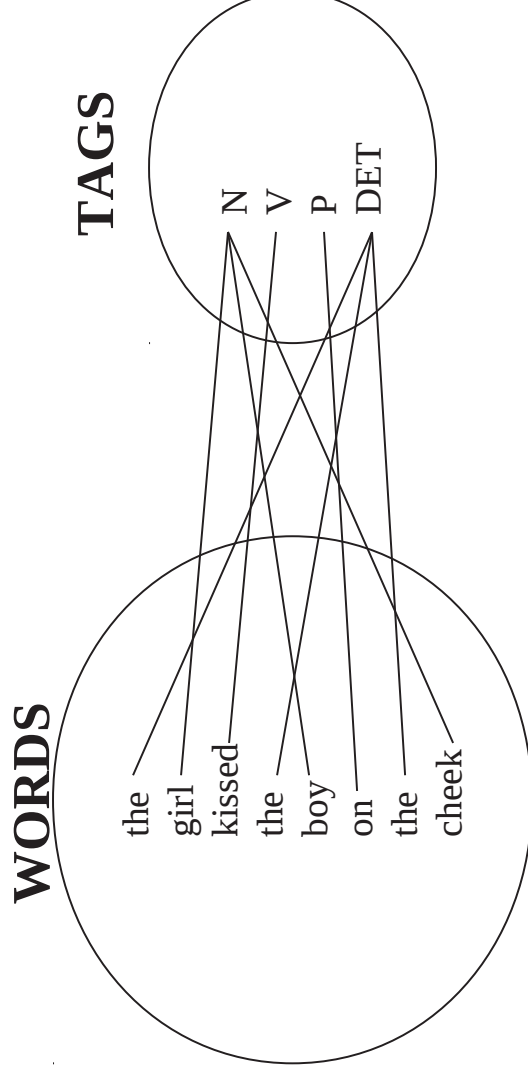


# POS Tagging - Definition

- The process of assigning a part-of-speech or other lexical class marker to each word in a corpus” (Jurafsky and Martin)



# POS Tagging - 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

# Motivation

- Speech synthesis — pronunciation
- Speech recognition — class-based N-grams
- Information retrieval — stemming, selection high-content words
- Word-sense disambiguation
- Corpus analysis of language & lexicography

# Tagsets for English

- There are a small number of popular tagsets for English, many of which evolved from the 87-tag tagset used for the Brown corpus.
  - Three commonly used
    - **The small 45-tag Penn Treebank tagset**
    - The medium-sized 61 tag C5 tagset used by the Lancaster UCREL project's CLAWS tagger to tag the British National Corpus, and
    - The larger 146-tag C7 tagset

# Tagsets for English

Tag	Description	Example	Tag	Description	Example
CC	Coordin. Conjunction	<i>and, but, or</i>	SYM	Symbol	<i>+, %, &amp;</i>
CD	Cardinal number	<i>one, two, three</i>	TO	"to"	<i>to</i>
DT	Determiner	<i>a, the</i>	UH	Interjection	<i>uh, oops</i>
EX	Existential 'there'	<i>there</i>	VB	Verb, base form	<i>eat</i>
FW	Foreign word	<i>meu culpa</i>	VBD	Verb, past tense	<i>ate</i>
IN	Preposition/sub-conj	<i>of, in, by</i>	VBG	Verb, gerund	<i>eating</i>
JJ	Adjective	<i>yellow</i>	VCN	Verb, past participle	<i>eaten</i>
JJR	Adj., comparative	<i>bigger</i>	VBP	Verb, non-3sg pres	<i>eat</i>
JJS	Adj., superlative	<i>wildest</i>	VBZ	Verb, 3sg pres	<i>eats</i>
LS	List item marker	<i>1, 2, One</i>	WDT	Wh-determiner	<i>which, that</i>
MD	Modal	<i>can, should</i>	WP	Wh-pronoun	<i>what, who</i>
NN	Noun, sing. or mass	<i>llama</i>	WP\$	Possessive wh-	<i>whose</i>
NNS	Noun, plural	<i>llamas</i>	WRB	Wh-adverb	<i>how, where</i>
NNP	Proper noun, singular	<i>IBM</i>	\$	Dollar sign	<i>\$</i>
NNPS	Proper noun, plural	<i>Carolinus</i>	#	Pound sign	<i>#</i>
PDT	Predeterminer	<i>all, both</i>	"	Left quote	<i>(' or ")</i>
POS	Possessive ending	<i>'s</i>	"	Right quote	<i>(' or ")</i>
PP	Personal pronoun	<i>I, you, he</i>	(	Left parenthesis	<i>( [ , { , &lt;)</i>
PP\$	Possessive pronoun	<i>your, one's</i>	)	Right parenthesis	<i>( [ , } , &gt;)</i>
RB	Adverb	<i>quickly, never</i>	,	Comma	<i>,</i>
RBR	Adverb, comparative	<i>faster</i>	.	Sentence-final punc	<i>(. ! ?)</i>
RBS	Adverb, superlative	<i>fastest</i>	:	Mid-sentence punc	<i>( : ; ... -- )</i>
RP	Particle	<i>up, off</i>			

Penn Treebank POS tags

# Tagsets for English

Tag the following sentence using Penn Treebank tagset:

*Book that flight .*

*Book/VB that/DT flight/NN .*

# Tagsets for English

Tag the following sentence using Penn Treebank tagset:

*The grand jury commented on a number of other topics.*

# Tagsets for English

Tag the following sentence using Penn Treebank tagset:

*The grand jury commented on a number of other topics.*

The/DT grand/JJ jury/NN commented/VBD on/IN a/DT number/NN  
of/IN other/JJ topics/NNS ./.



# Tagsets for English

Tag the following sentence using Penn Treebank tagset:

*There are 70 children there*

There/EX are/VBP 70/CD children/NNS there/RB

# Tagsets for English

- Some tagging distinctions are quite hard.
- Prepositions (IN), particles (RP), and adverbs (RB) can have a large overlap. Word like *around* can be all three:

Mr./NNP John/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  
Apples/NNP costs/VBZ **around/RB** 250/CD

# Tagsets for English

- Particles often can either precede or follow a noun phrase object
- Prepositions can not follow their noun phrase

She told off/RP her friends  
She told her friends off/RP  
She stepped off/IN the train  
**\*She stepped the train off/IN**

- Words that can be adjective, proper nouns are tagged as common nouns when acting as modifiers:

Chinese/NN cooking/NN  
Pacific/NN waters/NNS

# Tagsets for English

- Distinguishing past participles (VBN) from adjectives (JJ)

They were **married**/VBN by the Justice of the Peace yesterday.  
At the time, he was already **married**/JJ

- Certain syntactic distinctions were not marked in the Penn

Treebank tagset

- *prepositions* and *subordinating conjunctions* were combined into the single tag *IN*, since the tree-structure of the sentence disambiguated them

# Tagsets for English

- Brown and C5 tagsets distinguish prepositions (IN) from subordinating conjunctions (CS)

after/**CS** spending/VBG a/AT few/AP days/NNS at/IN the/AT hotel/NN  
after/**IN** a/AT holiday/NN trip/NN to/IN Canada/NP

- Also contains two tags for word to – infinitive use as TO, prepositional use as IN

to/**TO** give/VB priority/NN to/**IN** teacher/NN pay/NN raises/NNS

- Which tagset to use depends on how much information the application needs

# Part-of-Speech Tagging

- **POS tagging (tagging)**
  - The process of assigning a part-of-speech or other lexical marker to each word in a corpus.
  - Also applied to punctuation marks
  - Thus, tagging for NL is the same process as **tokenization** for computer language, although tags for NL are much more ambiguous.
  - Taggers play an increasingly important role in speech recognition, NL parsing and IR

# Part-of-Speech Tagging

- Input to tag algorithm: a string of words (ex. *Book that flight, Penn*  
Treebank tagset)

- Output: a single best tag for each word

VB		NN
<u>Book</u> that flight.	Hand me that <u>book</u>	
DT		CS
Does <u>that</u> flight serve dinner ?	I thought <u>that</u> your flight was earlier	

- Automatically assigning a tag to a word is not trivial
  - For example, *book* is **ambiguous**: it can be a verb or a noun
  - Similarly, *that* can be a determiner, or a complementizer
- The problem of POS-tagging is to resolve these ambiguities, choosing the proper tag for the context.

# Part-of-Speech Tagging

- Many of the 40% ambiguous tokens are easy to disambiguate, because
  - The various tags associated with a word are not equally likely.
  - For example, *a* can be a determiner, or the letter *a* (part of acronym or initial), but the determiner sense of *a* is much more likely



# Part-of-Speech Tagging

- Many tagging algorithms fall into two classes:
  - Rule-based taggers
    - Involve a large database of hand-written disambiguation rule specifying, for example, that *an ambiguous word is a noun rather than a verb if it follows a determiner*. [EngCG tagger]
  - Stochastic taggers
    - Resolve tagging ambiguities by using a training corpus to *count the probability of a given word having a given tag in a given context*.
  - The Brill tagger, called the transformation-based tagger, shares features of both tagging architecture.

# References

- *Speech and Language Processing*, Jurafsky and H.Martin



**Thank you!**

