# **Chapter NLP:IV**

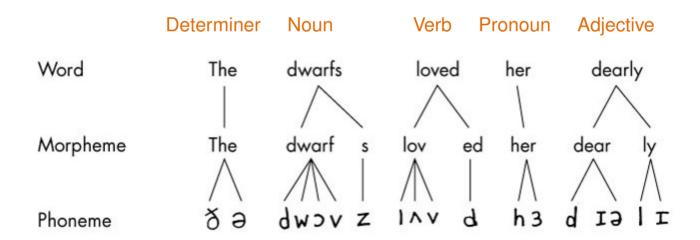
#### IV. Words

- □ Morphology
- □ Word Classes
- Named Entities

#### **Definition**

A word class is a set of lexical items with similar formal (grammatical) properties.

- Also called part of speech, grammatical category, lexical category, or syntactic category. roughly synonymous
- Common properties are morphology and semantic or syntactic behavior.
- Often serve an encoding of the formal properties of words for programs.



## Traditional grammar

The traditional English grammar lists 9 word classes split into two groups: function or content classes.

- 1. Content classes: or open/form/lexical
  - Accept new members, can hold infinitely many items per class.
  - Nouns, verbs, adjectives, adverbs.
- 2. Function classes: or closed/structure
  - Number of members is fixed.
  - Prepositions, pronouns, determiners, conjunctions and interjections.
  - □ As language evolves, changes may also happen in closed classes.

```
The dwarfs loved her dearly
TG determiner noun verb pronoun adverb
PENN
UD
```

Traditional grammar: Example

Word Class: Noun

**Definition:** A noun is a word used for naming some person or thing. [Cambridge]

**Examples:** Paris, man, house, height.

- The definition is incomplete: What about places? What about abstract qualities (beauty) and actions (a thump)?
- No reference is made to morphology or syntax.
- The class does not differentiate (grammatical) properties: Plural nouns and proper nouns are all just **noun**.

#### Remarks:

- An **adjective** is a word used to qualify a noun [...] to restrict the application of a noun by adding something to its meaning. Examples: fine, brave, three, the.
  - The definition is vague and allows many elements with different grammatical properties (the, my, all), and even nouns in certain constructions (her brother the butcher).
  - No reference is made to morphology or syntax.
- □ A verb is a word used for saying something about some person or thing. Examples: make, know, buy, sleep.
  - On this definition, there is little difference between a verb and an adjective. Some grammars prefer to talk about 'doing words' or 'action words', but this seems to exclude the many state verbs, such as know, remember, be.
  - No reference is made to morphology or syntax.
- □ An adverb is a word used to qualify any part of speech except a noun or pronoun. Examples: today, often, slowly, very.
  - Adverbs are often said to qualify (or 'modify') verbs which is inadequate for such words as very and however.
  - This definition hardly applies to interjections or examples like the very man and slovenly me.
  - No reference is made to morphology or syntax.

#### Remarks:

- A **pronoun** is a word used instead of a noun or noun-equivalent (i.e. a word which is acting as a noun). Examples: this, who, mine.
  - Pronouns are used instead of noun phrases, not just nouns. He refers to the whole of the phrase the big lion, not just the word lion, since we cannot say the big he.
  - No reference is made to morphology or syntax.
- A preposition is a word placed before a noun or noun-equivalent to show in what relation the person or thing stands to something else. Examples: on, to, about, beyond.
  - This gives a clear syntactic criterion. However, prepositions go before noun phrases, rather than nouns, and may also be used in other parts of the sentence. As with nouns, more than just persons and things are involved.
- A conjunction is a word used to join words or phrases together, or one clause to another clause, Examples: and, before, as well as.
  - Prepositions also have a joining function (the man in the garden).
  - Conjunction Junction, what's your function?
- An interjection is a word or sound thrown into a sentence to express some feeling of the mind. Examples: Oh!, Bravo!, Fie!.
  - Interjections do not enter into the construction of sentences. Despite the emotional function of these words, they still need to be considered as part of sentence classification.

## **Tagsets**

The classes of the traditional grammar are not suited for language processing.

- Several advanced class schemes (called tagsets) exist.
- □ They distinguish between 17 up to 100+ word classes.
  - Penn Treebank tagset 36 tags
  - Universal POS tags
     17 tags
  - CLAWS tagsets
     CLAWS1: 132, CLAWS2: 166, C5: 60, C6: 160, C8, ...
  - Brown Corpus 87 tags
  - Penn Treebank II41 tags
  - British National Corpus 61 tags
  - BNC Sampler 146 tags
- Corpora for part of speech tagging are manually annotated, for example:
  - The 1 million word Brown corpus in the 1960s.
  - The 100 million word British National Corpus.

Penn Treebank tagset [upenn]

**Idea:** Assign a tag to each combination of class of the traditional grammar and their observed grammatical properties.

NN	Noun in singular
NNS	Noun in plural
NNP	Proper noun
NNPS	Proper noun in plural

## Penn Treebank tagset [upenn]

**Idea:** Assign a tag to each combination of class of the traditional grammar and their observed grammatical properties for all classes.

NN	Noun, singular or mass		
NNS	Noun, plural		
NNP	Proper noun, singular		
NNPS Proper noun, plural			
VB	Verb, base form		
VBD	Verb, past tense		
VBG	Verb, gerund or present participle		
VBN	Verb, past participle		
VBP	Verb, non-3rd person singular present		
VBZ	Verb, 3rd person singular present		
JJ	Adjective		
JJR	Adjective, comparative		
JJS	Adjective, superlative		
RB	Adverb		
RBR	Adverb, comparative		
RBS	Adverb, superlative		
WRB	Wh-adverb		
	·		

TO	to (Preposition)
RP	Particle
POS	Possessive ending
MD	Modal
PRP	Personal pronoun
PRP\$	Possessive pronoun
WP	Wh-pronoun
WP\$	Possessive wh-pronoun
DT	Determiner
PDT	Predeterminer
WDT	Wh-determiner
CC	Coordinating conjunction
IN	Preposition or subordinating conjunction
UH	Interjection

#### Penn Treebank tagset [upenn]

**Idea:** Assign a tag to each combination of class of the traditional grammar and their observed grammatical properties for all classes. Add classes for everything else we find while annotating text.

NN	Noun, singular or mass		
NNS	Noun, plural		
NNP	Proper noun, singular		
NNPS Proper noun, plural			
VB	Verb, base form		
VBD	Verb, past tense		
VBG	Verb, gerund or present participle		
VBN	Verb, past participle		
VBP	Verb, non-3rd person singular present		
VBZ	Verb, 3rd person singular present		
JJ	Adjective		
JJR	Adjective, comparative		
JJS	Adjective, superlative		
RB	Adverb		
RBR	Adverb, comparative		
RBS	Adverb, superlative		
WRB	Wh-adverb		

TO	to
RP	Particle
POS	Possessive ending
MD	Modal
PRP	Personal pronoun
PRP\$	Possessive pronoun
WP	Wh-pronoun
WP\$	Possessive wh-pronoun
DT	Determiner
PDT	Predeterminer
WDT	Wh-determiner
CC	Coordinating conjunction
IN	Preposition or subordinating conjunction
UH	Interjection
CD	Cardinal number
EX	Existential there
FW	Foreign word
LS	List item marker
SYM	Symbol

Penn Treebank tagset [upenn]

The Penn tagset covers more grammatical properties and is still frequently used.

The Penn tagset is also English-centric:

What does it mean and why is this a problem?



Penn Treebank tagset [upenn]

The Penn tagset covers more grammatical properties and is still frequently used.

The Penn tagset is also English-centric:

- □ With this strategy, each language needs it's own tagset.
- □ A different tagger needs to be developed for each language. Very difficult for low-resource languages, see electronic colonialism.
- It violates Chomsky's theory of universal grammar, which is very beloved by computer scientists.



Universal Dependencies tagset [UD]

**Idea:** Seperate form from function: Use only few, universal POS tags and many lexical and grammatical properties which can be freely assigned to any tag.

Part-of-speech Tag		
NOUN	Noun	
PROPN	Proper noun	
VERB	Verb	
AUX	Auxiliary	
ADJ	Adjective	
ADV	Adverb	
ADP	Adposition (Preposition)	
PART	Particle	
PRON	Pronoun	
DET	Determiner	
CCONJ	Coordinating conjunction	
SCONJ	Subordinating conjunction	
INTJ	Interjection	
NUM	Numeral	
PUNCT	Punctuation	
SYM	Symbol	
X	Other	

<b>Lexical Features</b>
PronType
NumType
Poss
Reflex
Foreign
Abbr
Туро

Inflextional	<b>Features</b>
Nominal	Verbal
Gender	VerbForm
Animacy	Mood
NounClass	Tense
Number	Aspect
Case	Voice
Definite	Evident
Degree	Polarity
	Person
	Polite
	Clusivity

Which of these do not exist in English?

Universal Dependencies tagset [UD]

**Idea:** Seperate form from function: Use only few, universal POS tags and many lexical and grammatical properties which can be freely assigned to any tag.

	The	dwarfs	loved	her	dearly
TG	determiner	noun	verb	pronoun	adverb
PENN	DT	NNS	VBD	PRP	RB
UD	DET; PronType: Art	NOUN; Number: Plur	VERB; Tense: Past	PRON	ADV

## **Ambiguities**

- About 85% of the vocubulary (types)
   belong to only one word class.
- The others are ambiguous.

The back door  $\rightarrow$  adjective, JJ

On my back  $\rightarrow$  noun, NN

Win the voters back  $\rightarrow$  adverb, RB

Said to back the bill  $\rightarrow$  verb, VB

 However, ambiguous types are much more frequent. About half of the tokens are ambiguous.

## 10 REASONS WHY ENGLISH IS WEIRD

- 1) The bandage was wound around the wound.
- 2) The farm was used to produce produce.
- 3) The dump was so full that it had to refuse more refuse.
- 4) We must polish the Polish furniture.
- 5) He could lead if he would get the lead out.
- 6) The soldier decided to desert his dessert in the desert.
- 7) Since there is no time like the present, he thought it was time to present the present.
- 8) A bass was painted on the head of the bass drum.
- 9) When shot at, the dove dove into the bushes.
- 10) I did not object to the object.

#### Remarks:

□ About the termininology, the Cambridge Encyclopedia of the English Language (CUP) writes:

"When linguists began to look closely at English grammatical structure in the 1940s and 1950s, they encountered so many problems of identification and definition that the term part of speech soon fell out of favor, word class being introduced instead. Word classes are equivalent to parts of speech, but defined according to strict linguistic criteria." — (Crystal (2003)

- "There is no single correct way of analyzing words into word classes... Grammarians disagree about the boundaries between the word classes, and it is not always clear whether to lump subcategories together or to split them. For example, in some grammars pronouns are classed as nouns, whereas in other frameworks they are treated as a separate word class." Aarts, Chalker, Weiner (2014) The Oxford Dictionary of English Grammar. OUP
- □ The 9 word classes of the traditional grammar are usually attributed to *Dionysius Thrax of Alexandria* (100 B.C.), who wrote in his *Techne* on Greek about eight parts of speech: noun, verb, pronoun, preposition, adverb, conjunction, participle, and article.

## Part-of-Speech Tagging

A tagger is a program that assigns tags from a tagset to each token in a sequence. Tagging is a sequence labeling problem: the input is a sequence of tokens  $w_1, \ldots, w_n$  and a tagset T and the output is a sequence of tags  $c_1, \ldots, c_n$ ,  $c_i \in T$  of equal length.

Part-of-speech tags can inform us about syntax and semantics of a sequence:

- the intended sense of a word
  - apple (single noun, NN) vs. Apple (proper noun, NNP)
- the applied morphemes (lemmatization)
  - sigh (verb base form, VB), sighed (verb past tense or past participle, VBD or VBN)
- □ the meaning of a sentence (shallow parsing)
- the correct pronunciation (speech synthesis)
  - OBject vs. obJECT, CONtent vs. conTENT

Part-of-Speech Tagging: Maximum Likelihood Estimate

**Idea**: Most types are unambiguous. Most ambiguous types have one very likely tag. So we can do a Maximum Likelihood Estimate (MLE):

 $\Box$  Tag each token  $w_i$  with the word class  $c_i$  it appears in most often:

$$c_i = \operatorname*{argmax}_{c_j \in T} \frac{C(c_j, w_i)}{C(w_i)}$$

- Unknown words are often tagged as proper nouns.
- □ This is sometimes called the most frequent class baseline.
- □ This baseline works extremely well; It achieves 92% accuracy on UD over 15 languages [Wu and Dredze 2019].
- Humans reach roughly 97% accuracy.

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 $<sup>{}^{1}</sup>C(w)$  is the count function.

Part-of-Speech Tagging: Brill Tagger [Brill 1992]

Idea: Improve the most frequent class baseline by applying rules for special cases.

The Brill Tagger is an "error-driven transformation-based tagging". It iteratively applies rules T1 T2 <Premise>.

- 1. Initially tag a sequence with MLE.
- 2. If a token is tagged with T1 and fulfills the <Premise>, replace T1 with T2.
- 3. Repeatedly iterate the sequence and apply the rules in order until the stopping criterion is reached.

The rules are learned from errors made on a pre-tagged corpus.

- □ If a false tag T1 is encounterd in the pre-tagged corpus, create several rules with the correct tag T2 and a fulfilled <Premise>.
- Correct a test corpus with each new rule.
- □ If a rule does not increase accuracy, discard it.

Part-of-Speech Tagging: Brill Tagger [Brill 1992]

## Premises are constructed from templates:

context x A word in context is tagged x. The word has a certain property.

context property A word in context has a certain property.

context One or any of  $i \in [1,3]$  preceding or following word(s).

#### Example rules:

TO ΙN next-tag AT prev-word-is-cap TRUE **VBN** VBD VBD VBN prev-1-or-2-or-3-tag HVD VB NNprev-1-or-2-tag AT VB prev-tag TO NNTO TNnext-word-is-cap TRUE NN VB prev-tag MD

Part-of-Speech Tagging: Brill Tagger [Brill 1994]

Idea: Create rules to tag unknown tokens. T1 may be UNK for unknown.

## Premise templates for unknown tokens:

affix x constraint
context word
char x

Token fulfills constraint regarding affix of at most 4 chars.

A word appears in context. Character x occurs in word.

constraint

When deleting or adding affix x, word found in dictionary.

Else, affix x occurs in token.

## Example rules:

NN	NNS	suffix -s occurs
NN	CD	char .
NN	JJ	char -
NN	VBN	suffix -ed occurs
NN	VBG	suffix -in occurs
UNK	ADJ	suffix -ly addition
UNK	RB	suffix -ly occurs

## Part-of-Speech Tagging

## Original text

A relevant document will describe marketing strategies carried out by U.S. companies for their agricultural chemicals, report predictions for market share of such chemicals, or report market statistics for agrochemicals, pesticide, herbicide, fungicide, insecticide, fertilizer, predicted sales, market share, stimulate demand, price cut, volume of sales.

#### Tagged text with Brill tagger

A/DT relevant/JJ document/NN will/MD describe/VB marketing/NN strategies/NNS carried/VBD out/IN by/IN U.S./NNP companies/NNS for/IN their/PRP\$ agricultural/JJ chemicals/NNS ,/, report/NN predictions/NNS for/IN market/NN share/NN of/IN such/JJ chemicals/NNS ,/, or/CC report/NN market/NN statistics/NNS for/IN agrochemicals/NNS ,/, pesticide/NN ,/, herbicide/NN ,/, fungicide/NN ,/, insecticide/NN ,/, fertilizer/NN ,/, predicted/VBN sales/NNS ,/, market/NN share/NN ,/, stimulate/VB demand/NN ,/, price/NN cut/NN ,/, volume/NN of/IN sales/NNS ./.

#### Remarks:

- □ Part-of-speech tagging can be solved with any generic sequence labeling model, like Hidden Markov Models, CRFs, RNNs, and transformer models.
- □ The Brill Tagger can be seen as a special case of a sequence classifier, where the premises are the features and the order of the rules are the weights.
- □ The state of the art in part of speech tagging can be reviewed at <u>aclweb.org</u>, <u>paperswithcode</u>, or <u>NLPprogress</u>. Most taggers reported are based on statistical sequence models rather than rules. However, many taggers proposed are not included, including the Brill tagger.