

Part-of-Speech Tagging

Parts of Speech (POS)

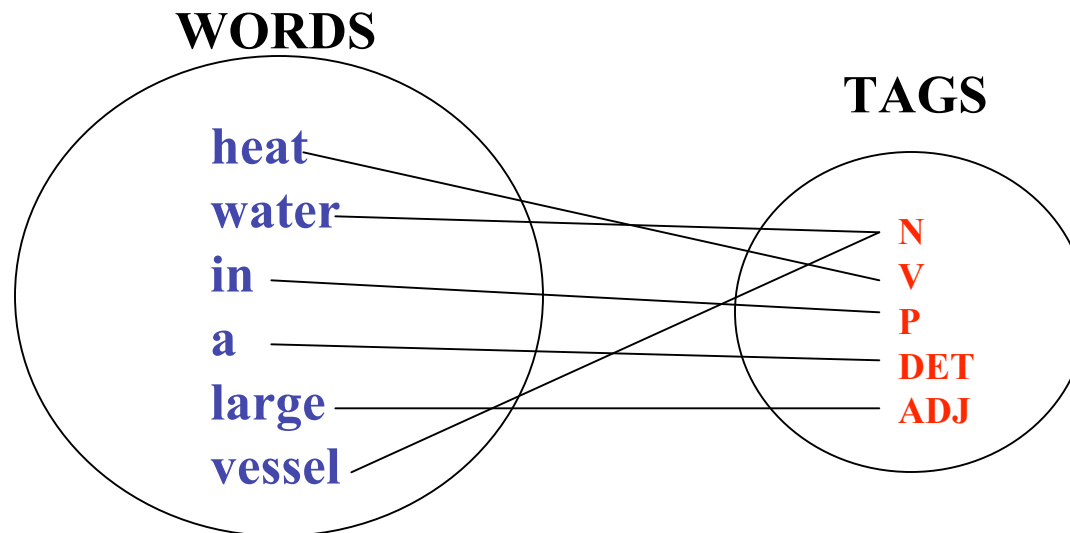
- Categories to which words are assigned according to their function.
 - Noun, verb, adjective, preposition, adverb, article, pronoun, conjunction, etc.
 - This idea has been around for over 2000 years (Dionysius Thrax of Alexandria, 100 B.C.)
 - Called: parts-of-speech, lexical category, word classes, morphological classes, lexical tags, POS

POS examples

- N noun chair, bandwidth, pacing
- V verb study, debate, munch
- ADJ adjective purple, tall, ridiculous
- ADV adverb unfortunately, slowly,
- P preposition of, by, to
- PRO pronoun I, me, mine
- DET determiner the, a, that, those

POS Tagging

- The process of assigning a part-of-speech to each word in a sentence



Example

Word

Tag

heat

verb (noun)

water

noun (verb)

in

prep (noun, adv)

a

det (noun)

large

adj (noun)

vessel

noun

What is POS tagging good for?

- Useful in
 - Information Retrieval
 - Text to Speech: OBject(N) vs. obJECT(V);
DIScount(N) vs. disCOUNT(V)
 - Word Sense Disambiguation
- Useful as a preprocessing step of parsing
 - Unique tag to each word reduces the number of parses

Choosing a tagset

- Need to choose a standard set of tags to do POS tagging
 - One tag for each parts-of-speech
- Could pick very coarse tagset
 - N, V, Adj, Adv, Prep.
- More commonly used set is finer-grained
 - E.g., the *UPeen TreeBank* tagset, 45 tags
 - PRP, PRP\$, VBG, VBD, JJR, JJS ...
- Even more finely-grained tagsets exist

Why is POS tagging hard?

- Ambiguity
 - **Plants**/N need light and water.
 - Each one **plant**/V one.
- The number of tags used by different systems varies a lot. Some systems use < 20 tags while others use >70.

Methods for POS tagging

- Rule-Based POS tagging
 - ENGTWOL [Voutilainen, 1995]
- Stochastic (Probabilistic) tagging
 - TNT [Brants, 2000]
- Transformation-based tagging
 - Brill's tagger [Brill, 1995]

Stochastic Tagging

- Based on probability of certain tag occurring, given various possibilities
 - Necessitates a *training corpus*
 - A collection of sentences that have already been tagged
 - Several such corpora exist

Approach 1

- Assign each word its most likely POS tag
 - Success: 91% for English

- Example

heat :: noun/89, verb/5

Approach 2

- Given: sequence of words

$$W = w_1, w_2, \dots, w_n \text{ (a sentence)}$$

- Assign tags:

$$T = t_1, t_2, \dots, t_n$$

- Find T that maximizes $P(T | W)$

Practical Statistical Tagger

- By Bayes' Rule,

$$P(T | W) = P(W|T) P(T) / P(W) = \alpha P(W|T) P(T)$$

- So find T that maximizes $P(T | W) P(T)$

$$P(T) = P(t_1) P(t_2 | t_1) P(t_3 | t_1, t_2) \dots P(t_n | t_1, t_2, \dots t_{n-1})$$

– As an approximation, use

$$P(T) \approx P(t_1) P(t_2 | t_1) P(t_3 | t_2) \dots P(t_n | t_{n-1})$$

- Assume each word is dependent only on its own POS tag: given its POS tag, it is independent of the other words around it. Then

$$P(W|T) = P(w_1 | t_1) P(w_2 | t_2) \dots P(w_n | t_n)$$

- So

$$P(T) P(W|T) \approx P(t_1) P(t_2|t_1) \dots P(t_n|t_{n-1}) P(w_1|t_1) P(w_2|t_2) \dots P(w_n|t_n)$$

Getting the Conditional Probabilities

- How do we compute $P(w_i|t_i)$?
 - $c(w_i, t_i)/c(t_i)$
- How do we compute $P(t_i|t_{i-1})$?
 - $c(t_{i-1}, t_i)/c(t_{i-1})$

Example

- Secretariat/NNP is/VBZ expected/VBN **to/TO race/VB** tomorrow/NN
 - to/TO race/???
- People/NNS continue/VBP to/TO inquire/VB the DT reason/NN for/IN **the/DT race/NN** for/IN outer/JJ space/NN
 - the/DT race/???
- $t_i = \operatorname{argmax}_j P(t_j|t_{i-1})P(w_i|t_j)$
 - i = number of word in sequence, j = number among possible tags
 - $\max[P(\text{VB}|\text{TO})P(\text{race}|\text{VB}) , P(\text{NN}|\text{TO})P(\text{race}|\text{NN})]$
- From the Brown corpus
 - $P(\text{NN}|\text{TO}) = .021$ $P(\text{race}|\text{NN}) = .00041$
 - $P(\text{VB}|\text{TO}) = .34$ $P(\text{race}|\text{VB}) = .00003$
- So
 - $P(\text{NN}|\text{TO}) P(\text{race}|\text{NN}) = .021 \times .00041 = .000007$
 - $P(\text{VB}|\text{TO}) P(\text{race}|\text{VB}) = .34 \times .00003 = .00001$

