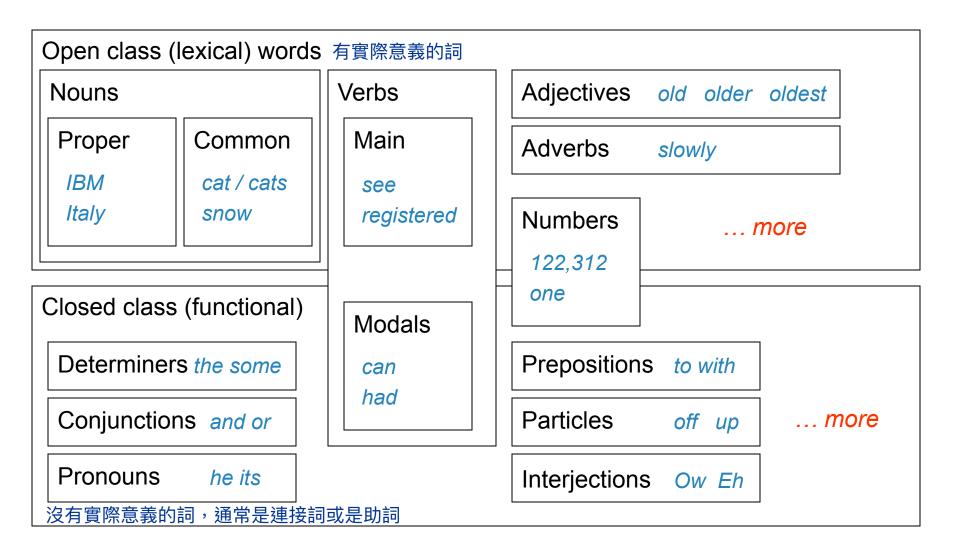
# Part-of-speech tagging

A simple but useful form of linguistic analysis



# Parts of Speech 詞性辨識

- Perhaps starting with Aristotle in the West (384–322 BCE), there
  was the idea of having parts of speech
  - a.k.a lexical categories, word classes, "tags", POS
- It comes from Dionysius Thrax of Alexandria (c. 100 BCE) the idea that is still with us that there are 8 parts of speech
  - But actually his 8 aren't exactly the ones we are taught today
    - Thrax: noun, verb, article, adverb, preposition, conjunction, participle, pronoun
    - School grammar: noun, verb, adjective, adverb, preposition, conjunction, pronoun, interjection





### **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**

同一個字可能會有很多詞性,所以要看前後文才能決定

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



## **POS Tagging**

一個字會有很多種詞性,POS Tagging的用途是找出正確的詞性組合

Penn

Treebank

POS tags

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



## **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

困難的POS Tagging句子例子

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

# Part-of-speech tagging

A simple but useful form of linguistic analysis

# Part-of-speech tagging revisited

A simple but useful form of linguistic analysis



### Sources of information

### **FEATURES**

- What are the main sources of information for POS tagging?
  - Knowledge of neighboring words
    - Bill saw that man yesterday
    - - 但有很多其實是不合理的
    - VB VB(D) IN VB NN 例如Bill很少當作VB、that後面不會接VB
  - Knowledge of word probabilities
    - man is rarely used as a verb....
- The latter proves the most useful, but the former also helps

Christopher Manning



# More and Better Features Feature-based tagger

雖然很多字有多種詞性,但很多詞其實用字根就可以直接看出詞性

- Can do surprisingly well just looking at a word by itself:
  - Word the: the  $\rightarrow$  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 中間有一個dash的就是形容詞
- 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:
  - Trigram HMM:
  - Maxent P(t|w): 最直覺的方法
  - TnT (HMM++):
  - MEMM tagger:
  - Bidirectional dependencies:
  - Upper bound:

overall / unknown words

~90% / ~50%

~95% / ~55%

93.7% / 82.6%

96.2% / 86.0%

96.9% / 86.9%

97.2% / 90.0%

~98% (human agreement)

Most errors on unknown words



## How to improve supervised results?

Build better features!

那要怎麼修正這個錯誤?
RB 可以在標記時多看下一個字
PRP VBD IN RB IN PRP VBD . as soon
They left as soon as he arrived . as是拿來修飾soon(RB)

所以會是副詞(RB)而非(IN)

答案應該是RB

但卻誤標為IN

We could fix this with a feature that looked at the next word

JJ對於沒看過的字IntrinsicNNPNNSVBDVBN因為首字母大寫,先標記成名詞Intrinsic flaws remained undetected<br/>就再改成形容詞

We could fix this by linking capitalized words to their lowercase versions



## **Tagging Without Sequence Information**



Model	Features	Token	Unknown	Sentence
Baseline	56,805	93.69%	82.61%	26.74%
3Words	239,767	96.57%	86.78%	48.27%

Using words only in a straight classifier works as well as a basic (HMM or discriminative) sequence model!!



# **Summary of POS Tagging**

- For tagging, the change from generative to discriminative model **does not by itself** result in great improvement
- One profits from models for specifying dependence on **overlapping features of the observation** such as spelling, suffix analysis, etc.
- An MEMM allows integration of rich features of the observations, but can suffer strongly from assuming independence from following observations; this effect can be relieved by adding dependence on following words
- This additional power (of the MEMM ,CRF, Perceptron models) has been shown to result in improvements in accuracy
- The **higher accuracy** of discriminative models comes at the price of **much** slower training

# Part-of-speech tagging revisited

A simple but useful form of linguistic analysis