

Open access and open source tools for corpus linguistics: Wmatrix version 7 and PyMUSAS

CL2025 Pre-conference workshop: 29th June 2025

Slides at https://ucrel.lancs.ac.uk/paul/

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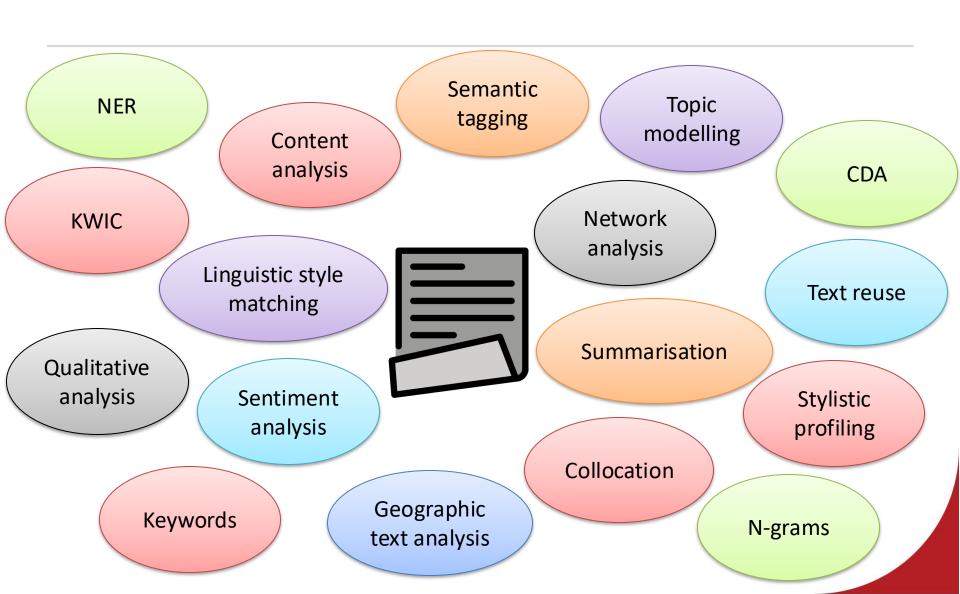




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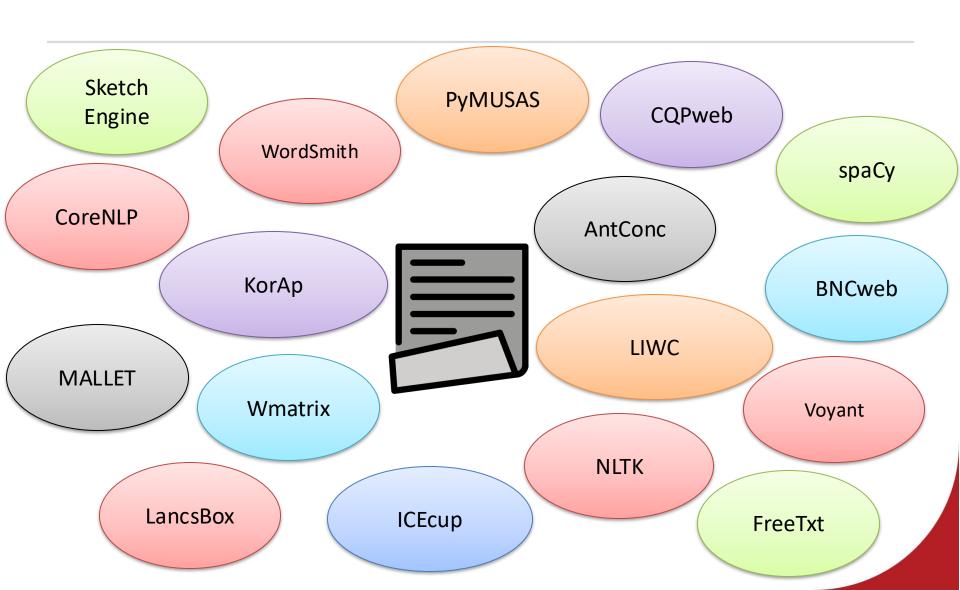


A myriad of NLP and CL methods ...





A myriad of NLP and CL tools ...





Importance of open source and open access tools & resources

Vital for reproducibility and replicability of corpus linguistics studies

Explainability of annotation c.f. generative AI / LLM methods, many of which do not declare their training materials or methods

Extensibility and unrestricted free access to tools and resources in the global south and low resource contexts





Model Name	Availability				Documentation				Access					
	Base Data	Fine Tuning Data	Base Weight s	Fine Tuning Weights	Code	Code Docs	Model Architecture	Pre- print	Paper	Model Card	Datasheet	Package	API	License
OLMo	Y	Υ	Y	Y	Υ	Y	Y	Y	Y	Υ	Υ	Υ	~	Y
Deepseek V3	N	N	N	Υ	~	~	~	Y	N	~	N	Υ	Υ	~
Phi	N	N	Y	Y	N	N	Υ	Y	N	Υ	N	Υ	N	Υ
Mistral	N	N	Y	Y	~	~	~	~	N	N	N	Υ	Υ	Υ
Gemma	N	N	~	~	~	N	Y	Y	N	Y	N	Y	N	N
Chat-GPT	N	N	N	N	N	N	N	~	N	N	N	N	~	N
	N = No			~ = Partial			Y = Yes							

Based off the European Open-Source Al Index:

https://osai-index.eu/the-index

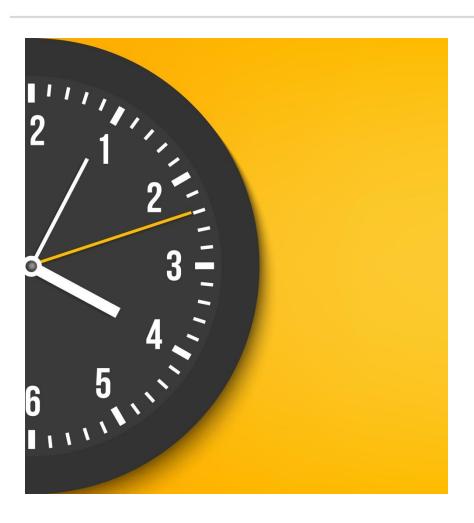
Recent releases



- All semantic lexicons are now open access (CC-BY-NC-SA-4.0)
 - https://github.com/UCREL/Multilingual-USAS
- Python reimplementation of semantic tagger, PyMUSAS is now open source (Apache 2.0)
 - https://pypi.org/project/pymusas/
- Wmatrix7, with PyMUSAS tagging for 8 languages, is now free & open access for researchers worldwide albeit with filestore limits, and a new indexing system built on SQLite



A workshop of five parts



- 1. Semantic annotation (tagging)
 - a little bit computational
- 2. Key semantic tags (key domains)
 - a little bit of statistics
- 3. Wmatrix and PyMUSAS software
 - Hands on practical
- 4. Refreshments (15:00–15:30)
- 5. Current and future developments
 - UCREL-Hex 'medium' scale compute cluster
 - 4D Picture application and PyMUSAS
- 6. Yet more hands on practical ...
 - And your chance to provide feedback and influence future plans!

Lexical ambiguity



- Question:
 - How many senses does spring have?
- Answer:
 - A: 3
 - − B: 4
 - C: 5
 - − D: 31

to arrange the escape of (someone) from prison. 17 (intr) Archaic or poetic. (of daylight or dawn) to begin to appear. \blacklozenge n 18 the act or an instance of springing. 19 a leap, jump, or bound. 20a the quality of resilience; elasticity. 20b (as modifier): spring steel. 21 the act or an instance of moving rapidly back from a position of tension. 22a a natural outflow of ground water, as forming the source of a stream. 22b (as modifier): spring water. 23a a device, such as a coil or strip of steel, that stores potential energy when it is compressed, stretched, or bent and releases it when the restraining force is removed. 23b (as modifier): a spring mattress. 24 a structural defect such as a warp or bend. 25a (sometimes cap.) the season of the year between winter and summer, astronomically from the March equinox to the June solstice in the N hemisphere and from the September equinox to the December solstice in the S hemisphere. 25b (as modifier): spring showers. Related adj: vernal. 26 the earliest or freshest time of something. 27 a source or origin. 28 one of a set of strips of rubber, steel,

etc., running down the inside of the handle of a cricket bat, hockey stick, etc. **29** Also called: **spring line**. *Nautical*. a mooring line, usually one of a pair that cross amidships. **30** a flock of teal. **31** *Architect*. another name for **springing**. [Old English *springan*; related to Old Norse *springa*, Old High German *springan*, Sanskrit *sprhayati* he desires, Old Slavonic *pragu* grasshopper] b 'springless *adj* b 'spring, like *adj*

spring balance or esp. U.S. spring scale n a device in which an object to be weighed is attached to the end of a helical spring, the extension of which indi-

cates the weight of the object on a calibrated scale.

Definition of *spring* from Collins English Dictionary (21st Century Edition)

Spring (https://dictionary.cambridge.org)



- spring was found in the Cambridge Advanced Learner's Dictionary at the entries listed below.
 - spring (MOVE QUICKLY)
 - spring (APPEAR SUDDENLY)
 - spring (SEASON)
 - spring (CURVED METAL)
 - spring (WATER)
 - box spring
 - spring chicken
 - spring-clean
 - spring greens
 - spring onion
 - spring roll
 - spring from sth
 - spring sth on sb
 - be full of the joys of spring
 - spring to life
 - spring to mind
 - a spring in your step

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What is Semantic Tagging?

- Semantic field annotation has applications for conceptual or topic tagging:
 - Last_T1.1.1 year_T1.1.1 was_A3+ the_Z5 UK_Z2 's_Z5 second_N4 warmest_O4.6+++ on_A11.2+ record_A11.2+ ,_PUNC according_Z5 to_Z5 provisional_T1.3- data_X2.2 from_Z5 the_Z5 Met_S3.1 Office_I2.1/H1c ._PUNC This_Z8 puts_X2.2- it_Z8 just_A14 behind_X2.2- 2022_N1 ,_PUNC which_Z8 recorded_Q1.2 an_Z5 average_A6.2+ temperature_O4.6 of_Z5 only_A14 0.06C_Z99 higher_N3.7++ ._PUNC

• A3+ = being; A6.2 = comparing; A11.2 = importance; A14 = exclusivisers; H1 = architecture, buildings; I2.1 = business; N1 = numbers; N3.7 = measurement; N4 = linear order; O4.6 = temperature; Q1.2 = documents, writing; S3.1 = relationship; T1.1.1 = Time past; T1.3 = time period; X2.2 = knowledge; Z2 = geographical names; Z5 = grammatical bin; Z8 = pronouns etc; Z99 = unmatched

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Multiword expressions: plain sailing?

- Phrasal verbs
 - Stubbed out
- Noun phrases
 - Riding boots
 - Pony nuts
- Proper names
 - United States of America
- Named entities
 - 23rd November 1963
 - British Broadcasting Corporation

- Multiword prepositions
 - In terms of
 - As soon as
- Idiomatic expressions
 - Spill the beans
 - A pain in the neck



UCREL Semantic Analysis System (USAS)

- Full text tagging, not just selected words (c.f. Diction, LIWC, RID)
- Tagging the coarse-grained sense in context, not just the word
- Not task specific categories
- Flexible category set with hierarchical structure
- Words and multi-word expressions (MWE) e.g. phrasal verbs (stubbed out), noun phrases (riding boots), proper names (United States of America), true idioms (living the life of Riley)
- https://ucrel.lancs.ac.uk/usas/
- Lexicons available free for academic use:
 - https://github.com/UCREL/Multilingual-USAS

The work of many hands ...



- Joint research with
 - Geoffrey Leech
 - Roger Garside
 - Jenny Thomas
 - Andrew Wilson
 - Dawn Archer
 - Scott Piao
 - Sheryl Prentice
 - Andrew Moore
 - Daisy Lal
 - Ignatius Ezeani





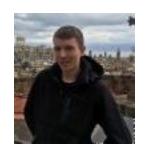
















Semantic fields



- AKA concepts, semantic domains
- 'groups together word senses that are related by virtue of their being connected at some level of generality with the same mental concept'
- Not only synonymy and antonymy but also hypernymy and hyponymy
- E.g. EDUCATION: academic, coaching, coursework, deputy head, exams, PhD, playschool, revision notes, studious, swot, viva



	Α	В	С	E
	General and abstract terms	The body and the individual	Arts and crafts	Emotion
	F Food and farming	G Government and public	H Architecture, housing and the home	l Money and commerce in industry
-	K Entertainment, sports and games	L Life and living things	M Movement, location, travel and transport	N Numbers and measurement
	O Substances, materials, objects and equipment	P Education	Q Language and communication	Social actions, states and processes
-	T Time	W World and environment	X Psychological actions, states and processes	Y Science and technology
	Z Names and grammar			

Lexical resources for English



- Lexicon of 56,316 items
 - presentation NN1 Q2.2 A8 S1.1.1 K4
- MWE list of 18,971 items
 - travel_NN1 card*_NN* M3/Q1.2
- A small wildcard lexicon
 - *kg NNU N3.5
- Unknown words using WordNet synonym lookup



English Disambiguation methods (1)

- 1. POS tag
 - spring noun [season sense] [coil sense]
 - spring verb [jump sense]
- 2. General likelihood ranking for single-word and MWE tags
 - green referring to [colour] is generally more frequent than green meaning [inexperienced]
- 3. Overlapping MWE resolution
 - Heuristics applied: semantic MWEs override single word tagging, length and span of MWE also significant



English Disambiguation methods (2)

- 4. Domain of discourse
 - adjective battered
 - [Violence] (e.g. battered person)
 - [Judgement of Appearance] (e.g. battered car)
 - [Food] (e.g. battered cod)
- 5. Text-based disambiguation
 - one sense per text
- 6. Template rules
 - Auxiliary verbs (be/do/have)
 - account of NP [narrative]
 - balance of xxx account [financial]

Evaluation (English data)

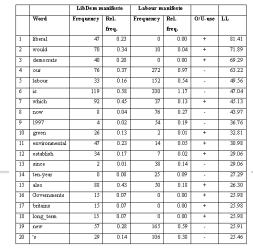


- Hand tagged test corpus of 124,839 words
- Error rate of 8.95%
- Ambiguity ratio 47.73%
- Reduced to 17.06% by disambiguation
- Not all ambiguity is resolved, but 1st choice tag selection gives 91% accuracy.



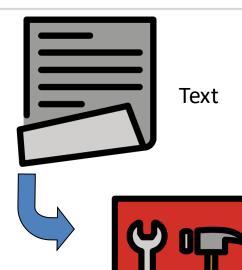
KEY SEMANTIC DOMAINS AND FURTHER APPLICATIONS

Umatrix

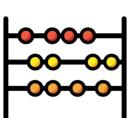


Keywords









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1	/

by

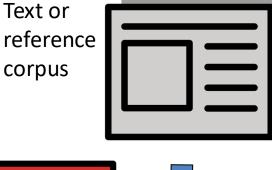
words

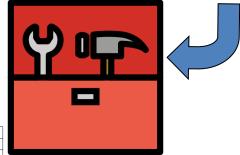
phrases

that

words

1		
		_
the	351	
of	243	
а	221	
and	153	
to	139	
in	134	
is	123	
be	83	
for	81	
phrase	69	
that	67	•
which	66	





Word frequency list



Word frequency list

Significance and effect size



- Log-likelihood (LL) Wizard online at:
 - https://ucrel.lancs.ac.uk/llwizard.html
- Spreadsheet and code also available for download
 - https://github.com/UCREL/SigEff
- Very important to consider dispersion and effect size measures (depending on your corpus) – included in Wmatrix frequency lists and keyness measures
 - See the work of Hardie, Gabrielatos, Brezina and others
 - Rayson and Potts (2021)



Figure 1: keywords in LibDem 2010 manifesto

2020 2050 affordable allow banking banks believe better Britain budget businesses carbon change child climate create crime cut deficit democrats developing_countries economy education emissions energy ensure environment establish EU every fair fairness financial for funding future give global government health help homes improve increase infrastructure insulate introduce jobs justice liberal OCA local_authorities long-term manifesto money mutuals need NHS our over_time paid pay people politics polluting power protect public reduce reducing reform reforming renewable replace restore review Savings schools scrap seek services spending state_pension such_as support sustainability Sustainable system target targets tax taxes to UK UN Unfair We will



Figure 2: key domains (semantic fields) in LibDem 2010 manifesto

```
Able/intelligent Alive Allowed Attentive Business Business: Generally Chance, Juck Change Cheap Confident
Constraint Crime Danger Degree Deserving Education_in_general Entire;_maximum Ethical
Ethical Evaluation: Good Evaluation: Good Evaluation: Authentic Exceed; waste
Getting_and_giving;_possession Giving Government Green_issues
Health_and_disease Helping Hindering
                                                       Important
                                                                   Inclusion
                                                                             Interested/excited/energetic
Law and order Lawful Location and direction Long tall and wide Medicines and
Money_and_pa Law_and_order: law, prison(s, ers), loopholes, security, police (force, officer, station, services) ...
Money:_Affluence Money:_Lack Money:_Affluence NO_CONSTRAINT No_obligation_or_necessity
Other_proper_names Participating People Places Politics Putting_pulling_pushing_transporting Quantities:_little
Quantities: little Quantities: many/much Relationship Residence Safe Safe Science_and_technology_in_general Social_Actions, States_And_Processes
Strong_obligation_or_necessity Success The_Media The_universe Time_period:_long Time: Future
Time:_Ending Time:_New_and_young Time:_Beginning Time:_Beginning Tough/strong Tough/strong Unethical Wanted Weather
Work and employment: Generally
```

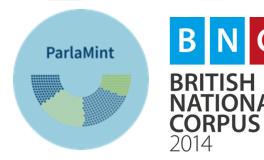


Applications of semantic analysis

100+ papers listed at https://ucrel.lancs.ac.uk/wmatrix/

- Analysis of market research interview transcripts
- Intelligent dictionaries
- Assistance for human translators
- Software Engineering domain understanding
- Language profiling for online child protection
- Actionability
- Corpus stylistics
- Prediction of real-world events from social media
- Metaphor and end-of-life care
- Pattern analysis of the language of psychopaths
- Political discourse analysis
- Describing the language of extremism and counter-extremism
- UK General Election Manifestos (Rayson 2008)









Metaphor, cancer and end of life care (MELC)

- Analysis of metaphorical language used to talk about cancer, dying and death: people 'fight' their cancer, 'win' or 'lose' their 'battle' against it, hope for a positive end to their cancer 'journey', and so on.
- 1.5M word corpus of interviews and online forum posts from patients, carers and healthcare professionals
- Methods: Manual analysis
 (MIP) and Wmatrix (Semantic analysis & concordancing)
- http://wp.lancs.ac.uk/melc/

G3 Warfare (e.g. *fight* as a verb, *battle*)

A1.1.1 General actions, making (e.g. *blast*, *confront*)

A1.1.2 Damaging and destroying (e.g. *destroy*, *shatter*)

E3- Violent/angry (e.g. hit, attack)

S8+ Helping (e.g. *defend*, *protect*)

S8– Hindering (e.g. *fight* as a noun)

X8+ Trying hard (e.g. struggle)

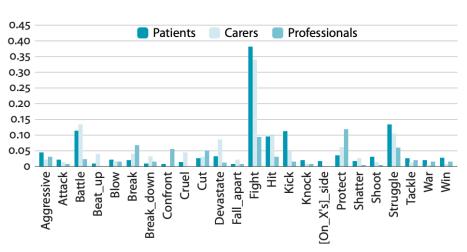


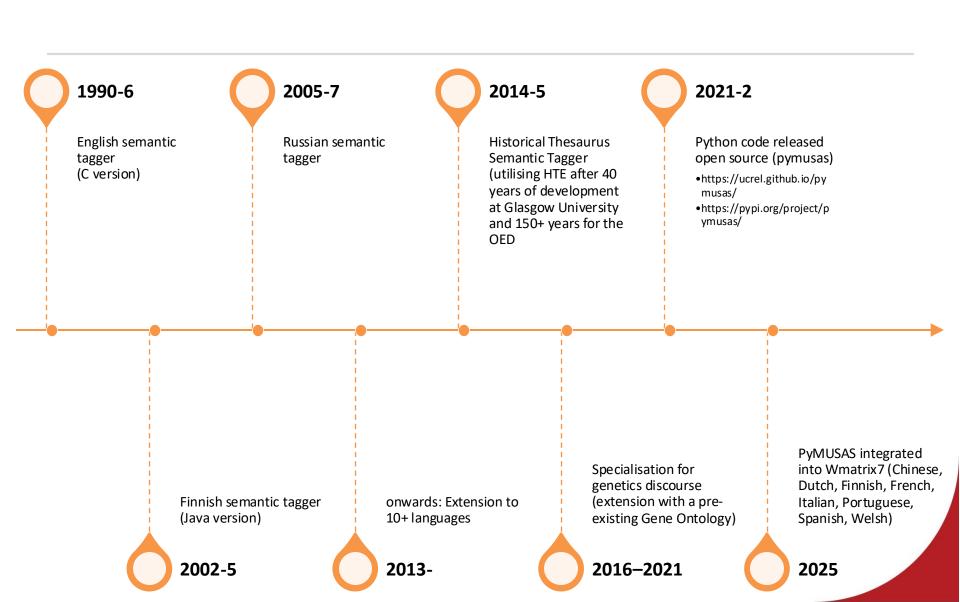
Figure 3. Relative use of most frequent Violence metaphors by each stakeholder group (per 1,000 tokens): Online forum posts



Qualitative survey analysis: FreeTxt/TestunRhydd project (2022-3)

- Surveys are widely used in many areas of professional practice, e.g. staff development, professional training, product design, testing as well as for many types of hotel, movie and product reviews
- Very little support for bilingual free-text survey and questionnaire data analysis in English and Welsh
- Follow on funding impact project building on CorCenCC project (National Corpus of Contemporary Welsh), we will develop an
 - open access user friendly online interface
- Partners: National Trust Wales, Cadw and National Museum Wales
- https://ucrel.lancs.ac.uk/freetxt/







Recipe for creating a tagger in a new language

- 1. re-evaluate USAS semantic tagset for new language context
- 2. find freely available (open source if possible) POS tagger & lemmatiser
- 3. integrate these into USAS Multilingual software framework (PyMUSAS)
 - a. consider whether other new components are needed e.g. tokeniser or compound tool
- 4. develop single-word semantic lexicon and MWE dictionary
 - a. bilingual dictionary
 - b. parallel aligned corpus (Moses / Giza)
 - c. machine translation / translation memory
 - d. crowdsourcing by non-experts
 - e. named entity recognition and gazetteers
 - f. vector-based approaches
 - g. multi-task & deep learning
 - h. manual checking and editing by experts
- 5. extend disambiguation routines
- 6. release lexicons with CC-BY-NC-SA licence
- 7. release software as REST API and/or open-source licence



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https://pypi.org/project/pymusas/

- Open source Apache License Version 2.0
- Open resources Creative
 Commons licence version 4
- Rule based tagger
- Identify and tag Multi Word Expressions (MWE)
- Supports multiple languages through downloadable spaCy pipelines
- Supports Indonesian and Welsh via other POS taggers (TreeTagger for Indonesian and CyTag for Welsh)

Language (BCP 47 language code)	MWE Support	Size
Mandarin Chinese (cmn)	~	1.28MB
Welsh (cy)	~	1.09MB
Spanish, Castilian (es)	~	0.20MB
French (fr)	×	0.08MB
Indonesian (id)	×	0.24MB
Italian (it)	~	0.50MB
Dutch, Flemish (nl)	×	0.15MB
Portuguese (pt)	~	0.27MB

PyMUSAS – Language Support



Each language that we support has a guide on how to semantically tag text for that language:

https://ucrel.github.io/pymusas/usage/how-to/tag-text

Tag Text

In this guide we are going to show you how to tag text using the PyMUSAS RuleBasedTagger so that you can extract token level USAS semantic tags from the tagged text. The guide is broken down into different languages, for each guide we are going to:

- 1. Download the relevant pre-configured PyMUSAS RuleBasedTagger spaCy component for the language.
- Download and use a Natural Language Processing (NLP) pipeline that will tokenise, lemmatise, and Part Of Speech (POS) tag. In
 most cases this will be a spaCy pipeline. Note that the PyMUSAS RuleBasedTagger only requires at minimum the data to be
 tokenised but having the lemma and POS tag will improve the accuracy of the tagging of the text.
- Run the PyMUSAS RuleBasedTagger.
- 4. Extract token level linguistic information from the tagged text, which will include USAS semantic tags.
- 5. For Chinese, Italian, Portuguese, Spanish, and Welsh taggers which support Multi Word Expression (MWE) identification and tagging we will show how to extract this information from the tagged text as well.

Chinese

▶ Expand			
Dutch			

▶ Expand

French

Expand



Try this Python Notebook during the hands on session:

https://github.com/UCREL/pymusas_notebook

Chinese Dutch French

ench alian

Portuguese Spanish Welsh



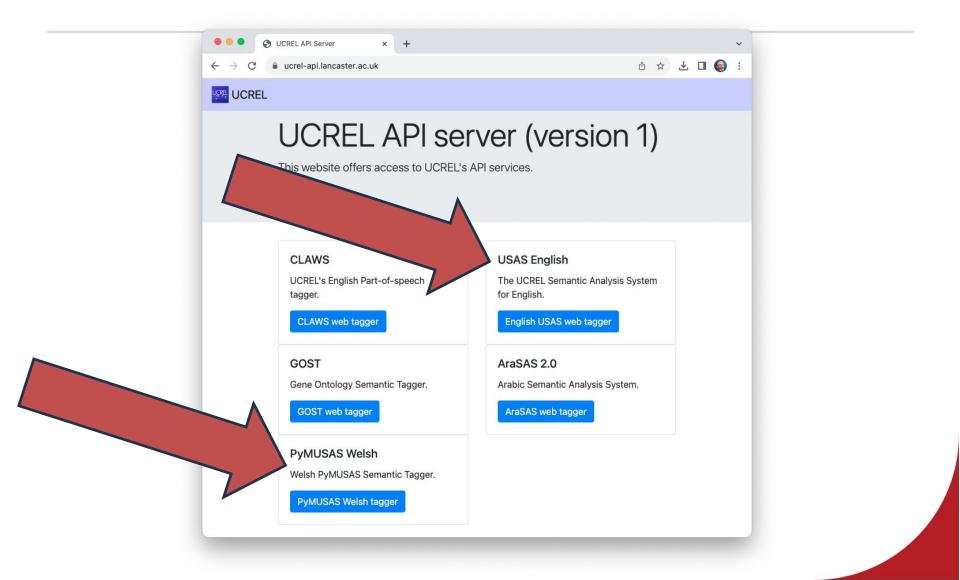
Recent developments in new languages

- Used the IgboAPI dataset (33 distinct Igbo dialects, 5,095 Igbo words with 17,979 unique dialectal word variations, complemented by 27,816 example parallel sentences) to bootstrap a lexicon for the Igbo semantic tagger
 - https://aclanthology.org/2024.lrec-main.1384/
- Creation of high quality linguistic resources (MWE lexicon) via LLMs to retrieve MWE definitions facilitating accurate translation from English to Danish lexicons, coverage evaluation and manual annotation for metaphor analysis in 4D Picture project
 - Puts et al. (2025) Pushing the boundaries: creating a Danish semantic tagger for metaphor analysis of cancer narratives.
 Corpus Linguistics 2025, Birmingham, UK.

https://ucrel-api.lancaster.ac.uk/

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You can also test USAS without a login for Wmatrix









WMATRIX VERSION 7

Key points



- Web-based (c.f. BNCweb, CQPweb, SketchEngine)
- Dedicated server, Secure HTTPS access
- You can load your own data (Multilingual in v7)
- Incorporates main methods in corpus linguistics toolbox
 - frequency lists, concordances, key words, collocations, n-grams
- Adds two levels of linguistic annotation (NLP methods)
 - POS tagging, Semantic field tagging
- Novelty
 - key domain analysis, semantic collocations

Hands on practical





- 2005 UK general election
 - Liberal Democrat party manifesto
 - Labour party manifesto
- 2010 UK general election
 - manifestos for all three main parties
- 2015, 2017, 2019 and 2024 UK general elections
 - manifestos for seven parties
- Aims:
 - To help you understand the basic Wmatrix features and key domains method
 - To give you some awareness of the semantic tagset



Version 7 compared to version 5

	Wmatrix5	Wmatrix7
Indexing system	Bespoke from 1990s	SQLite
Folders / Corpus	Single file, up to 1M words	Multiple files (zip), tested up to 30M words
Concordances	Corpus order	Various sort options
N-grams and collocations	NSP and Java code	SQLite
Language	USAS English, Spanish beta	PyMUSAS for Chinese, Dutch, Finnish, French, Italian, Portuguese, Spanish, and Welsh
MWEs	Tagged, displayed in frequency lists	Tagged but not yet displayed in frequency lists
Optional features	Domain and My Tag Wizard, Metaphor features, folder sharing	

Open two web-browser windows or tabs





- All URLs linked from Wmatrix home page:
 - https://ucrel.lancs.ac.uk/wmatrix/
- Wmatrix tutorials
 - https://ucrel.lancs.ac.uk/wmatrix/tutorial7/
- Wmatrix tool:
 - https://ucrel-wmatrix7.lancaster.ac.uk/
 - Apply for login now if you haven't already got one

Your tasks!!





- https://ucrel.lancs.ac.uk/wmatrix/tutorial7/
- On your own or in small groups ...
 - Do tutorials A and B (you can either upload the manifesto documents yourself into Wmatrix, or use the ones I made earlier in the corpus library)
 - Do tutorial C (key words, key domains and concordances)
 - For the keen ones amongst you, move on to the other tutorials
 - You can use your own data if you wish
 - Ask questions any time!



Beyond Wmatrix:

Parallel and Cluster Systems

Or "This is taking too long..." and "Why is my computer really hot?"



Working with truly vast amounts of data...

If any of these apply, then you may need to start seriously thinking about scaling your software:

- Does everything take too long?
 - Are you 'at risk' for process instability?
- Are you running out of RAM?
 - "Isn't 128GiB enough?"
- Are you resource bound?
 - "But I need a GPU for every process..."

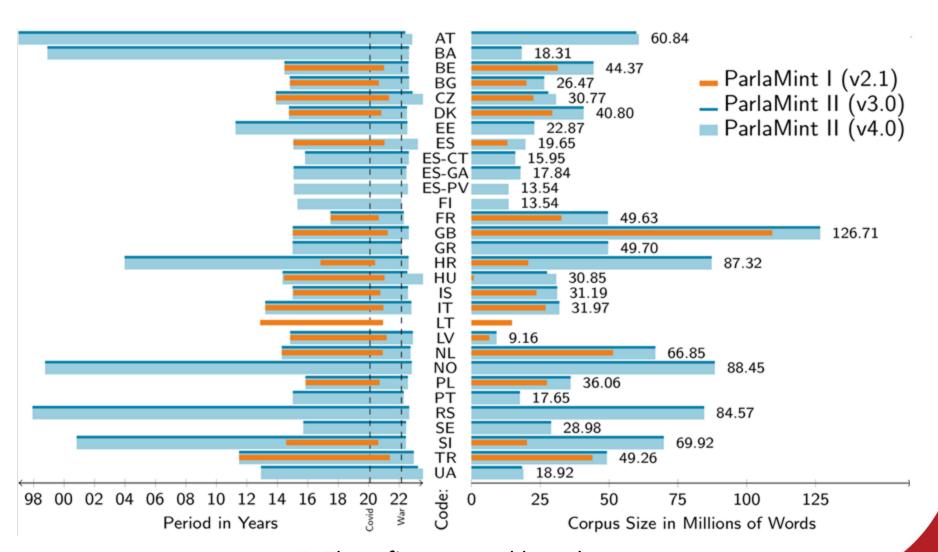


An Example: ParlaMint

- CLARIN flagship project (<u>https://www.clarin.eu/parlamint</u>)
- Creation of comparable, uniformly annotated, CC-BY corpora of parliamentary debates across Europe
- Two project stages (2020-21, 2022-23)
- V4.0 released October 2023: http://hdl.handle.net/11356/1859
- English MT annotated version released November 2023: http://hdl.handle.net/11356/1864
- 7.8M utterances; 1.2B words; 29 languages



ParlaMint: Endlessly Growing Data

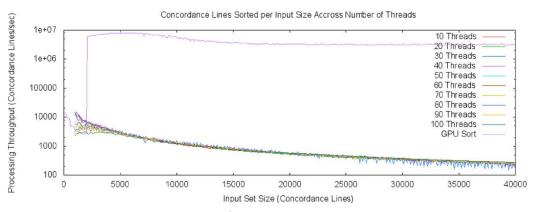


NB. These figures are old now!



Data Dependencies

- If dataset 'A' never interacts with dataset 'B' we can process them independently
 - We can often cheat this too (kinda!)
- In general, loops over large collections are good candidates for handling in parallel
 - This is formalised as the 'map' semantics from map/reduce





Data Dependencies

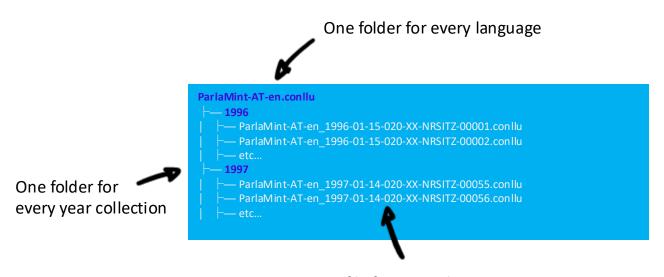
The ParlaMint dataset is *highly* independent

- Three levels of iteration:
 - 1. By Language
 - 2. By Year
 - 3. By File

But... the corpus is also *very large™*, so rather than attempting to parallelise everywhere, instead focus on "easy wins"

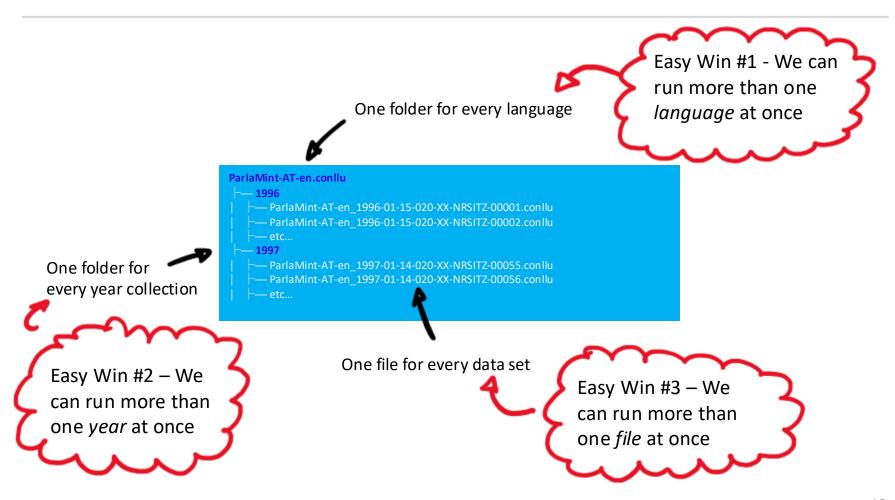
For reference, let us take a quick look at the folder structure



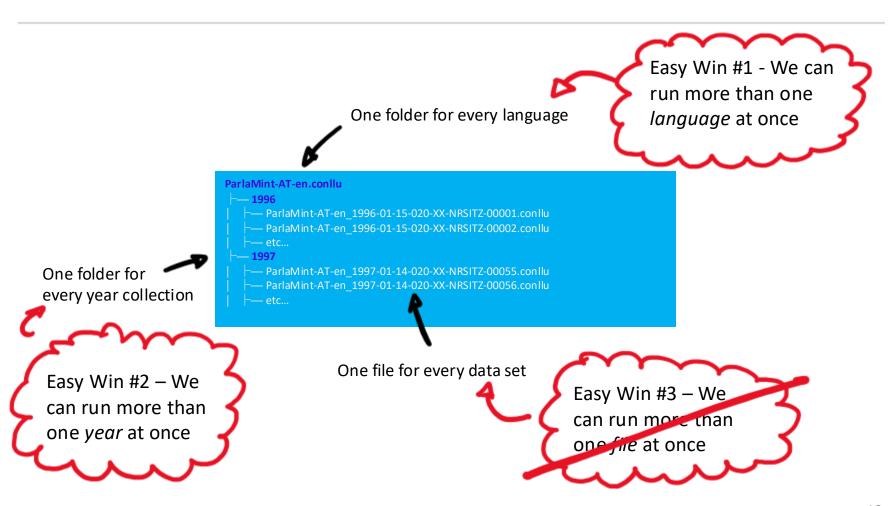


One file for every data set











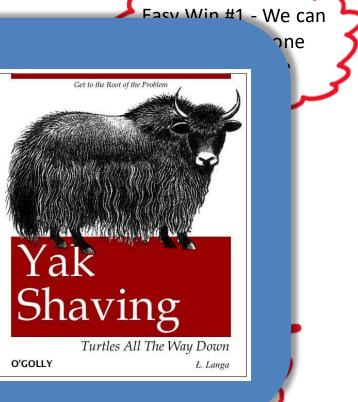
Why not #3?

We may run the risk of over producing very short-lived jobs; which might not be a problem but could easily result in the majority of our time spent running being taken up by starting/stopping the jobs rather than actually running them.

en

One folder for every year collecti

Easy Win #2 – \can run more t one *year* at onc.



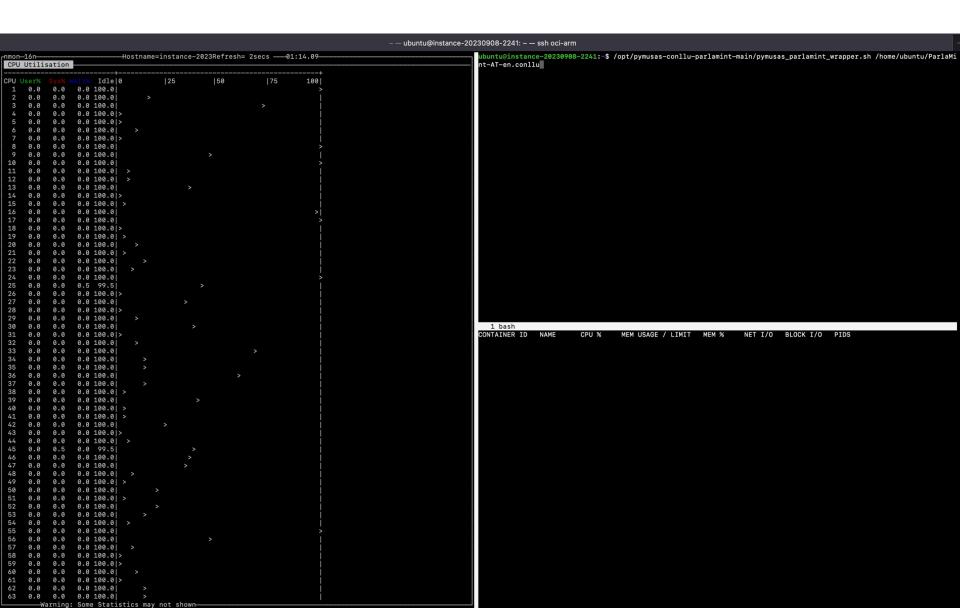


Tagging Parlamint: How it began

	~ — ubuntu@instance-20230908-2241: ~ — ssh oci-arm									
rnmon—1	L6n		Host	name=instance	e-2023Refresh	= 2secs	01:03.30	ubuntu@instance-20230908-2241:-\$ ls ParlaMint-AT-en.conllu		
CPU Utilisation										
CPU Us	er%		 /ait%	25	50	 75	100	1997 1999 2001 2003 2005 2007 2009 2011 2013 2015 2017 2019 2021 ubuntu@instance-20230908-2241:~\$ ls ParlaMint-AT-en.conllu wc -l		
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10	0.0	0.0	0.0 100.0							
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			0.0 99.5							
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18	0.0	0.0	0.0 100.0							
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22	0.5	0.0	0.0 99.5							
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			0.0 99.5							
27	0.0	0.0	0.0 100.0							
	0.0 0.5		0.0 100.0 0.0 99.5							
	0.5		0.0 99.5					2 bash		
31	0.5	0.0	0.0 99.5					CONTAINER ID NAME CPU % MEM USAGE / LIMIT MEM % NET I/O BLOCK I/O PIDS		
32	0.0	0.0	0.0 100.0							
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35	0.0	0.0	0.0 100.0							
			0.0 99.5							
	0.0		0.0 100.0 0.0 100.0				>			
39	0.0		0.0 100.0							
40	0.5	0.0	0.0 99.5							
	0.5 0.0		0.0 99.5 0.0 100.0				>			
43			0.0 99.5							
44	0.5	0.0	0.0 99.5							
			0.0 100.0 0.0 100.0							
			0.0 100.0				>			
48	0.5	0.0	0.0 99.5							
			0.0 99.5							
			0.0 100.0 0.0 100.0				>			
52	0.0	0.0	0.0 100.0							
53	0.0		0.0 100.0							
			0.0 100.0 0.0 99.5				>			
56	0.0	0.0	0.0 100.0							
57	0.5	0.0	0.0 99.5							
			0.0 100.0 0.0 99.5				>			
			0.0 99.5				>			
61	0.0	0.0	0.0 100.0							
62	0.5	0.0	0.0 99.5							
03	0.5	0.5	0.0 99.0				>			



Tagging Parlamint: Afterwards





ParlaMint: A Summary

Shortened a 18-day runtime to a **7-hour** one!

For all languages!



What tools can we use to achieve this in notebooks?

- Serial Operations:
 - https://tqdm.github.io/
- Parallel Operations:
 - https://pypi.org/project/p-tqdm/

(This may not be the best solution out there, but definitely one of the more simple and effective!)





https://slurm.schedmd.com/

Slurm is a job batch and queuing system on many supercomputer clusters:

- BEDE: https://n8cir.org.uk/bede/
- HEC: https://lancaster-hec.readthedocs.io
- Hex: https://www.lancaster.ac.uk/scc/research/research-facilities/hex/



```
vidlerj@login:~$ squeue

JOBID PARTITION NAME USER ST TIME NODES NODELIST(REASON)

2126_53 a2000-6h slurmRun hylandr R 1:02:59 1 hex-p3-g1

2126_52 a5000-6h slurmRun hylandr R 1:41:36 1 hex-vm-002

2126_50 a5000-6h slurmRun hylandr R 3:32:30 1 hex-vm-001

2126_[54-75%3] a5000-6h, slurmRun hylandr PD 0:00 1 (JobArrayTaskLimit)
```

One of our users using 3 GPUs at once!



Parallel Processing Corpus Data: Tips!

- Think about which parts of your program(s) are accessing which data
 - Can parts be split apart to run in parallel without 'side effects?'
- Avoid 'state' and especially 'shared state'
 - If a process has state, it means it cannot run without external information
- Consider 'idempotency':
 - Multiple operations can be called, but only the initial result will be used.

```
If alreadyRun:
    return result
result = processThings()
return result
```

Roughly this, but there are nuance here... check with the libraries you're using to see how to do this safely!

Other Tools



- Docker! https://www.docker.com/
 - Isolate your strange dependencies, so you can deploy on anything ©
 - https://docs.docker.com/guides/text-classification/
- My own tutorials (slightly old)
 - https://johnvidler.co.uk/blog/docker-101/
 - https://johnvidler.co.uk/blog/docker-102/

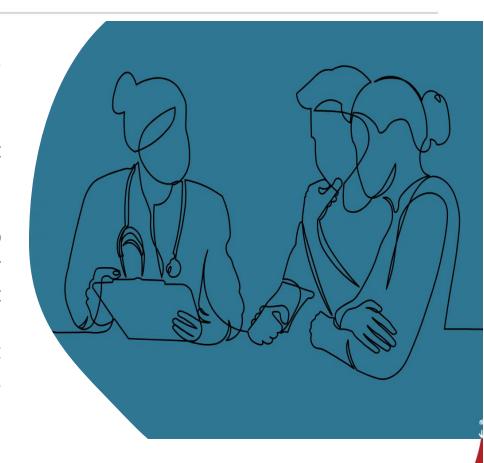




USAS-based Sentiment Analysis 4DPicture

The 4D PICTURE project aims to help cancer patients, their families, and healthcare providers better understand their options. It supports their treatment and care choices, at each stage of disease.

The project's primary objective is to improve decision-making about cancer treatment by better predicting treatment outcomes by developing data-driven algorithms, resulting in decision-support tools for people with breast cancer, prostate cancer, or melanoma.



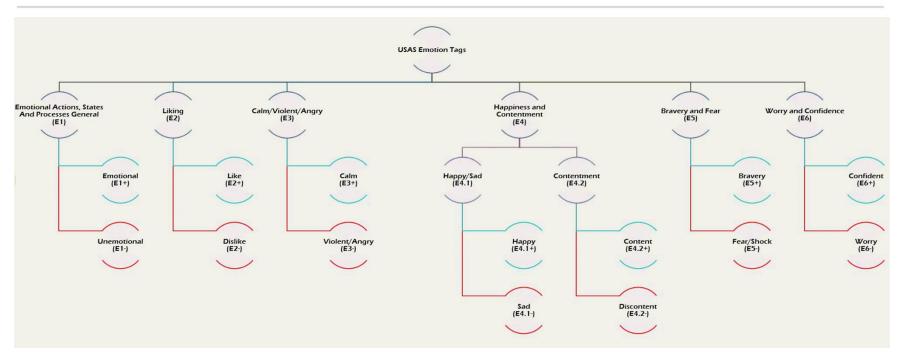
Design-based Data-Driven Decision-support (4D) Tools

Producing Improved Cancer Outcomes Through User-centered Research



USAS-based Sentiment Analysis

Code



https://drive.google.com/drive/folders/1Yaqr-PrtaB14Oqhx8etdqoBlleVuQ8n8?usp=sharing



Continue with your tasks!!





- https://ucrel.lancs.ac.uk/wmatrix/tutorial7/
- On your own or in small groups ...
 - Do tutorials A and B (you can either upload the manifesto documents yourself into Wmatrix, or use the ones I made earlier in the corpus library)
 - Do tutorial C (key words, key domains and concordances)
 - Ask questions any time!
 - Chance to provide feedback and influence future plans!



Thanks for listening!



Questions and comments?

PyMUSAS collaboration for existing and new languages welcome!!

- Contact:
 - Email: p.rayson@lancaster.ac.uk
 - https://bsky.app/profile/perayson.bsky.social

Icons from https://openmoji.org/

Key papers



- Rayson, P., Archer, D., Piao, S. L., McEnery, T. (2004). The UCREL semantic analysis system. In proceedings of the workshop on Beyond Named Entity Recognition Semantic labelling for NLP tasks in association with 4th International Conference on Language Resources and Evaluation (LREC 2004), 25th May 2004, Lisbon, Portugal, pp. 7-12.
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 - http://www.lrec-conf.org/proceedings/lrec2016/pdf/257 Paper.pdf
- El-Haj, M., Rayson, P., Piao, S., & Wattam, S. (2017). Creating and validating multilingual semantic representations for six languages: expert versus non-expert crowds. In Proceedings of the 1st Workshop on Sense, Concept and Entity Representations and their Applications: Proceedings of the 15th Conference of the European Chapter of the Association for Computational Linguistics. (pp. 61-71). Association for Computational Linguistics.
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References ...



- Wmatrix, CLAWS and USAS websites:
 - https://ucrel.lancs.ac.uk/wmatrix/
 - https://ucrel.lancs.ac.uk/claws/
 - https://ucrel.lancs.ac.uk/usas/
- Semantic lexicon expansion
 - Sheryl Prentice, Paul Rayson, Jo Knight, Mahmoud El-Haj, Solly Elstein (2021) A Domain Based Approach to Semantic Lexicon Expansion, International Journal of Lexicography. https://doi.org/10.1093/ijl/ecab028
- Useful background reading (keyness, annotation and MWE):
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Further reading ...



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