

Learner Corpus Research Summer School

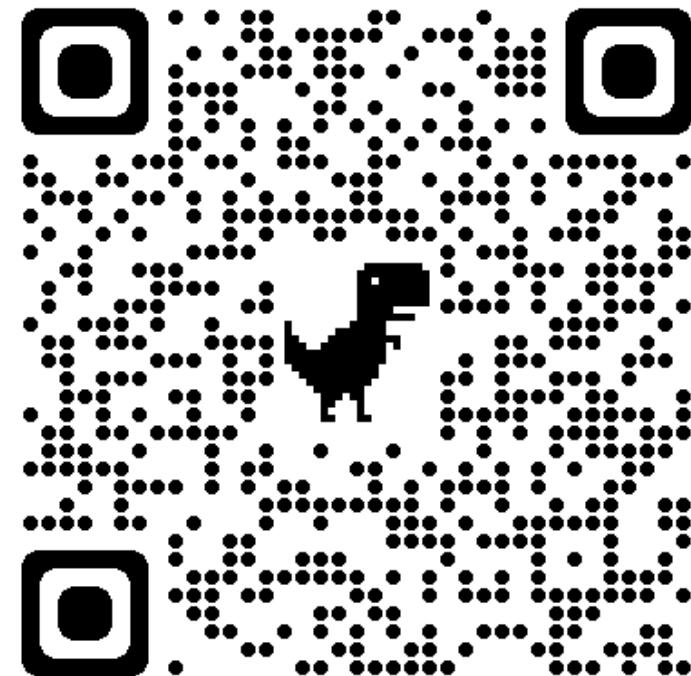
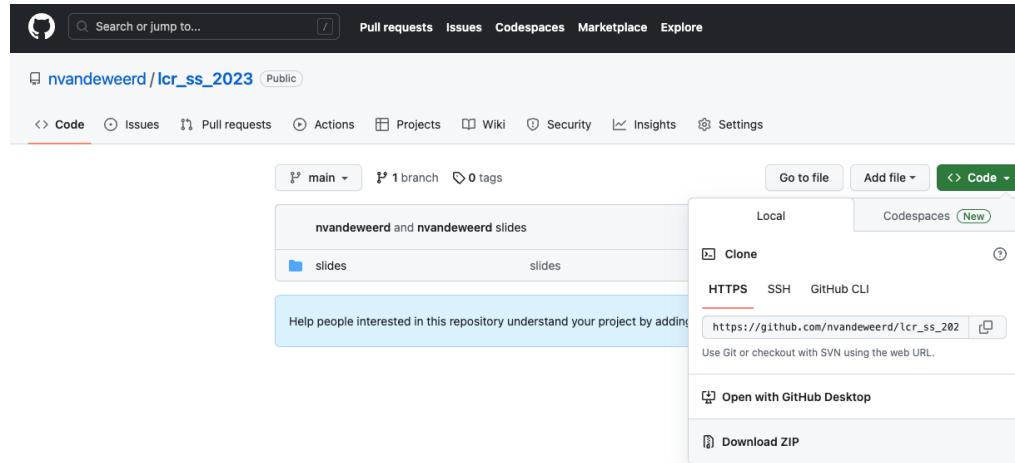
Automatic annotation of learner corpus data

Dr. Nathan Vandeweerd

Louvain-la-Neuve, Belgium

July 4th, 2023

All materials and slides available on GitHub



1. Click on < > Code .
2. Download ZIP to download all files.

L

Introduction

About me

☰ Radboud University Nijmegen

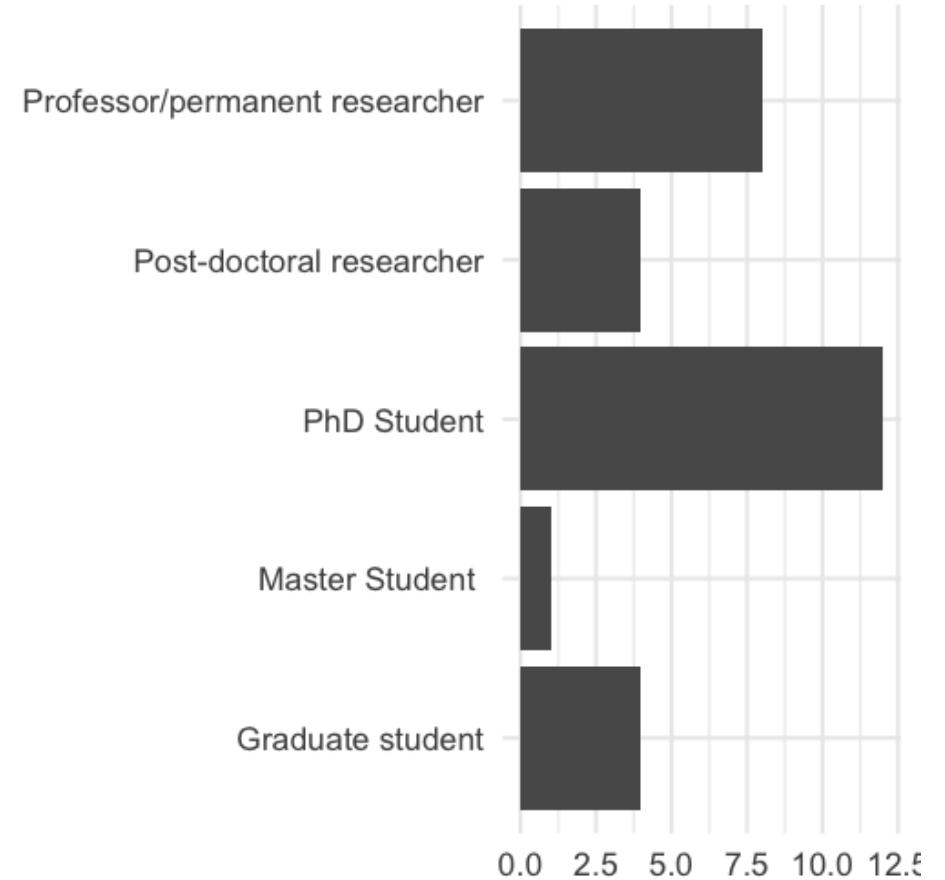
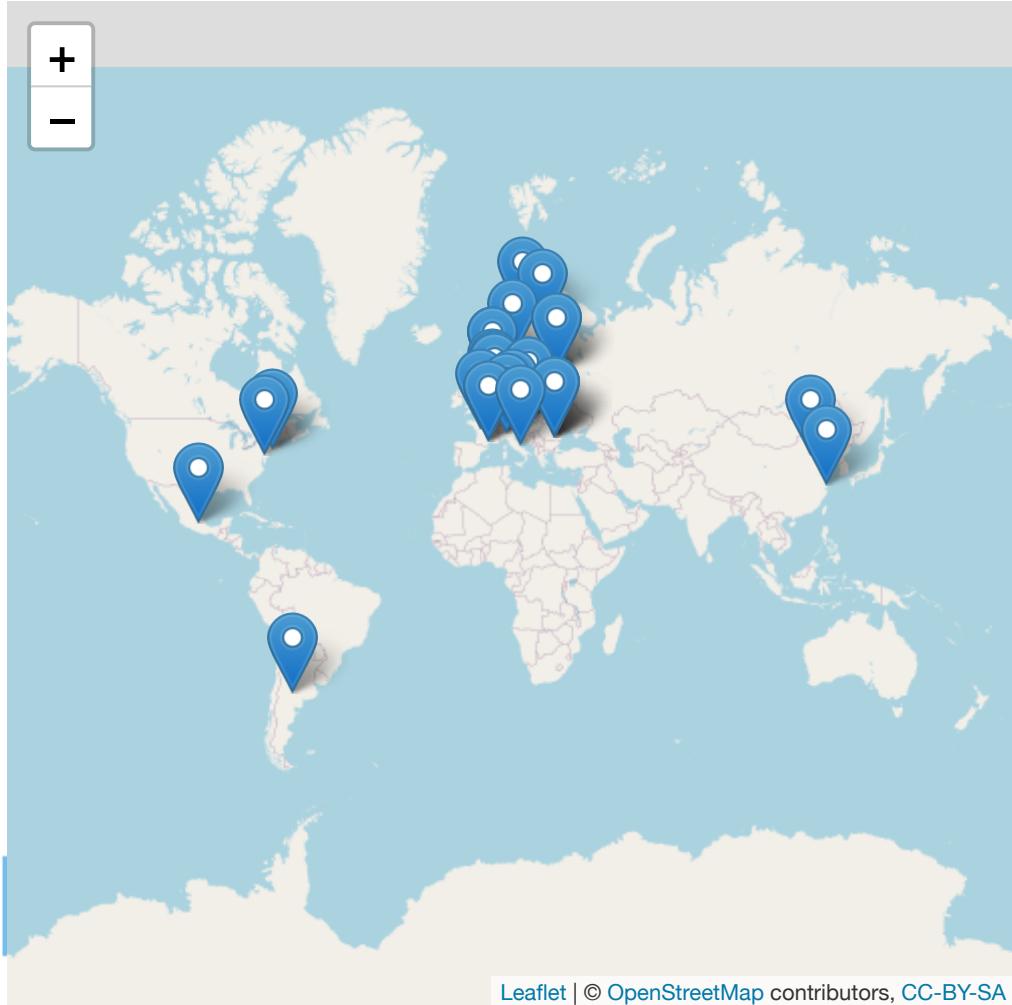
☒ Assistant professor in *Language and Communication*

❑ Research interests:

- Phraseological complexity in L2 French
- Accuracy of automatic transcription software for L2 data
- Language development during study abroad
- Crowdsourcing language assessment project



About you



Your goals

"the correlation between corpus building and its exploitation"

"most appropriate scientific approaches that one may use in analysing data from a learner corpus"

"I would like to know which new (or less new) software might be used to analyse the corpus."

"how to build my own machine-learning scheme when using topic modelling?"

"Secondly, I wish to get some theoretical knowledge & practical skills on programming"



What types of automatic annotation exist?

1. Part of Speech (POS) tagging



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(1) We_PPIS2 find_VV0 that_CST in_II fact_NN1 these_DD2 people_NN are_VBR the_AT most_RGT exposed_JJ to_II media_NN not_XX to_TO mension_VVI the_AT fact_NN1 that_CST there_EX is_VBZ forever_RT AIDS_NP1 awareness_NN1 campaigns_NN2 launget_VVN through_RP out_RP the_AT county_NN1.

(ICLE-TS-NOUN-0005.1)

(van Rooy, 2015: 80)

Q Your research:

- Spanish articles
- morpho-syntactic and syntactic labels
- morphology
- semantic, syntactic and discourse features
- verb valency patterns
- grammatical complexity
- grammatical case
- cohesion and cohesive devices
- stance features
- verb aspect
- adverbs of degree and negation
- verb phrase ellipsis
- colour terms
- meta-discourse markers
- formulaic language



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What types of automatic annotation exist?

1. Part of Speech (POS) tagging
2. Syntactic parsing

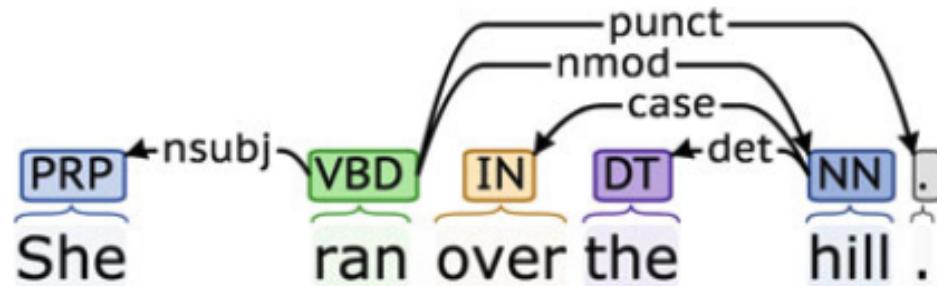
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(Newman and Cox, 2021: 32)

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What types of automatic annotation exist?

1. Part of Speech (POS) tagging
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3. Semantic annotation

- (6) a. The_Z5 ending_T2- of_Z5 the_Z5 poem_Q3 may_A7+ seem_A8 to_Z5
be_A3+ contradictory_A6.1- because_Z5/A2.2 both_N5 girls_S2.1f
marry_S4 and_Z5 have_A9+ children_S2mf/T3- ;_PUNC thereby_Z5
filling_N5.1+ the_Z5 traditional_S1.1.1 female_S2.1 role_I3.1 ._PUNC
b. at_T1.1.2[i165.3.1 a_T1.1.2[i165.3.2 time_T1.1.2[i165.3.3

(Newman and Cox, 2021: 35)

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What types of automatic annotation exist?

1. Part of Speech (POS) tagging
2. Syntactic parsing
3. Semantic annotation



What are the advantages of automatic annotation?

:

1. speed
2. reliability



Today's session

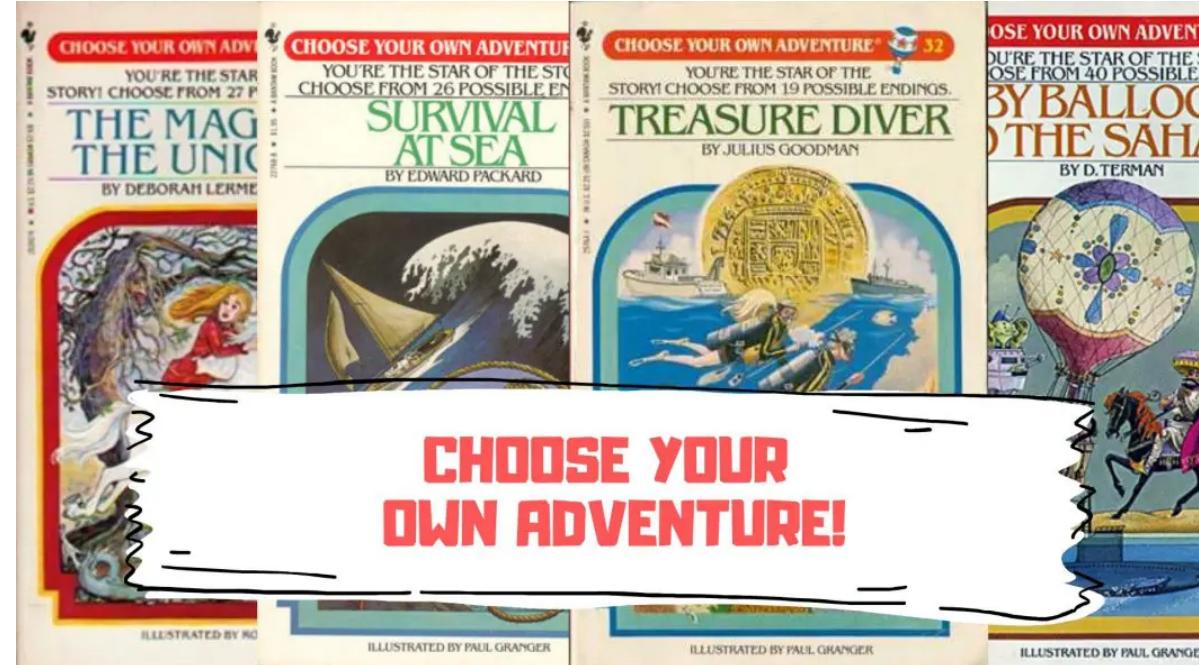
- Text-preprocessing
- POS-tagging and lemmatization
- Hands on activity: *POS-tagging and lemmatization*
- Syntactic annotation

Coffee break

- Hands on activity: *Syntactic annotation*
- Accuracy measures
- Final remarks



How this workshop will work



Option 1. Webtools/Excel

Option 2. R



Text-preprocessing

What do you notice about this text?

I agree that successful people try <e>news</e> things and take risk rather than only doing what they already know how to do well, for these reasons; By trying new things allow you to be curious to know how someone did it and you will find out how to do it too, that way it make you make a research. By trying new things allow you to be positive in your mind and to have a great desire to succed no matter how difficult is the situation. By trying new things you no that you should be openminded, go through dissussion with people who have done the same thing to learn their ways of doing thing, you should meet or have conversation with a lot of these people in other to learn from their experiences. By trying new things you take a big risks, like in " french we say if you don 't risk you don't have anything", risk in a goog way to takle something. We never know if we might succed or not the only way to do it is to risk. Since we do not loose anything when we risk.ed As for me doing the same thing every day become boring, i can say is a waste of energy and time, To conclude, people who succed try new things and take risks rather than only doing what they already know how to do well.



⚠ Ignore spelling mistakes for the time being...

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I agree that successful people try <e>news</e> things and take risk rather than only doing what they already know how to do well, for these reasons; By trying new things allow you to be curious to know how someone did it and you will find out how to do it too, that way it make you make a research. By trying new things allow you to be positive in your mind and to have a great desire to succed no matter how difficult is the situation. By trying new things you no that you should be

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(Non-exhaustive) list of things that can cause issues for automatic tools:

- (Inconsistent) file encoding
- Spacing
 - Lack of space between words
 - Unnecessary space between words
 - Double space between words
- 'Stylized' apostrophes or quotation marks
- Accented characters (e.g., à)
- Special characters (e.g., *, %, |)
- Inconsistent spelling rules (e.g., email/e-mail)
- Coding schemes (e.g., XML)



What do you notice about this text?

I agree that successful people try <e>news</e> things and take risk rather than only doing what they already know how to do well, for these reasons; By trying new things allow you to be curious to know how someone did it and you will find out how to do it too, that way it make you make a research. By trying new things allow you to be positive in your mind and to have a great desire to succed no matter how difficult is the situation. By trying new things you no that you should be

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⚠ Importance of **knowing your corpus...**

Methods of text cleaning/preprocessing

+ Time intensive

- Replicable

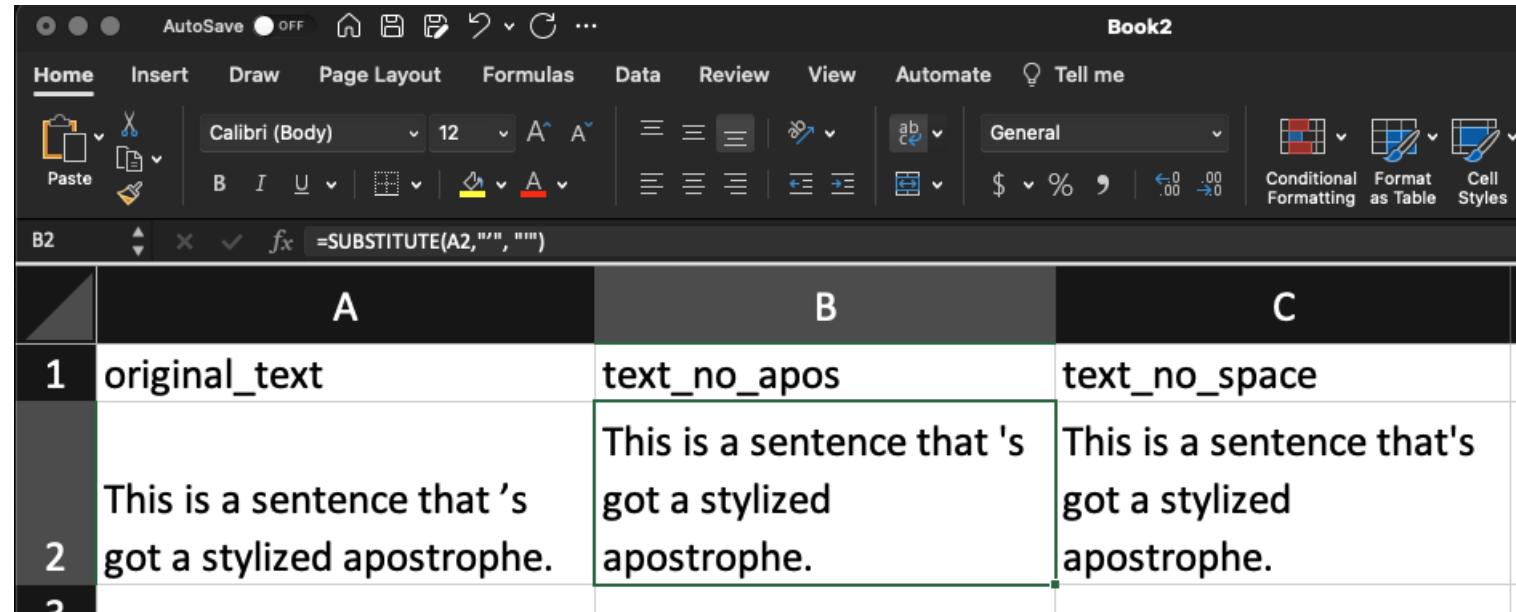
1. Manually
2. Semi-manually using search and replace (e.g., in Excel)
3. Semi-manually using regular expressions (e.g., in a text editor)
4. Semi-automatically using a script with regular expressions (e.g., in R or python)

- Time intensive

+ Replicable



Text pre-processing in Excel



	A	B	C
1	original_text	text_no_apos	text_no_space
2	This is a sentence that 's got a stylized apostrophe.	This is a sentence that 's got a stylized apostrophe.	This is a sentence that's got a stylized apostrophe.

SUBSTITUTE() function (For more details see [here](#))

```
=SUBSTITUTE(text, old_text, new_text, [instance_num])  
=SUBSTITUTE(A2,"'", "")
```



Text pre-processing in R

```
library(stringr); library(dplyr)

text <- "This is a sentence that 's got a stylized apostrophe."

text %>%
  # replace stylized apostrophes
  str_replace_all("'", "") %>%
  # remove spaces before apostrophes
  # if they are:
  # - preceded by the beginning of a string, space or final punctuation
  # - followed by only one or two letters (e.g., don't, they're)
  # - and then followed by either a space, final punctuation or end of string
  str_replace_all("[\\s\\.\\?\\!]'([^\s]{1,2}[\\s\\.\\?\\!$])", "'\\1")  
## [1] "This is a sentence that's got a stylized apostrophe."
```



With both approaches

⚠ Test (and re-test) your pre-processing pipeline!

- Be careful of inadvertent changes (especially when using regular expressions)
E.g. *don't* vs. *he said 'don't worry'*
- Be aware the different tools require different approaches (e.g., contractions separate or apart)
- Never edit the original corpus files!
- Keep track of any changes you make

```
text <- "'He 'll be comin' round the 'mountain' when he comes,' I said."  
  
text %>%  
  str_replace_all("", "") %>%  
  str_replace_all("[\\s\\.\\?\\!^]([^\s]{1,2}[\\s\\.\\?\\!$])", '\\1')  
  
## [1] "'He'll be comin' round the 'mountain' when he comes,' I said."
```



So why is pre-processing so important?

```
[1] "I"                      "agree"
[3] "that"                   "successful"
[5] "people"                 "try"
[7] "&lt;e&gt;news&lt;/e&gt;" "things"
[9] "and"                    "take"
[11] "risk"                   "rather"
[13] "than"                   "only"
[15] "doing"                  "what"
[17] "they"                   "already"
[19] "know"                   "how"
[21] "to"                     "do"
[23] "well,for"               "these"
[25] "reasons;"               ""
[27] "By"                     "trying"
[29] "new"                    "things"
```



Tokenization

Token: "the smallest unit of a corpus" (Krause, Lüdeling, Odebrecht, and Zeldes, 2012: 2)

= words, numbers, punctuation marks, quotation marks etc.

= syllable, phoneme, etc...

Easiest method (for English):

- split tokens at spaces
- split sentences at periods (.), exclamation marks (!) or question marks (?)

What problems do you see with this method?

Potential problems:

- clitics (isn't, ain't)
- missing whitespace
- periods (etc., U.S.A., fig.)
- ordinal numbers
- multiword expressions (New York-based, 10 000, as well as)
- word-internal punctuation (relationship(s), "Rambo"-type)
- (de)hyphenation (preprocessing vs. pre-processing)
- quoted speech ("You still don't have an accountant?" Ellis said.)
- ideographic languages (e.g., Chinese, Japanese)

Q For more information about tokenization
see Zeldes (2020) and Schmid (2008).



POS-tagging and lemmatization

Overview

Each word in the corpus is 'tagged' (labelled) with information about its grammatical category.

🔧 Under the hood:

1. All tokens with unambiguous POS labels are assigned tags (e.g., on the basis of a dictionary)
2. Contextual features (e.g., surrounding tags, morphological endings) used in a statistical model to predict the tags of ambiguous items

(Kyle, 2021: 6)

She 🏷 pronoun
sells 🏷 verb
seashells 🏷 noun
by 🏷 preposition
the 🏷 determiner
seashore 🏷 noun
. 🏷 punctuation

Tagsets for English

- CLAWS (Constituent Likelihood Automatic Word-tagging System)
 - CLAWS 5 = 60 tags
 - CLAWS 7 = 160 tags
- PENN Treebank Tagset
- BNC Tagset
- Universal POS tags

Different tagsets, (subtly) different theories of grammar

Table 2.1 Four tagging solutions for English *rid*

	<i>I am now completely rid of such things</i>	<i>You are well rid of him</i>	<i>I got rid of the rubbish</i>
CLAWS7 tagger ^a	Past participle	Past participle	Past participle
Infogistics ^b	Verb base	Verb base	Past participle
FreeLing ^c	Adjective	Verb base	Past participle
(Brill-based) GoTagger ^d	Adjective	Adjective	Adjective

(Newman and Cox, 2021: 21)

💡 Select the tagset that is right for your data/research question.

Example

- (1) We_PPIS2 find_VV0 that_CST in_II fact_NN1 these_DD2 people_NN are_VBR the_AT most_RGT exposed_JJ to_II media_NN not_XX to_TO mension_VVI the_AT fact_NN1 that_CST there_EX is_VBZ forever_RT AIDS_NP1 awareness_NN1 campaigns_NN2 launaged_VVN through_RP out_RP the_AT county_NN1..

(ICLE-TS-NOUN-0005.1)

- (2) We find that in fact these people are the most exposed to media not to mension the fact that there is forever AIDS awareness campaigns launaged through out the county.

(van Rooy, 2015: 80-81)

Verbs: launaged_VVD, find_VV0, is_VBZ, are_VBR

Nouns: media_NN, fact_NN1, AIDS_NN1, county_NN1, awareness_NN1, people_NN, mension_NN1, campaigns_NN2, fact_NN1

= CLAWS C7 Tagset

② What does the tag PPIS2 refer to?

② What do you notice about the learner errors? (e.g., 'mension', 'is awareness campaigns')



Types of output: *Horizontal*

We_PPIS2 find_VV0 that_CST in_II fact_NN1 these_DD2 people_NN are_VBR the_AT most_RGT exposed_JJ to_II media_NN not_XX to_II mension_NN1 the_AT fact_NN1 that_CST there_EX is_VBZ forever_RT AIDS_NN1 awareness_NN1 campaigns_NN2 launaged_VVD through_RP out_RP the_AT county_NN1 ._.



Types of output: (*Pseudo*)-XML

```
<w id="2.1" pos="PPIS2">We</w>
<w id="2.2" pos="VV0">find</w>
<w id="2.3" pos="CST">that</w>
<w id="2.4" pos="II">in</w>
<w id="2.5" pos="NN1">fact</w>
<w id="2.6" pos="DD2">these</w>
<w id="2.7" pos="NN">people</w>
<w id="2.8" pos="VBR">are</w>
<w id="2.9" pos="AT">the</w>
<w id="2.10" pos="RGT">most</w>
<w id="2.11" pos="JJ">exposed</w>
<w id="2.12" pos="II">to</w>
<w id="2.13" pos="NN">media</w>
<w id="2.14" pos="XX">not</w>
<w id="2.15" pos="II">to</w>
<w id="2.16" pos="NN1">mension</w>
<w id="2.17" pos="AT">the</w>
<w id="2.18" pos="NN1">fact</w>
<w id="2.19" pos="CST">that</w>
<w id="2.20" pos="EX">there</w>
```



Types of output: *Vertical*

```
##   idx sntc  token tag lttr      wclass
## 1    1    1     We  PP    2    pronoun
## 2    2    1    find VBP    4    verb
## 3    3    1    that IN    4 preposition
## 4    4    1     in IN    2 preposition
## 5    5    1    fact NN    4    noun
## 6    6    1   these DT    5 determiner
## 7    7    1  people NNS   6    noun
## 8    8    1     are VBP   3    verb
## 9    9    1     the DT    3 determiner
## 10   10   1    most RBS   4    adverb
## 11   11   1 exposed VBN   7    verb
## 12   12   1     to TO    2    to
## 13   13   1   media NNS   5    noun
## 14   14   1     not RB    3    adverb
## 15   15   1     to TO    2    to
## 16   16   1 mension NN    7    noun
## 17   17   1     the DT    3 determiner
## 18   18   1    fact NN    4    noun
## 19   19   1    that IN    4 preposition
## 20   20   1   there EX    5 existential
```



Lemmatization

Lemma: "a 'base form', which provides a level of abstraction from any inflection that might appear in the original orthographic word."

(Newman and Cox, 2021: 29)

- (2) We find that in fact these people are the most exposed to media not to mention the fact that there is forever AIDS awareness campaigns launched through out the county.

we find that in fact these people be the most expose to medium not to mention the fact that there be forever AIDS awareness campaign launched through out the county .

Example

```
##   idx sntc  token tag      lemma lttr      wclass
## 1    1    1     We  PP       we     2  pronoun
## 2    2    1    find VBP     find    4    verb
## 3    3    1    that IN      that    4 preposition
## 4    4    1     in  IN      in     2 preposition
## 5    5    1   fact  NN      fact    4    noun
## 6    6    1   these DT     these   5 determiner
## 7    7    1  people NNS    people   6    noun
## 8    8    1     are VBP     be     3    verb
## 9    9    1     the DT      the     3 determiner
## 10   10   1    most RBS    most    4 adverb
## 11   11   1 exposed VBN   expose   7    verb
## 12   12   1     to  TO      to     2    to
## 13   13   1   media NNS   medium   5    noun
## 14   14   1     not RB     not    3 adverb
## 15   15   1     to  TO      to     2    to
## 16   16   1 mension NN <unknown> 7    noun
## 17   17   1     the DT      the    3 determiner
## 18   18   1   fact  NN      fact    4    noun
## 19   19   1     that IN     that    4 preposition
## 20   20   1   there EX     there   5 existential
```



Webtools



TreeTagger

Online TreeTagger

Annotate your texts with part-of-speech and lemma information using [TreeTagger](#).

Type a text Upload a file

Text to process*

Language of your text*

CLAWS

UCREL API

Free CLAWS web tagger

Our free web tagging service offers access to the latest version of the tagger, CLAWS4, which was used to POS tag c.100 million words of the original [British National Corpus \(BNC1994\)](#), the [BNC2014](#), and all the English corpora in Mark Davies' [BYU corpus server](#). You can choose to have output in either the smaller [C5 tagset](#) or the larger [C7 tagset](#).

[CLAWS POS tagger](#) | [Obtaining a licence](#) | [Tagging service](#)

If you would like to use our free WWW tagger, please complete the form below. You can enter up to 100,000 words of English running text. If you enter more, it will be cut off at the word limit. [Input format guidelines](#) are available. To tag the text you have entered click the button below the form.

Select tagset: C5 C7

Select output style: Horizontal Vertical Pseudo-XML

Type (or paste) your text to be tagged into this box.





```
library(koRpus); library(koRpus.lang.en)

file <- "data/example_texts/ICLE-TS-NOUN-0005.1.txt"

treetag(
  file,
  treetagger="manual",
  lang="en",
  TT.options=list(
    # Change this to the location where TreeTagger is installed
    path="/Applications/tree-tagger",
    preset="en"
  ),
  doc_id=basename(file)
)
```

⚠ Note: For this to work, both TreeTagger and the appropriate tagsets must first be installed locally on your computer. See instructions [here](#)

💡 See this vignette for more information about the **koRpus** package.

Activity 1: POS-tagging



Webtools/Excel Option

Open `activity_02_pos-tagging.docx` and follow the instructions.

R Option

Open `activity_02_pos-tagging.R` and follow the instructions.

💡 Remember that all materials and sides available on GitHub.

1. Click on `< > Code`.
2. Download ZIP to download all files.



Activity 1: POS-tagging



Webtools/Excel Option

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Questions

1. What do you notice about the ICLE texts? What pre-processing steps (if any) might be necessary before using automatic annotation tools?
2. R: What 'special characters' do you notice?
3. R: How many 'words' and 'sentences' does BRFF1065.txt contain?
4. R: What is the average sentence length of BRFF1065.txt?
5. WT: How many tokens were tagged with CLAWS (v5)?
6. WT: What is the tag for the base form of a lexical verb in the C5 tagset?
7. WT: What is the tag for the base form of a lexical verb in the C7 tagset?
8. WT: What is meant by '[VVZ/86] NN2/14'?
9. WT: What does the code '\@card\@' mean?
10. How many adjectives (JJ) are there in BRFF1065.txt?
11. How many common nouns (NN, NNS) are there in BRFF1065.txt?
12. How should you best deal with unknown lemmas?
13. What are the most frequent verb lemmas in the corpus?



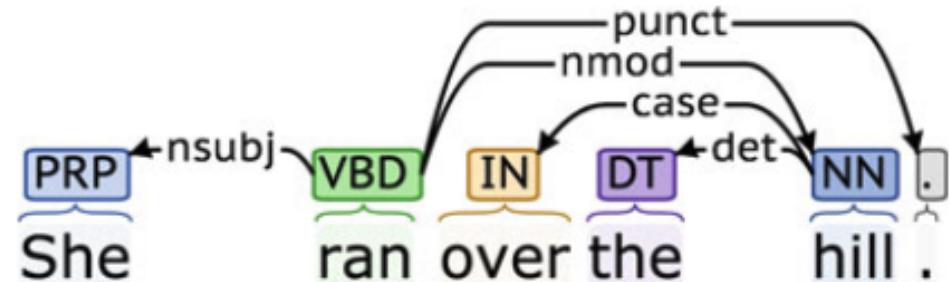
Syntactic parsing

Overview

Labels of the syntactic connections between words (heads and dependents)

🔧 Under the hood:

1. Texts are POS-tagged.
2. POS tags used in conjunction with phrase-structure rules (generated from training algorithms on large corpora) to generate several possible *parse trees* for each sentence.
3. Statistical or machine learning algorithms are used to select the most probable parse tree for a given sentence.



(Kyle, 2021: 7)



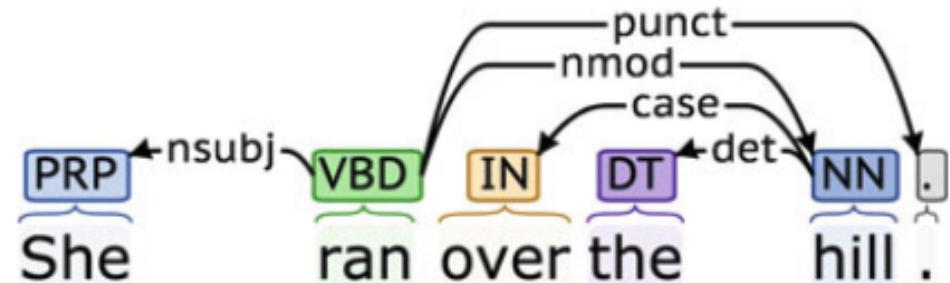
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(Kyle, 2021: 7)



(3) Parse of *She ran over the hill.*
(ROOT
(S
 (NP (PRP She))
 (VP (VBD ran))
 (PP (IN over))
 (NP (DT the) (NN hill)))
 (. .)))

Dependency models for English

- Stanford CoreNLP
- spaCy
- Universal dependencies

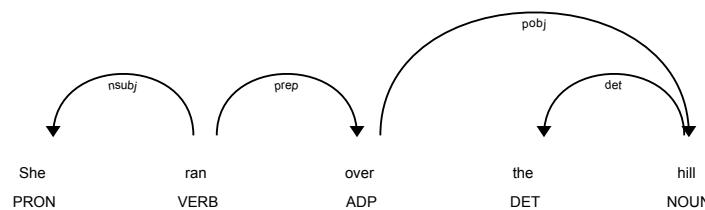
Different models, (subtly) different theories of grammar



Types of output: CoNLL

= Conference on Natural Language Learning

doc_id	sentence_id	token_id	token	lemma	pos	head_token_id	dep_rel
1	text1	1	1	She	she	PRON	2 nsubj
2	text1	1	2	ran	run	VERB	2 ROOT
3	text1	1	3	over	over	ADP	2 prep
4	text1	1	4	the	the	DET	5 det
5	text1	1	5	hill	hill	NOUN	3 pobj
6	text1	1	6	.	.	PUNCT	2 punct



Types of output: *FoLiA XML*

= Format for Linguistic Annotation

```
<p xml:id="example.p.1">
  <t>He hits Mr. Smith. That came quite expected!</t>
  <s xml:id="example.p.1.s.1">
    <t offset="0">He hits Mr. Smith.</t>
    <w xml:id="example.p.1.s.1.w.1"><t offset="0">He</t></w>
    <w xml:id="example.p.1.s.1.w.2"><t offset="3">hits</t>
      <morphology>
        <morpheme class="lexical" function="lexical">
          <t offset="0">hit</t>
        </morpheme>
        <morpheme class="suffix" function="inflectional">
          <t offset="3">s</t>
        </morpheme>
      </morphology>
    </w>
    <w xml:id="example.p.1.s.1.w.3"><t offset="8">Mr.</t></w>
    <w xml:id="example.p.1.s.1.w.4" space="no"><t offset="10">Smith</t></w>
    <w xml:id="example.p.1.s.1.w.5"><t offset="15">. </t></w>
  </s>
  ..
</p>
```



Types of output: json

```
[1] {"  
[2]   "text": "She ran over the hill",  
[3]   "ents": [],  
[4]   "sents": [  
[5]     {"  
[6]       "start": 0,  
[7]       "end": 21  
[8]     }  
[9]   ],  
[10]  "tokens": [  
[11]    {"  
[12]      "id": 0,  
[13]      "start": 0,  
[14]      "end": 3,  
[15]      "tag": "PRP",  
[16]      "pos": "PRON",  
[17]      "morph": "Case=Nom|Gender=Fem|Number=Sing|Person=3|PronType=Prs",  
[18]      "lemma": "she",  
[19]      "dep": "nsubj",  
[20]      "head": 1  
[21]    },  
[22]    {"  
[23]      "id": 1,  
[24]      "start": 4,  
[25]      "end": 7,  
[26]      "tag": "VBD",  
[27]      "pos": "VERB",  
[28]      "morph": "Tense=Past|VerbForm=Fin",  
[29]      "lemma": "run",  
[30]      "dep": "ROOT",  
[31]      "head": 1
```

💡 Can easily be converted into CSV/Excel format.

Webtools



CoreNLP

The screenshot shows the CoreNLP web interface. At the top is the CoreNLP logo with three orange arches and the text "CoreNLP" and "version 4.4.0". Below is a text input field containing the sentence "e.g., The quick brown fox jumped over the lazy dog.". Underneath the input field are three buttons: "Annotations", "Language" (set to English), and "Submit".

⚠ Note: May not be available due to planned outage from June 24th-July 3rd.

Hugging Face spaCy visualizer

The screenshot shows the Hugging Face spaCy visualizer. At the top is the Hugging Face logo and search bar. Below is a sidebar with "spaCy" and "Explore trained spaCy pipelines" for the "en_core_web_sm" model. The main area displays the sentence "Apple is looking at buying U.K. startup for \$1 billion" with a dependency parse diagram. The diagram shows arrows indicating grammatical relationships between words: "Apple" (PROPN) is the subject ("nsubj" arrow to "is" (AUX)); "is" is a verb ("verb" arrow to "looking" (VERB)); "looking" is the object ("dobj" arrow from "looking" to "buying" (VERB)); "buying" is the object ("dobj" arrow from "buying" to "U.K." (ADP)); "at" (ADP) is a preposition ("prep" arrow from "at" to "looking"); and "\$1 billion" (NOUN) is the object ("dobj" arrow from "buying" to "\$1 billion").



R

```
library(spacyr)  
spacy_initialize(model = "en_core_web_sm")  
text <- "She ran over the hill."  
spacy_parse(text, dependency = TRUE)
```

⚠ Note: For this to work, spaCy must be installed locally on your computer but this can be done within the spaCy package using the `spacy_install()` function.

💡 See this vignette for more information about the `spacyr` package.

L

Activity 2: Syntactic parsing

Webtools/Excel Option

Open `activity_02_parsing.docx` and follow the instructions.

R Option

Open `activity_02_parsing.R` and follow the instructions.

💡 Remember that all materials and sides available on GitHub.

1. Click on `< > Code`.
2. Download ZIP to download all files.



Activity 2: Syntactic parsing

Webtools/Excel Option

Open `activity_02_parsing.docx` and follow the instructions.

R Option

Open `activity_02_parsing.R` and follow the instructions.

Questions

1. Which word is the final period dependent on?
2. What type of dependency relationships are marked by 'amod' and 'dobj'?
3. How many amod dependencies are there in this text?
4. What adjective modifies the word 'difference'?
5. What is the object of the verb 'control'?
6. R: What is the most frequency dependency relation in the corpus?
7. R: What is the average length of noun phrases in the corpus?

 Remember that all materials and sides available on GitHub.

1. Click on `< > Code`.
2. `Download ZIP` to download all files.



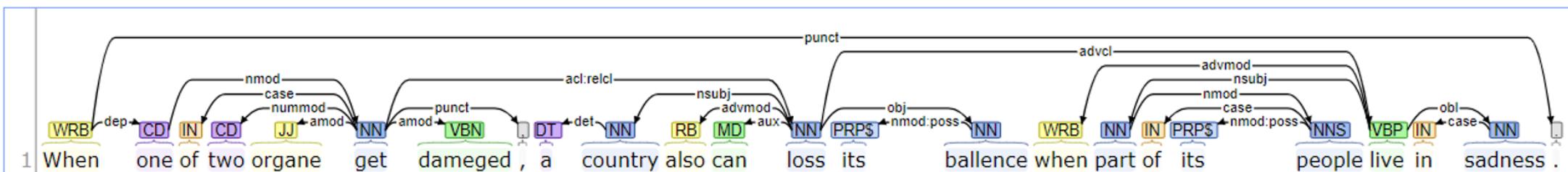
Accuracy

The effect of learner errors

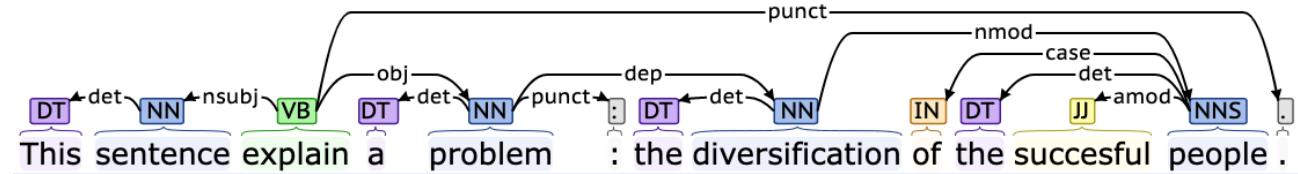
- (4) a. ... like body that will **loss** its **ballence** when one of two **organe** get **dameged**, a country also can **loss** its **ballence** when part of its people live in sadness.
- b. ... **peoples** who I met is not good ...
- c. So when I **admit** to korea university, I decide what i find my own way.
- d. Also, the people in it very friendly.

(Ragheb and Dickinson, 2012)

What problems do you see here?



Not all errors are equally as problematic



(Ragheb and Dickinson, 2012)

Evaluating the *reliability* of automatic tools

Confusion matrix: A tabulation of the agreement between manual and automatic annotation.

Example:

```
##      unit human computer
## 1    apple   noun     noun
## 2      run   verb     noun
## 3     cake   noun     verb
## 4   coffee   noun     noun
## 5    drink   verb     noun
## 6    swim   verb     verb
## 7     tart   noun     verb
## 8    walk   verb     noun
## 9     ice   noun     noun
## 10    pen   noun     noun
```

		Actual Values	
		Positive (1)	Negative (0)
Predicted Values	Positive (1)	TP	FP
	Negative (0)	FN	TN

Evaluating the *reliability* of automatic tools

Confusion matrix: A tabulation of the agreement between manual and automatic annotation.

		Reference	
		Prediction	noun
		noun	4
		verb	3
			1

Example:

```
##      unit human computer
## 1    apple noun     noun
## 2      run verb     noun
## 3     cake noun     verb
## 4   coffee noun     noun
## 5   drink verb     noun
## 6    swim verb     verb
## 7     tart noun     verb
## 8    walk verb     noun
## 9     ice noun     noun
## 10    pen noun     noun
```



Evaluating the *reliability* of automatic tools

Precision: the extent to which the retrieved objects in a query are correctly tagged

Recall/Sensitivity: the extent to which the objects matching the query retrieve all the target objects in the corpus

F-score/F1-score the balance between precision and recall (what is normally reported)



Evaluating the *reliability* of automatic tools

Precision: the extent to which the retrieved objects in a query are correctly tagged

$$\frac{TP}{(TP + FP)}$$

Recall/Sensitivity: the extent to which the objects matching the query retrieve all the target objects in the corpus

$$\frac{TP}{(TP + FN)}$$

F-score/F1-score the balance between precision and recall (what is normally reported)

$$\frac{2 * (P * R)}{P + R}$$

Reliability of automatic annotation for fsca tool
(Vandeweerd, 2021: 265)

Unit	Precision	Recall	F-score
Sentences	0.96	0.97	0.97
Clauses	0.75	0.72	0.74
Dependent Clauses	0.63	0.58	0.60
Coordinated Clauses	0.54	0.51	0.53
T-units	0.83	0.82	0.83
Noun Phrases	0.84	0.84	0.84
Verb Phrases	0.78	0.78	0.78

Webtools



Confusion Matrix Online Calculator

Confusion Matrix About Measures Example Contact

Confusion Matrix

Online Calculator

		True Positive	True Negative
Predicted Positive	True Positive		False Positive
	False Negative		True Negative
Predicted Negative			

Calculate



```
library(caret)  
confusionMatrix(data, reference)
```

💡 See this vignette for more information about the **caret** package.

```
# Confusion Matrix and Statistics  
#  
#           Reference  
# Prediction M R  
#           M 21 7  
#           R  6 17  
#  
#           Accuracy : 0.745  
#                 95% CI : (0.604, 0.857)  
#   No Information Rate : 0.529  
#   P-Value [Acc NIR] : 0.00131  
#  
#           Kappa : 0.487  
#  
#   Mcnemar's Test P-Value : 1.00000  
#  
#           Sensitivity : 0.778  
#           Specificity : 0.708  
#   Pos Pred Value : 0.750  
#   Neg Pred Value : 0.739  
#           Prevalence : 0.529  
#   Detection Rate : 0.412  
# Detection Prevalence : 0.549  
#       Balanced Accuracy : 0.743  
#  
#   'Positive' Class : M  
#
```

Final remarks

Recap

L



Your goals



Q

& Eh?

L

Further reading

Kyle, K. (2021). Natural language processing for learner corpus research. *International Journal of Learner Corpus Research*, 7(1), 1–16.

Newman, J., & Cox, C. (2021). Corpus annotation. In *A practical handbook of corpus linguistics* (pp. 25–48). Springer. <https://doi.org/10.4324/9780429269035-7>

Ragheb, M., & Dickinson, M. (2012). Defining syntax for learner language annotation. *Coling2012*. 965–974.

Schmid, H. (2008). Tokenizing and part-of-speech tagging. In *Corpus Linguistics: An International Handbook*. de Gruyter.

van Rooy, B. (2015). Annotating learner corpora. In S. Granger, G. Gilquin, & F. Meunier (Eds.), *The Cambridge handbook of learner corpus research* (pp. 79–106). Cambridge University Press.
<https://doi.org/10.1017/CBO9781139649414.005>

Zeldes, A. (2020). Corpus Architecture. In M. Paquot & S. Th. Gries (Eds.), *A Practical Handbook of Corpus Linguistics* (pp. 49–73). Springer International Publishing. https://link.springer.com/10.1007/978-3-030-46216-1_3

