# Lecture 3: Part-of-Speech Tagging and Name Entity Recognition

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

Figure 8.1 The 17 parts of speech in the Universal Dependencies tagset (Nivre et al., 2016a). Features can be added to make finer-grained distinctions (with properties like number, case, definiteness, and so on).

Figure 1: The common part-of-speech in English.

## 1 Concepts

This chapter mainly covers the **part-of-speech**, i.e. **POS**. We also covers the term **named entity** for, roughly speaking, anything that can be referred to with a proper name: a person, a location, an organization, although as well see the term is commonly extended to include things that arent entities per se. In this chapter well introduce the task of **part-of-speech tagging**, taking a sequence of words and assigning each word a part of speech like NOUN or VERB, and the task of **named entity recognition (NER)**, assigning words or phrases tags like PERSON, LOCATION, or ORGANIZATION.

Such tasks in which we assign, to each word  $x_i$  in an input word sequence, a label  $y_i$ , so that the output sequence Y has the same length as the input sequence X are called **sequence labeling** tasks.

#### 1.1 Word Classes in English

The common part-of-speech tag can be seen in Figure 1. This includes **closed class** words and **open class** words: Closed classes are those with relatively fixed membership, such as prepositionsnew prepositions are rarely coined. By contrast, nouns and verbs are open classes, since new names and actions are invented all the time.

The open class words include:

- noun (NOUN): words for person, things, places, etc. can be used for concrete terms, or abstract terms. Common nouns can be divided into count nouns or mass nouns. Count nouns can occur in the *singular* and *plural* (*goat/goats*, *relationship/relationships*) and can be *counted* (one goat, two goats). Mass nouns are used when something is conceptualized as a *homogeneous group*.
- verb (VERB): words for action and processes. English verbs have inflections (non-third-person-singular (eat), third-person-singular (eats), progressive (eating), past participle (eaten)).
- adjective (ADJ): noun modifiers, describing noun
- adverb (ADV): verb modifiers, of manner, place, time etc. Directional adverbs or locative adverbs (home, here, downhill) specify the direction or location of some action; degree adverbs (extremely, very, somewhat) specify the extent of some action, process, or property; manner adverbs (slowly, slinkily, delicately) describe the manner of some action or process; and temporal adverbs describe the time that some action or event took place (yesterday, Monday).
- proper noun (PROPN): name of place, person etc.
- interjection (INTJ): exclamation, greeting, yes/no response, etc.

The closed class words include:

- preposition/postposition (ADP): mark a noun's spacial, temporal, or other relation
- **pronoun (PRON)**: a shorthand for referring to an entity or event such as *I*, *me*, *he*, *she*, *you*. **Wh-pronouns** (*what*, *who*, *whom*, *whoever*) are used in certain question forms, or act as **complementizers** (*Frida*, *who married Diego...*).
- auxiliary (AUX): such as can, make, need, may, should etc. Auxiliary verbs mark semantic features of a main verb such as its tense, whether it is completed (aspect), whether it is negated (polarity), and whether an action is necessary, possible, suggested, or desired (mood). English auxiliaries include the copula verb be, the two verbs do and have, forms, as well as modal verbs used to mark the mood associated with the event depicted by the main verb: can indicates ability or possibility, may permission or possibility, must necessity.
- determiner (DET): marks noun phrase properties such as this, that, a, an, the.
- coordinating conjunction (CCONJ): such as and, but, however,
- subordinating conjunction (SCONJ): such as which, where, that etc. Subordinating conjunctions like that which link a verb to its argument in this way are also called complementizers
- numeral (NUM): numbers, one, two, three etc.
- particle (PART): a proposition-like form used together with verb, down, up, off, on, at, by, out, in. A particle resembles a preposition or an adverb and is used in combination with a verb. Particles often have extended meanings that arent quite the same as the prepositions they resemble, as in the particle over in she turned the paper over. A verb and a particle acting as a single unit is called a phrasal verb. The meaning of phrasal verbs is often non-compositionalnot predictable from the individual meanings of the verb and the

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

Figure 8.2 Penn Treebank part-of-speech tags.

Figure 2: The part-of-speech tag in Penn Treebank.

particle.

Closed class words are generally **function words** like of, it, and, or you, which tend to be very short, occur frequently, and often have structuring uses in grammar.

Figure 2 shows the symbol of part-of-speech tag from Penn Treebank [Jurafsky and Martin, 2014].

## 2 Part-of-speech tagging

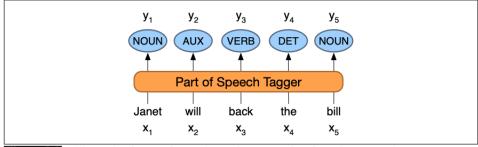
**Part-of-speech tagging** is the process of assigning a part-of-speech to each word in a text. The input is a sequence  $x_1, x_2, \ldots x_n$  of (tokenized) words and a tagset, and the output is a sequence  $y_1, y_2, \ldots, y_n$  of tags, each output  $y_i$  corresponding exactly to one input  $x_i$ , as shown in the intuition in Figure 3.

Tagging is a **disambiguation** task; words are ambiguous have more than one possible part-of-speech and the goal is to find the correct tag for the situation. The goal of POS-tagging is to resolve these ambiguities, choosing the proper tag for the context (i.e. **ambiguity resolution**). Nonetheless, many words are easy to disambiguate, because their different tags arent equally likely. For example, "a" can be a determiner or the letter a, but the determiner sense is much more likely.

This idea suggests a useful **baseline**: given an ambiguous word, choose the tag which is most **frequent** in the training corpus. This is a key concept:

Most Frequent Class Baseline: Always compare a classifier against a baseline at least as good as the most frequent class baseline (assigning each token to the class it occurred in most often in the training set).

The most-frequent-tag baseline has an accuracy of about 92%. The standard technique behinds POS tagging is **sequence labeling**, i.e. for an input sequence of texts, output a sequence of labels,



**Figure 8.3** The task of part-of-speech tagging: mapping from input words  $x_1, x_2, ..., x_n$  to output POS tags  $y_1, y_2, ..., y_n$ .

Figure 3: The part-of-speech tagging process.

Citing high fuel prices, [ORG United Airlines] said [TIME Friday] it has increased fares by [MONEY \$6] per round trip on flights to some cities also served by lower-cost carriers. [ORG American Airlines], a unit of [ORG AMR Corp.], immediately matched the move, spokesman [PER Tim Wagner] said. [ORG United], a unit of [ORG UAL Corp.], said the increase took effect [TIME Thursday] and applies to most routes where it competes against discount carriers, such as [LOC Chicago] to [LOC Dallas] and [LOC Denver] to [LOC San Francisco].

Figure 4: An example of name entities in a paragraph.

each for a text.

## 3 Named Entities and Named Entity Tagging

A named entity is, roughly speaking, anything that can be referred to with a proper name: a person, a location, an organization. The task of named entity recognition (NER) is to find spans of text that constitute proper names and tag the type of the entity. Four entity tags are most common: PER (person), LOC (location), ORG (organization), or GPE (geo-political entity). However, the term named entity is commonly extended to include things that arent entities per se, including dates, times, and other kinds of temporal expressions, and even numerical expressions like prices. Figure 4 shows name entities in a paragraph.

Named Entity Recognition can be used as pre-processing steps for *sentiment analysis*, *question* answering as well as learning *semantic representations* such as event extraction and relationship inference.

NER is more challenging than POS tagging. For POS tagging, each word has only one part-of-speech tag but for NER, there is a *segmentation* problem. The task of NER is to find and label the *span* of text. The ambiguity of segmentation brings additional challenge: we need to decide whats an entity and what isnt, and where the boundaries are.

The standard approach that convert the span-recognition to sequence labeling is **BIO tagging** [Jurafsky and Martin, 2014]. This is a method that allows us to treat NER like a word-by-word sequence labeling task, via **tags** that capture both the boundary and the named entity type. Similarly,

Words	IO Label	BIO Label	BIOES Label			
Jane	I-PER	B-PER	B-PER			
Villanueva	I-PER	I-PER	E-PER			
of	0	0	0			
United	I-ORG	B-ORG	B-ORG			
Airlines	I-ORG	I-ORG	I-ORG			
Holding	I-ORG	I-ORG	E-ORG			
discussed	0	0	0			
the	0	0	0			
Chicago	I-LOC	B-LOC	S-LOC			
route	0	0	0			
•	0	0	0			
Figure 8.7 NE	Figure 8.7 NER as a sequence model, showing IO, BIO, and BIOES taggings.					

Figure 5: BIO, IO and BIOES tagging for NER.

#### there are IO tagging and BIOES tagging.

In BIO tagging we label any token that begins a span of interest with the label  $\mathbf{B}$ , tokens that occur inside a span are tagged with an  $\mathbf{I}$ , and any tokens outside of any span of interest are labeled  $\mathbf{O}$ . There is only one  $\mathbf{O}$  tag to indicate not a named entity. On the other hand, we can have distinct  $\mathbf{B}$  and  $\mathbf{I}$  for each named entity type, i.e.  $\mathbf{B}\text{-}\mathbf{PER}$ , I-PER, B-ORG, I-ORG etc. For n distinct named entity types, there are 2n+1 distinct tags. BIO tagging can represent exactly the same information as the bracketed notation, but has the advantage that we can represent the task in the same simple sequence modeling way as part-of-speech tagging. In BIOES tagging, we add  $\mathbf{E}$  tag for end of a span, and a span tag  $\mathbf{S}$  for a span consist of only one word.

A sequence labeler (HMM, CRF, RNN, Transformer, etc.) is trained to label each token in a text with tags that indicate the presence (or absence) of particular kinds of named entities.

## References

Dan Jurafsky and James H Martin. Speech and language processing. vol. 3. US: Prentice Hall, 2014.