



Application of Big Data in Social Science

week 5

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Recap from last week

Natural Language Processing

- we used nltk library
- tokenization: to break down or split text into smaller pieces
 - word tokenization
 - sentence tokenization
- we made sure to make into lower cases when counting frequencies of words.
- stopwords: words that do not have much contribution to analyzing data
 - eg) a, the, of, and, he, she etc.
- List comprehension
- Regular Expressions



List comprehension

List comprehension with append option

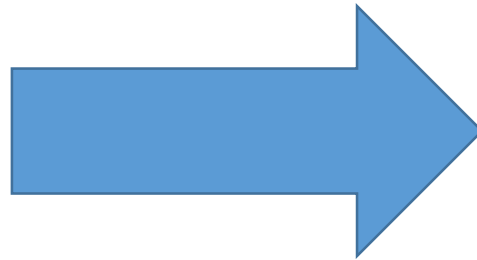
```
>>> new_list= [expression for item in list (if conditional)]
```

```
>>> x= []
```

```
>>> my_list = [1,2,3]
```

```
>>> for a in my_list:
```

```
>>>     x.append(a*2)
```



```
x=[a*2 for a in my_list]
```



Regular Expressions

[Regex documentation](#)

Regex Syntax	What it means
[^...]	matches a character not present in the square brackets after the ^ symbol
	OR operator
+	matches one or more cases of the previous mentioned regex before the + symbol
[A-Za-z]	matches all alphabets (upper and lower cases)
[a-zA-Z0-9_]	matches alpha-numeric characters (\w)
[^a-zA-Z0-9_]	matches non-alpha-numeric characters (\W)

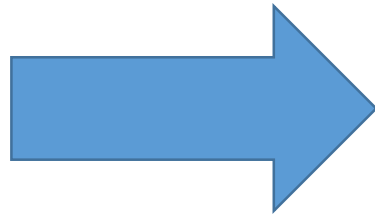
`re.sub(pattern, repl, string, count=0, flags=0)`

This method is to substitute a specified regex pattern in a string with a replacement string.



Order of NLP

- HTML
- HTML parsing
- sentence segmentation
- word tokenization
- stopwords
- POS tagging
- text lemmatization



today's portion!



Part of speech (POS)

- There are eight parts of speech in the English language:
 - **Noun**... a person, place, thing, or idea (tree, Heidi, notebook, Seoul)
 - **Pronoun**... word used in place of a noun (she, we, they, it)
 - **Verb**... expresses action or being (go, write, teach, code)
 - **Adjective**... describes a noun or pronoun (small, old, blue, smart)
 - **Adverb**... describes a verb, an adjective or another adverb (extremely, well, carefully)
 - **Preposition**... placed before a noun to form a phrase modifying another word.. (by, with, about, until, in, on, of,)
 - **Conjunction**... joins words, phrases, or clauses (and, but, or, while, because)
 - **Interjection**...used to express emotions (oh! wow! oops!)



Part of speech (POS)

Open class (lexical) words

Nouns

Proper

John, Seoul,
Hanyang

Common

dog, cats

Verbs

Main

code,
play, drink

Adjectives younger, young, youngest

Adverbs slowly, carefully, hardly

Number 1, 3, 10, five, one

Preposition of, in, about, that

Particles off, out, away, down

Interjections wow, oh, gosh

Closed class (functional) words

Determiners a, an, the, this, these

Conjunctions and, or, but

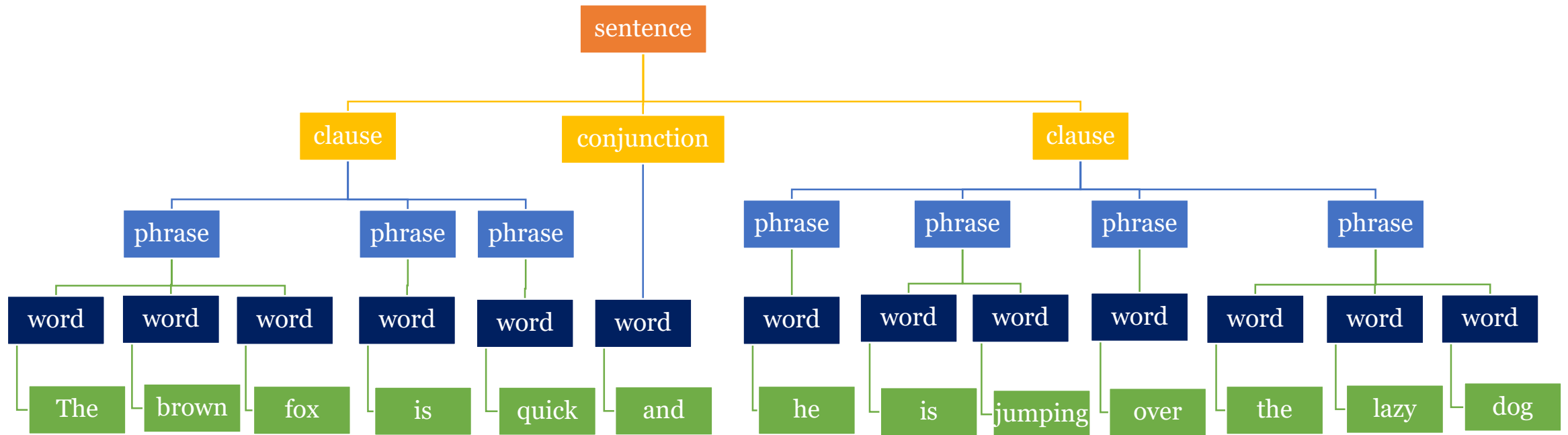
Pronouns I, we, you, he, she, they

Modals

could, should

Language Syntax and Structure

The brown fox is quick and he is jumping over the lazy dog.



Hierarchical tree:
Sentence -> clauses -> phrases -> words

Words :POS tagging

- Words are the smallest unit in a language.
- It is useful to annotate and tag words then analyze them into their POS to see the major syntactic categories.

	0	1	2	3	4	5	6	7	8	9	10	11	12
0	The	brown	fox	is	quick	and	he	is	jumping	over	the	lazy	dog
1	DT	JJ	NN	VBZ	JJ	CC	PRP	VBZ	VBG	IN	DT	JJ	NN

CC	Coordinating conjunction	NNS	Noun, plural	UH	Interjection
CD	Cardinal number	NNP	Proper noun, singular	VB	Verb, base form
DT	Determiner	NNPS	Proper noun, plural	VBD	Verb, past tense
EX	Existential there	PDT	Predeterminer	VBG	Verb, gerund or present
FW	Foreign word	POS	Possessive ending		participle
IN	Preposition or subordinating conjunction	PRP	Personal pronoun	VBN	Verb, past participle
		PRP\$	Possessive pronoun	VBP	Verb, non-3rd person singular
JJ	Adjective	RB	Adverb		present
JJR	Adjective, comparative	RBR	Adverb, comparative	VBZ	Verb, 3rd person singular
JJS	Adjective, superlative	RBS	Adverb, superlative		present
LS	List item marker	RP	Particle	WDT	Wh-determiner
MD	Modal	SYM	Symbol	WP	Wh-pronoun
NN	Noun, singular or mass	TO	to	WP\$	Possessive wh-pronoun
				WRB	Wh-adverb

Phrases

- Group of words make up phrases.
- Phrases are assumed to have at least two or more words.
- However, a phrase can be a single word or a combination of words based on the syntax and position of the phrase in a clause.

Noun phrase (NP): phrases where a noun acts as the head word. NPs act as a subject or object to a verb.

eg) “the lazy dog”, “the brown fox”

Verb phrase (VP): phrases where a verb acts as the head word.

eg) “It has been a while” => ‘has been’

Adjective phrase (ADJP):

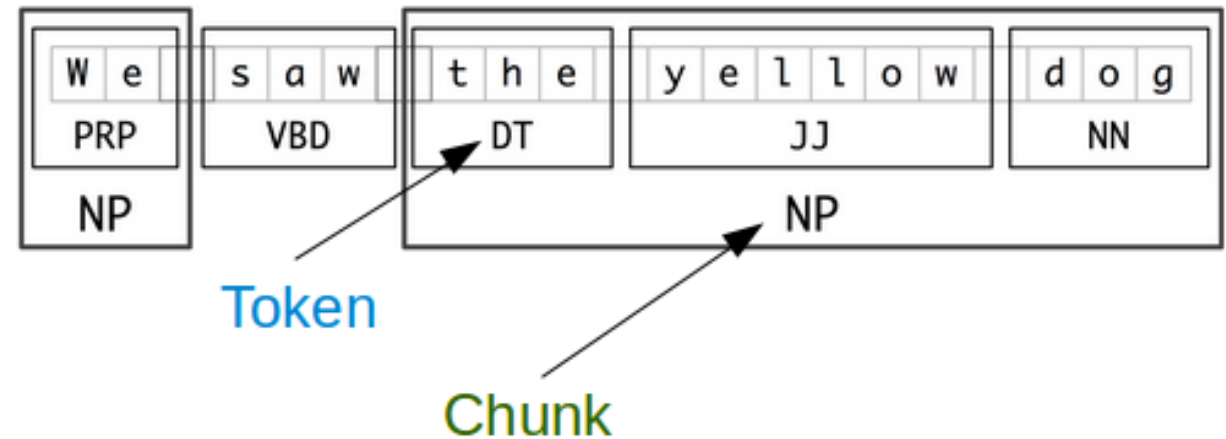
eg) “You are losing weight too fast” => ‘too fast’

Adverb phrase (ADVP):

eg) “I think it is pretty much the same” => ‘pretty much’

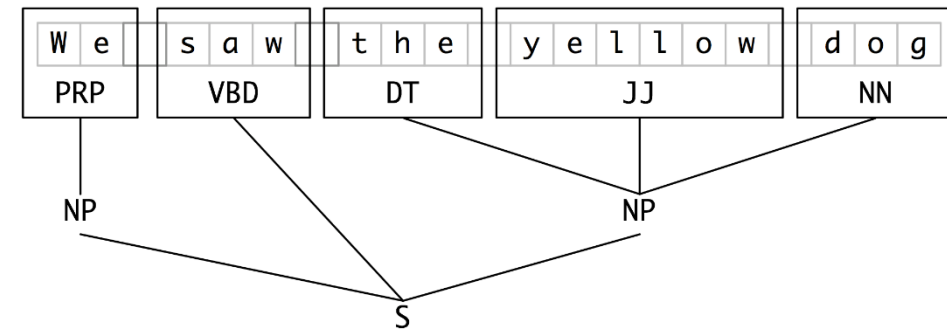
Prepositional phrase (PP):

eg) “Go down the hallway” => ‘down’



Chunking

- In part of speech tagging, we tag individual words.
- **Chunking** works on top of POS tagging and it chunks together set of tokens like Verb phrase or Noun phrase. The process of extracting phrases, or 'chunks' of texts from unstructured text.
- It is a very important concept if you are working with unstructured data and you want to obtain information from it.
- A common group of chunking is the noun phrase chunk (NP chunk).
- POS tags are used to create chunk grammar.
- Regex is used to make chunk grammar rules.



Chunking

- Noun Phrase Chunk:
- It follows a rule which determines if the context it takes into consideration represents a Noun phrase.
- If the function finds a **Determiner** followed by an **Adjective** and then a **Noun** then the chunk will be tagged as a Noun Phrase.

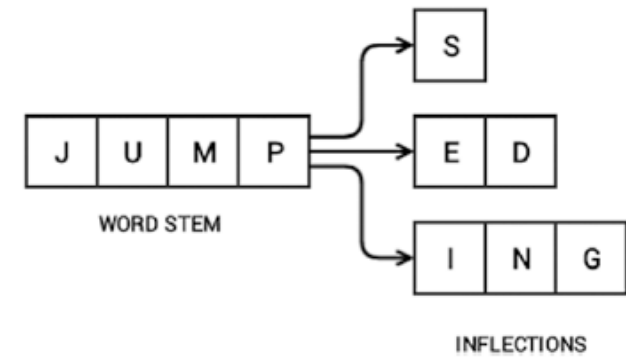
```
grammar = "NP: {<DT>?<JJ>*<NN>}"
```

Your chunks:

1. Start with an optional (?) determiner ('DT')
2. Can have any number (*) of adjectives (JJ)
3. End with a noun (<NN>)

Word stemming

- Word stem is the base form of a word where we can create new words by attaching affixes to them.
 - There are different algorithms, such as Porter Stemmer, Snowball Stemmer, Lancaster Stemmer etc.
 - **Porter Stemmer:** The Porter stemming algorithm (or “Porter stemmer”) uses suffix-stemming to produce stems.
 - **Snowball Stemmer:** Upgraded version, a.k.a. Porter2 Stemmer.
 - **Lancaster Stemmer:** Compared to snowball and porter stemming, lancaster is the most aggressive stemming algorithm because it tends to over-stem a lot of words. It tries to reduce the word to the shortest stem possible.
- => Stemming is fast, it is not 100% accurate.



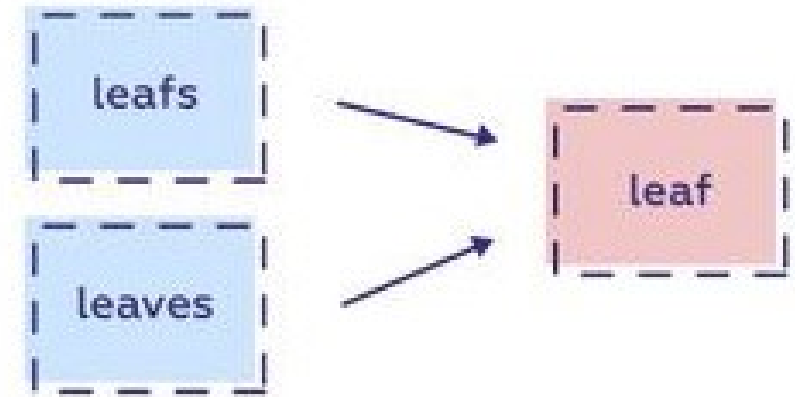
Note!

The purpose of the Porter stemmer is not to produce complete words but to find variant forms of a word.

Lemmatization

- Lemmatization is similar to stemming, but base form for lemmatization is the root word.
- In Lemmatization, the parts of speech(POS) will be determined first, unlike stemming which stems the word to its root form without considering the context.
- Lemmatization always considers the context and converts the word to its meaningful root/dictionary(WordNet) form called Lemma.
- It is important to do POS tagging before using this algorithm or it would assume every word as a noun.
- we will use NLTK package module 'WordNet' for lemmatization.

Lemmatization



Word stemming & lemmatization

- **Stemming** usually refers to a crude heuristic process that chops off the ends of words in the hope of achieving this goal correctly most of the time, and often includes the removal of derivational affixes.
- **Lemmatization** usually refers to doing things properly with the use of a vocabulary and morphological analysis of words, normally aiming to remove inflectional endings only and to return the base or dictionary form of a word, which is known as the lemma.

Stemming

adjustable → adjust
formality → formaliti
formaliti → formal
airliner → airlin ⚠

Lemmatization

was → (to) be
better → good
meeting → meeting

Stemming vs Lemmatization

change
changing
changes
changed
changer

→

chang

change
changing
changes
changed
changer

→

change

Let's go to Jupyter notebook 😊

