# CSE528 <br> <br> Natural Language Processing <br> <br> Natural Language Processing <br> Venue:ADB-405 <br> Topic: PartsOfSpeach Tagging 

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## Definition

The process of assigning a part-of-speech or other lexical class marker to each word in a corpus.


## Definition

$\square$ Annotate each word in a sentence with a part-of-speech marker.
$\square$ Lowest level of syntactic analysis.
$\square$ Useful for subsequent syntactic parsing and word sense disambiguation.
$\square$ Example
John saw the saw and decided to take it to the table. NNP VBD DT NN CC VBD TO VB PRPIN DT NN

## An Example

| WORD | LEMMA | TAG |
| :--- | :--- | :--- |
|  |  |  |
| the | the | +DET |
| girl | girl | +NOUN |
| kissed | kiss | +VPAST |
| the | the | +DET |
| boy | boy | +NOUN |
| on | on | +PREP |
| the | the | +DET |
| cheek | cheek | + NOUN |

## English POS Tagsets

$\square$ Original Brown corpus used a large set of 87 POS tags.
$\square$ Most common in NLP today is the Penn Treebank set of 45 tags.
$\square$ Reduced from the Brown set for use in the context of a parsed corpus (i.e. treebank).
$\square$ The C5 tagset used for the British National Corpus (BNC) has 61 tags.

## Word Classes

Basic word classes: Noun, Verb, Adjective, Adverb, Preposition, ...
Open vs. Closed classes

- Open:
- Nouns, Verbs, Adjectives, Adverbs.
- Why "open"?
- Closed:
- determiners: a, an, the
- pronouns: she, he, I
- prepositions: on, under, over, near, by, ...


## Closed vs. Open Class

Closed class categories are composed of a small, fixed set of grammatical function words for a given language.
$\square$ prepositions: on, under, over, ...
$\square$ particles: up, down, on, off, ...
$\square$ determiners: a, an, the, ...
$\square$ pronouns: she, who, I, ..
$\square$ conjunctions: and, but, or, ...
$\square$ auxiliary verbs: can, may should, ...

## Closed vs. Open Class

Open class categories have large number of words and new ones are easily invented.
$\square$ Nouns new nouns: Internet, website, URL, CD-ROM, email, newsgroup, bitmap, modem, multimedia
-New verbs have also : download, upload, reboot, right-click, doubleclick,
$\square$ Verbs (Google),
$\square$ Adjectives (geeky)
$\square$ Abverb (chompingly)

## English Parts of Speech (Nouns)

Noun (person, place or thing)
$\square$ Singular (NN): dog, fork
$\square$ Plural (NNS): dogs, forks
$\square$ Proper (NNP, NNPS): John, Springfields
$\square$ Personal pronoun (PRP): I, you, he, she, it
$\square$ Wh-pronoun (WP): who, what

## English Parts of Speech (Nouns)

Proper nouns (Penn, Philadelphia, Davidson)
$\square$ English capitalizes these.
Common nouns (the rest).
Count nouns and mass nouns
$\square$ Count: have plurals, get counted: goat/goats,
Mass: don't get counted (snow, salt, water,)

## English Parts of Speech (Verbs)

Verb (actions and processes)
$\square$ Base, infinitive (VB): eat
$\square$ Past tense (VBD): ate
$\square$ Gerund (VBG): eating
$\square$ Past participle (VBN): eaten
$\square$ Non $3^{\text {rd }}$ person singular present tense (VBP): eat
$\square 3^{\text {rd }}$ person singular present tense: (VBZ): eats
$\square$ Modal (MD): should, can
$\square$ To (TO): to (to eat)

## English Parts of Speech (Adjectives)

Adjective (modify nouns, identify properties or qualities of nouns)
$\square$ Basic (JJ): red, tall
$\square$ Comparative (JJR): redder, taller
$\square$ Superlative (JJS): reddest, tallest
Adjective ordering restrictions in English:
$\square$ Old blue book, not Blue old book
$\square$ the 44th president
$\square$ a green product
$\square$ a responsible investment
$\square$ the dumbest, worst leader

## English Parts of Speech (Adverbs)

Adverb (modify verbs)
$\square$ Basic (RB): quickly
$\square$ Comparative (RBR): quicker
$\square$ Superlative (RBS): quickest
Unfortunately, John walked home extremely slowly yesterday
$\square$ Directional/locative adverbs (here, downhill)
$\square$ Degree adverbs (extremely, very, somewhat)
$\square$ Manner adverbs (slowly, slinkily, delicately)
$\square$ Temporal adverbs (yesterday, tomorrow)

## English Parts of Speech (Determiner)

Is a word that occurs together with a noun or noun phrase and serves to express the reference of that noun or noun phrase in the context.
That is, a determiner may indicate whether the noun is referring to a definite or indefinite element of a class, to a closer or more distant element, to an element belonging to a specified person or thing, to a particular number or quantity, etc.

## English Parts of Speech(Determiner)

Common kinds of determiners include
$\square$ definite and indefinite articles (the, a, an)
$\square$ demonstratives (this, that, these)
$\square$ possessive determiners (my, their)
$\square$ quantifiers (many, few , several).

## English Parts of Speech ( preposition)

Preposition (IN): a word governing, and usually preceding, a noun or pronoun and expressing a relation to another word or element in the clause, as in 'the man on the platform', 'she arrived after dinner'.

Ex: on, in, by, to, with

## English Parts of Speech

Coordinating Conjunction (CC): that connects words, sentences, phrases or clauses.
the truth of nature, and the power of giving interest
Ex: and, but, or.
Particle (RP): a particle is a function word that must be associated with another word or phrase to impart meaning, i.e., does not have its own lexical definition.

Ex: off (took off), up (put up)

## POS tagging

$\square$ POS Tagging is a process that attaches each word in a sentence with a suitable tag from a given set of tags.
$\square$ Tagging is the assignment of a single part-of-speech tag to each word (and punctuation marker) in a corpus.
$\square$ The set of tags is called the Tag-set.
$\square$ Standard Tag-set : Penn Treebank (for English).

## POS tagging

$\square$ There are so many parts of speech, potential distinctions we can draw.
$\square$ To do POS tagging, we need to choose a standard set of tags to work with.
$\square$ Could pick very coarse tag sets.
$\square \mathrm{N}, \mathrm{v}, \mathrm{Adj}, \mathrm{Adv}$.
$\square$ More commonly used set is finer grained (Penn TreeBank, 45 tags) $\square$ PRP\$, WRB, WP\$, VBG

## POS Tag Ambiguity

$\square$ Deciding on the correct part of speech can be difficult even for people.
$\square$ In English : I bank1 on the bank2 on the river bank3 for my transactions.
$\square$ Bank1 is verb, the other two banks are nouns
$\square$ In Hindi :
$\square$ "Khaanaa" : can be noun (food) or verb (to eat)

## Measuring Ambiguity

|  | 87-tag Original Brown | 45-tag Treebank Brown |
| :---: | :---: | :---: |
| Unambiguous (1 tag) | $\mathbf{4 4 , 0 1 9}$ | $\mathbf{3 8 , 8 5 7}$ |
| Ambiguous (2-7 tags) | $\mathbf{5 , 4 9 0}$ | $\mathbf{8 8 4 4}$ |
| Details: 2 tags | 4,967 | 6,731 |
|  | 3 tags | 411 |
| 4 tags | 91 | 1621 |
| 5 tags | 17 | 357 |
| 6 tags | 2 (well, beat) | 90 |
| 7 tags | 2 (still, down) | 32 |
|  |  | 6 (well, set, round, |
|  |  | 4 open, fit, down) |
|  |  | 3 ('s, half, back, a) |
| 9 tags |  | 3 (that, more, in) |

## How Hard is POS Tagging?

$\square$ About $11 \%$ of the word types in the Brown corpus are ambiguous with regard to part of speech
$\square$ But they tend to be very common words
$\square 40 \%$ of the word tokens are ambiguous

## Penn TreeBank POS Tagset

| Tag | Description | Example | Tag | Description | Example |
| :---: | :---: | :---: | :---: | :---: | :---: |
| CC | coordin. conjunction | and, but, or | SYM | symbol | $+, \%, d$ |
| CD | cardinal number | one, two, three | TO | "to" | to |
| DT | determiner | a, the | UH | interjection | ah, oops |
| EX | existential 'there' | there | VB | verb, base form | eat |
| FW | foreign word | mea culpa | VBD | verb, past tense | ate |
| IN | preposition/sub-conj | of, in, by | VBG | verb, gerund | eating |
| JJ | adjective | yellow | VBN | verb, past participle | eaten |
| JJR | adj-, comparative | bigger | VBP | verb, non-3sg pres | eat |
| JJS | adj., superlative | wildest | VBZ | verb, 3sp pres | eats |
| LS | list item marker | 1, 2, One | WDT | wh-determiner | which, that |
| MD | modal | can, should | WP | wh-pronoun | what, who |
| NN | noun, sing. or mass | llama | WPS | possessive wh- | whose |
| NNS | noun, plural | llamas | WRB | wh-adverb | how, where |
| NNP | proper noun, singular | IBM | \$ | dollar sign | \$ |
| NNPS | proper noun, plural | Carolinas | \# | pound sign | \# |
| PDT | predeterminer | all, both | c | left quote | c Or ${ }^{\circ}$ |
| POS | possessive ending | 's | 3 | right quote | ${ }^{\text {, }} \mathrm{Or}{ }^{\prime}$ |
| PRP | personal pronoun | I, you, he | ( | left parenthesis | $[, C,\{,<$ |
| PRP\$ | possessive pronoun | your, one's | $)$ | right parenthesis | $]),,\},>$ |
| RB | adverb | quickly, never | , | comma |  |
| RBR | adverb, comparative | faster | . | sentence-final punc | - ! ? |
| RBS | adverb, superlative | fastest | : | mid-sentence punc | : ; ... - |
| RP | particle | up, off |  |  |  |

## Using the Penn Tagset

$\square$ The/DT grand/JJ jury/NN commmented/VBD on/IN a/DT number/NN of/IN other/JJ topics/NNS ./.
$\square$ Prepositions and subordinating conjunctions marked IN ("although/IN I/PRP..")
$\square$ Except the preposition/complementizer "to" is just marked "TO".

## Process

$\square$ List all possible tag for each word in sentence.
$\square$ Choose best suitable tag sequence.
$\square$ Example
$\square$ "People jump high".
$\square$ People : Noun/Verb
$\square$ jump : Noun/Verb
$\square$ high : Noun/Verb/Adjective
$\square$ We can start with probabilities.

## Example



## Why POS

$\square$ POS tell us a lot about a word (and the words near it).
$\square$ E.g, adjectives often followed by nouns
$\square$ personal pronouns often followed by verbs
$\square$ possessive pronouns by nouns
$\square$ Pronunciations depends on POS, e.g.
$\square$ object (first syllable NN, second syllable VM), content, discount
$\square$ First step in many NLP applications

## Rule-Based Tagging

$\square$ Start with a dictionary.
$\square$ Assign all possible tags to words from the dictionary.
$\square$ Write rules by hand to selectively remove tags.
$\square$ Leaving the correct tag for each word.

## Step1: Start with a Dictionary

| she: | PRP |
| :--- | :--- |
| promised: | VBN,VBD |
| to: | TO |
| back: | VB, JJ, RB, NN |
| the: | DT |
| bill: | NN, VB |
| Etc... for the $\sim 100,000$ words of English with more than 1 tag |  |

## Step2: Assign Every Possible Tag



## Step3: Write Rules to Eliminate Tags

Eliminate VBN if VBD is an option when VBN|VBD follows "<start> PRP"

NN

|  |  | RB |  |  |
| :--- | :--- | :--- | :--- | :--- |
|  | VBN |  | JJ | VB |
| PRP | VBD | TO | VB | DT |
| She | promised | to | back | the |
| Sill |  |  |  |  |

## Simply assign each word its most likely POS.

Success rate: 91\%!

| Word | POS listings in Brown |  |  |
| :---: | :---: | :---: | :---: |
| heat | noun/89 | verb/5 |  |
| oil | noun/87 |  |  |
| in | prep/20731 | noun/1 | adv/462 |
| a | det/22943 | noun/50 | noun-proper/30 |
| large | adj/354 | noun/2 | adv/5 |
| pot | noun/27 |  |  |

## END

