO FW

Table 11.3: A fragment of head percolation rules

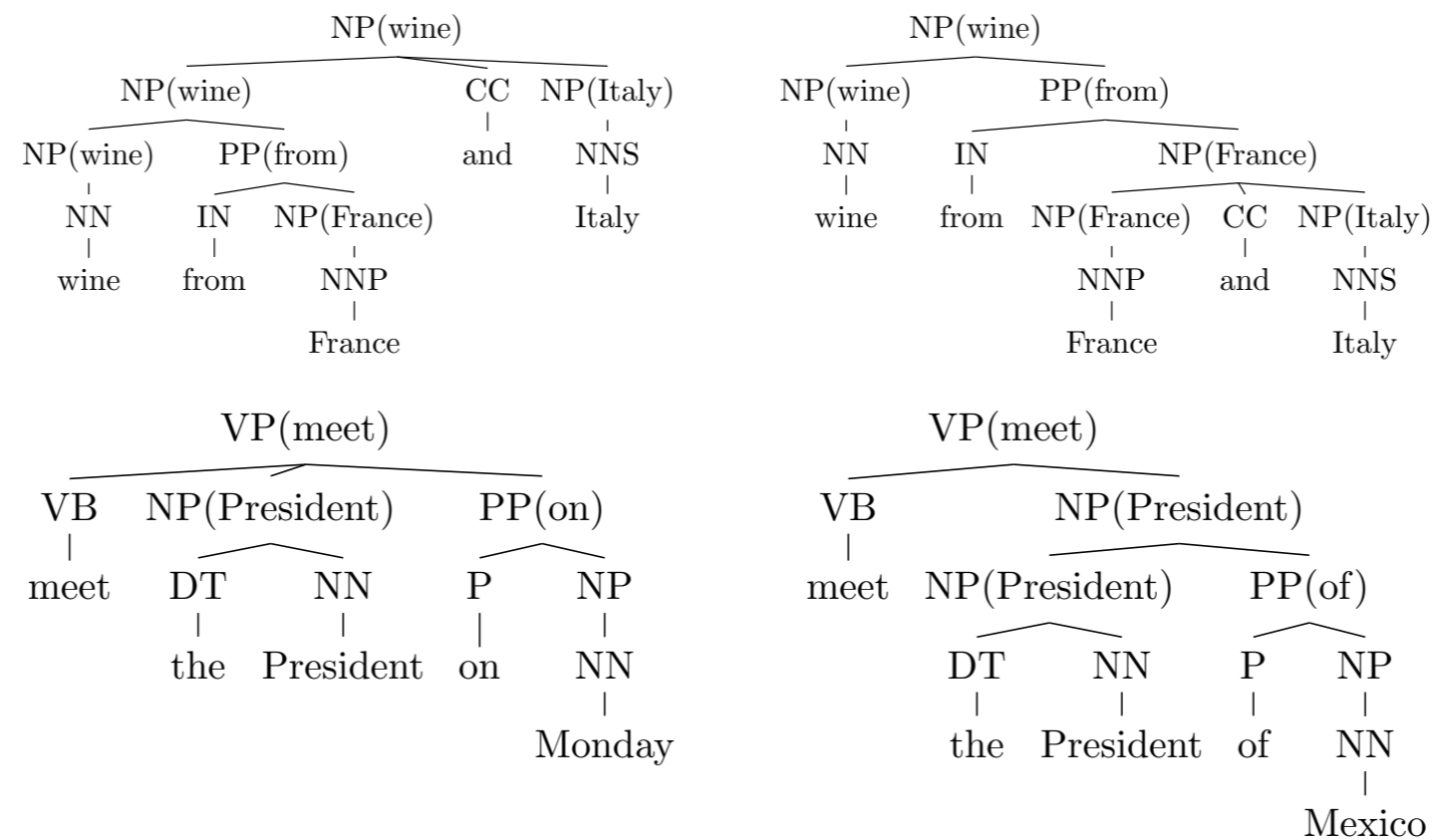
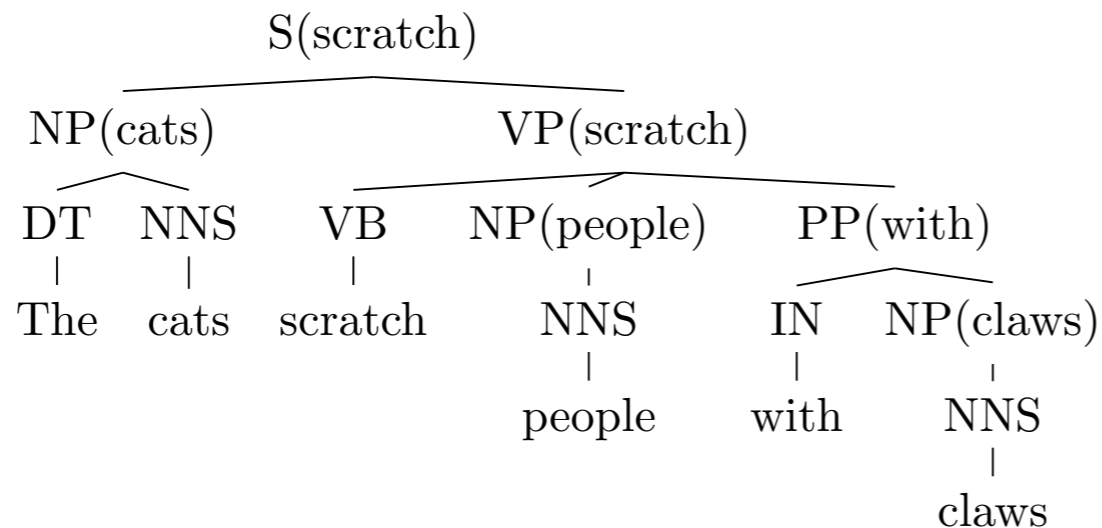
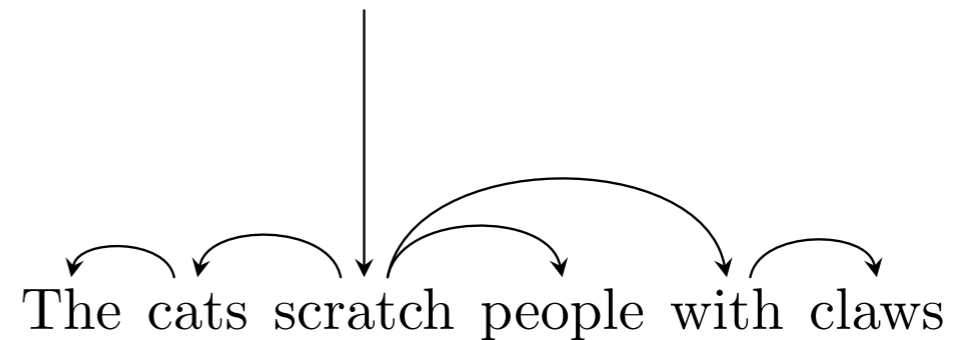


Figure 11.9: Lexicalization can address ambiguity on coordination scope (upper) and PP attachment (lower)

From constituency structure to dependency graphs



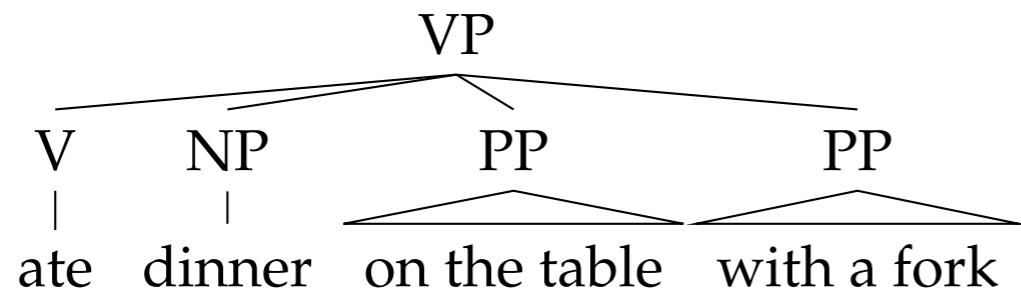
(a) lexicalized constituency parse



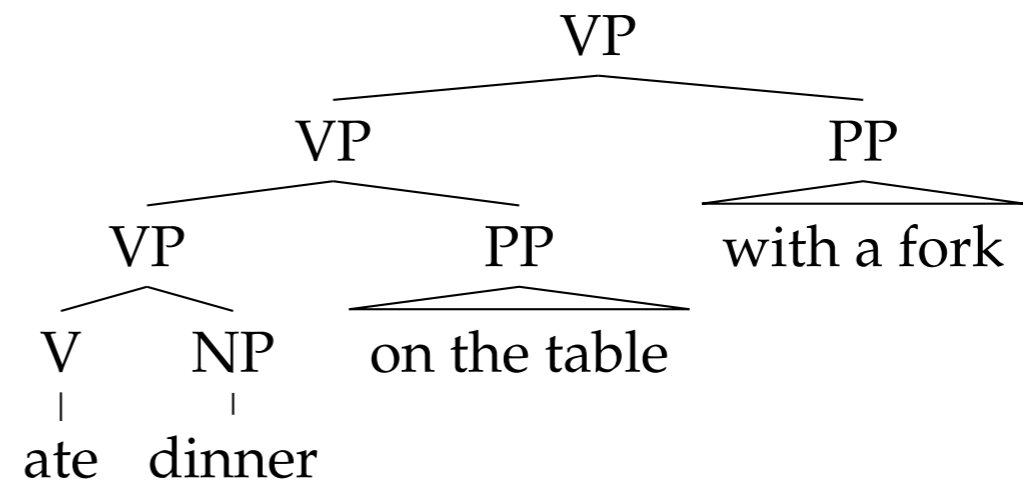
(b) unlabeled dependency tree

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* → *claws*.)

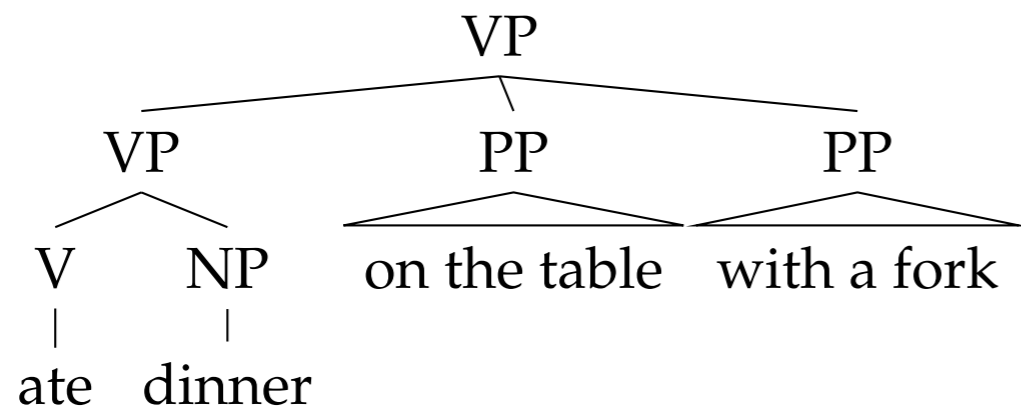
- Dependencies tend to be less specific than constituent structure



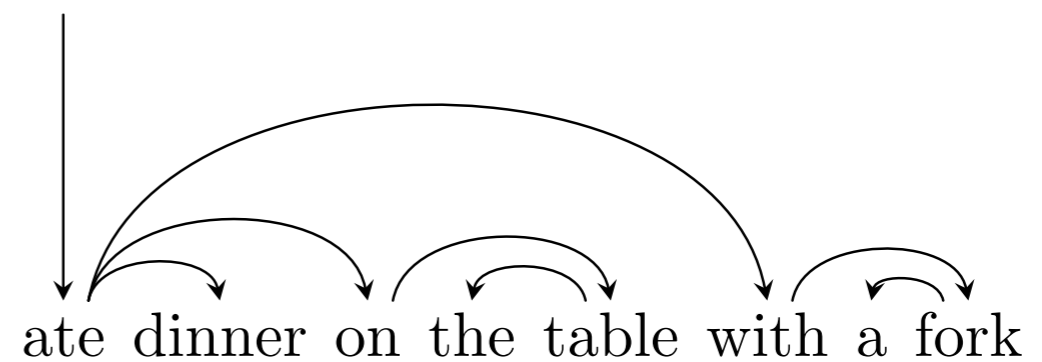
(a) Flat



(b) Two-level (PTB-style)



(c) Chomsky adjunction



(d) Dependency representation

Headedness for *phrase* relations

- Is a given word *X* the subject of verb *Y*?
- Is a given *phrase* *X* the subject of verb *Y*?

Universal Dependencies

- Dependency treebanks are available for *many* different languages
 - <https://universaldependencies.org/>
- Many open-source dependency parsers (and tagging/POS/morphology) trained on them are also widely available; e.g. Stanza, SpaCy, etc.
- They typically directly predict dependencies with another parsing algorithm (shift-reduce, not CKY)

Dependency applications

- Dependencies can be used as less sparse alternative to n-grams
- Sometimes helps, sometimes doesn't
- Dependency relations can be selected for semantic relationships...

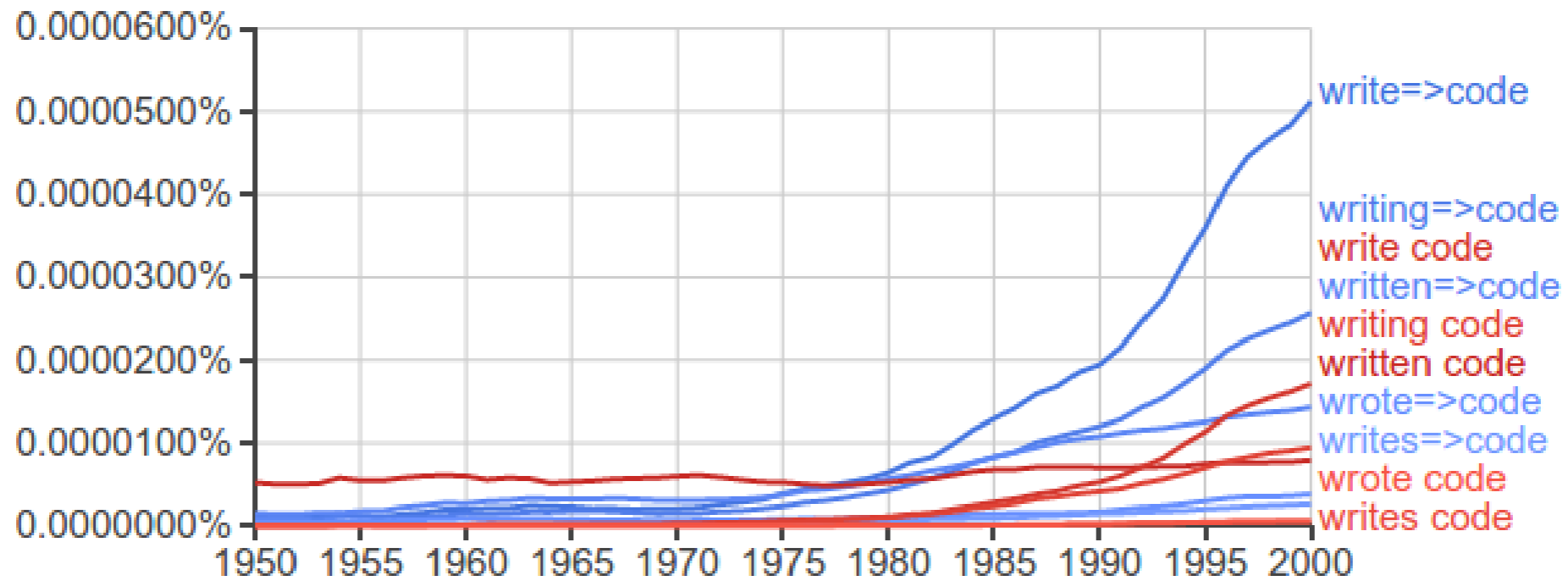


Figure 12.8: Google n-grams results for the bigram *write code* and the dependency arc *write => code* (and their morphological variants)

- *Goldberg & Orwant 2013*: historical dependencies from google books (<https://books.google.com/ngrams/>)

Dependency pattern statistics

4.3.1 IS_A

The IS_A relation covers any nominal or adjectival properties stated to directly pertain to the target entity, represented using the following patterns:⁵

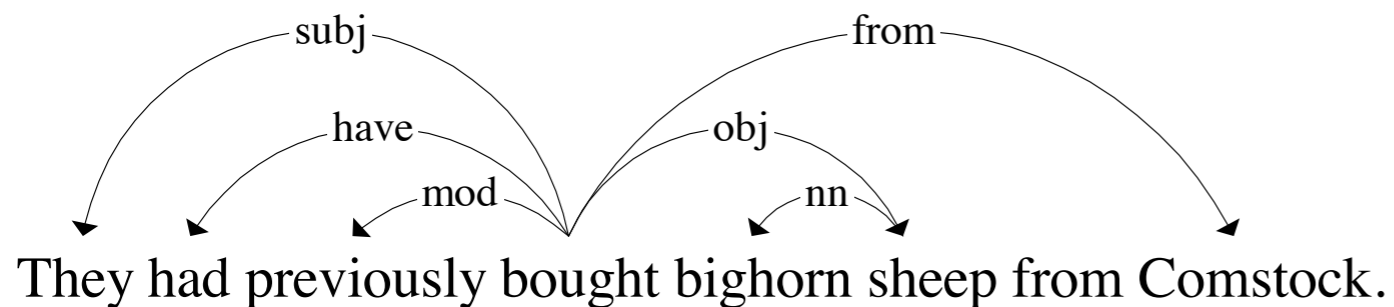
1. target $\xleftrightarrow{\text{nsubj}}$ property_{nom}
2. property_{adj} $\xrightarrow{\text{nsubj}}$ target
3. target $\xleftrightarrow{\text{appos}}$ property_{nom}
4. target $\xrightarrow{\text{compound}}$ property_{nom}
5. target $\xrightarrow{\text{amod}}$ property_{adj}
6. target $\xleftrightarrow{\text{nsubj}}$ property_{nom} $\xrightarrow{\text{amod}}$ property_{adj}
7. target $\xleftrightarrow{\text{appos}}$ property_{nom} $\xrightarrow{\text{amod}}$ property_{adj}

Relation	Trump-Leaning ($t < -2$)	Biden-Leaning ($t > 2$)
IS_A(fauci, <i>property</i> _{nom})	murderer ^{**} , joke ^{**} , hack [*] , fraud [*] , rat [*] , flip [*] , idiot, flop, state, prison, fake, jail	nih ^{**} , hero, md, director, president
IS_A(fauci, <i>property</i> _{adj})	fake [*] , little [*] , deep, liberal, wrong, corrupt	beloved, optimistic, best
AS_AGENT(fauci, <i>verb</i>)	sweat ^{**} , force ^{**} , need [*] , help [*] , read [*] , lie [*] , know [*] , let [*] , not_fund [*] , not_understand [*] , flip, predict, write, make, stick, hold, prove, want, not_say, admit, not_get, demand, issue, laugh, state, put, spread, pull	speak ^{**} , join [*] , warn [*] , throw, not_recommend, offer, provide, respond, consider, debunk, fail, reveal
AS_PATIENT(fauci, <i>verb</i>)	not_trust ^{***} , screw, prosecute, grill, keep to, arrest, expose, lock, do to, remove, accord to, look like, mean, blast, read	know [*] , feature, discredit, threaten, worship, join, insult
HAS_A(fauci, <i>object</i>)	friend [*] , nih [*] , family, mind, hand, ex-employee, involvement, fraud, mask	guidance, time
AS_CONJUNCT(fauci, <i>conj.</i>)	gates ^{***} , obama ^{**} , bill gates [*] , biden [*] , brix, cdc, rest, covid, nih, company, government	director, experts

Table 5: TweetIE extractions with at least 20 unique users with a county-level political valence t -statistic outside of $[-2, 2]$. Results are reported in decreasing absolute value t -statistic. * $|t| > 3$, ** $|t| > 4$, *** $|t| > 5$.

- From geo-located tweets, Mar-Dec 2020

Dependency paths



The paths extracted from this sentence and their meanings are:

- (a) $N:\text{subj}:V \leftarrow \text{buy} \rightarrow V:\text{from}:N$
 $\equiv X \text{ buys something from } Y$
- (b) $N:\text{subj}:V \leftarrow \text{buy} \rightarrow V:\text{obj}:N$
 $\equiv X \text{ buys } Y$
- (c) $N:\text{subj}:V \leftarrow \text{buy} \rightarrow V:\text{obj}:N \rightarrow \text{sheep} \rightarrow N:\text{nn}:N$
 $\equiv X \text{ buys } Y \text{ sheep}$
- (d) $N:\text{nn}:N \leftarrow \text{sheep} \leftarrow N:\text{obj}:V \leftarrow \text{buy} \rightarrow V:\text{from}:N$
 $\equiv X \text{ sheep is bought from } Y$
- (e) $N:\text{obj}:V \leftarrow \text{buy} \rightarrow V:\text{from}:N$
 $\equiv X \text{ is bought from } Y$

An inverse path is also added for each one above.

- **Dependency path corresponds to a lexico-syntactic pattern**

Distributional similarity

- “You shall know a word by the company it keeps” [Firth, 1957]
- Simple single-word (lexical semantics) example:
“duty” vs “responsibility”
adj. modification, verbs they’re arguments of?
 - *duty* can be modified by adjectives such as *additional*, *administrative*, *assigned*, *assumed*, *collective*, *congressional*, *constitutional*, ..., so can *responsibility*;
 - *duty* can be the object of verbs such as *accept*, *articulate*, *assert*, *assign*, *assume*, *attend to*, *avoid*, *become*, *breach*, ..., so can *responsibility*.