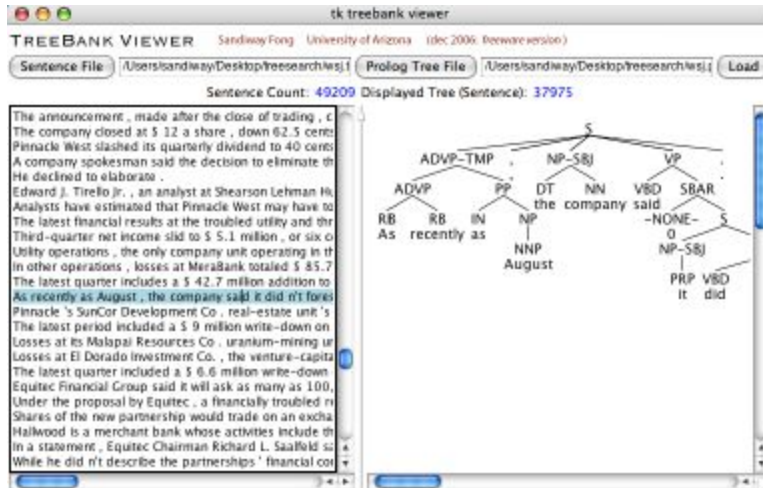


# Part-of-speech tagging



A simple but useful form of linguistic analysis

## Open class (lexical) words

### Nouns

#### Proper

*IBM*  
*Italy*

#### Common

*cat / cats*  
*snow*

### Verbs

#### Main

*see*  
*registered*

### Adjectives

*old older oldest*

### Adverbs

*slowly*

### Numbers

*122,312*  
*one*

*... more*

## Closed class (functional)

Determiners *the some*

Conjunctions *and or*

Pronouns *he its*

### Modals

*can*  
*had*

Prepositions *to with*

Particles *off up*

*... more*

Interjections *Ow Eh*

# Open vs. Closed classes

- Open vs. Closed classes
  - Closed:
    - determiners: *a, an, the*
    - pronouns: *she, he, I*
    - prepositions: *on, under, over, near, by, ...*
    - Why “closed”?
  - Open:
    - Nouns, Verbs, Adjectives, Adverbs.

# POS Tagging

- Words often have more than one POS: *back*
  - The back door = JJ
  - On my back = NN
  - Win the voters back = RB
  - Promised to back the bill = VB
- The POS tagging problem is to determine the POS tag for a particular instance of a word.

	Tag	Description	Example
Open Class	<b>ADJ</b>	Adjective: noun modifiers describing properties	<i>red, young, awesome</i>
	<b>ADV</b>	Adverb: verb modifiers of time, place, manner	<i>very, slowly, home, yesterday</i>
	<b>NOUN</b>	words for persons, places, things, etc.	<i>algorithm, cat, mango, beauty</i>
	<b>VERB</b>	words for actions and processes	<i>draw, provide, go</i>
	<b>PROPN</b>	Proper noun: name of a person, organization, place, etc..	<i>Regina, IBM, Colorado</i>
	<b>INTJ</b>	Interjection: exclamation, greeting, yes/no response, etc.	<i>oh, um, yes, hello</i>
Closed Class Words	<b>ADP</b>	Adposition (Preposition/Postposition): marks a noun's spacial, temporal, or other relation	<i>in, on, by, under</i>
	<b>AUX</b>	Auxiliary: helping verb marking tense, aspect, mood, etc.,	<i>can, may, should, are</i>
	<b>CCONJ</b>	Coordinating Conjunction: joins two phrases/clauses	<i>and, or, but</i>
	<b>DET</b>	Determiner: marks noun phrase properties	<i>a, an, the, this</i>
	<b>NUM</b>	Numeral	<i>one, two, first, second</i>
	<b>PART</b>	Particle: a preposition-like form used together with a verb	<i>up, down, on, off, in, out, at, by</i>
	<b>PRON</b>	Pronoun: a shorthand for referring to an entity or event	<i>she, who, I, others</i>
Other	<b>SCONJ</b>	Subordinating Conjunction: joins a main clause with a subordinate clause such as a sentential complement	<i>that, which</i>
	<b>PUNCT</b>	Punctuation	<i>, , ()</i>
	<b>SYM</b>	Symbols like \$ or emoji	<i>\$, %</i>
	<b>X</b>	Other	<i>asdf, qwfg</i>

Tag	Description	Example	Tag	Description	Example	Tag	Description	Example
CC	coord. conj.	<i>and, but, or</i>	NNP	proper noun, sing.	<i>IBM</i>	TO	“to”	<i>to</i>
CD	cardinal number	<i>one, two</i>	NNPS	proper noun, plu.	<i>Carolinas</i>	UH	interjection	<i>ah, oops</i>
DT	determiner	<i>a, the</i>	NNS	noun, plural	<i>llamas</i>	VB	verb base	<i>eat</i>
EX	existential ‘there’	<i>there</i>	PDT	predeterminer	<i>all, both</i>	VBD	verb past tense	<i>ate</i>
FW	foreign word	<i>mea culpa</i>	POS	possessive ending	<i>’s</i>	VBG	verb gerund	<i>eating</i>
IN	preposition/ subordin-conj	<i>of, in, by</i>	PRP	personal pronoun	<i>I, you, he</i>	VBN	verb past partici- ple	<i>eaten</i>
JJ	adjective	<i>yellow</i>	PRP\$	possess. pronoun	<i>your, one’s</i>	VBP	verb non-3sg-pr	<i>eat</i>
JJR	comparative adj	<i>bigger</i>	RB	adverb	<i>quickly</i>	VBZ	verb 3sg pres	<i>eats</i>
JJS	superlative adj	<i>wildest</i>	RBR	comparative adv	<i>faster</i>	WDT	wh-determ.	<i>which, that</i>
LS	list item marker	<i>1, 2, One</i>	RBS	superlatv. adv	<i>fastest</i>	WP	wh-pronoun	<i>what, who</i>
MD	modal	<i>can, should</i>	RP	particle	<i>up, off</i>	WP\$	wh-possess.	<i>whose</i>
NN	sing or mass noun	<i>llama</i>	SYM	symbol	<i>+, %, &amp;</i>	WRB	wh-adverb	<i>how, where</i>

# POS Tagging

- Input: Plays well with others
- Ambiguity: NNS/VBZ UH/JJ/NN/RB IN NNS
- Output: Plays/VBZ well/RB with/IN others/NNS
- Uses:
  - Text-to-speech (how do we pronounce “lead”?)
  - Can write regexps like (Det) Adj\* N+ over the output for phrases, etc.
  - As input to or to speed up a full parser
  - If you know the tag, you can back off to it in other tasks

Penn  
Treebank  
POS tags

# POS tagging performance

- How many tags are correct? (Tag accuracy)
  - About 97% currently
  - But baseline is already 90%
    - Baseline is performance of stupidest possible method
      - Tag every word with its most frequent tag
      - Tag unknown words as nouns
- Partly easy because
  - Many words are unambiguous
  - You get points for them (*the*, *a*, etc.) and for punctuation marks!



# Deciding on the correct part of speech can be difficult even for people

- Mrs/NNP Shaefer/NNP never/RB got/VBD around/RP to/TO joining/VBG
- All/DT we/PRP gotta/VBN do/VB is/VBZ go/VB around/IN the/DT corner/NN
- Chateau/NNP Petrus/NNP costs/VBZ around/RB 250/CD

# How difficult is POS tagging?

- About 11% of the word types in the Brown corpus are ambiguous with regard to part of speech
- But they tend to be very common words. E.g., *that*
  - I know *that* he is honest = IN
  - Yes, *that* play was nice = DT
  - You can't go *that* far = RB
- 40% of the word tokens are ambiguous

# Sources of information

- What are the main sources of information for POS tagging?
  - Knowledge of neighboring words
    - Bill saw that man yesterday
    - NNP NN DT NN NN
    - VB VB(D) IN VB NN
  - Knowledge of word probabilities
    - *man* is rarely used as a verb....
- The latter proves the most useful, but the former also helps

# More and Better Features □ Feature-based tagger

- Can do surprisingly well just looking at a word by itself:
  - Word            the: the → DT
  - Lowercased word    Importantly: importantly → RB
  - Prefixes            unfathomable: un- → JJ
  - Suffixes            Importantly: -ly → RB
  - Capitalization    Meridian: CAP → NNP
  - Word shapes    35-year: d-x → JJ
- Then build a maxent (or whatever) model to predict tag
  - Maxent  $P(t|w)$ :    93.7% overall / 82.6% unknown

# Overview: POS Tagging Accuracies

- Rough accuracies:

- Most freq tag:

~90% / ~50%

- Trigram HMM:

~95% / ~55%

- Maxent  $P(t|w)$ :

93.7% / 82.6%

Most errors  
on unknown  
words

