Heidelberg-Boston @ SIGTYP 2024 Shared Task: Enhancing Low-Resource Language Analysis With Character-Aware Hierarchical Transformers

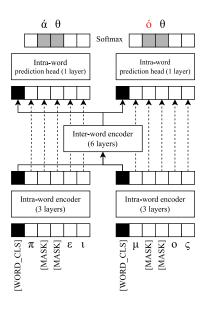
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Tokenization

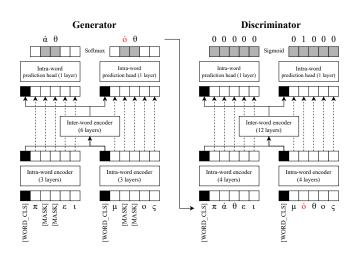
- word-level
 - \times low-resource context \rightarrow rare occurrence of words
 - ✗ inflected languages → unique word forms exceedingly rare
- subword-level
 - X much character information lost
 - X no explicit word representations
- character-level
 - X long sequence lengths
 - no explicit word representations

Hierarchical Language Model¹



¹Sun et al. 2023.

Replaced Token Detection



Masking Strategy

- whole-word masking
- character masking
- character n-gram masking

Lemmatization

- ► sequence-to-sequence task
- ► character-based T5 model

Morphological Tagging

- concatenate intra- and inter-word embeddings
- classifier for each feature

$$\mathcal{L}_{\mathsf{morph}} = rac{1}{k} \sum_{m=0}^{k-1} \mathcal{L}_m$$

PoS Tagging

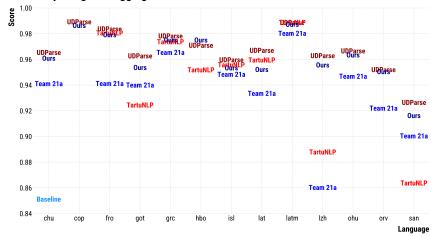
- concatenate intra- and inter-word embeddings
- ► multi-task learning with POS + morph

$$\mathcal{L}_{\mathsf{UPoS}} + \mathcal{L}_{\mathsf{morph}}$$

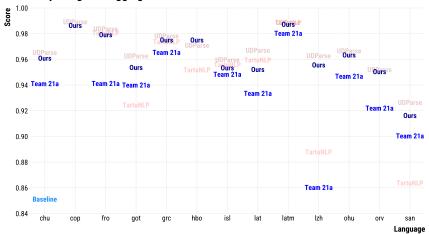
Lemmatization

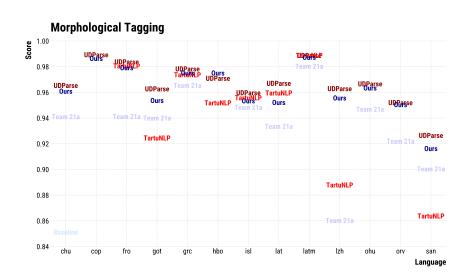
Input	Label
quem PRON	quis
me PRON	ego
arbitramini VERB	arbitror
esse AUX	sum
non ADV	non
sum AUX	sum
ego PRON	ego

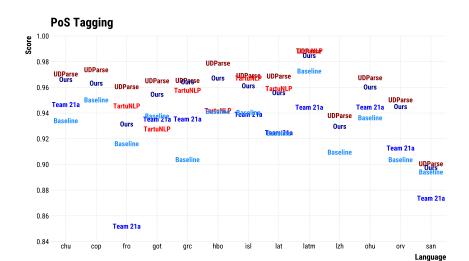


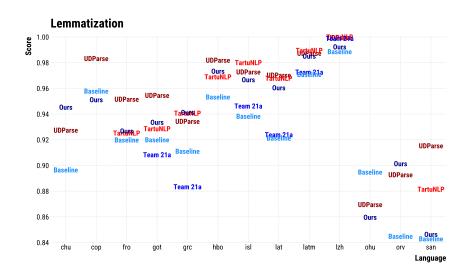






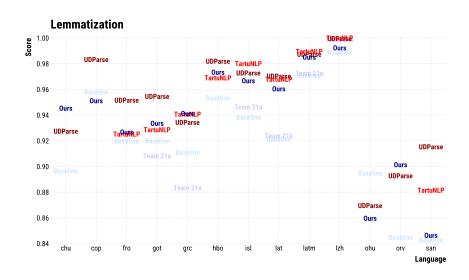












Negative Results

- ► multi-task learning
- ► tall models
 - narrower and deeper architecture
 - benefits for Masked Language Modeling but not for Replaced Token Detection

Take-aways

- ► Hierarchical Language Model to learn efficiently from every character
- ► DeBERTa-V3 with Replaced Token Detection
- ► lemmatization as character-level sequence-to-sequence problem



Thank you for your attention!