# Creating and Analysing Multilingually Comparable Text Corpora

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**Language Technology Initiative** 2.3.1.1.i.0/1/22/I/CFLA/002

# **Agenda**

A showcase: the **ParlaMint** corpora

Universal Dependencies: cross-linguistic grammatical annotation and analysis

## The task and the toolkit

Under the hood: small-scale yet efficient LMs for tagging & parsing - BERT

## Hands-on work

- The grunt work: data acquisition via web scraping
- The actual work: parsing unstructured text into structured data
- Enjoying the fruits: running corpus queries

## Comparable and interoperable parliamentary corpora

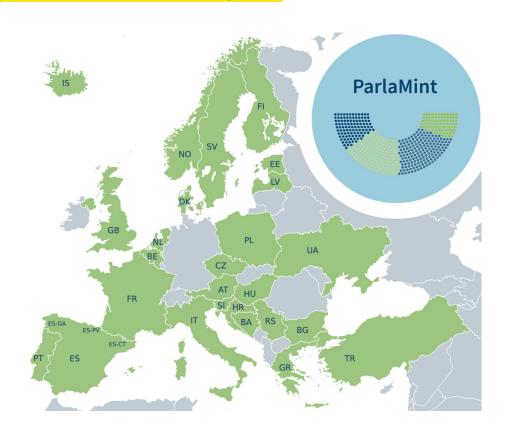
## https://www.clarin.eu/parlamint

Version **4.1** contains corpora for **29** countries and autonomous regions

Linguistically annotated corpora: **morphology**, **syntax**, named entities

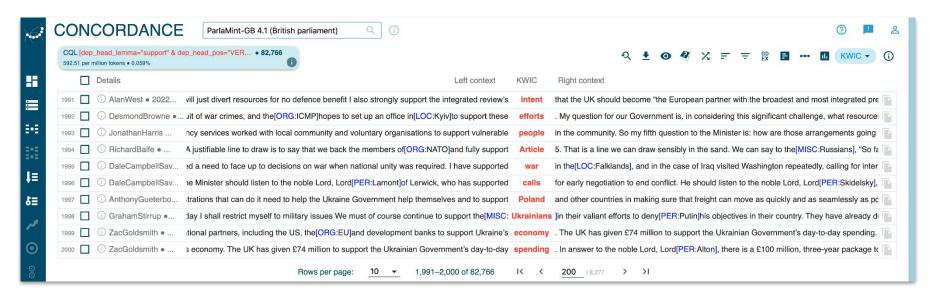
Interoperable\* corpora: TEI, **UD**, **VERT** 

Open data: <a href="http://hdl.handle.net/11356/1912">http://hdl.handle.net/11356/1912</a>
Open access: <a href="https://www.clarin.si/ske/#open">https://www.clarin.si/ske/#open</a>



<sup>\*</sup> FAIR: findability, accessibility, interoperability, reusability

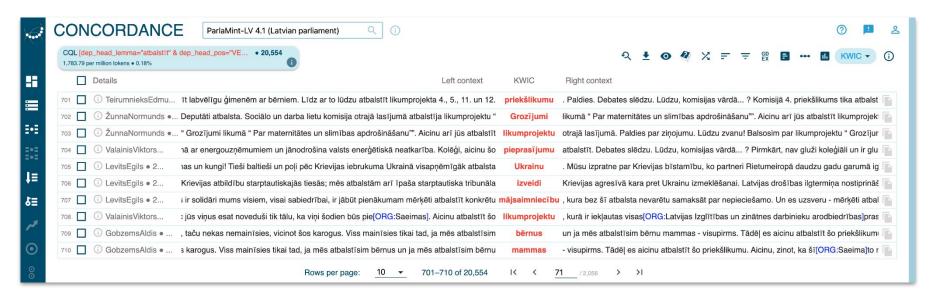
# Uniform querying in the ParlaMint corpora



```
[dep_head_lemma="support"
  & dep_head_pos="VERB"
  & dep="obj" & pos="NOUN|PROPN"]
```



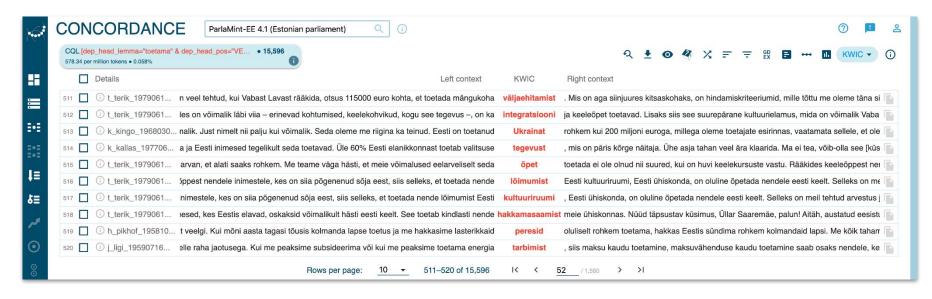
# Uniform querying in the ParlaMint corpora



```
[dep_head_lemma="atbalstīt"
  & dep_head_pos="VERB"
  & dep="obj" & pos="NOUN|PROPN"]
```



# Uniform querying in the ParlaMint corpora

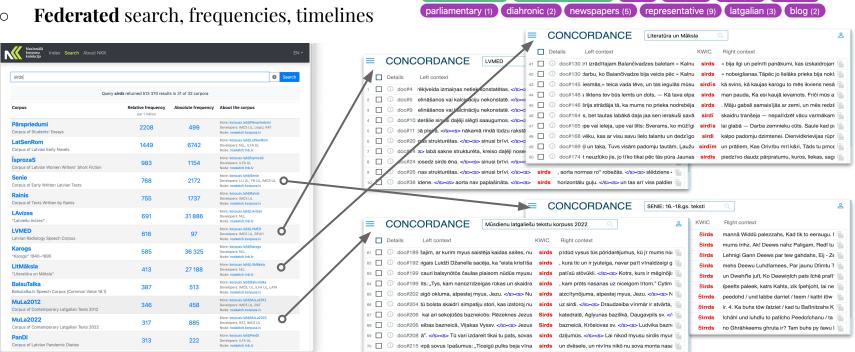


```
[dep_head_lemma="toetama"
  & dep_head_pos="VERB"
  & dep="obj" & pos="NOUN|PROPN"]
```



# Korpuss.lv: Latvian National Corpora Collection

- **39** corpora by **13** institutions, **2.8B** tokens
- **Unified** morpho-syntactic annotations



speech (7)

error annotation (2)

general (11)

manually annotated (5)

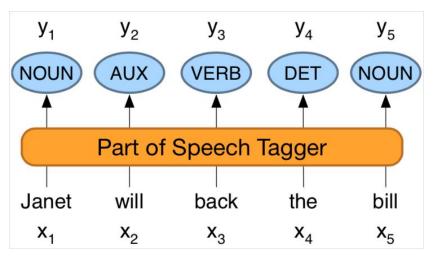
specialised (26)

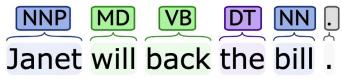
morphology (31)

learner (2)

# Part-of-Speech tagging

Sentence splitting » Tokenization » Tagging & Lemmatization » Parsing State-of-the-art POS taggers: above 95% accuracy





#### https://corenlp.run

INDEX	FORM	LEMMA	UPOSTAG	XPOSTAG			
#text=Dženeta atbalstīs likumprojektu .							
1	Dženeta	Dženeta	PROPN	npfsn4			
2	atbalstīs	atbalstīt	VERB	vmnift330an			
3	likumprojektu	likumprojekts	NOUN	ncmsa1			
4			PUNCT	zs			

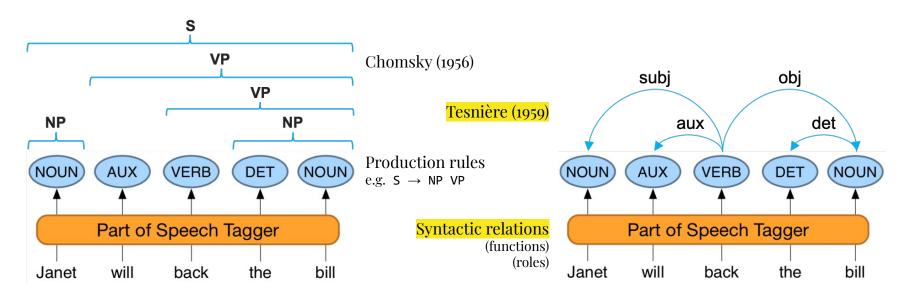
https://nlp.ailab.lv

https://web.stanford.edu/~jurafsky/slp3/

# **Syntactic parsing**

Constituency (CFG/PSG) vs. dependency (UD) parsing

State-of-the-art parsers: above 90% accuracy



# **Universal Dependencies**

## A Universal Part-of-Speech Tagset

Slav Petrov<sup>1</sup> Dipanjan Das<sup>2</sup> Ryan McDonald<sup>1</sup>

<sup>1</sup>Google Research, New York, NY, USA, {slav,ryanmcd}@google.com <sup>2</sup>Carnegie Mellon University, Pittsburgh, PA, USA, dipanjan@cs.cmu.edu

## **LREC 2012**

Language and treebank specific POS tagsets → 12 universal tags (currently: 17)

To facilitate future research in unsupervised inc consists of twelve universal part-of-speech categ to this universal set. As a result, when combin consisting of common parts-of-speech for 22 di compare tagging accuracies across languages, ( part-of-speech tags, and (3) use the universal tag

## Universal Stanford Dependencies: A cross-linguistic typology

Marie-Catherine de Marneffe°, Timothy Dozat\*, Natalia Silveira\*, Katri Haverinen\*, Filip Ginter\*, Joakim Nivre√, Christopher D. Manning\*

°Linguistics Department, The Ohio State University
\*Linguistics and °Computer Science Departments, Stanford University
•Department of Information Technology, University of Turku

□Department of Linguistics and Philology, Uppsala University

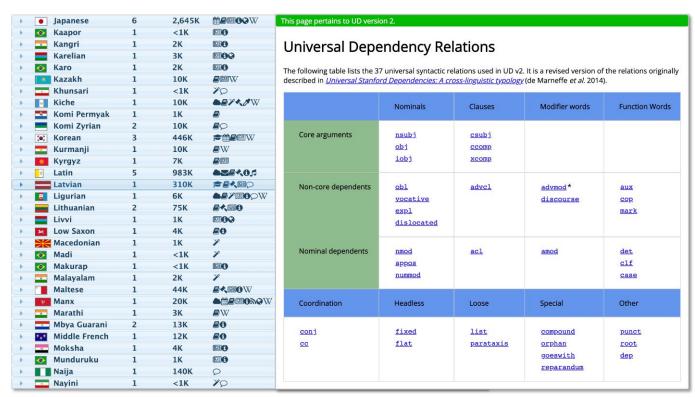
#### **Abstract**

Revisiting the now de facto standard Stanford dependency representation, we propose an improved taxonomy to capture grammatical relations across languages, including morphologically rich ones. We suggest a two-layered taxonomy: a set of broadly attested universal grammatical relations, to which language-specific relations can be added. We emphasize the lexicalist stance of the Stanford Dependencies, which leads to a particular, partially new treatment of compounding, prepositions, and morphology. We show how existing dependency schemes for several languages map onto the universal taxonomy proposed here and close with consideration of practical implications of dependency representation choices for NLP applications, in particular parsing.

## **LREC 2014**

A set of broadly attested universal syntactic relations (currently: 37)

## **Universal Dependencies**



~150 languages~250 treebanks500+ contributors

Released twice a year (via CLARIN)

## Used for:

- o linguistic research
- training of language models

https://universaldependencies.org

https://universaldependencies.org/u/pos/ https://universaldependencies.org/u/dep/

# The CoNLL-U format (i.e., the data fields)

1	Workers	worker	NOUN	Number=Plur	2	nsubj
2	dumped	dump	VERB	Mood=Ind Tense=Past VerbForm=Fin	0	root
3	sacks	sack	NOUN	Number=Plur	2	dobj
4	into	into	ADP	_	6	case
5	а	a	DET	Definite=Ind PronType=Art	6	det
6	bin	bin	NOUN	Number=Sing	2	nmod
7	•	•	PUNCT	_	2	punct

https://universaldependencies.org/format.html

# UD parsing: CoNLL 2018 Shared Task, etc.

#### Big treebanks only

Macro-average LAS-F1 of the 61 big treebanks: af\_afribooms, grc\_perseus, grc\_proiel, ar\_padt, eu\_bdt, bg\_btb, ca\_ancora, hr\_set, cs\_cac, cs\_fictree, cs\_pdt, da\_ddt, nl\_alpino, nl\_lassysmall, en\_ewt, en\_gum, en\_lines, et\_edt, fi\_ftb, fi\_tdt, fr\_gsd, fr\_sequoia, fr\_spoken, gl\_ctg, de\_gsd, got\_proiel, el\_gdt, he\_htb, hi\_hdtb, hu\_szeged, zh\_gsd, id\_gsd, it\_isdt, it\_postwita, ja\_gsd, ko\_gsd, ko\_kaist, la\_ittb, la\_proiel, lv\_lvtb, no\_bokmaal, no\_nynorsk, fro\_srcmf, cu\_proiel, fa\_seraji, pl\_lfg, pl\_sz, pt\_bosque, ...

1.	HIT-SCIR (Harbin)	software1-P	84.37
2.	Stanford (Stanford)	software2	83.03
3.	TurkuNLP (Turku)	software1-P	81.85
4.	UDPipe Future (Praha)	software1-P	81.83
5.	ICS PAS (Warszawa)	software1-P	81.72
6.	CEA LIST (Paris)	software1-P	81.66
7.	LATTICE (Paris)	software1-P	80.97
8.	NLP-Cube (București)	software1-P	80.48
9.	ParisNLP (Paris)	software1-P	80.29
10.	Uppsala (Uppsala)	software1-P	80.25
11.	SLT-Interactions (Bengaluru)	software2-P	79.67
12.	AntNLP (Shanghai)	software1-P	79.61
13.	LeisureX (Shanghai)	software1-P	77.98
14.	UniMelb (Melbourne)	software1-P	77.69
15.	IBM NY (Yorktown Heights)	software1-P	77.55
16.	Fudan (Shanghai)	software5-P	75.42
17.	KParse (İstanbul)	software1-P	74.84
18.	BASELINE UDPipe 1.2 (Praha)	software1-P	74.14
19.	Phoenix (Shanghai)	software1-P	73.93
20.	BOUN (İstanbul)	software2-P	72.85
21.	CUNI x-ling (Praha)	software1-P	71.54
22.	ONLP lab (Ra'anana)	software3-P	67.08
23.	iParse (Pittsburgh)	software1-P	66.55
24.	HUJI (Yerushalayim)	software1-P	62.07
25.	ArmParser (Yerevan)	software1-P	58.14

#### lv\_lvtb

1.	HIT-SCIR (Harbin)	software1-P	83.97
2.	Stanford (Stanford)	software2	81.85
3.	TurkuNLP (Turku)	software1-P	80.81
4.	ICS PAS (Warszawa)	software1-P	80.71
5.	CEA LIST (Paris)	software1-P	80.29
6.	UDPipe Future (Praha)	software1-P	79.32
7.	NLP-Cube (București)	software1-P	78.18
8.	ParisNLP (Paris)	software1-P	78.16
	SLT-Interactions (Bengaluru)	software2-P	78.16
10.	Uppsala (Uppsala)	software1-P	76.97
11.	LATTICE (Paris)	software1-P	76.91
12.	AntNLP (Shanghai)	software1-P	75.56
13.	UniMelb (Melbourne)	software1-P	75.28
14.	IBM NY (Yorktown Heights)	software1-P	73.17
15.	LeisureX (Shanghai)	software1-P	73.13
16.	KParse (İstanbul)	software1-P	72.33
17.	Fudan (Shanghai)	software5-P	70.04
18.	BASELINE UDPipe 1.2 (Praha)	software1-P	69.43
19.	Phoenix (Shanghai)	software1-P	69.06
20.	BOUN (İstanbul)	software2-P	68.47
21.	CUNI x-ling (Praha)	software1-P	67.23
22.	ONLP lab (Ra'anana)	software3-P	59.67
23.	ArmParser (Yerevan)	software1-P	57.88
24.	HUJI (Yerushalayim)	software1-P	55.19
25.	SParse (İstanbul)	software1-P	0.00
26.	iParse (Pittsburgh)	software1-P	0.00

# **UD parsing: Stanza's pretrained UD models**

						Search: Lat		
Language 🛊	Treebank ‡	UPOS ‡	XPOS	UFeats ‡	AllTags 🗦	Lemmas ‡	UAS ‡	LAS
Latin	ITTB	98.76	95.72	96.69	94.63	99.09	88.82	86.80
Latin	LLCT	99.57	96.65	96.79	96.47	98.09	96.13	94.88
Latin	Perseus	91.40	78.09	81.83	76.37	83.57	72.07	63.16
Latin	PROIEL	96.95	97.15	91.58	90.59	96.86	77.98	73.99
Latin	UDante	90.08	74.55	81.64	73.07	86.95	68.00	58.84
Latvian	LVTB	96.70	89.72	94.73	89.29	96.12	88.91	85.77

Showing 1 to 6 of 6 entries (filtered from 138 total entries)

https://stanfordnlp.github.io/stanza/performance.html

# UD parsers: Stanza, UDpipe, spaCy, LV-PIPE, etc.

Stanza: a Python package, supports **70**+ languages (out of the box)

- https://stanfordnlp.github.io/stanza/models.html
- o easy to use

UDpipe: a REST web-service, supports **70**+ languages

• https://ufal.mff.cuni.cz/udpipe/2/models

spaCy: a Python package, supports 25+ languages (out of the box)

- https://spacy.io/models
- o industry-strength

LV-PIPE: a REST web-service, supports only Latvian

- https://nlp.ailab.ly » https://korpuss.ly, https://proza.lnb.ly, etc.
- ~90% LAS score (based on LV-BERT, 2020-2022)

# The Task

Create a teeny-tiny UD annotated text corpus of ~10 currently trending news articles

- o In a language of your choice
- Together, you will create a multilingual corpus

Get a list of such articles (links)

- Use Europe Media Monitor as a source
- Extract the links form an **RSS** feed, using feedparser

Extract text from the webpages, using bs4

Parse the texts, using stanza, and create a **VERT** file

We will concatenate your VERT files into a single file

• It will be uploaded to a **NoSketch Engine** instance for running **CQL** queries over this dataset

https://github.com/LUMII-AILab/NLP Course/blob/main/notebooks/BSSDH2024.ipynb

# **Assignment & Grading**

## Submission: your final version of the BSSDH2024.ipynb template

o Improvements, extensions, comments, runtime output, etc.

## Criteria:

- o Grade **7–8**: you have managed to do basic plain-text extraction (which may still be noisy) and have produced a valid VERT file with basic UD annotations (lemma, pos, dep)
- Grade 8-9: the extracted plain-text is rather clean and rather well segmented,
   the VERT file contains extra UD annotations (dep\_head\_lemma, dep\_head\_pos, dep\_head\_dep)
- o Grade 10: surprise us 😉
  - » e.g., scale your corpus from 10 to 100 articles, or add named entity recognition, or use another framework, etc.
- +/-1 point for a well/poorly commented and documented notebook with some/no conclusions