Making Sentence Embeddings Robust to User-Generated Content



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PRAIRIE

Takeaways

Observation LASER has poor sentence representations for usergenerated content

Goal Make LASER more robust by reducing the standard-UGC distance in the space

√ is significantly more robust than LASER to user-generated content

RoLASER:

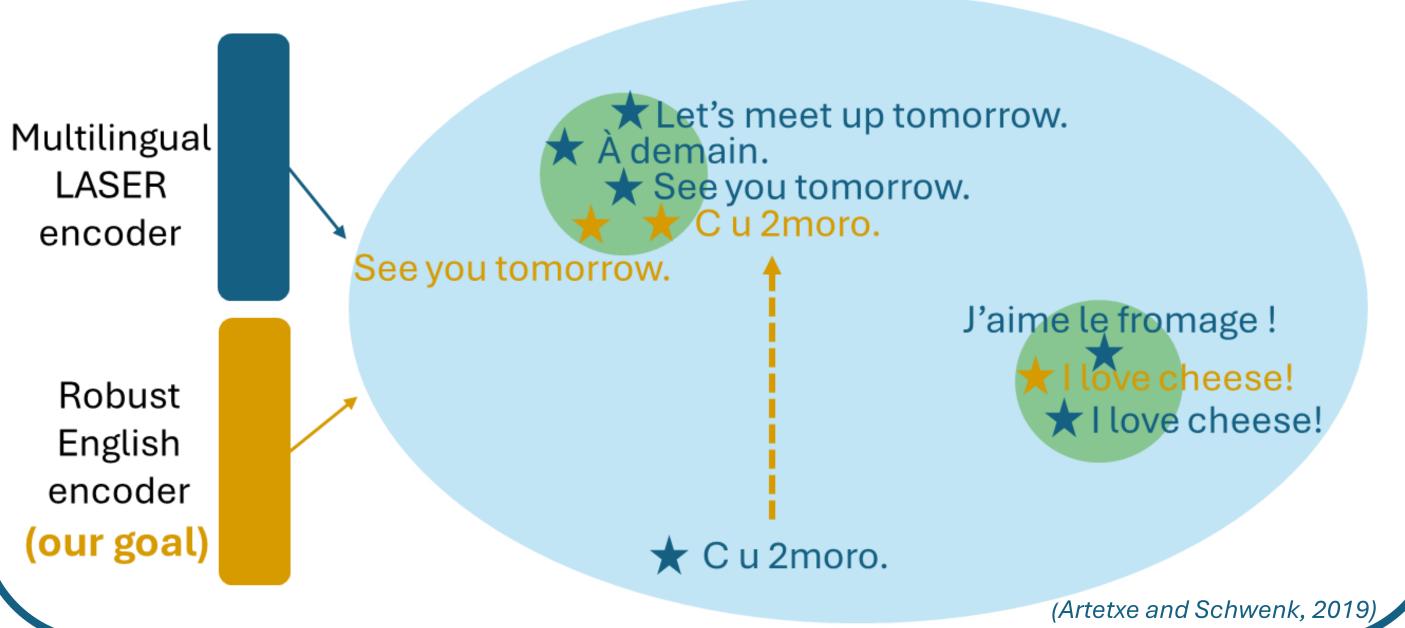
✓ matches/improves LASER's performance on standard text

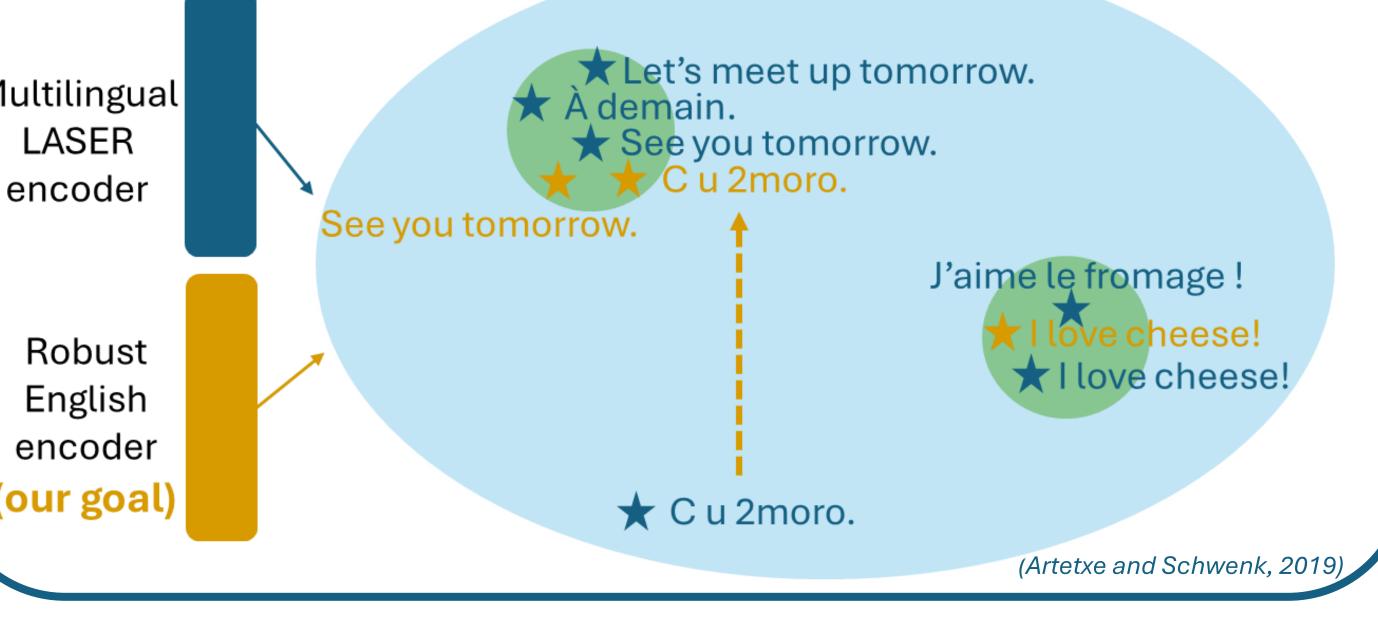
User-Generated Content (UGC)



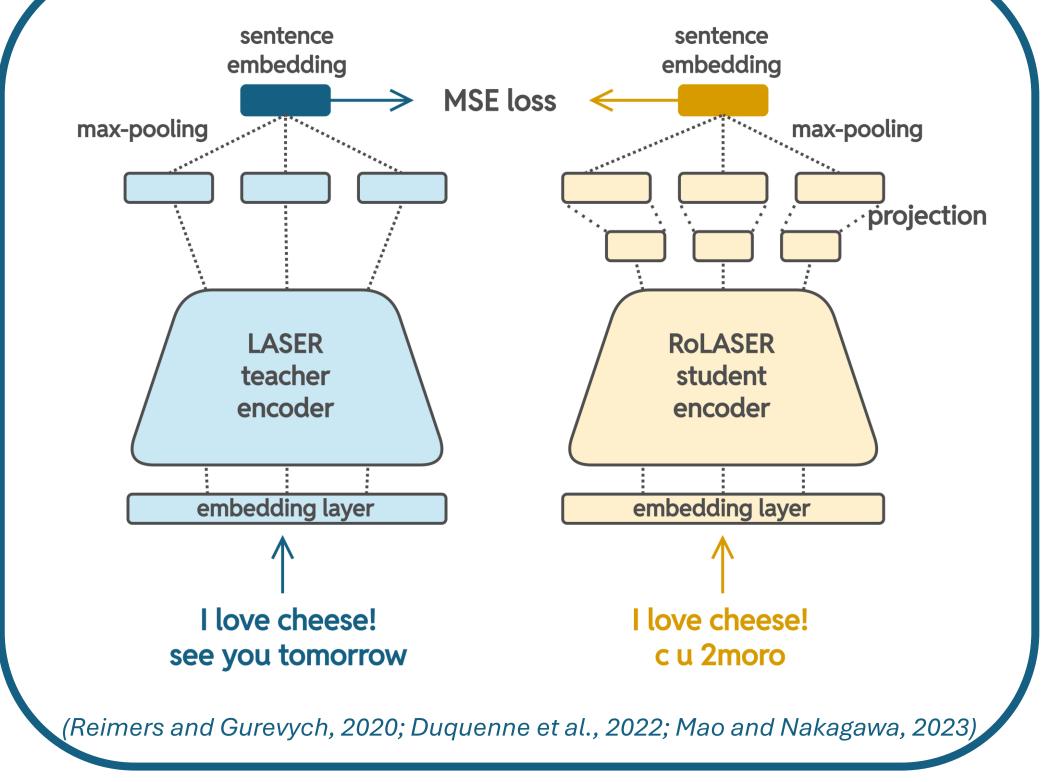
(Seddah et al., 2012; Zalmout et al., 2019; Sanguinetti et al., 2020)

LASER Sentence Embedding Space





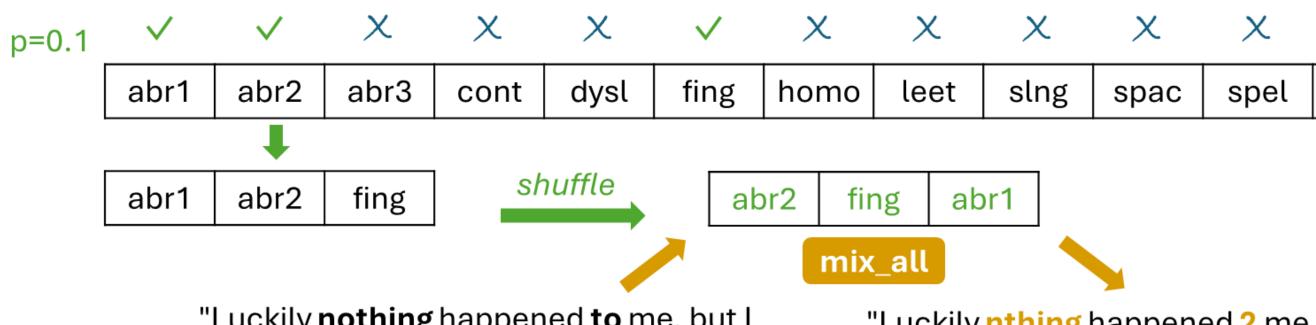
Teacher-Student Approach



Full paper here

week

Generating Artificial UGC Training Data



"Luckily **nothing** happened **to** me, but I saw a macabre scene, as people tried to break windows in order to get out."

"Luckily nthing happened 2 me, but I saw a macabre scene, as ppl triwd 2 break windows in order 2 gt out."

NL-Augmenter (Dhole et al., 2021)

Experimental Setup

Models

LASER (teacher)

- 5-layer bi-LSTM (45M params)
- fixed during training
- RoLASER [Robust LASER] (student)
- 12-layer Transformer (108M params)
- initialised with RoBERTa (Liu et al., 2019)

c-RoLASER (student)

- 12-layer Transformer (104M params)
- initialised with CharacterBERT (El Boukkouri et al., 2020)

Metrics

Average pairwise cosine distance

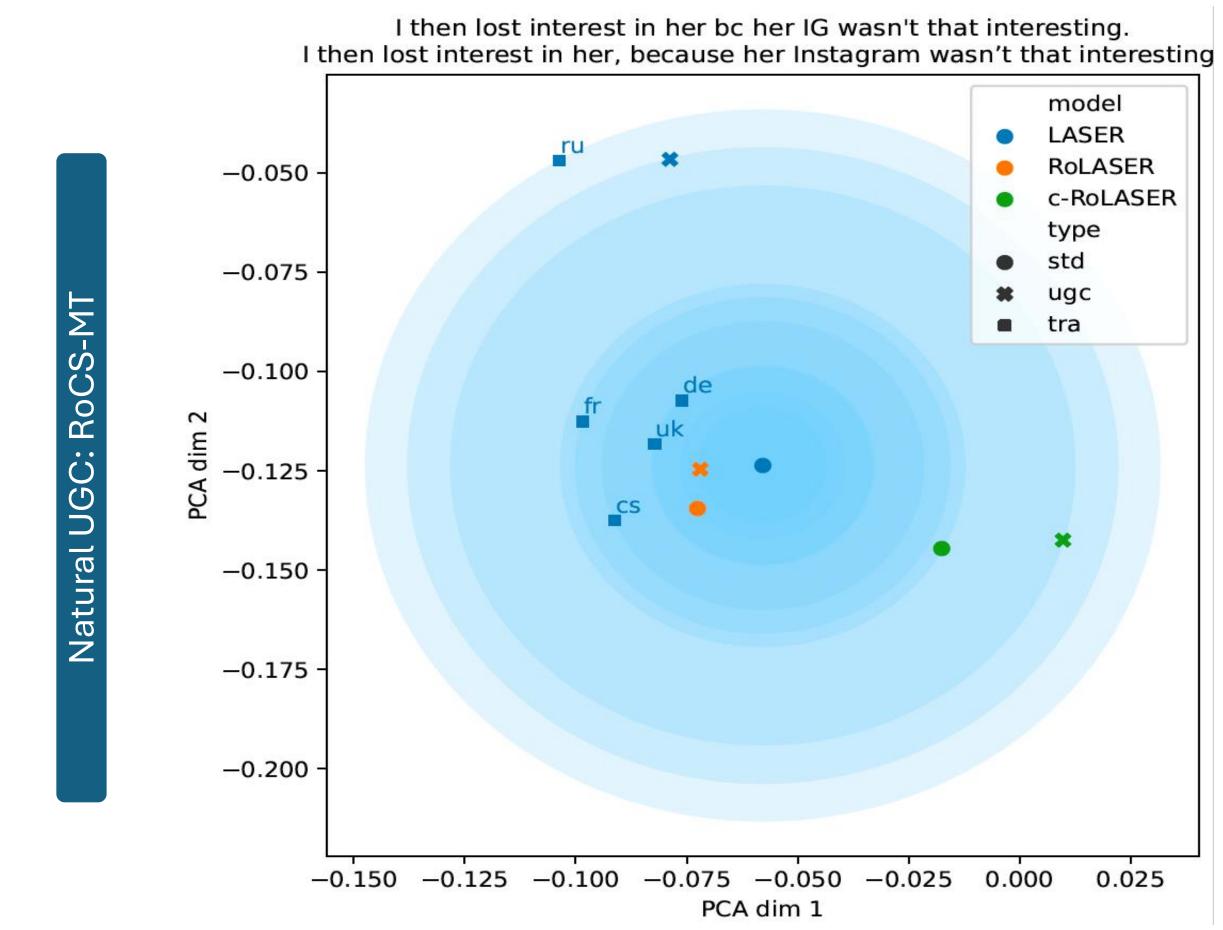
- •xSIM (Artetxe and Schwenk, 2019)
 - proxy metric for bitext mining
 - cross-lingual similarity search error rate of aligning translation pairs
- •xSIM++ (Chen et al., 2023)
- more challenging than xSIM

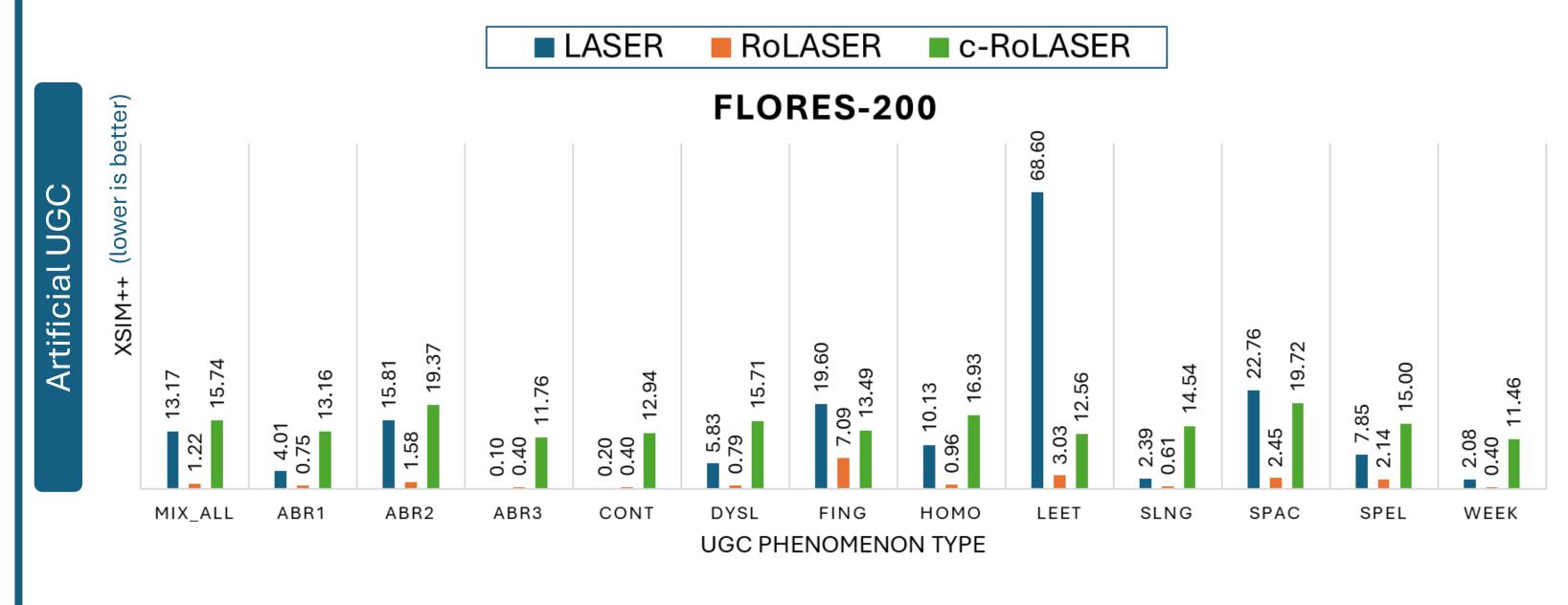
Data

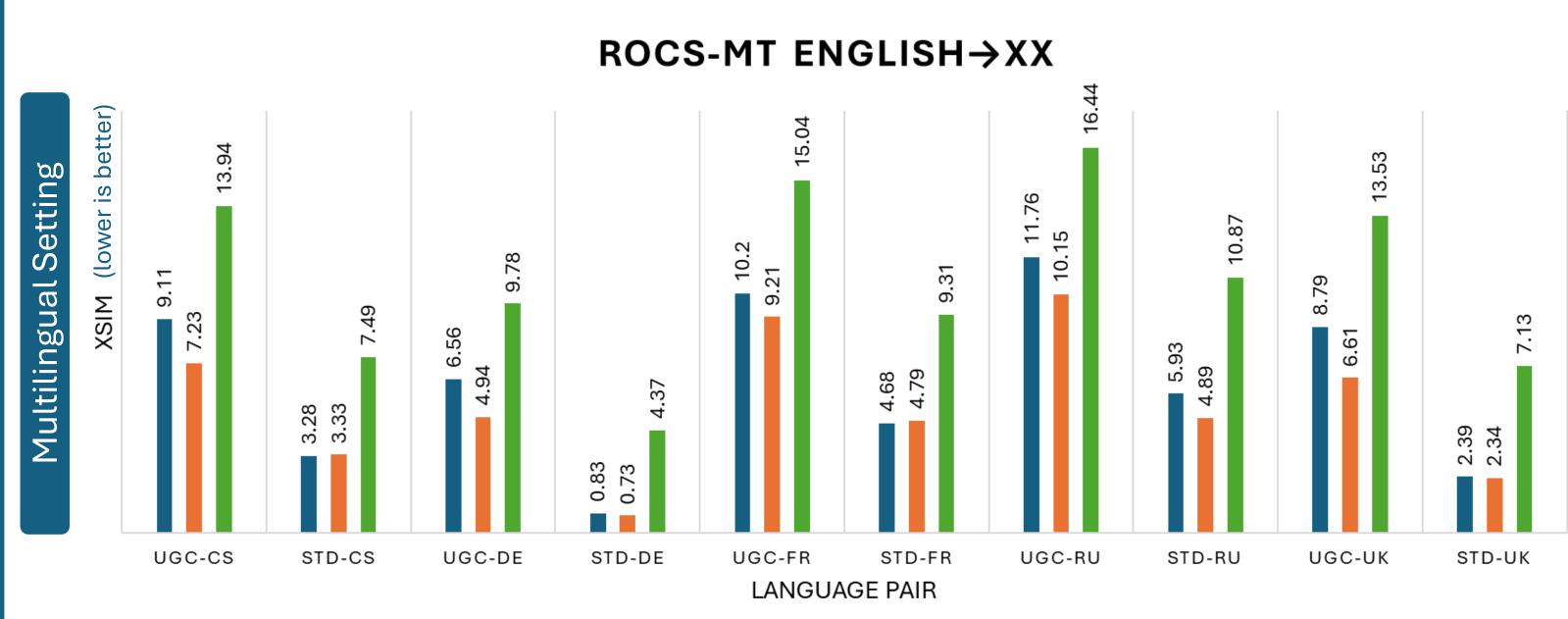
•OSCAR (Ortiz Suárez et al., 2019)

- 2M lines of standard English used for training
- artificially augmented with the mix_all transformation
- RoCS-MT (Bawden and Sagot, 2023)
 - 1922 standard ↔ UGC English sentences from Reddit
 - translations into 5 other languages
- •FLORES-200 (NLLB Team et al., 2022)
 - 1012 standard English sentences from WikiNews, WikiBooks, WikiVoyage
 - artificially augmented with mix_all and other NL-Augmenter transformations

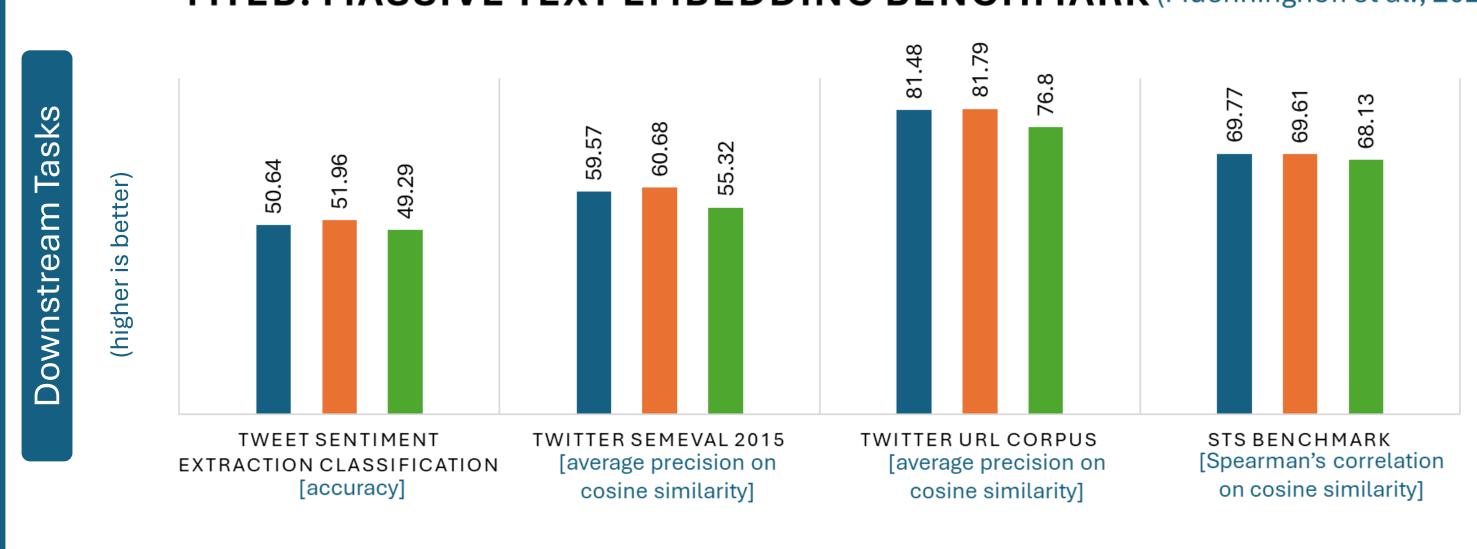
Results and Analysis







MTEB: MASSIVE TEXT EMBEDDING BENCHMARK (Muenninghoff et al., 2023)



RoLASER outperforms LASER on UGC

c-RoLASER struggles to map its standard embeddings to LASER's The most challenging UGC phenomena **shatter** subword tokenisation

Future work: Extend RoLASER to more languages