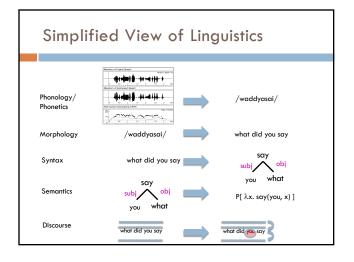
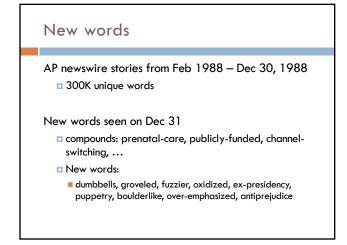
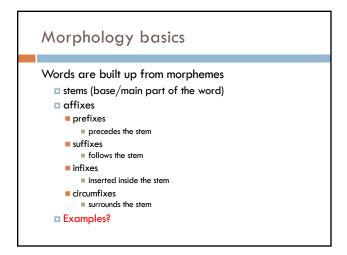


Admin Assignment 2 Quiz #1 Monday First 30 minutes of class (show up on time!) Everything up to, but not including today

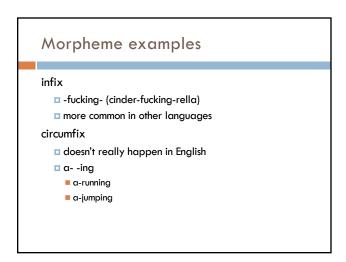


Morphology What is morphology? study of the internal structure of words morph-ology word-s jump-ing Why might this be useful for NLP? generalization (runs, running, runner are related) additional information (it's plural, past tense, etc) allows us to handle words we've never seen before smoothing?





prefix circum- (circumnavigate) dis- (dislike) mis- (misunderstood) com-, de-, dis-, in-, re-, post-, trans-, ... suffix -able (movable) -ance (resistance) -ly (quickly) -tion, -ness, -ate, -ful, ...



Agglutinative: Finnish

talo 'the-house' kaup-pa 'the-shop'
talo-ni 'my house' kaup-pa-ni 'my shop'
talo-ssa 'in the-house' kaup-a-ssa 'in the-shop'
talo-i-ssa-ni 'in my house' kaup-a-ssa-ni 'in my shop'
talo-i-ssa-ni 'in my houses' kaup-o-i-ssa 'in the-shops'
talo-i-ssa-ni 'in my houses' kaup-o-i-ssa-ni 'in my shops'

Stemming (baby lemmatization) Reduce a word to the main stem/morpheme automate automates automatic automatic automation run runs running

Stemming example

This is a poorly constructed example using the Porter stemmer.

This is a poorli construct example $\operatorname{\mathsf{us}}$ the Porter stemmer.

https://text-processing.com/demo/stem (or you can download versions online)

Porter's algorithm (1980)

Most common algorithm for stemming English

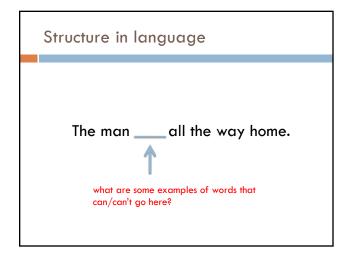
Results suggest it's at least as good as other stemming options

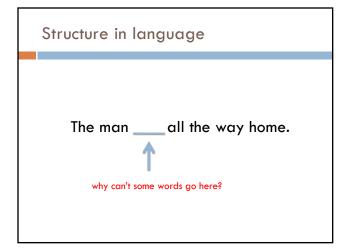
Multiple sequential phases of reductions using rules, e.g.

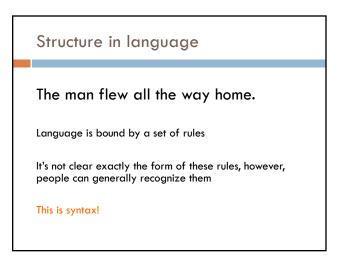
- ${\color{red}\square} \; \mathsf{sses} \to \mathsf{ss}$
- \square ies \rightarrow i
- ${\color{red}\square} \text{ ational} \rightarrow \text{ate}$
- \square tional \rightarrow tion

 $\underline{\text{http://tartarus.org/}{\sim}\text{martin/PorterStemmer/}}$

What is Syntax? Study of the structure of language Examine the rules of how words interact and go together Rules governing grammaticality I will give you one perspective no single correct theory of syntax still an active field of research in linguistics we will often use it as a tool/stepping stone for other applications







Syntax != Semantics

Colorless green ideas sleep furiously.

Syntax is only concerned with how words interact from a grammatical standpoint, not semantically (i.e. meaning)

Parts of speech What are parts of speech (think 3rd grade)?

Parts of speech

Parts of speech are constructed by grouping words that function similarly:

- with respect to the words that can occur nearby
- and by their morphological properties

The man ____ all the way home.

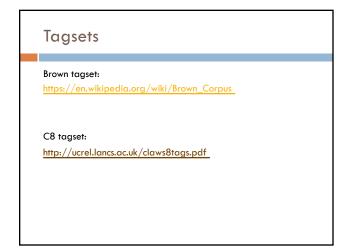
ran integrated washed forgave programmed ate drove shot walked shouted spoke succeeded survived learned understood read voted recorded

Parts of speech

What are the English parts of speech?

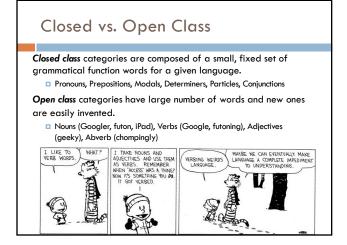
- 8 parts of speech?
 - Noun (person, place or thing)
 - Verb (actions and processes)
 - Adjective (modify nouns)
 - Adverb (modify verbs)
 - Preposition (on, in, by, to, with)
 - Determiners (a, an, the, what, which, that)
 - Conjunctions (and, but, or)
 - Particle (off, up)

English parts of speech Brown corpus: 87 POS tags Penn Treebank: ~45 POS tags Derived from the Brown tagset Most common in NLP Many of the examples we'll show us this one British National Corpus (C5 tagset): 61 tags C6 tagset: 148 C7 tagset: 146 C8 tagset: 171



English Parts of Speech Noun (person, place or thing) □ Singular (NN): dog, fork □ Plural (NNS): dogs, forks □ Proper (NNP, NNPS): John, Springfields □ Personal pronoun (PRP): I, you, he, she, it □ Wh-pronoun (WP): who, what Verb (actions and processes) □ Base, infinitive (VB): eat Past tense (VBD): ate □ Gerund (VBG): eating Past participle (VBN): eaten □ Non 3rd person singular present tense (VBP): eat $\hfill \ensuremath{\text{\footnotemath{\text{gr}}}} \ensuremath{3^{\text{rd}}}$ person singular present tense: (VBZ): eats □ Modal (MD): should, can □ To (TO): to (to eat)

English Parts of Speech (cont.) Adjective (modify nouns) Basic (JJ): red, tall Comparative (JJR): redder, taller Superlative (JJS): reddest, tallest Adverb (modify verbs) Basic (RB): quickety Comparative (RBR): quicker Superlative (RBS): quickest Preposition (IN): on, in, by, to, with Determiner: Basic (DT) a, an, the WH-determiner (WDT): which, that Coordinating Conjunction (CC): and, but, or, Particle (RP): off (took off), up (put up)



Part of speech tagging

Annotate each word in a sentence with a part-ofspeech marker

Lowest level of syntactic analysis

John saw the saw and decided to take it to the table. NNP VBD DT NN CC VBD TO VB PRP IN DT NN

Ambiguity in POS Tagging

I like candy.

(verb, non-3rd person, singular, present)

Time flies like an arrow.

(preposition)

Does "like" play the same role (POS) in these sentences?

Ambiguity in POS Tagging

I bought it at the shop around the corner.

I never got around to getting the car.

(particle... on, off)

The cost of a new Prius is around \$25K.

(adverb)

Does "around" play the same role (POS) in these sentences?

Ambiguity in POS tagging

Like most language components, the challenge with POS tagging is ambiguity

Brown corpus analysis

- 11.5% of word types are ambiguous (this sounds promising!), but...
- □ 40% of word appearances are ambiguous
- □ Unfortunately, the ambiguous words tend to be the more frequently used words

How hard is it?

If I told you had a POS tagger that achieved 90%accuracy would you be impressed?

 $\hfill\Box$ Shouldn't be... just picking the most frequent POS for a word gets you this

What about a POS tagger that achieves 93.7%?

□ Still probably shouldn't be... only need to add a basic module for handling unknown words

What about a POS tagger that achieves 100%?

- □ Should be suspicious... humans only achieve ~97%
- □ Probably overfitting (or cheating!)

POS Tagging Approaches

Rule-Based: Human crafted rules based on lexical and other linguistic knowledge

Learning-Based: Trained on human annotated corpora like the Penn Treebank

- Statistical models: Hidden Markov Model (HMM), Maximum Entropy
 Markov Model (MEMM), Conditional Random Field (CRF), log-linear
 models, support vector machines (SVMs), neural networks
 Rule learning: Transformation Based Learning (TBL)

The book discusses some of the more common approaches

Many publicly available:

- http://nlp.stanford.edu/links/statnlp.html (list 15 different ones mostly publicly available!)
- http://www.coli.uni-saarland.de/~thorsten/tnt/

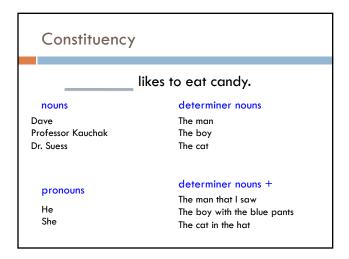
Constituency

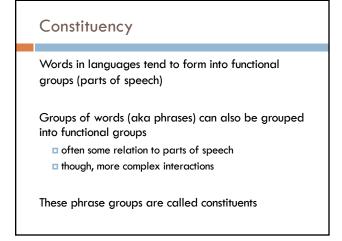
Parts of speech can be thought of as the lowest level of syntactic

Groups words together into categories

likes to eat candy.

What can/can't go here?





Common constituents

He likes to eat candy.

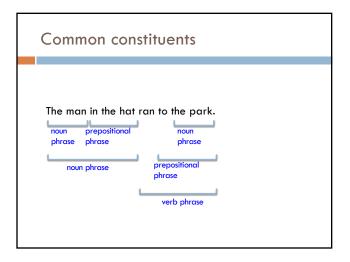
noun phrase verb phrase

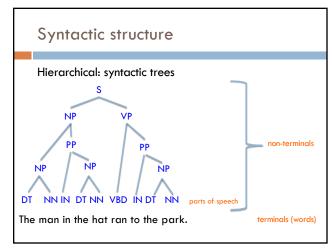
The man in the hat ran to the park.

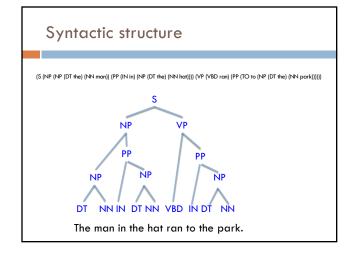
noun phrase verb phrase

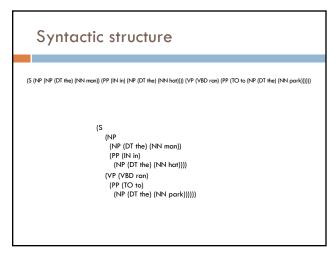
The man in the hat ran to the park.

noun prepositional phrase phrase phrase phrase verb phrase





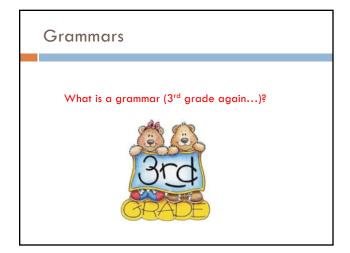




Syntactic structure

A number of related problems:

- □ Given a sentence, can we determine the syntactic structure?
- □ Can we determine if a sentence is grammatical?
- □ Can we determine how *likely* a sentence is to be grammatical? to be an English sentence?
- □ Can we generate candidate, grammatical sentences?



Grammars

Grammar is a set of structural rules that govern the composition of sentences, phrases and words

Lots of different kinds of grammars:

- regular
- □ context-free
- context-sensitive
- □ recursively enumerable
- $\ \square$ transformation grammars



Context free grammar

How many people have heard of them?

Look like:

 $S \rightarrow NP VP$

left hand side right hand side (single symbol) (one or more symbols)

Formally...

G = (NT, T, P, S)

NT: finite set of nonterminal symbols

T: finite set of terminal symbols, NT and T are disjoint

P: finite set of productions of the form $\mathsf{A} \to \alpha \text{, } \mathsf{A} \in \mathsf{NT} \text{ and } \alpha \in (\mathsf{T} \cup \mathsf{NT})^*$

 $S \in NT: start\ symbol$

CFG: Example

Many possible CFGs for English, here is an example (fragment):

 $S \rightarrow NP VP$

 $VP \rightarrow V NP$

 $NP \rightarrow DetP \ N \mid DetP \ AdjP \ N$

 $\mathrm{AdjP} \rightarrow \, \mathrm{Adj} \mid \mathrm{Adv} \, \mathrm{AdjP}$

 $N \rightarrow boy \mid girl$

 $V \rightarrow sees \mid likes$

 $\mathsf{Adj} \to \mathsf{big} \mid \mathsf{small}$

 $\mathsf{Adv} \to \mathsf{very}$

 $\text{DetP} \rightarrow \, \alpha \mid \text{the}$

CFG: Example

Many possible CFGs for English, here is an example (fragment):

 $S \rightarrow NP VP$

 $\mathrm{VP} \to \, \mathrm{V} \, \mathrm{NP}$

 $NT: \{S,\,NP,\,VP,\,DetP,\,N,\,AdjP,\,Adj,\,Adv\}$

T: {boy, girl, sees, likes, big, small, very, a, the}

 $\mathsf{NP} \to \mathsf{DetP}\;\mathsf{N}\;|\;\mathsf{DetP}\;\mathsf{AdjP}\;\mathsf{N}$

 $AdjP \rightarrow Adj \mid Adv AdjP$

 $N \rightarrow boy \mid girl$ $V \rightarrow sees \mid likes$

v / sees | likes

 $Adj \rightarrow \ big \ | \ small \qquad \qquad S: \ S$

 $\mathsf{Adv} \to \mathsf{very}$

 $\mathrm{DetP} \rightarrow \, \mathrm{a} \mid \mathrm{the}$

Often just specify the production rules

Grammar questions

Can we determine if a sentence is grammatical?

Given a sentence, can we determine the syntactic structure?

Can we determine how likely a sentence is to be grammatical? to be an English sentence?

Can we generate candidate, grammatical sentences?

Which of these can we answer with a CFG? How?

Grammar questions

Can we determine if a sentence is grammatical?

- $\hfill \square$ Is it accepted/recognized by the grammar
- □ Applying rules right to left, do we get the start symbol?

Given a sentence, can we determine the syntactic structure?

■ Keep track of the rules applied...

Can we determine how likely a sentence is to be grammatical? to be an

■ Not yet... no notion of "likelihood" (probability)

Can we generate candidate, grammatical sentences?

■ Start from the start symbol, randomly pick rules that apply (i.e. left hand side matches)