



# Is computational language modelling linguistics?

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7 October 2024

**Transformers** were invented for natural language processing (NLP)

processing = transforming

German sentence → English

any sentence → LABEL

## 6 Results

### 6.1 Machine Translation

Is this linguistics?

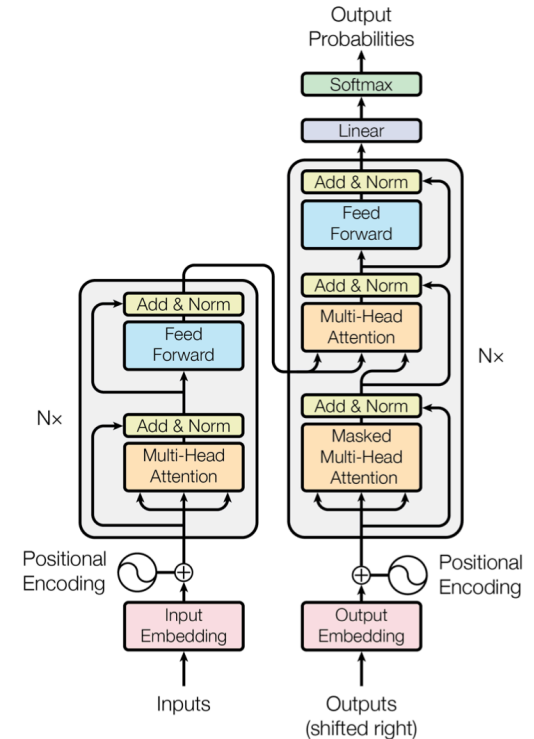


Figure 1: The Transformer - model architecture.

# Language and linguistics

## Language as art

- Speaking or writing "correctly"
- Speaking a foreign language
- Being able to translate
- Being able to teach a language
- Write dictionaries

## Philology

- Interpret the meaning of complex texts
- Knowing the particularities of a (single) language
- Knowing the associated culture and history

## Language as a cognitive capacity

Studied mostly in psychology

## Language as social phenomenon

Studied mostly in psychology

# General linguistics as typology



# General linguistics as syntactic theory



<http://mshang.ca/syntree/>



# Computational modelling as a means of integrating philological knowledge

**THE WORLD ATLAS OF LANGUAGE STRUCTURES ONLINE**



Home Features Chapters Languages References Authors

## Welcome to Glottolog 5.0

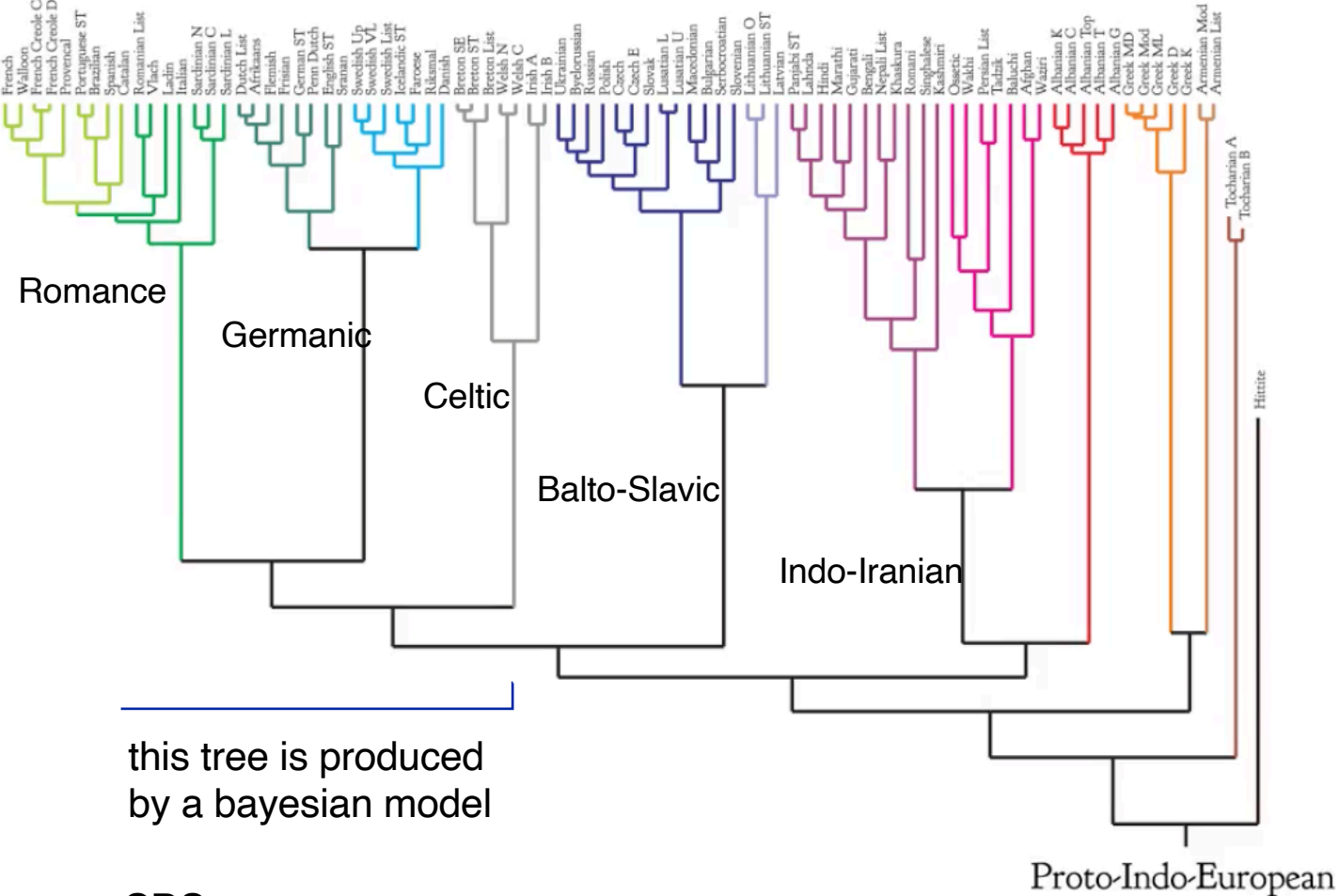
Comprehensive reference information for the world's languages, especially the lesser known languages.

Information about the different languages, dialects, and families of the world ('languoids') is available in the [Languages](#) and [Families](#) sections. The [References](#) section contains bibliographical information. You can query the bibliographical database by filtering the table view or using a [complex query](#) involving genealogical affiliation, document type, and macro-area.



### Catalogue of languages and families

**Glottolog** provides a [comprehensive catalogue](#) of the world's languages, language families and dialects. It assigns a unique and stable identifier (the Glottocode) to (in principle) all languoids, i.e. all families, languages, and dialects. Any variety that a linguist works on should eventually get its own entry. The languoids are organized via a genealogical classification (the Glottolog tree) that is based on available historical-comparative research (see also the [Languoids information](#) section).



this tree is produced by a bayesian model

**SDS group,  
Language and Space Lab, UZH**



# Computational modelling as a means of interpreting the meaning of texts

## Natural language processing (NLP)

$$\Psi(y, x)$$

$$p(y, x)$$

$x$  = the string "computational language modelling"  
(represented as a vector of features)

### Various predictions (NLP tasks)

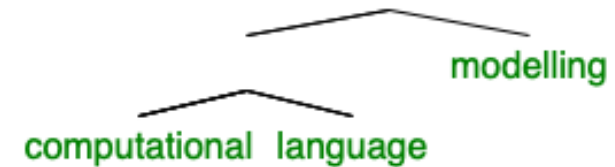
$y$  = a syntactic tree showing the relations between the words

$y$  = a synonym or a paraphrase

$y$  = sentiment (is this something good or bad?)

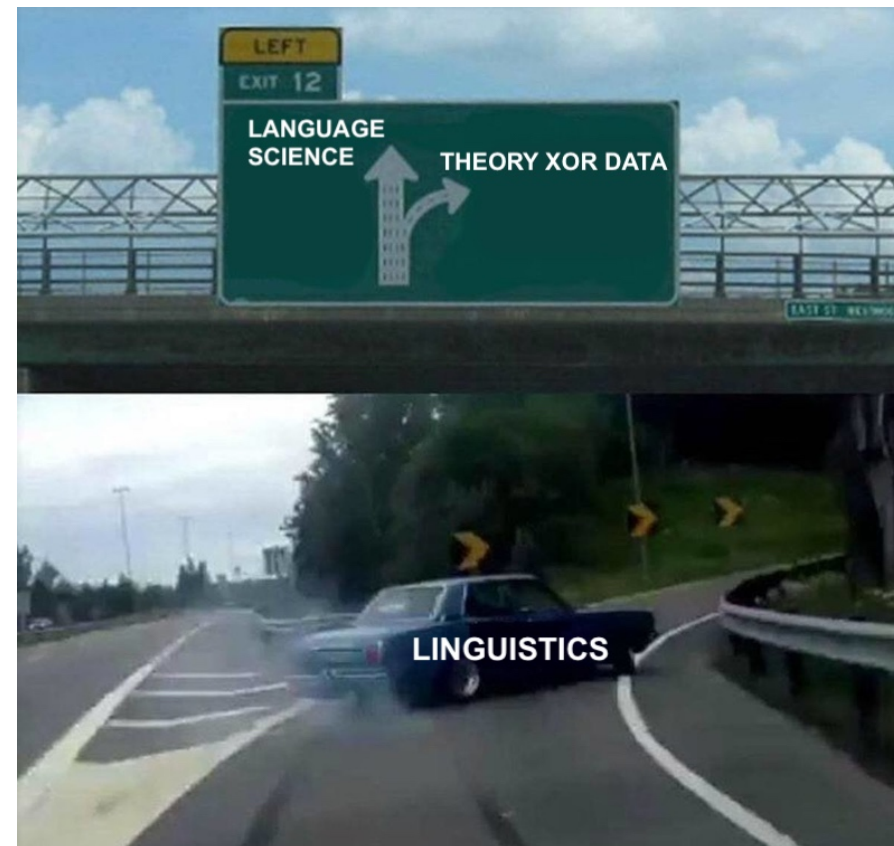
$y$  = translation to another language

many more ...

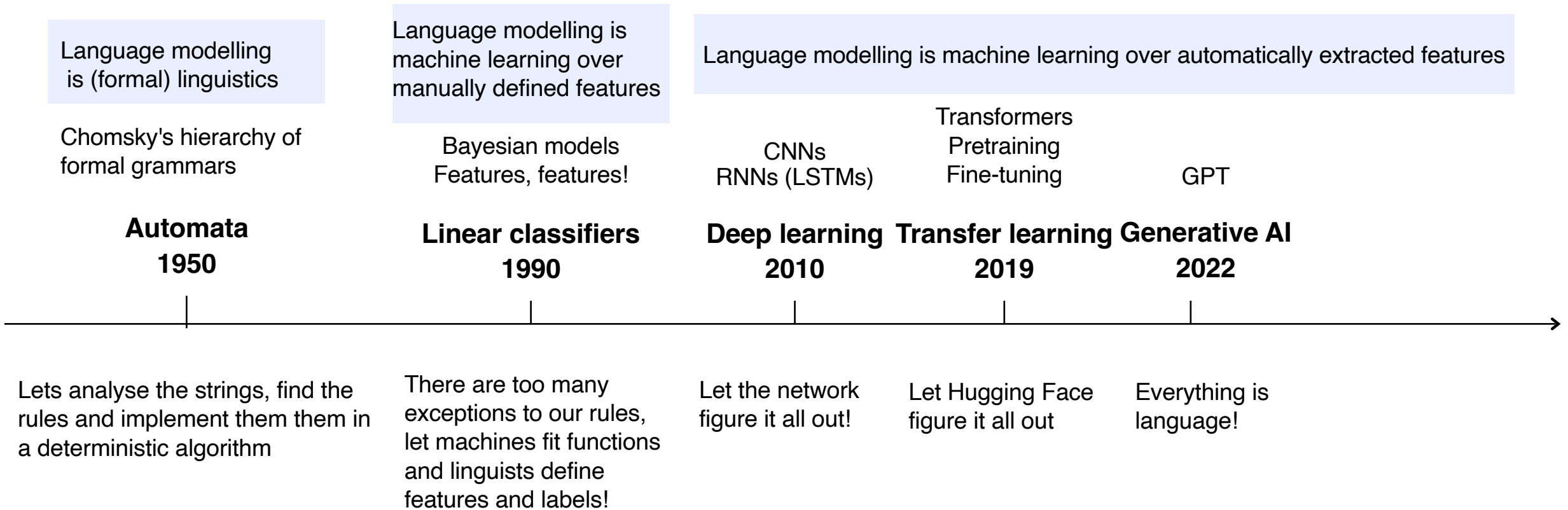


# Machine learning as a replacement for linguistics in NLP

"Statistical revolution" in AI in 1990s makes a split in NLP



# From linguistics to language

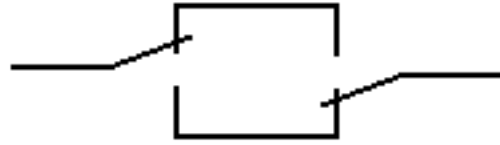
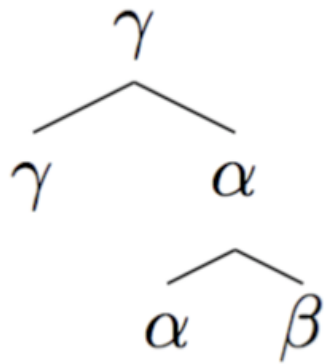




# Linguistics as theory XOR data

## Theory (Generative Grammar)

Counting observations is irrelevant to understanding language capacity!  
Especially since **1959**



## Typology

## Construction Grammar

## Corpus Linguistics

Let the data speak for themselves!

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Wikimedia Common

# Scientific language modelling is still possible as linguistics

## Theory AND data

For now, *information theory*, but likely relevant to syntax

## A lot of data

Text samples from many languages

## Computing

Simple "cheap" methods at the core of the research, transfer learning for tests

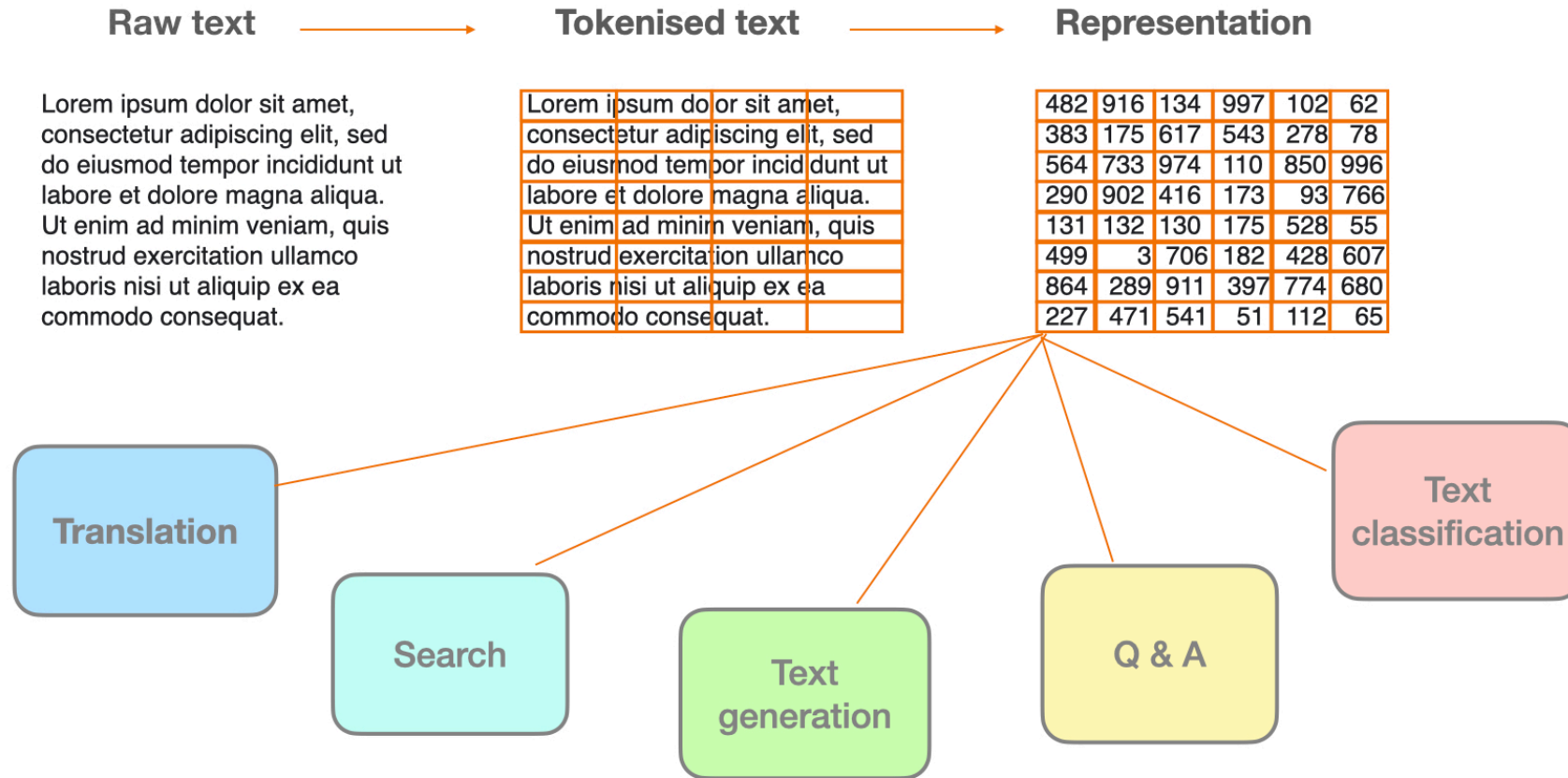
## Problem selection

An interesting theoretical problem, but also of an immediate use in practice

1. Text tokenisation
2. Cross-lingual transfer



# Text tokenisation as an interesting fundamental problem





# How to segment input text?

## Words are tokens

Traditional view strongly influenced by English

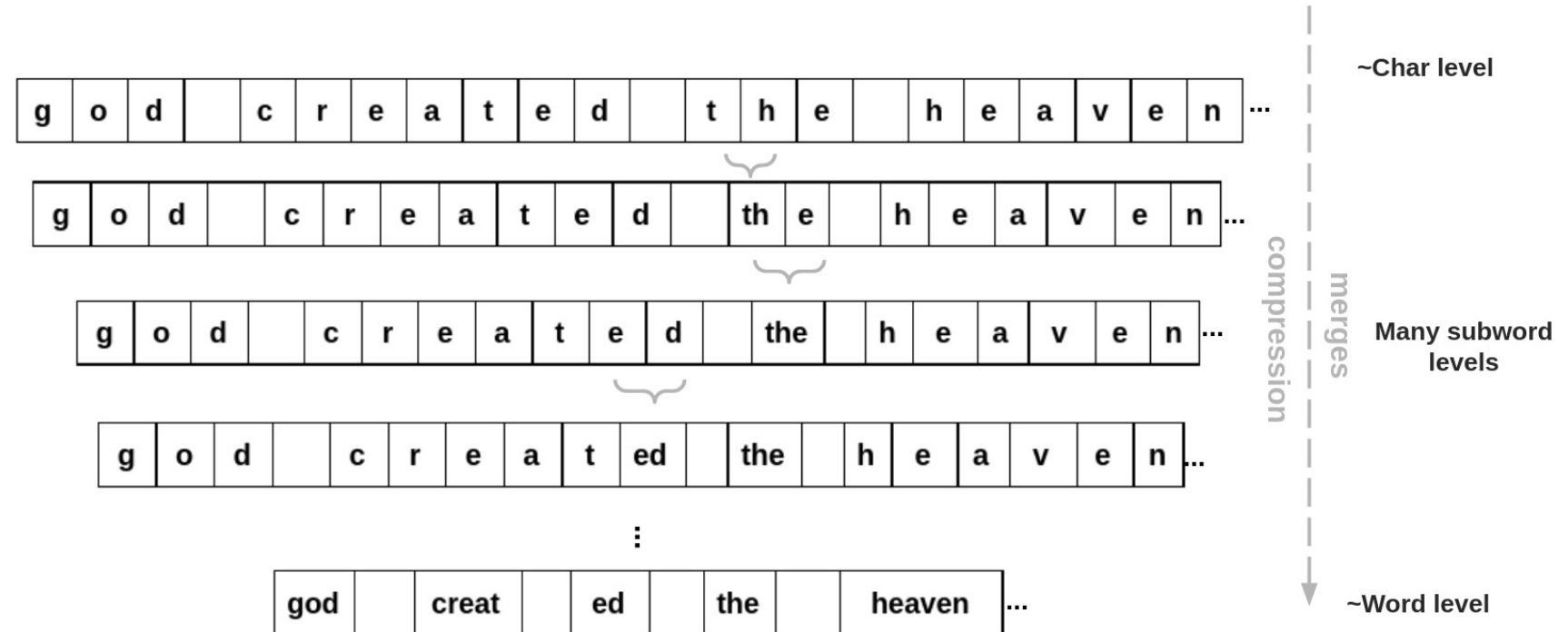
## Subword tokenisation

**BPE** Introduced in **2016**

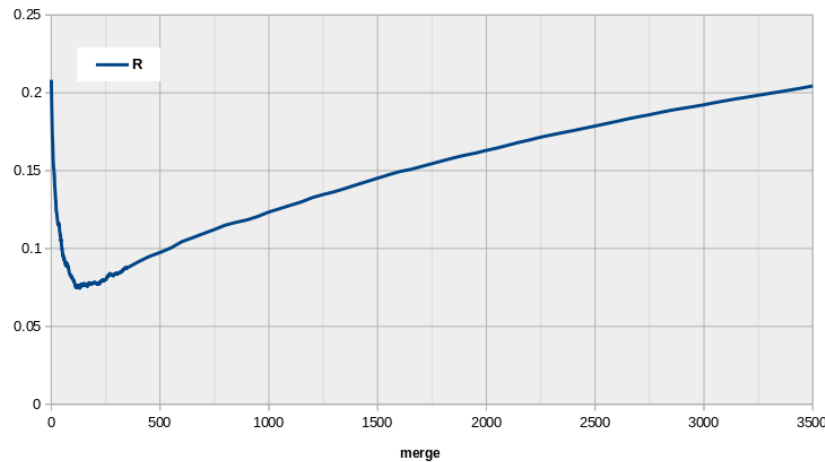
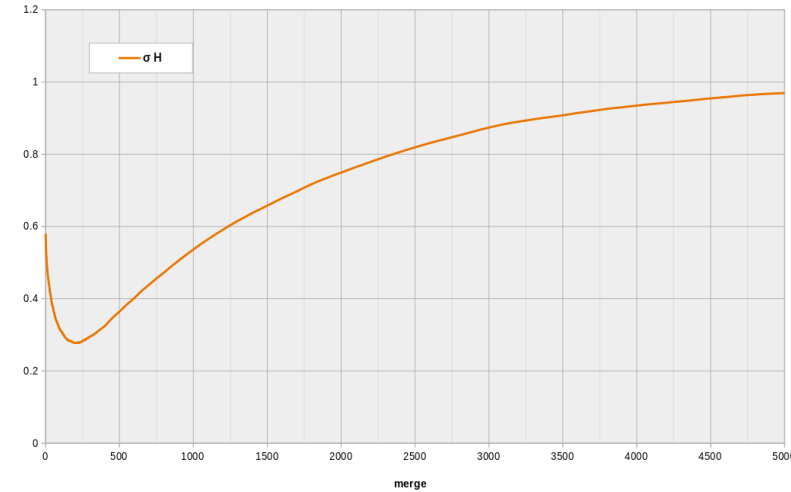
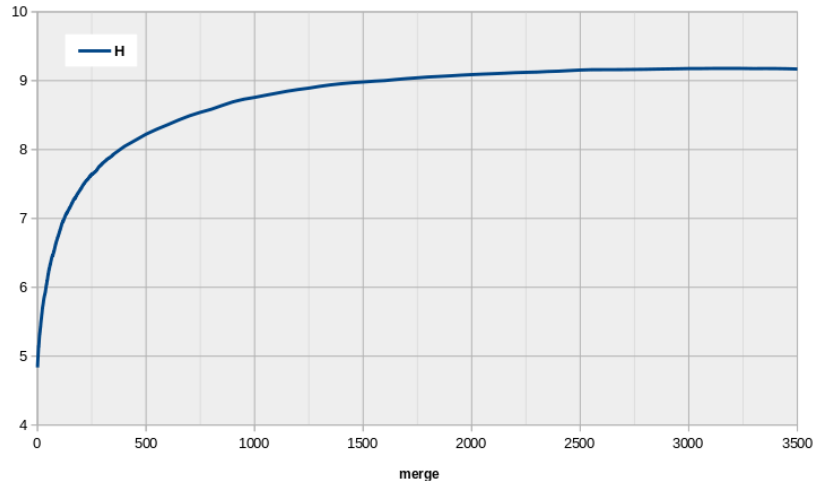
For some reason works with NNs

## Problems

- What subwords?
- Pre-trained models come with a selected (arbitrary) tokenisation
- Discrimination against languages other than English



# How about stopping BPE at minimum redundancy? (BPE-MR)



## Minimum redundancy, converging text entropy

There is an area of subword tokenisations where text redundancy is minimised. In the same area, text entropy growth slows down and the values across languages start converging

From characters to words: the turning point of BPE merges

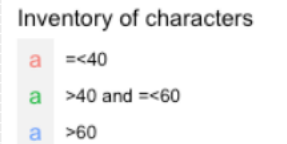
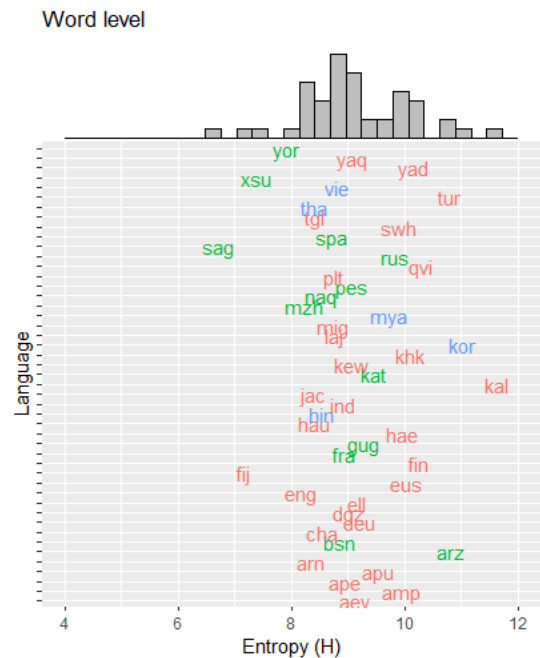
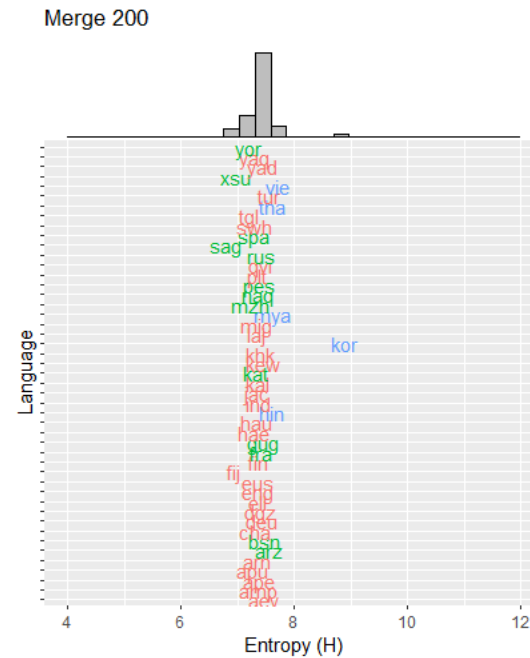
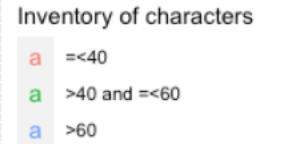
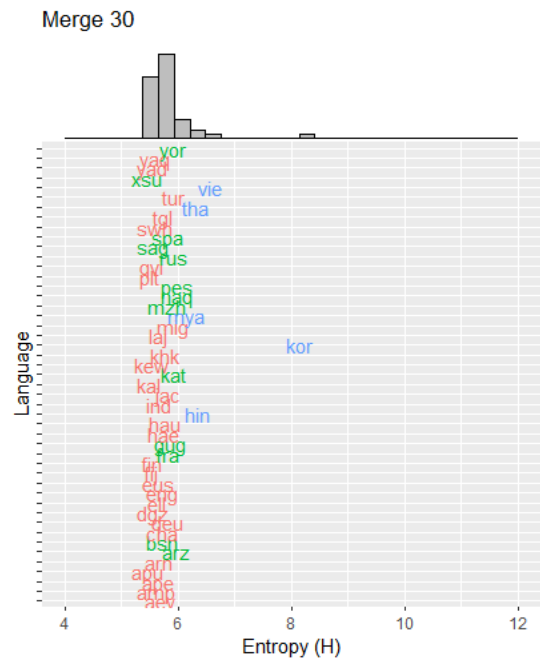
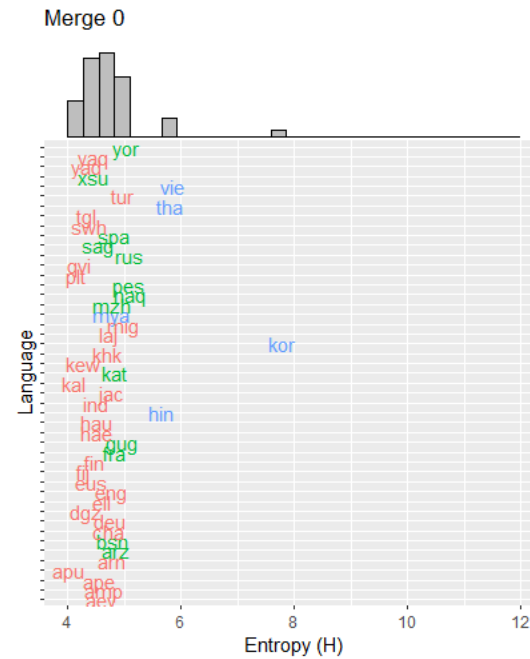
Ximena Gutierrez-Vasques, Christian Bentz, Olga Sozinova, Tanja Samardžić

EACL2021

# Text entropy across languages

## Early merges (200-350)

Text entropy almost the same across 47 languages in the parallel Bible corpus



From characters to words: the turning point of BPE merges

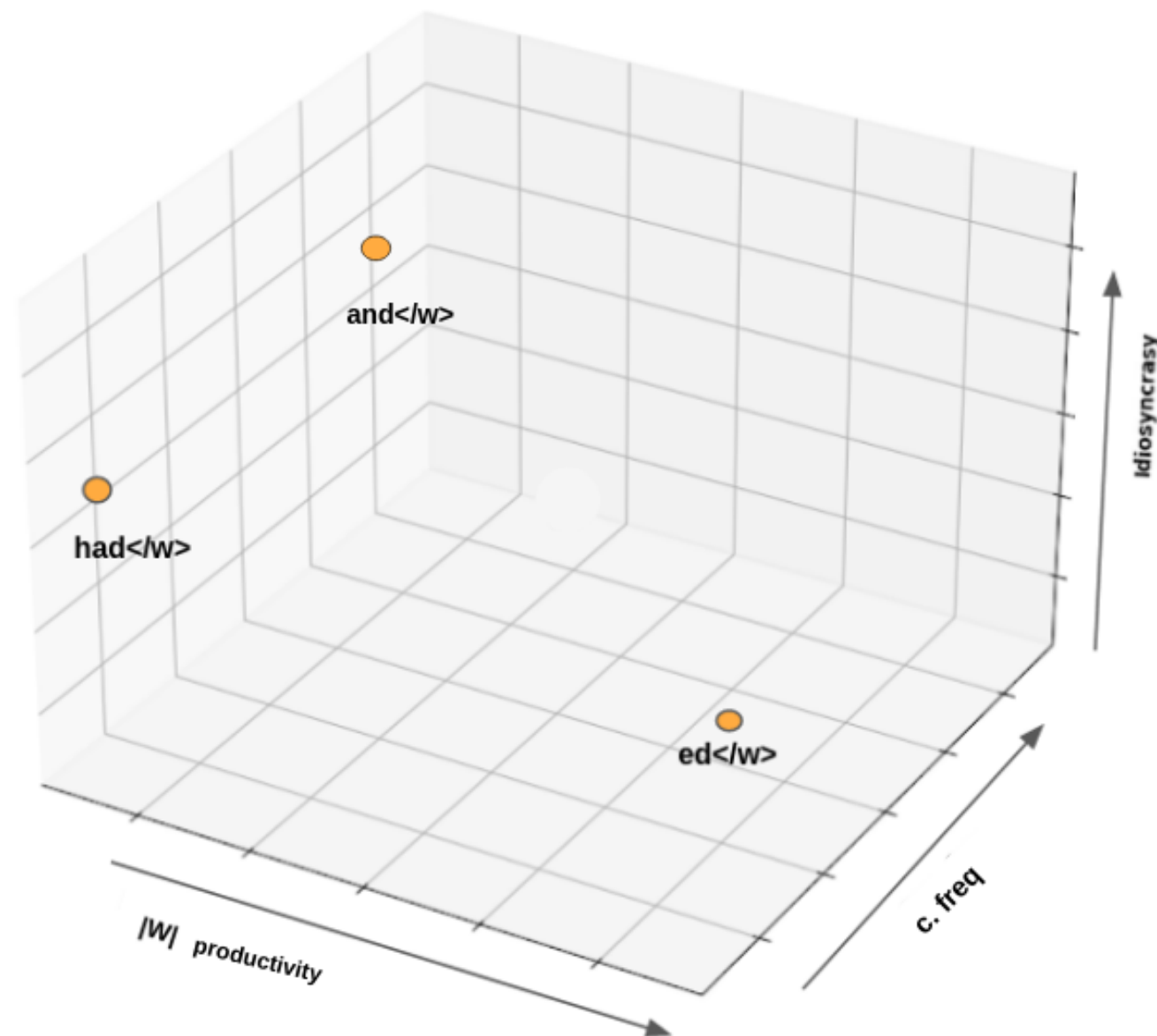
Ximena Gutierrez-Vasques, Christian Bentz, Olga Sozinova, Tanja Samardžić  
EACL2021



# What are the BPE units at minimum redundancy?

## Observations in a 3D space

It looks like we have productive affixes on the floor, and function words on the left-hand side wall



Languages through the Looking Glass of BPE

Compression

Ximena Gutierrez-Vasques, Christian Bentz, Tanja

Samardžić

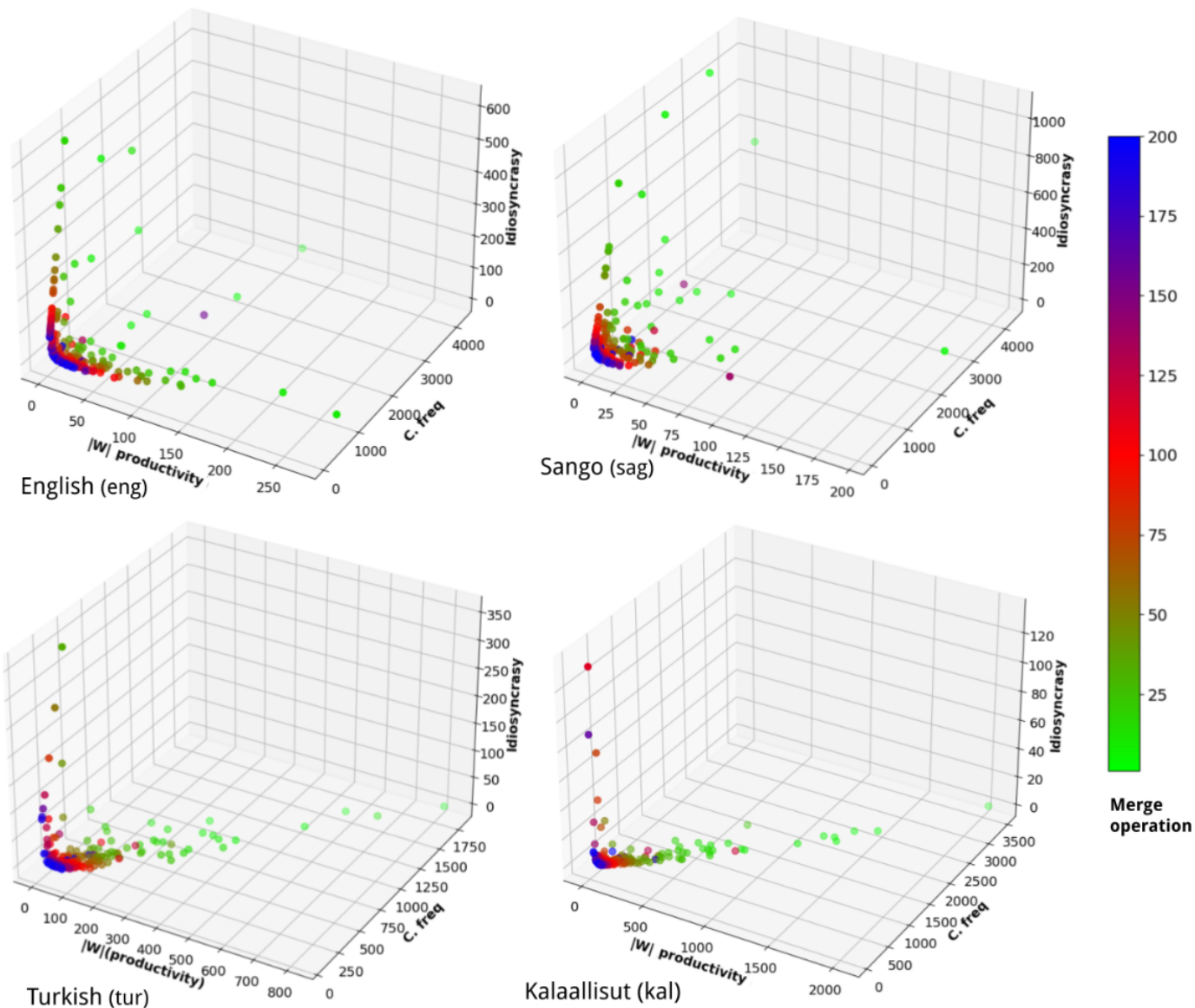
Computational Linguistics 2023

# How do languages look like in this 3D space?

## Languages have different shapes

Those with longer, more complex words tend to have most items on the floor, those with short words tend to have most items on the wall

This BPE units that are merged first are the most discriminative



Languages through the Looking Glass of BPE  
Compression

Ximena Gutierrez-Vasques, Christian Bentz,  
Tanja Samardžić

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# Implications for machine translation

## Translation from Spanish into 11 American indigenous languages

Hñähñu, Wixarika, Nahuatl, Guaraní, Bribri, Rarámuri, Quechua, Aymara, Shipibo-Konibo, Asháninka, Chatino

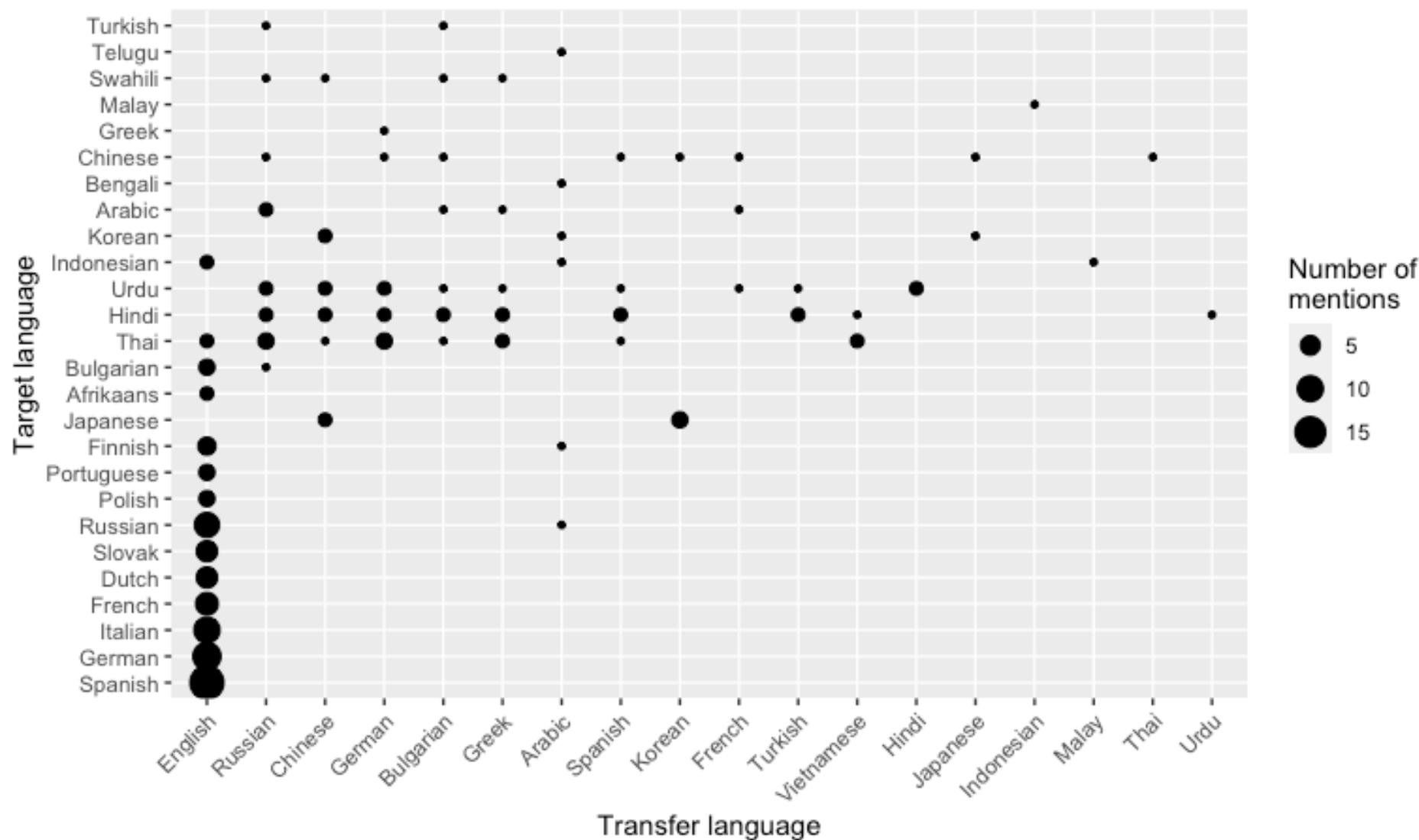
## NordicAlps

The only system that outperformed last year's winner

System Description of the NordicsAlps Submission  
Joseph Attieh, Zachary Hopton, Yves Scherrer,  
Tanja Samardžić  
AmericasNLP 2024



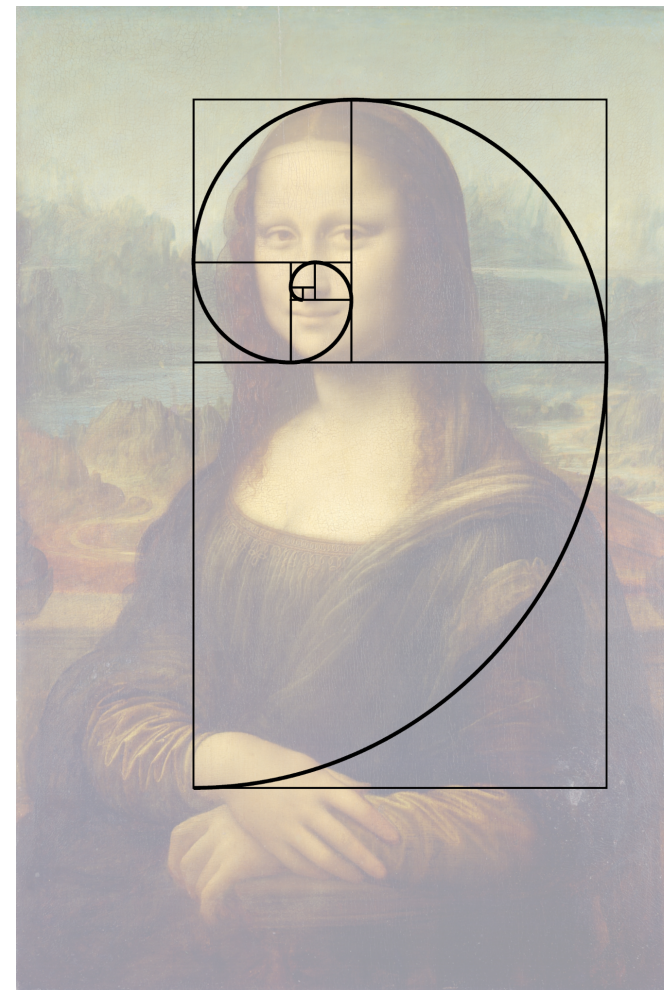
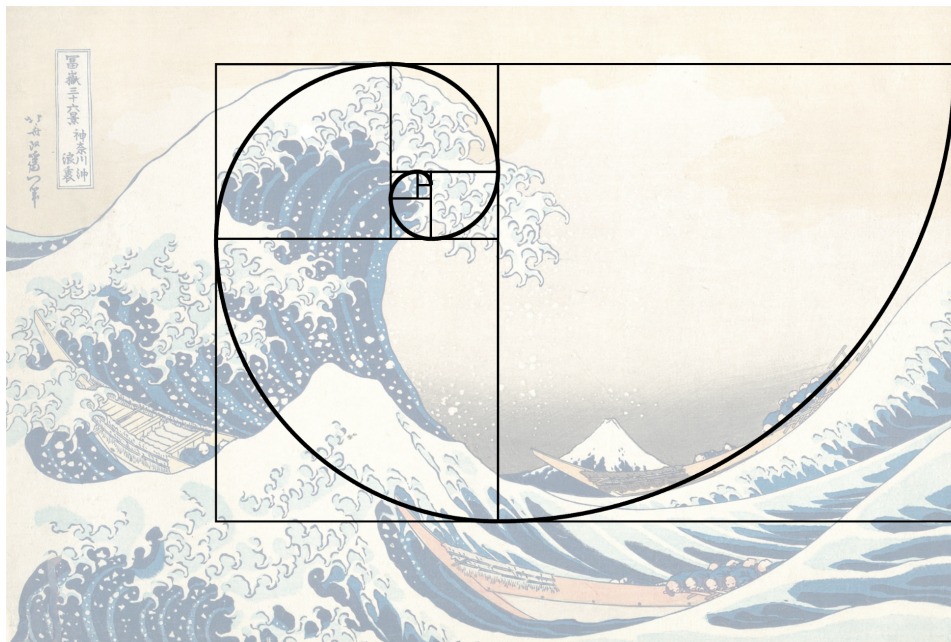
# Cross-lingual transfer of language models is unpredictable



Subword evenness (SuE) as a predictor of cross-lingual transfer to low-resource languages  
Olga Pelloni, Anastassia Shaitarova, Tanja Samardžić  
EMNLP 2022



# Subword geometry as a predictor of cross-lingual transfer



Subword evenness (SuE) as a predictor of cross-lingual transfer to low-resource languages  
Olga Pelloni, Anastassia Shaitarova, Tanja Samardžić  
EMNLP 2022



# Subwords as lines

## BPE-MR tokenisation

We stop BPE at minimum redundancy and look at what we get as geometric patterns

## Evenness

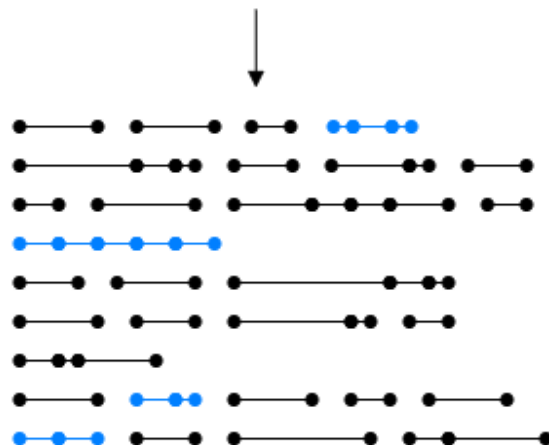
Some patterns are more even than others

Subword evenness (SuE) as a predictor of cross-lingual transfer to low-resource languages

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EMNLP 2022

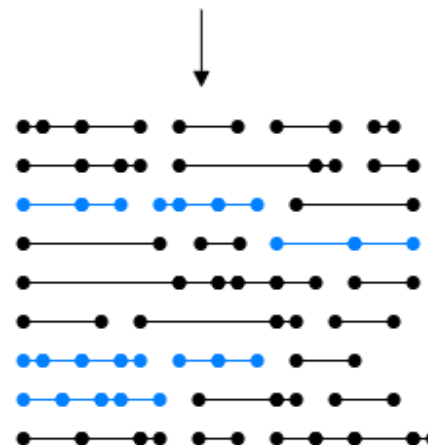
### English

Dogs come in m-an-y  
variet-ie-s and ther-e can  
be great diff-er-en-ces in  
ap-pe-ar-an-ce  
and even temperam-en-t  
from one variet-y to  
an-o-ther.  
They ar-e kept as both  
pe-ts and working an-im-als.



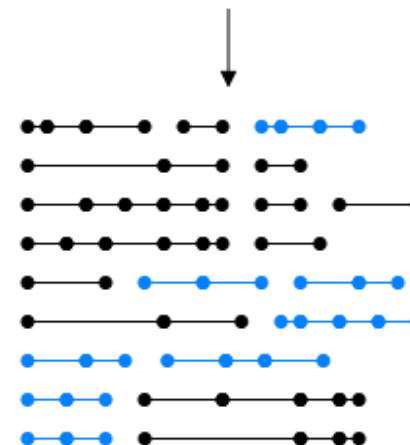
### Icelandic

H-un-dar eru til í  
fjö-ld-a afbrigð-a og  
get-ur v-er-ið mikill  
útlits- og jafn-vel  
skapgerð-ar-m-un-ur frá  
einu afbrigð-i til  
a-nn-ar-s. Þe-ir eru  
ha-ld-n-ir jafn-t sem  
gæl-udý-r og vi-nn-udý-r.



### Finnish

K-oi-ria on m-on-ia  
lajikke-ita, ja  
ulk-on-äö-ss-ä ja jopa  
lu-on-tee-ss-a voi  
olla suu-ria ero-ja  
lajikke-esta t-oi-se-en.  
Nii-tä pid-et-ään  
se-kä lemm-ikke-in-ä  
et-tä työeläim-in-ä.



# Predicting the perplexity of a transferred language model

## Languages

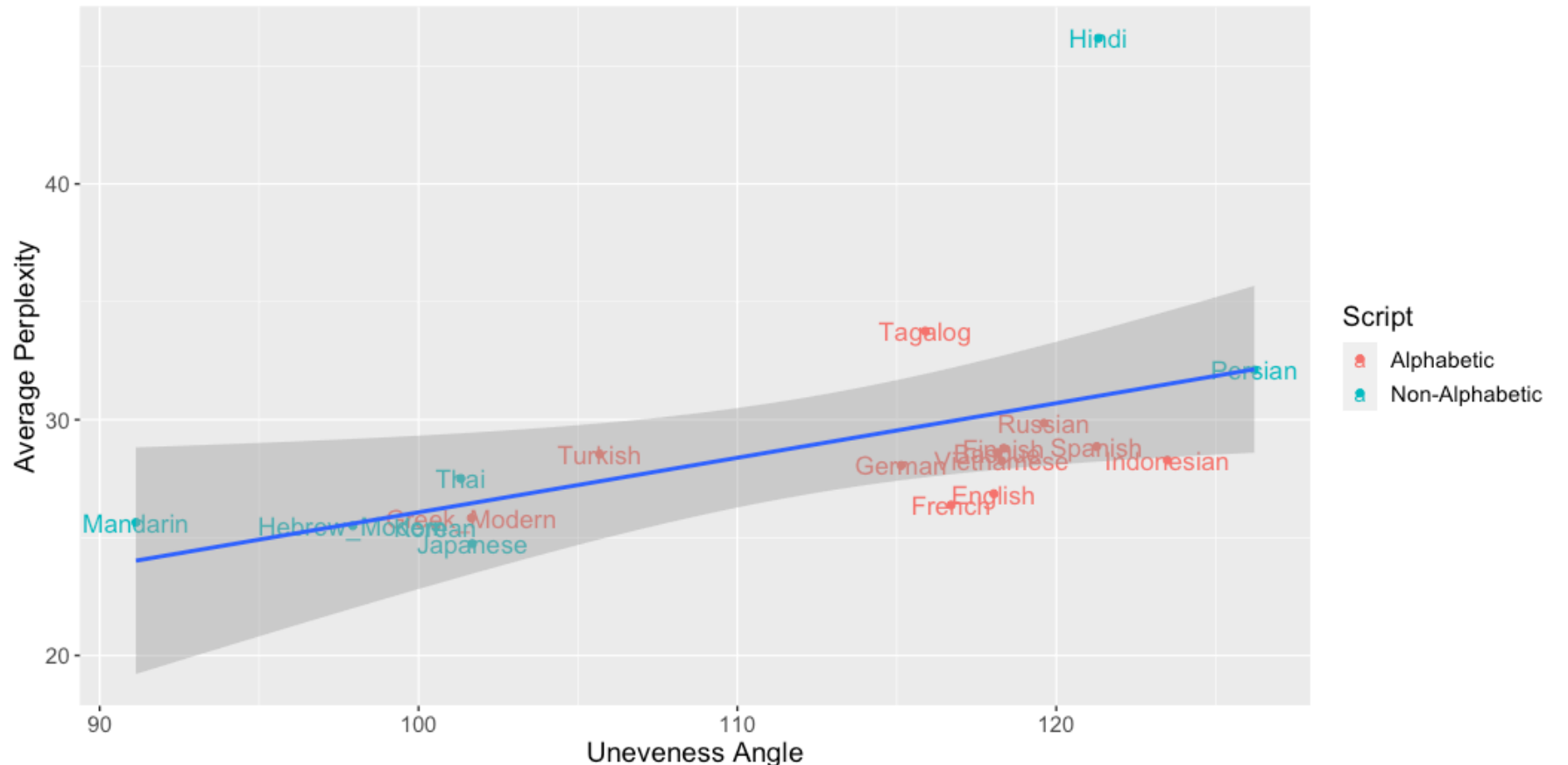
49 languages from the TeDDi sample

TeDDi Sample: Text Data Diversity Sample for Language Comparison and Multilingual NLP

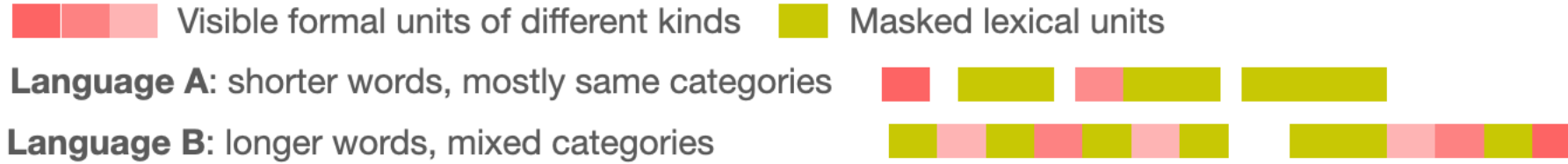
Steven Moran, Christian Bentz, Ximena Gutierrez-Vasques, Olga Pelloni, and Tanja Samardžić  
LREC 2022

Subword evenness (SuE) as a predictor of cross-lingual transfer to low-resource languages

Olga Pelloni, Anastassia Shaitarova, Tanja Samardžić  
EMNLP 2022



# Future work: separate form and content in any language



## Represent formal units

Can we learn the syntax of formal units by ignoring what is not formal?

## Efficient language embeddings

Can we learn language representations from sequences of formal units?

## General data efficiency

Can we model the meaning of texts better if we know at least something about their structure?

# Conclusion

## **Linguistics currently not part of computational language modelling**

This situation is due to complex history of the study of natural language, not a necessity

## **Linguistics might require computational language modelling**

Given the complexity of natural language, its scientific study might depend on our ability to deal with a lot of data and computational modelling. Computational modelling is a way to make explicit predictions and test them. It can be the way towards reuniting data and theory.

## **Partial overlap between NLP and linguistics**

Not all NLP has to serve linguistics, and not all linguistics has to rely on NLP, but there can and should be an overlap. Example topics are text tokenisation and cross-lingual transfer of language models.

## **SMASH**

An opportunity for a more scientific computational language modelling!