

Natural dialect processing

NLP for non-standardized language varieties

Verena Blaschke & Barbara Plank
MaiNLP lab, LMU Munich

Quantitative approaches in dialectology
and variationist sociolinguistics
December 7, 2023



Natural Language Processing

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Natural **Language** Processing

... but *which* languages?

NLP – but which “language(s)”?

- Many speakers, abundant data, standardization

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- Tricky for NLP!

Modern methods learn from massive amounts of data –
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NLP – but which “language(s)”?

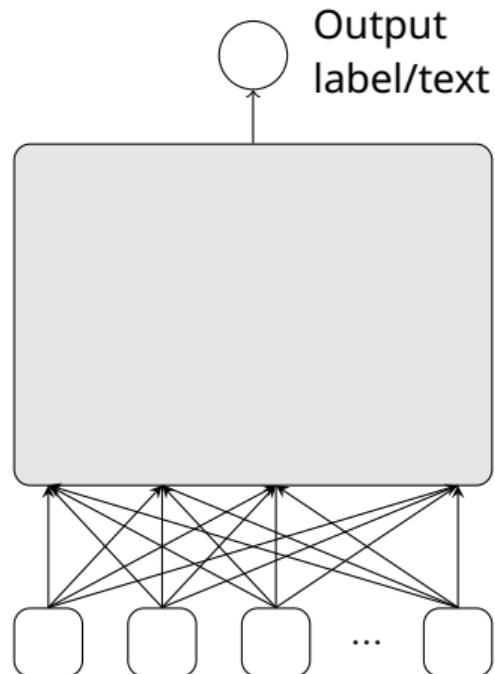
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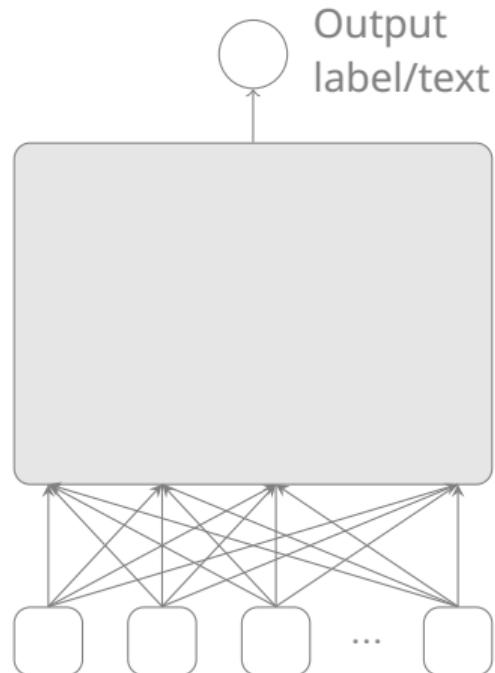
Overview of current challenges and approaches regarding
NLP & dialects

Overview



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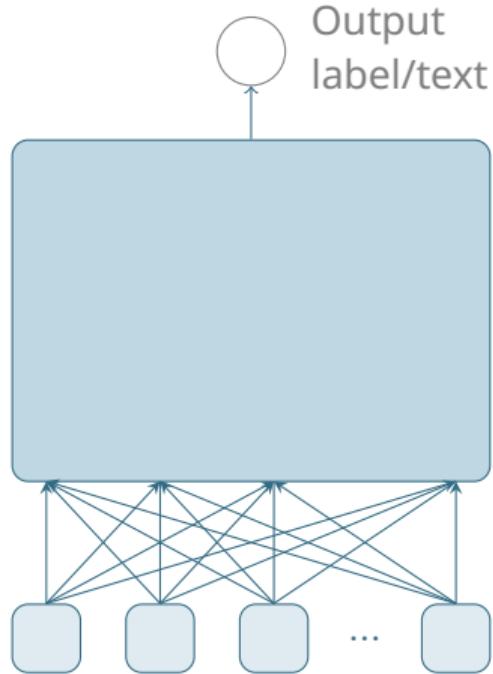
Overview



Input text sequence goes here

Available dialect data

Overview

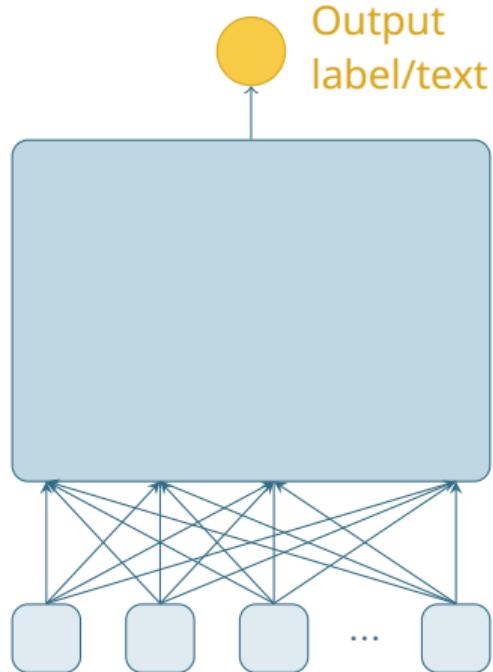


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⌚ Modelling non-standard data

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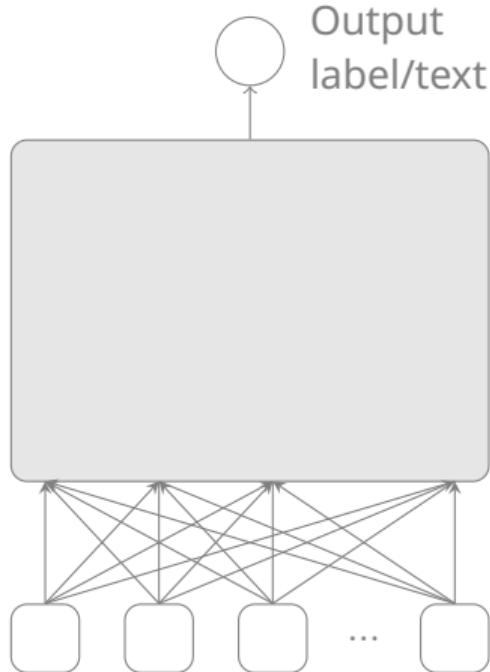
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👤 Human-centric NLP
(what tools and why?)

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(Lack of?) resources

Datasets for dialects and “small” languages

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- Two communities: variationists & NLP researchers

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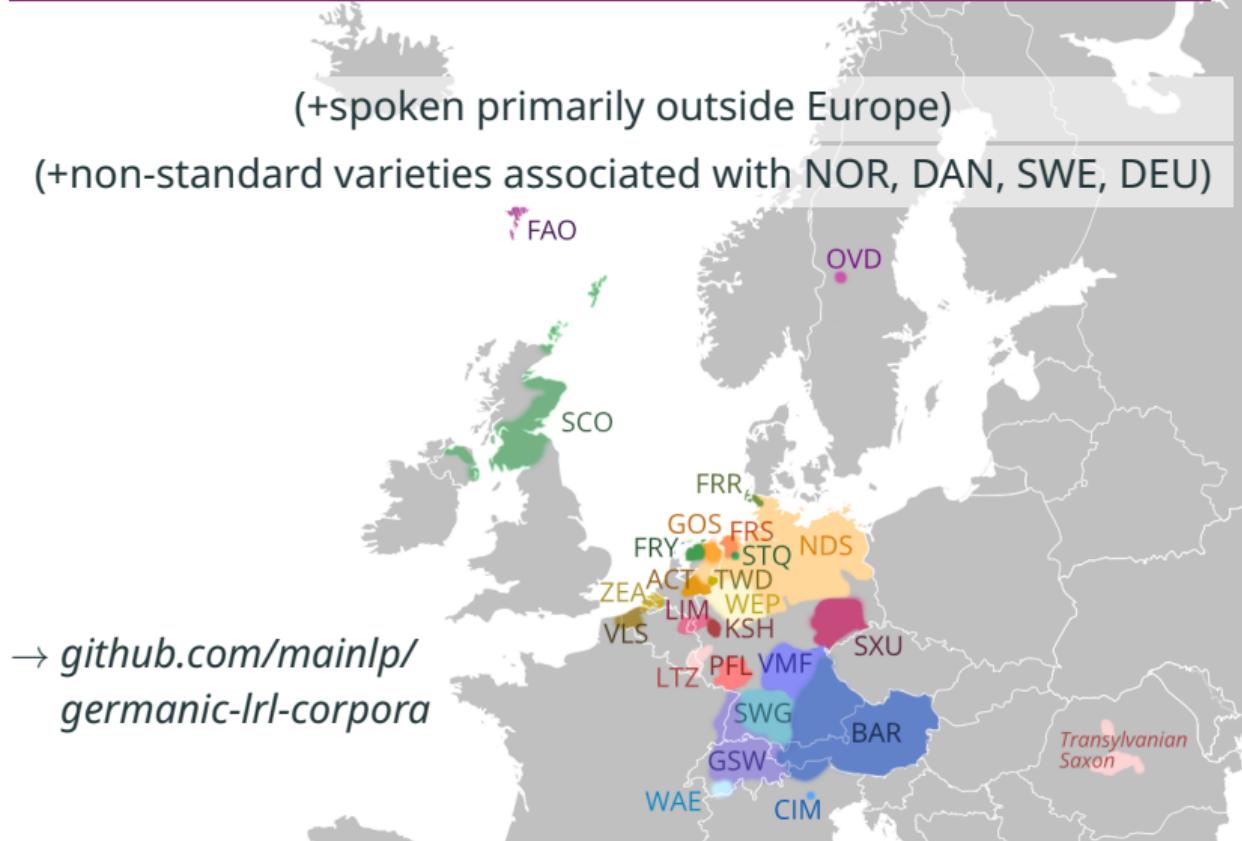
- Two communities: variationists & NLP researchers
- Findable; licenses allowing re-use

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Datasets for dialects and “small” languages

- Two communities: variationists & NLP researchers
- Findable; licenses allowing re-use
- Long-term storage + accessibility

Corpus overview (small/non-std Gmc varieties)



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100+ (mostly written) corpora for ~30 language varieties

Blaschke ea “A survey of corpora for Germanic low-resource languages and dialects” NoDaLiDa 2023

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- If annotated:
 - Geolocation, dialect group
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Recently: also uncurated, web-crawled ones

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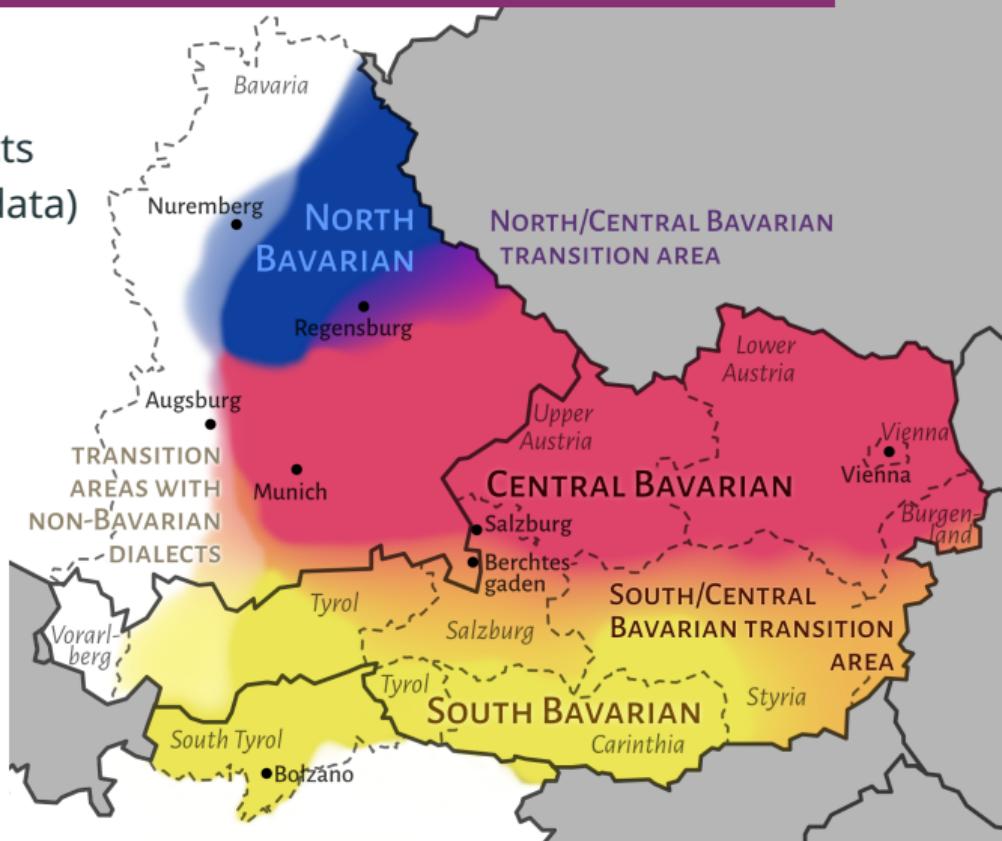
Data exchange between research communities?

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Bavarian dependency treebank

Work under review

- Different dialects
(location metadata)

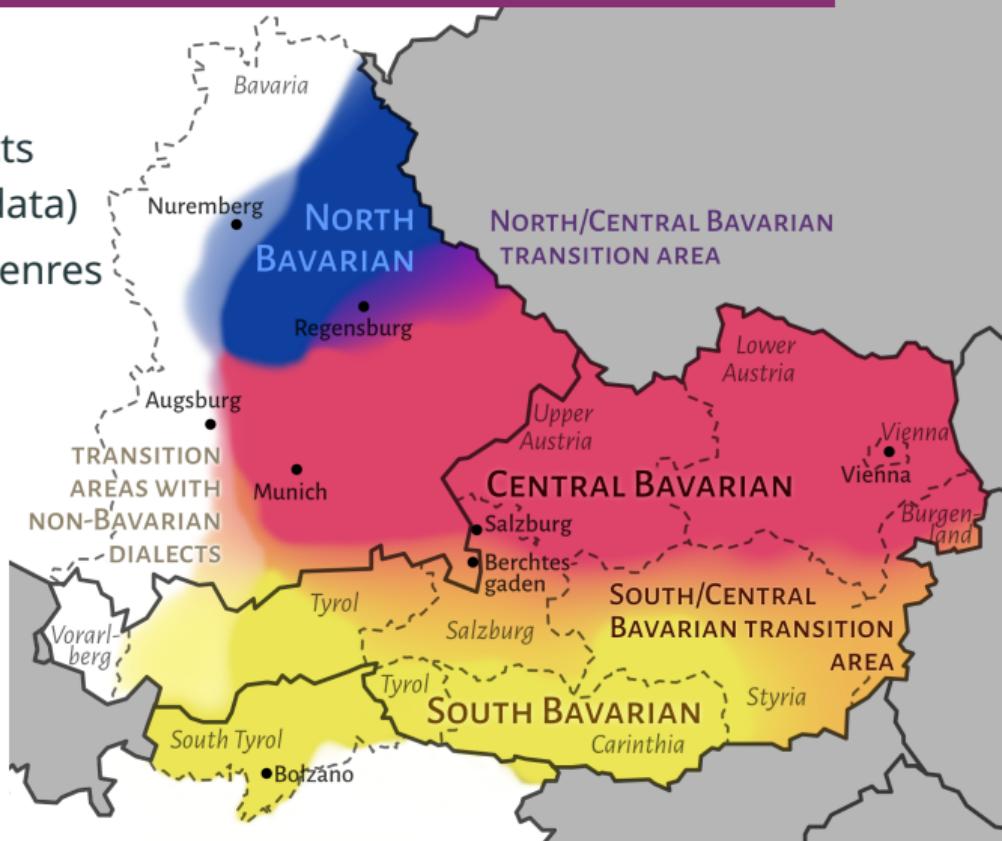


Map based on Wiesinger (1983, map 47.4)

Bavarian dependency treebank

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- Different dialects
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- Different text genres

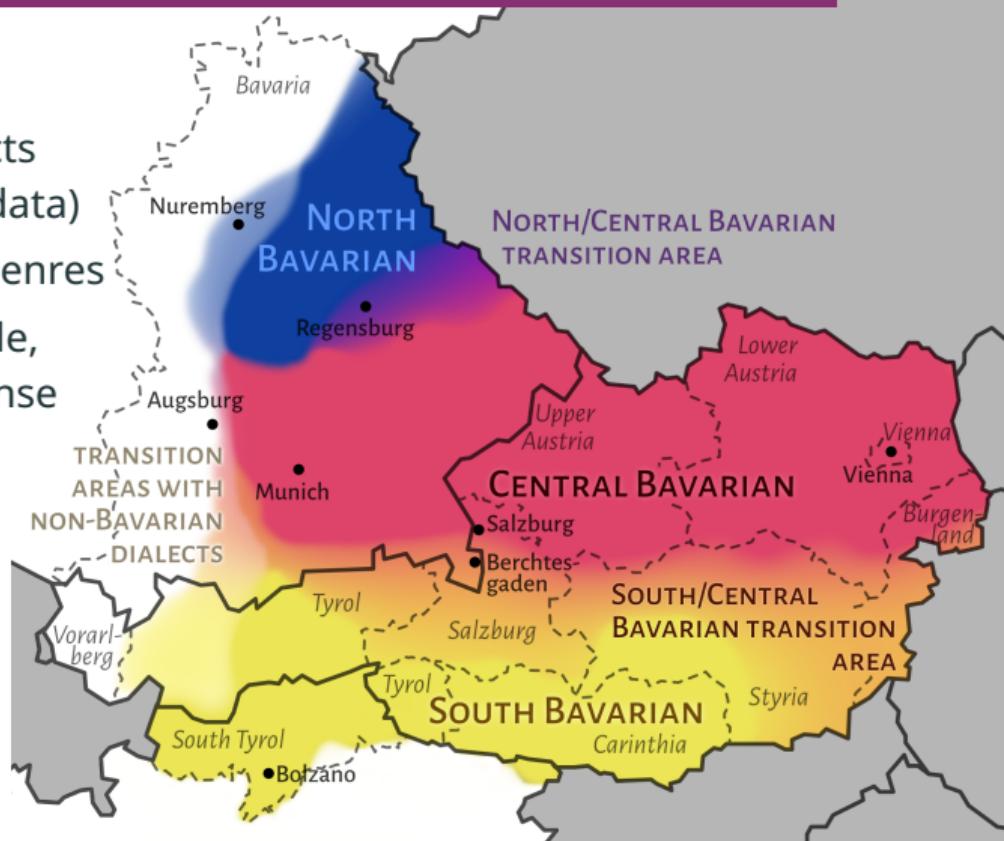


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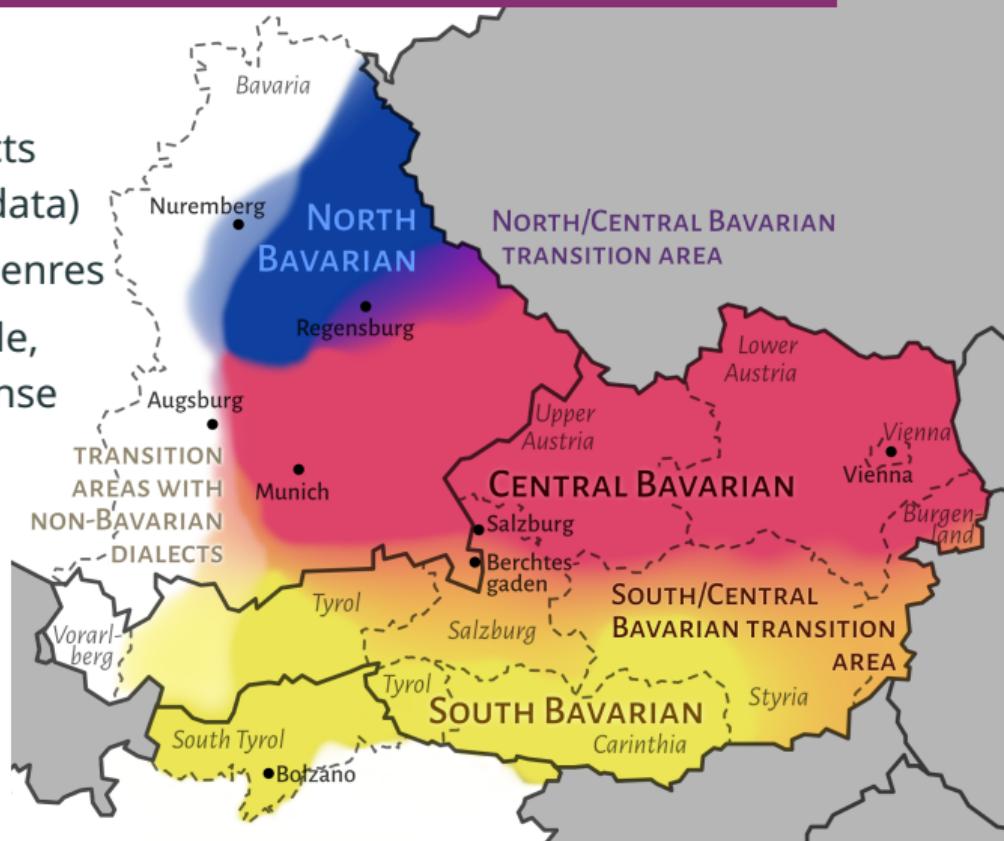


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- Different text genres
- Freely accessible,
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- ~ 11k words,
750 sentences

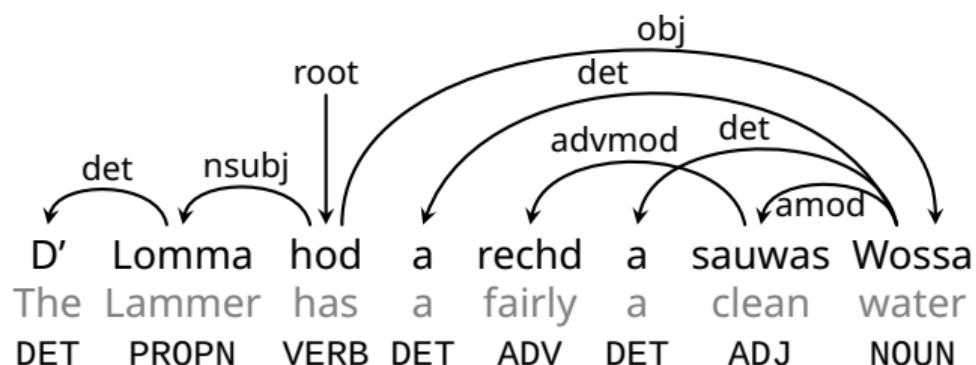


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Universal Dependencies



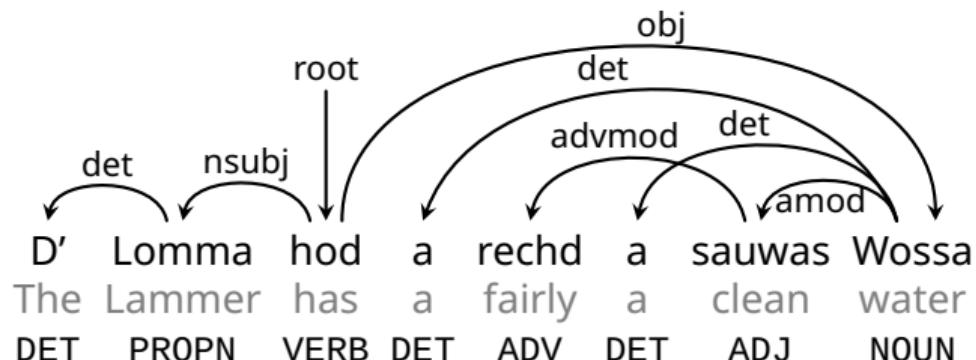
de Marneffe ea "Universal Dependencies" *Computational Linguistics* (2021)

Bavarian dependency treebank

Work under review

Universal Dependencies

- Cross-linguistic comparability (incl. DEU, GSW, NDS)



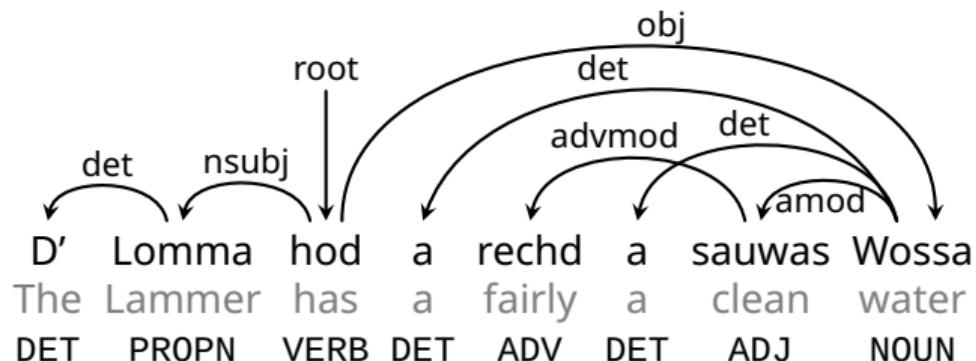
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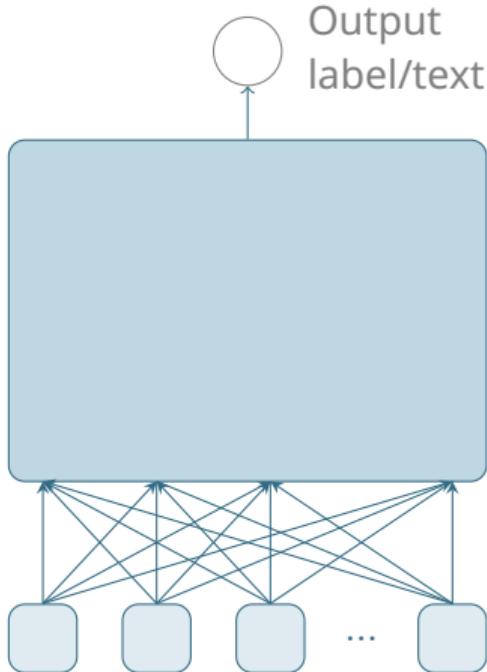
Universal Dependencies

- Cross-linguistic comparability (incl. DEU, GSW, NDS)
- Established for automatic annotation tasks



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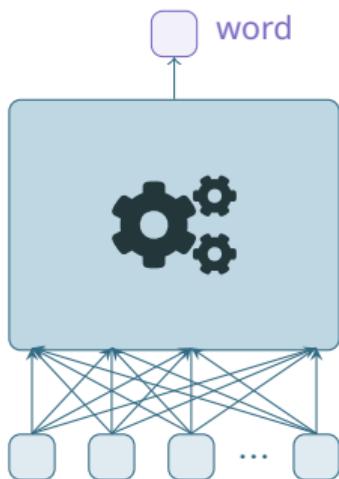
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Available dialect data

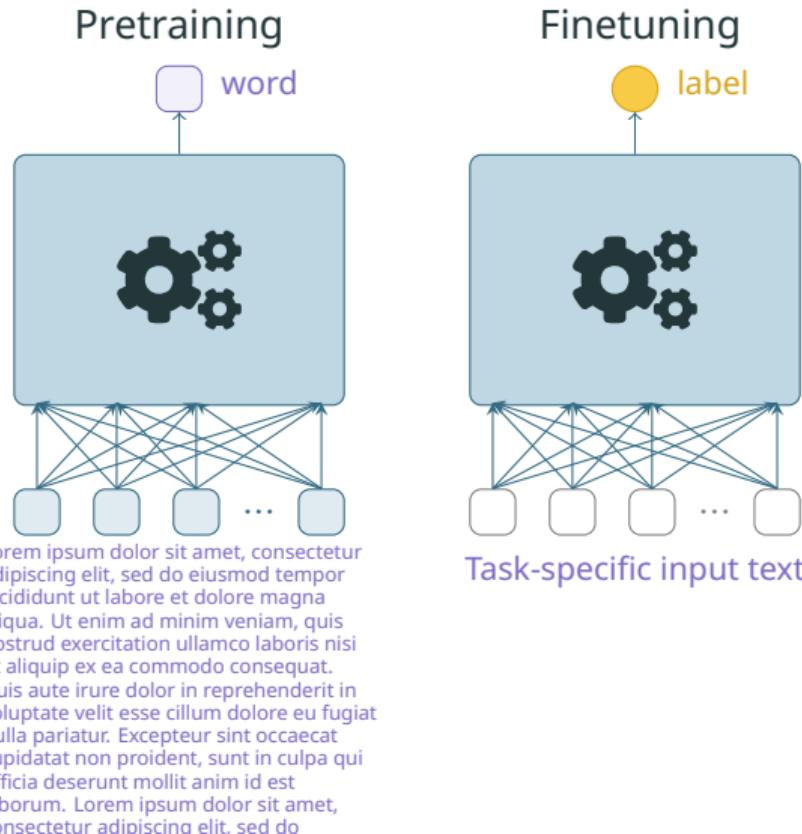
Pretrain – finetune – transfer

Pretraining



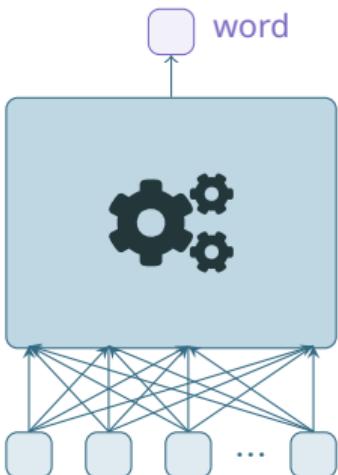
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Pretrain – finetune – transfer

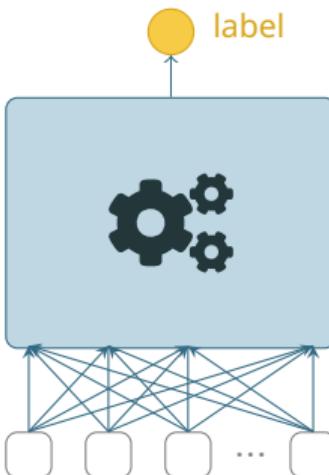


Pretrain – finetune – transfer

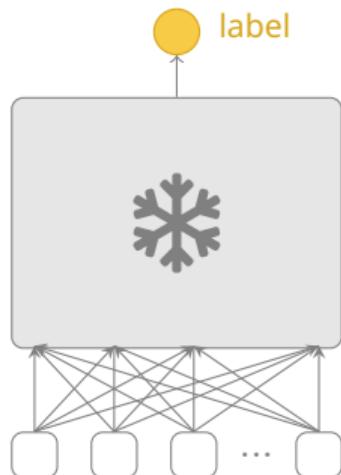
Pretraining



Finetuning



Transfer

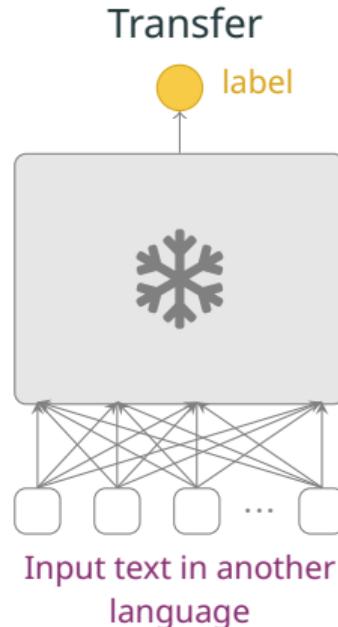
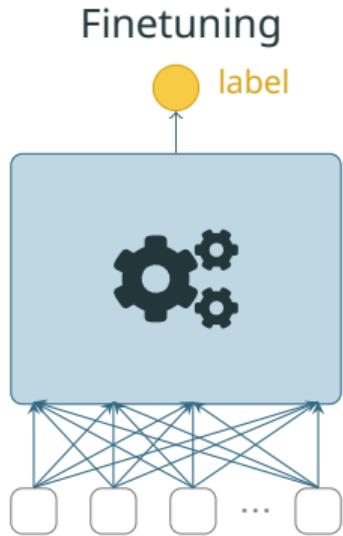
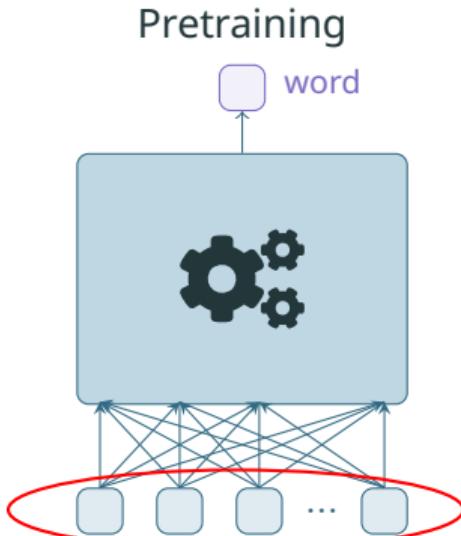


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Task-specific input text

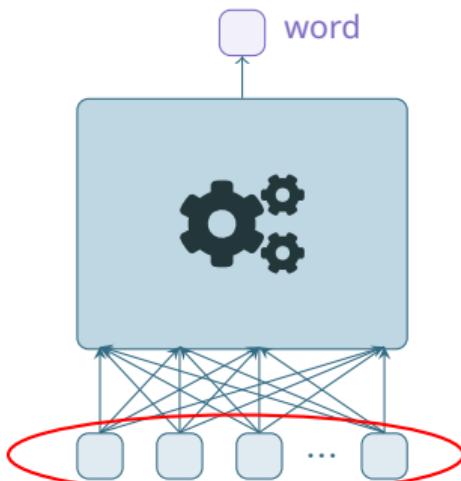
Input text in another language

Pretrain – finetune – transfer



Pretrain – finetune – transfer

Pretraining



Encoding input text

Map character sequences (“subword tokens”) to numeric representations

...
Lorem ipsum dolor sit amet, consectetur
adipiscing elit, sed do eiusmod tempor
incididunt ut labore et dolore magna
aliqua. Ut enim ad minim veniam, quis
nostrud exercitation ullamco laboris nisi
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Non-standard orthographies + tokenization

Die Lammer hat ein recht sauberes Wasser

D' Lomma hod a rechd a sauwas Wossa

Sentence via bar.wikipedia.org/wiki/Låmma
Chan ea "German's next language model" 2020

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Sidenote:

ChatGPT/GPT-4/etc also use such kinds of tokenization

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Solutions – ongoing work in the field

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- Move away from character-sequence-based representations altogether – requires training new models, experimental (overall performance might be worse)

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Solutions – ongoing work in the field

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- (How to) make existing models more robust to variation

Brittleness towards uncommon structures

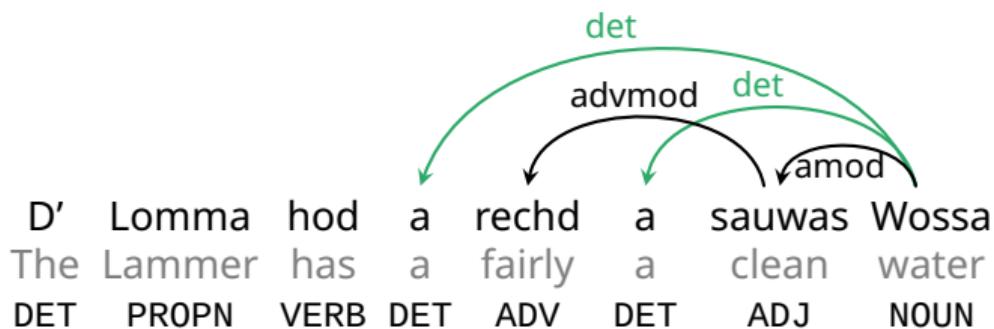
Work under review

D'	Lomma	hod	a	rechd	a	sauwas	Wossa
The	Lammer	has	a	fairly	a	clean	water
DET	PROPN	VERB	DET	ADV	DET	ADJ	NOUN

Sentence via bar.wikipedia.org/wiki/Låmma

Brittleness towards uncommon structures

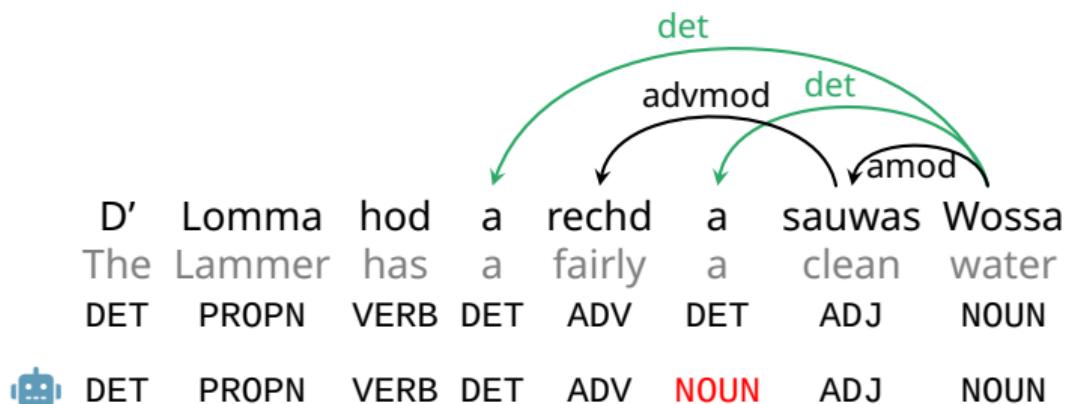
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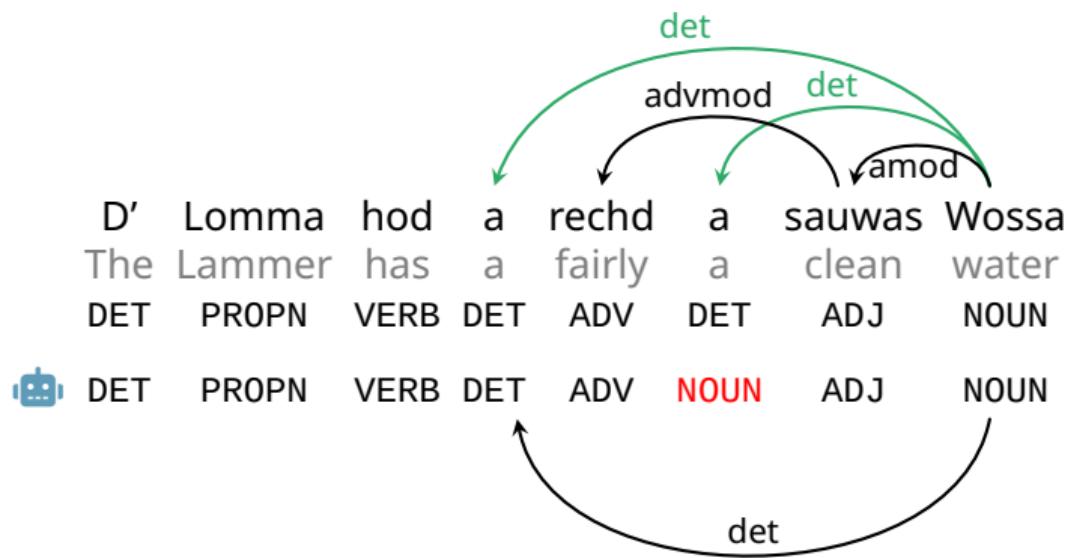
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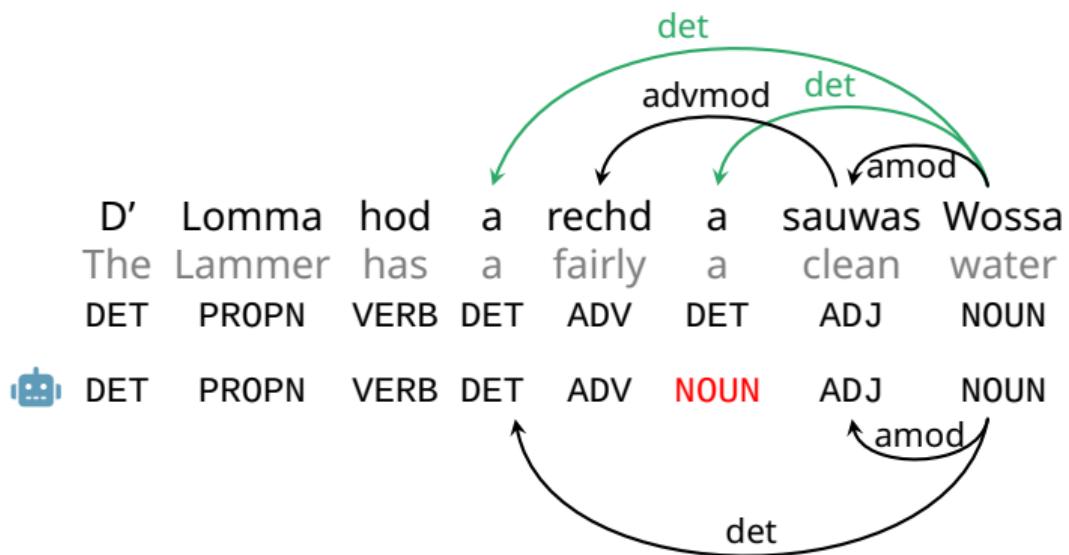
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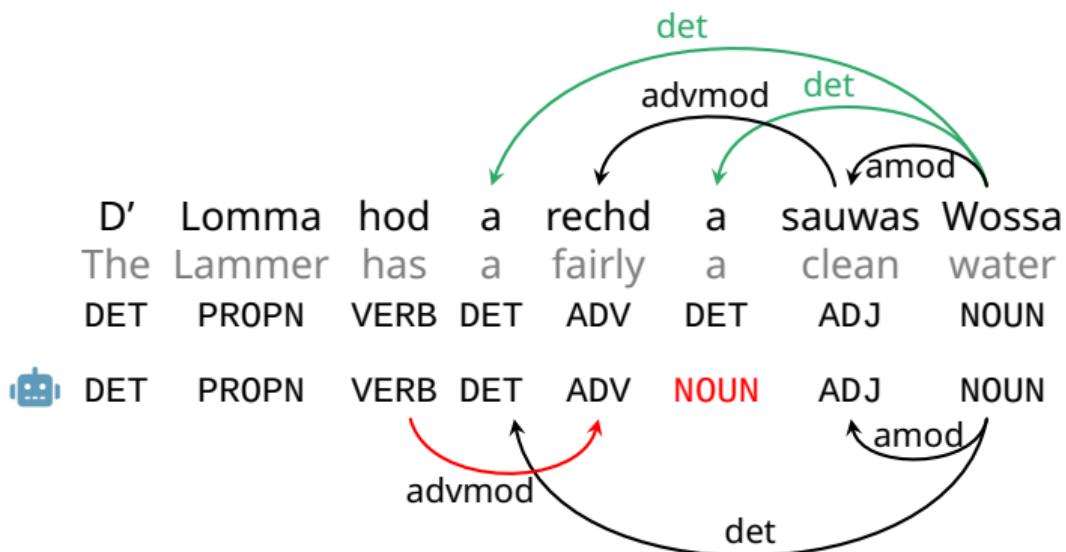
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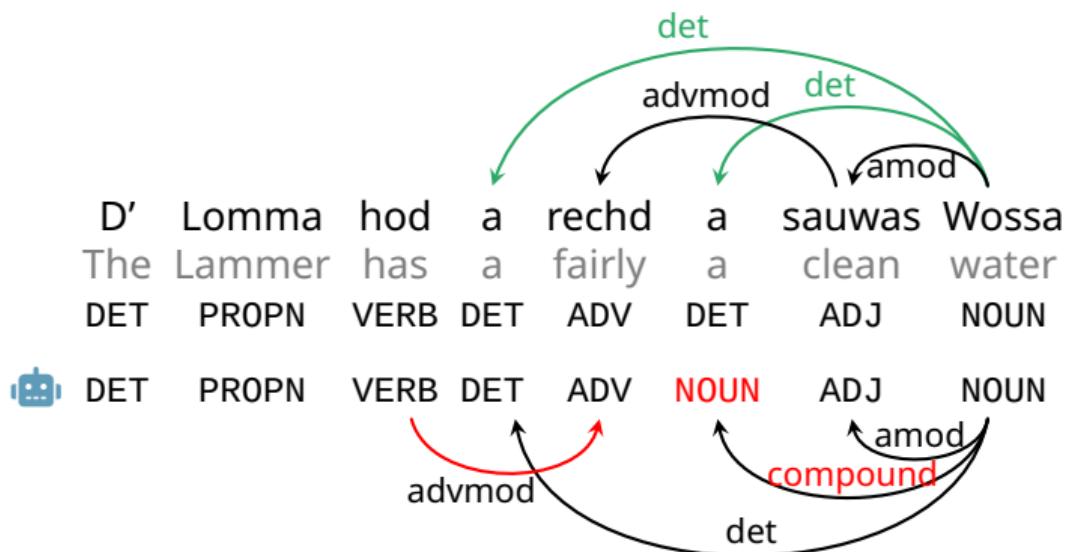
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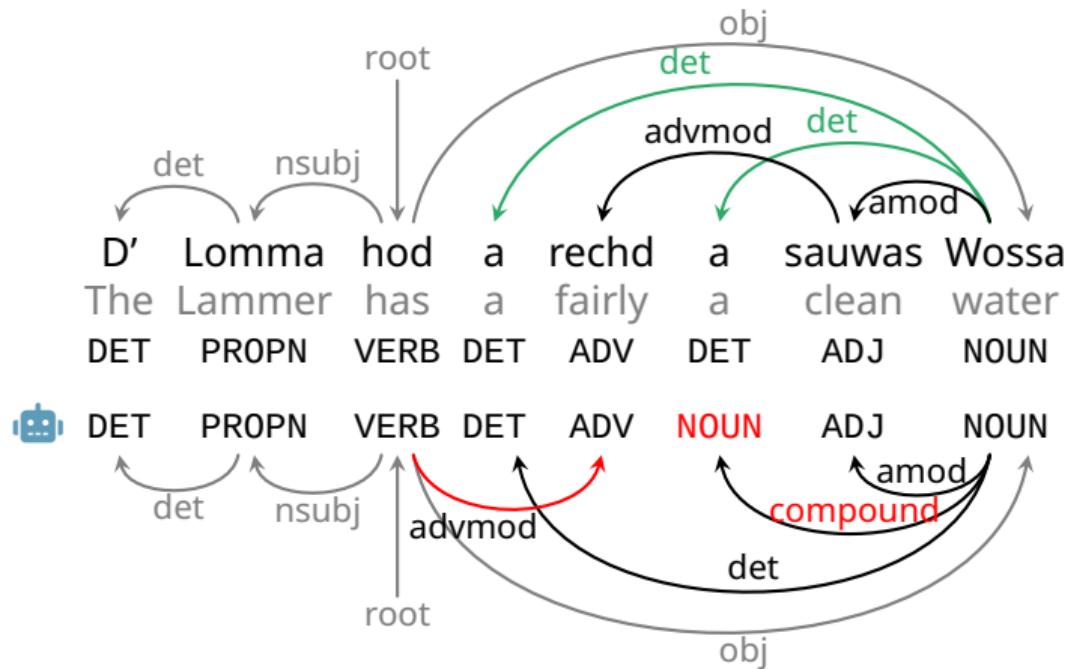
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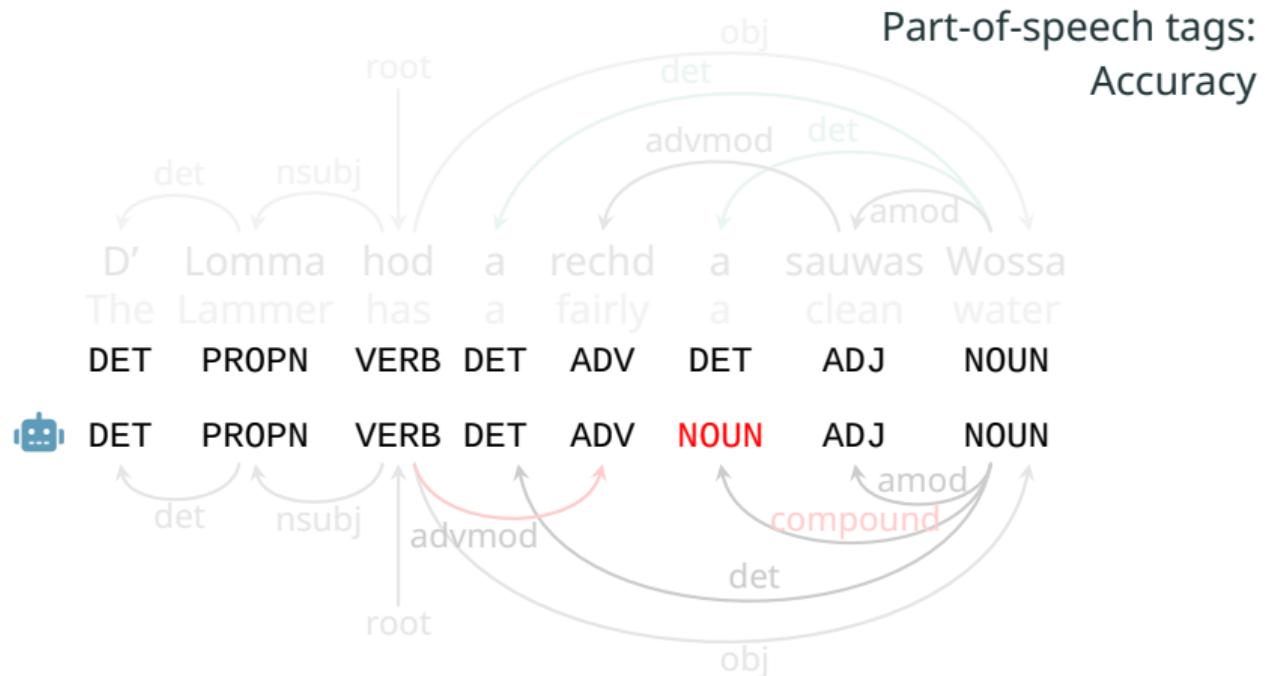
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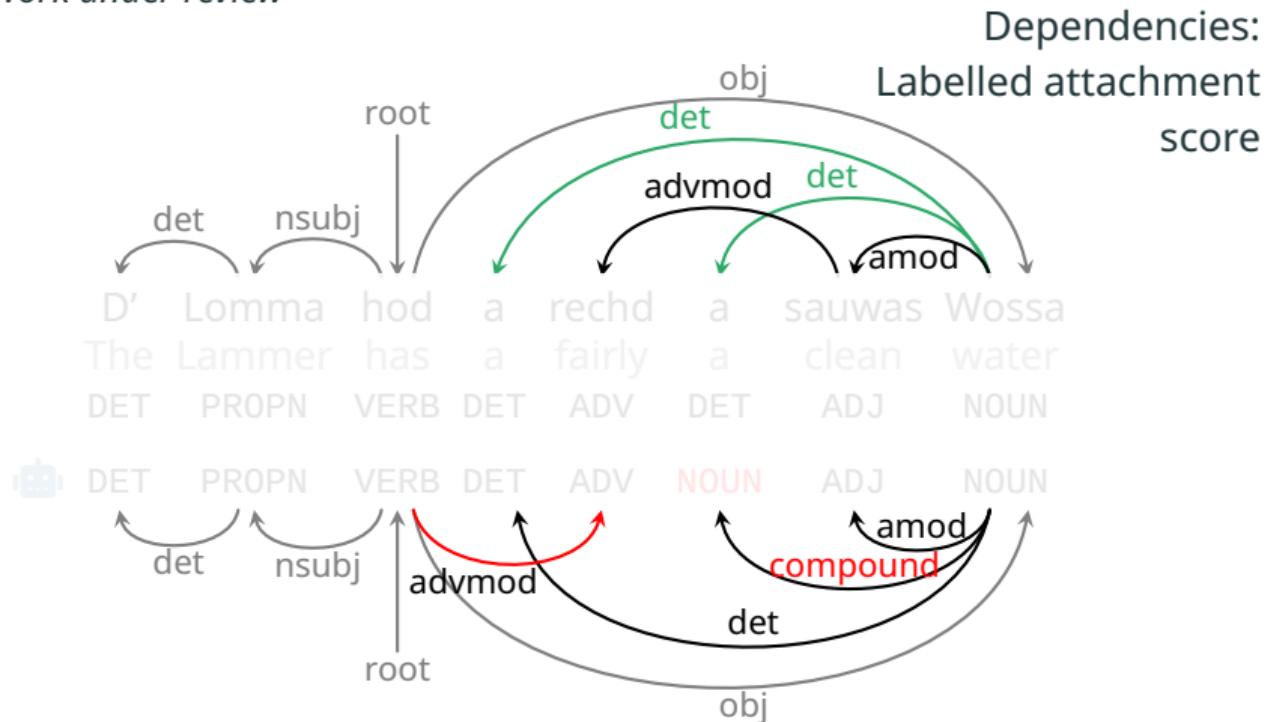
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Automatic tagging/parsing

Work under review

Train on German data (there is no Bavarian training data!),
test on German vs. Bavarian

Model	Test lang	Acc (%)	LAS (%)
Stanza	DEU	95.9	83.7
GBERT	DEU	96.8	83.1
UDPipe	DEU	96.5	84.9

Acc = accuracy (part-of-speech tags); LAS = labelled attachment score

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Stanza	BAR	40.9	23.1
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UDPipe	BAR	80.5	67.3

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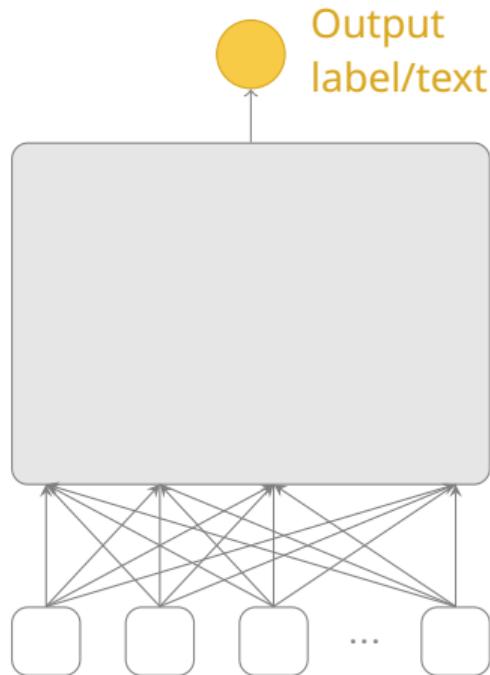
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GBERT	BAR	57.4	30.1	Subword tokens
UDPipe	BAR	80.5	67.3	Subword tok. + characters

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(what tools and why?)

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What NLP tools and why?

Computational linguistics & machine learning research

- Quantitative patterns
- How to learn from sparse + heterogeneous data?

Blokland ea “Language documentation meets language technology” *Septentrio conference series* (2015)

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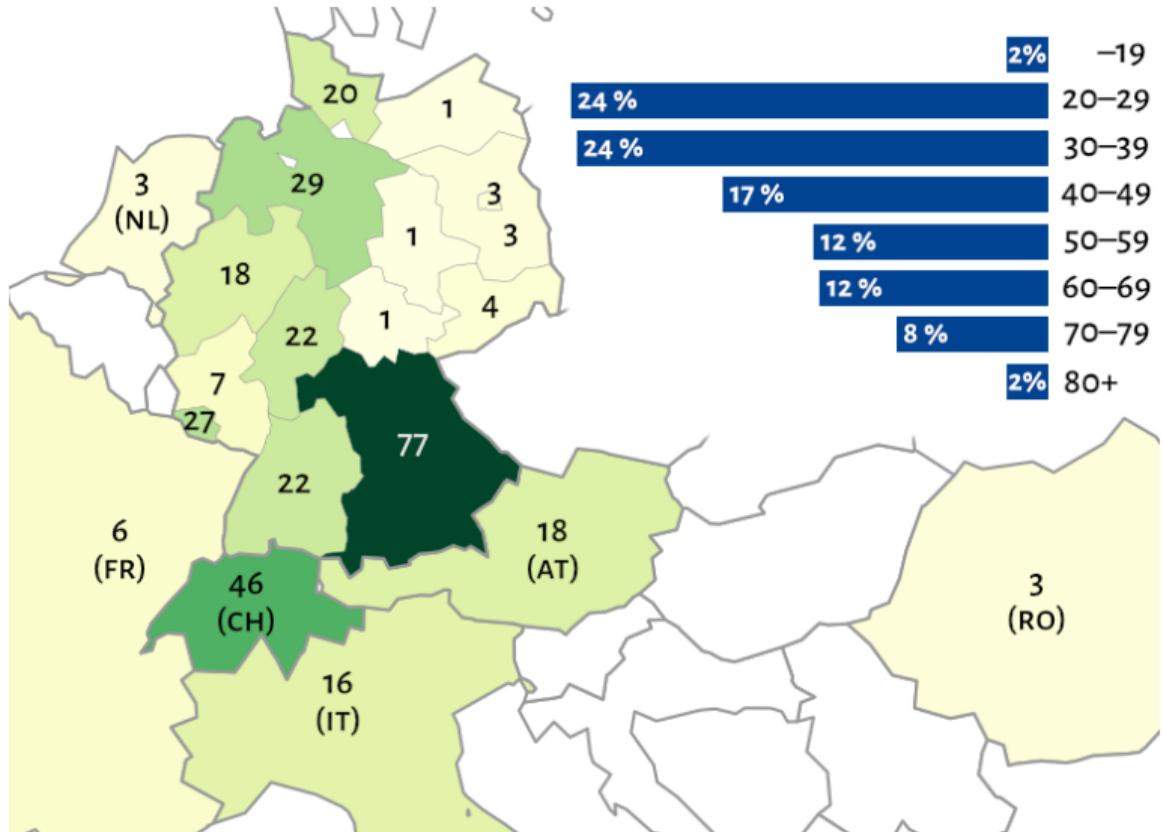
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NLP tools for linguists – also require dialogue between the communities!

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Language technology for dialect speakers

Work under review



Language technology for dialect speakers

Work under review

Self-identified dialect speakers

- How often / in what contexts do people speak/write their dialects?
- What do they think about various language-related technologies for their dialects? (e.g., spellcheckers, machine translation, digital assistants...)

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 - Between subgroups (e.g., GSW vs. NDS speakers)

Language technology for dialect speakers

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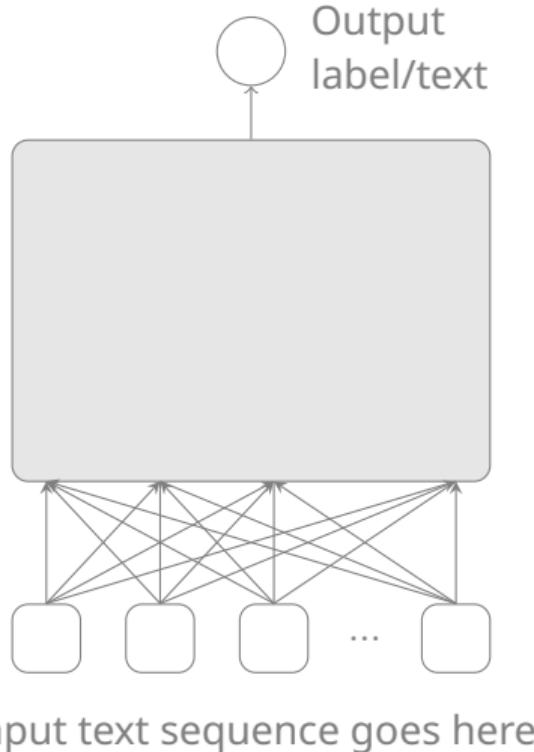
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 - Within subgroups

Summary



💡 Reflecting on
what tools we build

🤖 Representing/modelling
non-standard data

🧩 Data availability
→ [github.com/mainlp/
germanic-irl-corpora](https://github.com/mainlp/germanic-irl-corpora)

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