

On Cross-Language Entity Label Projection and Recognition

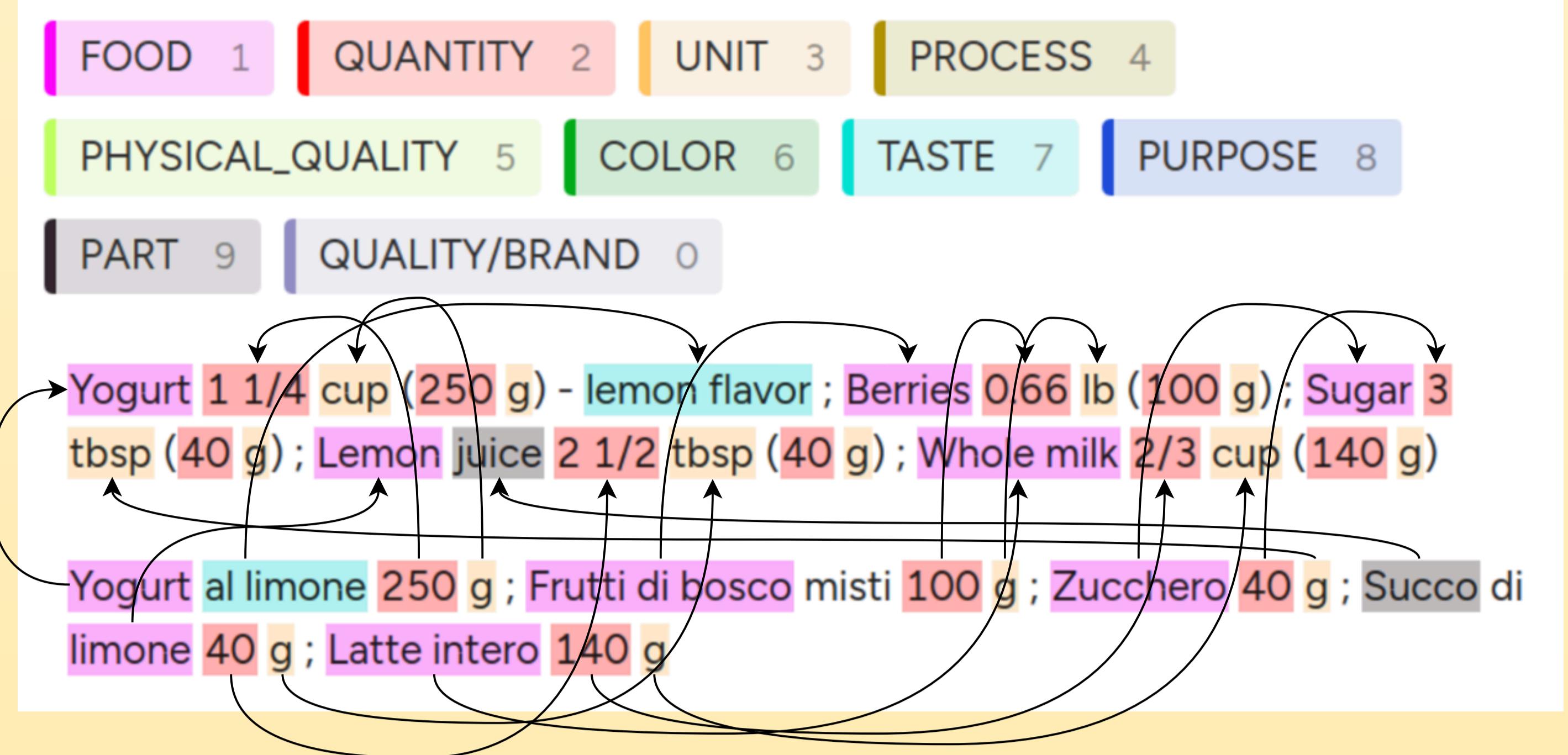
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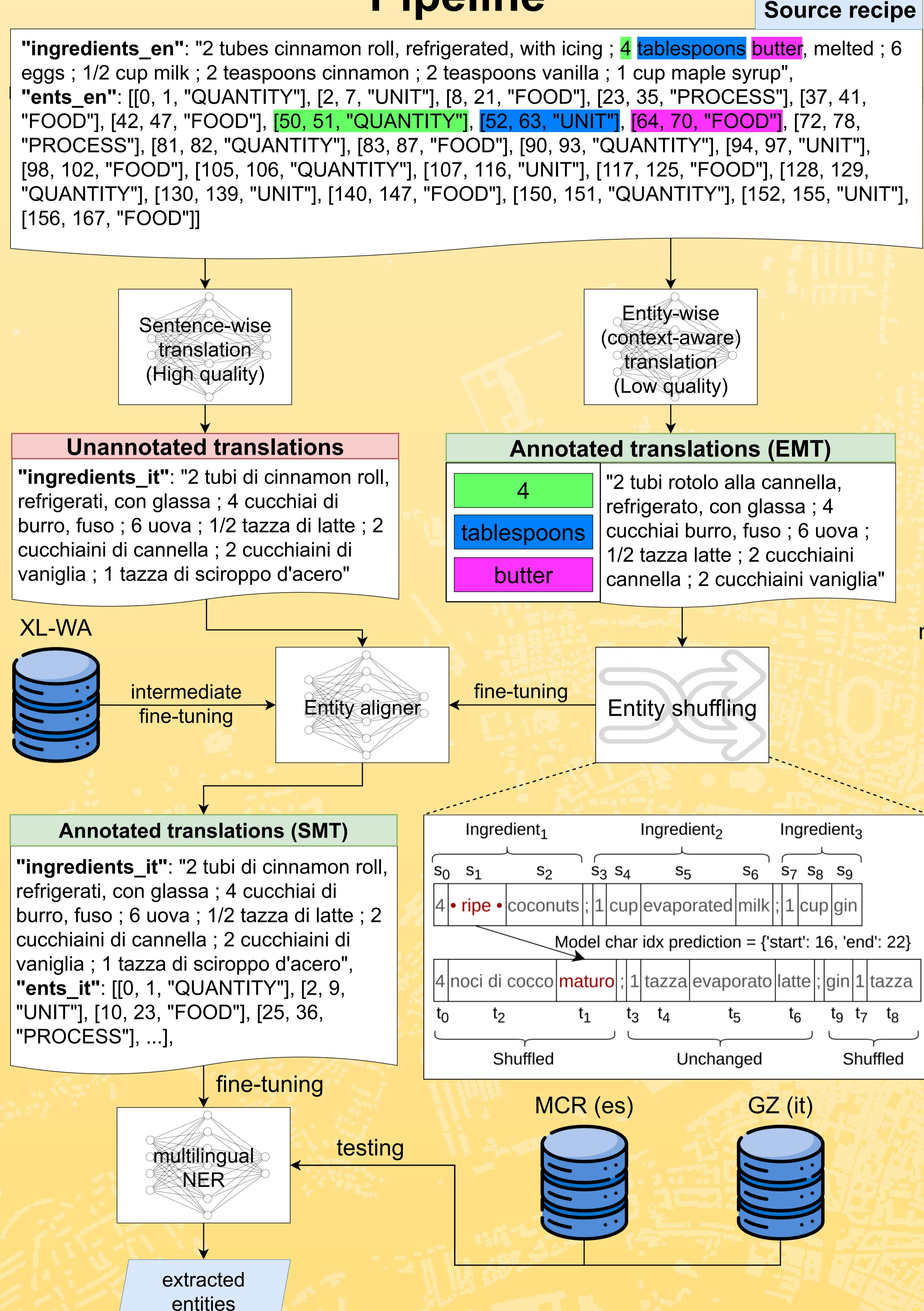
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Corpus	# Docs	Entities	Alignments
		en it es	
TASTEset (EN)	EMT	700	13,362
Wróblewska et al. (2022)	SMT	13,362	13,339 13,356
GialloZafferano (IT > EN)		597	26,631 20,272 /
Ours			9,842
My Colombian Recipes (EN > ES)		300	11,551 / 700
Ours			3,565
XL-WA (EN > IT / ES)		1,105	/
Martelli et al. (2023)		1,107	22,486 18,700

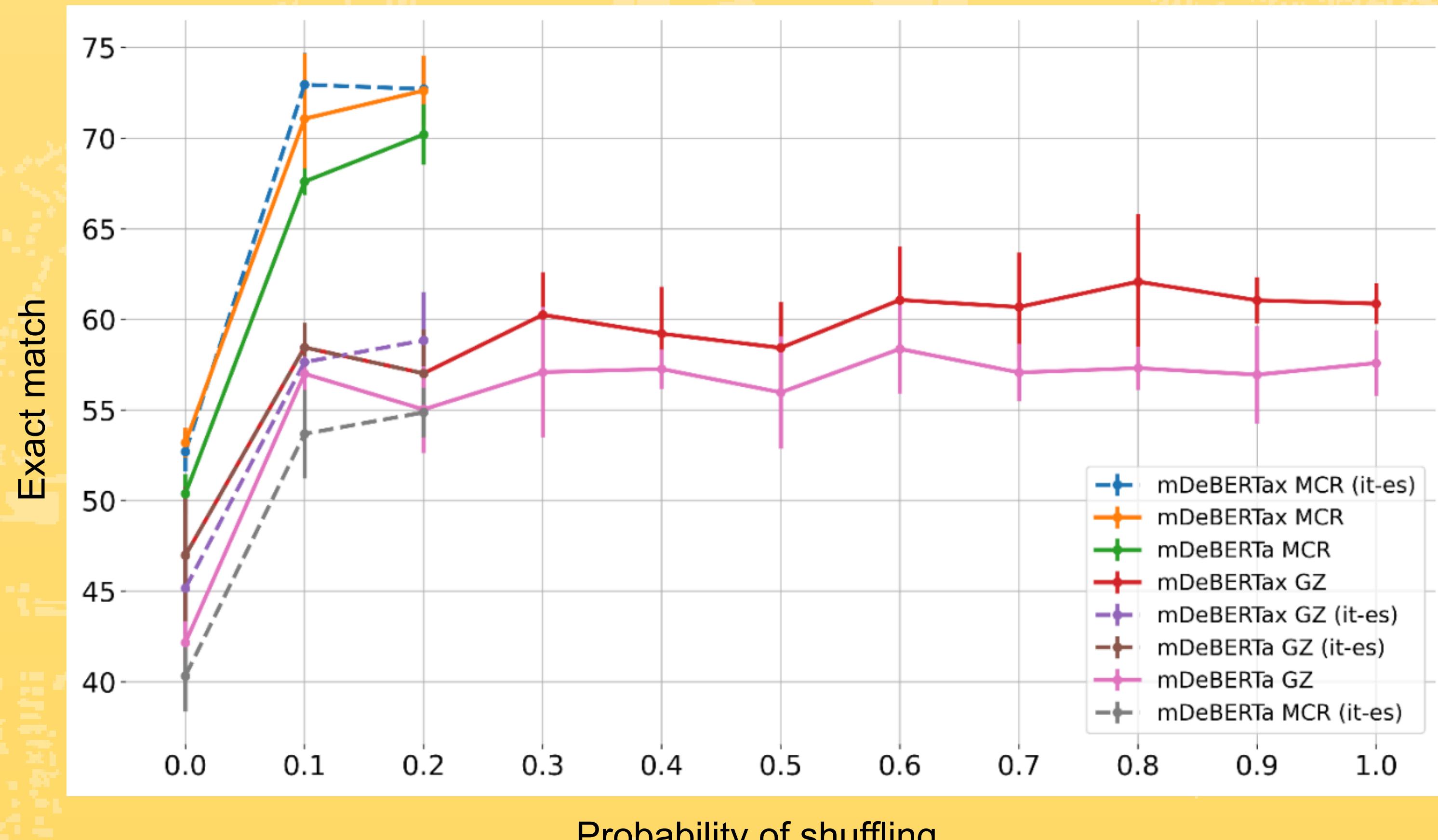
Data annotation (Label Studio)



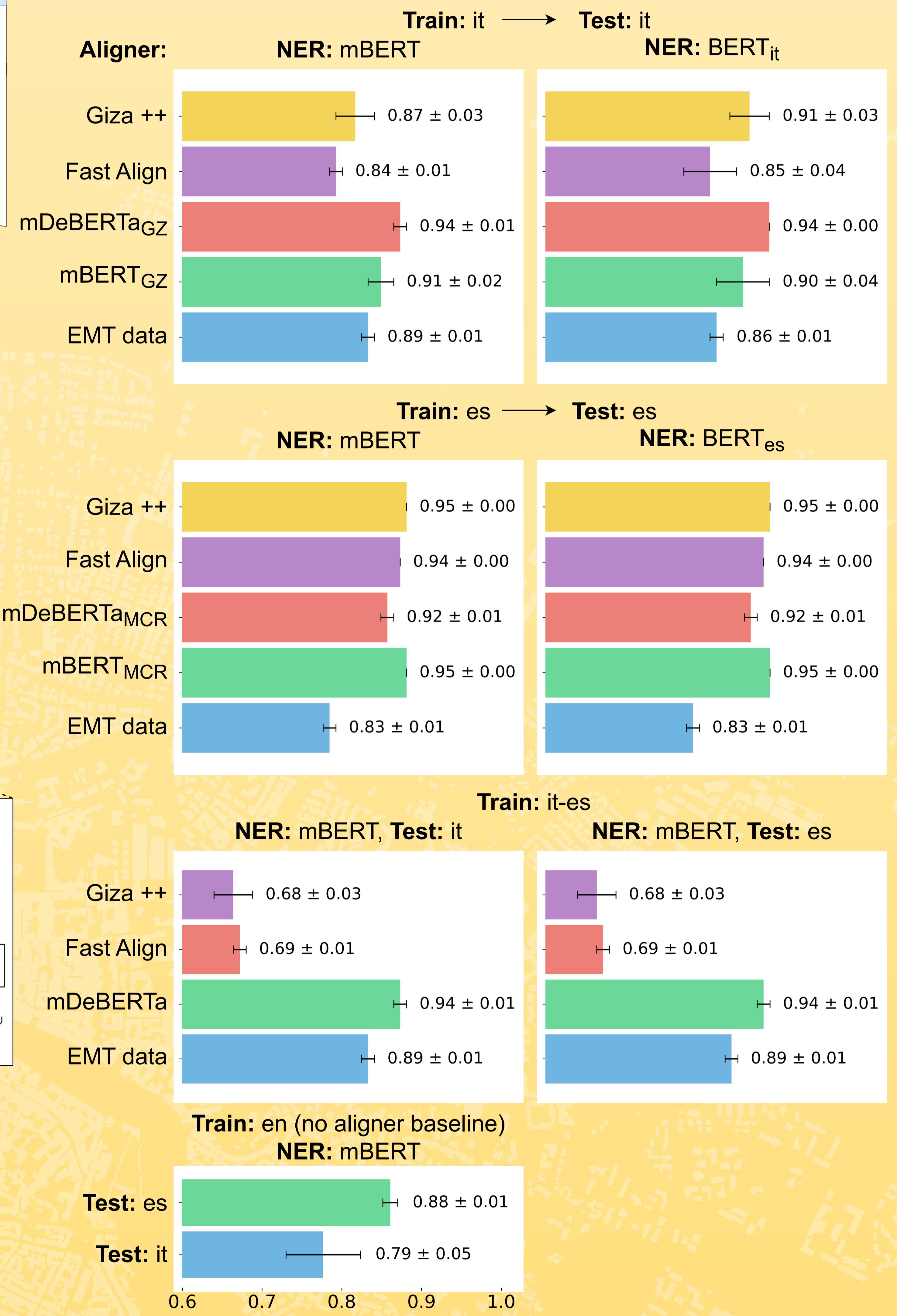
Pipeline



Alignment Results



NER Results



Takeaways

- Shuffling entities even with a small probability provides a large boost in alignment performance, as the model learns not to rely on the original order of the entities.
- Generating high quality synthetic data through good entity projection models leads to better NER models, compared to simply translating entities in place. Conversely, bad alignment models lead to NER models which are worse than the entity-wise translation baseline.
- A single multilingual BERT NER model can perform as well as multiple monolingual counterparts, which means less training and inference costs.

