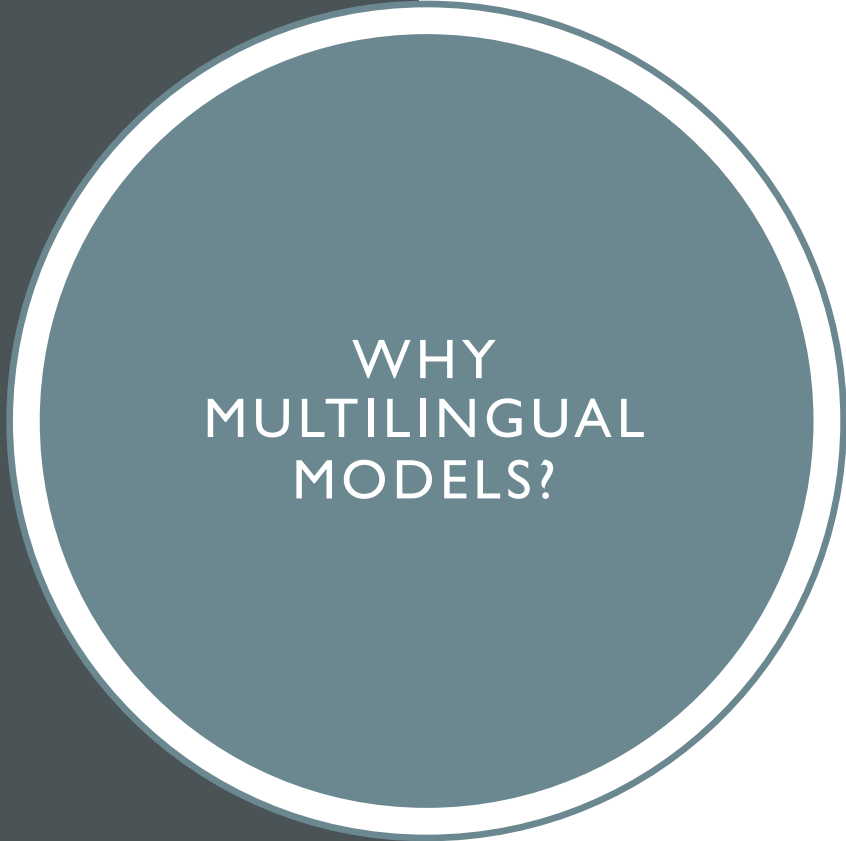


Samira Malek

Spring 2023

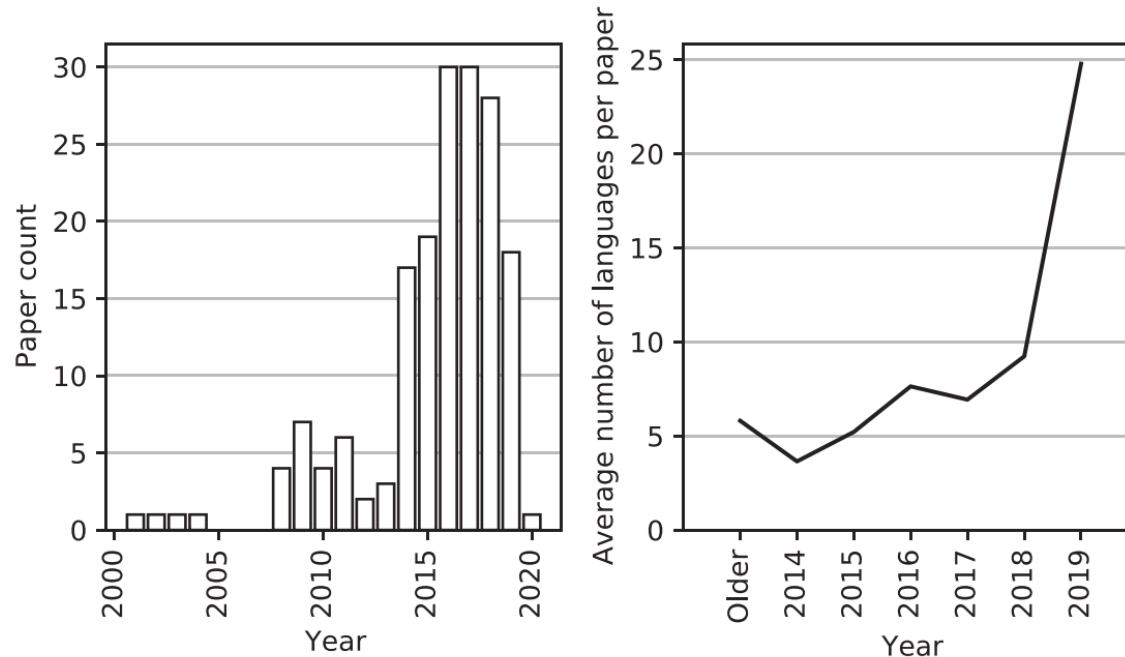
CROSS-LINGUAL LANGUAGE MODEL PRETRAINING



WHY MULTILINGUAL MODELS?

- English is the most researched language and it is the single language considered in more than 60% of the papers published.
- Why?
 - Business
 - Help (2010 Haitian earthquake)
 - Fairness

PROGRESS OF CROSS-LINGUAL LEARNING

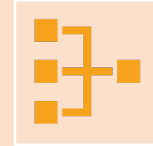


- Fig. Left: Number of surveyed papers per year. Right: Average number of languages used in papers per year¹.

CROSS-LINGUAL LANGUAGE MODELS (XLMS)

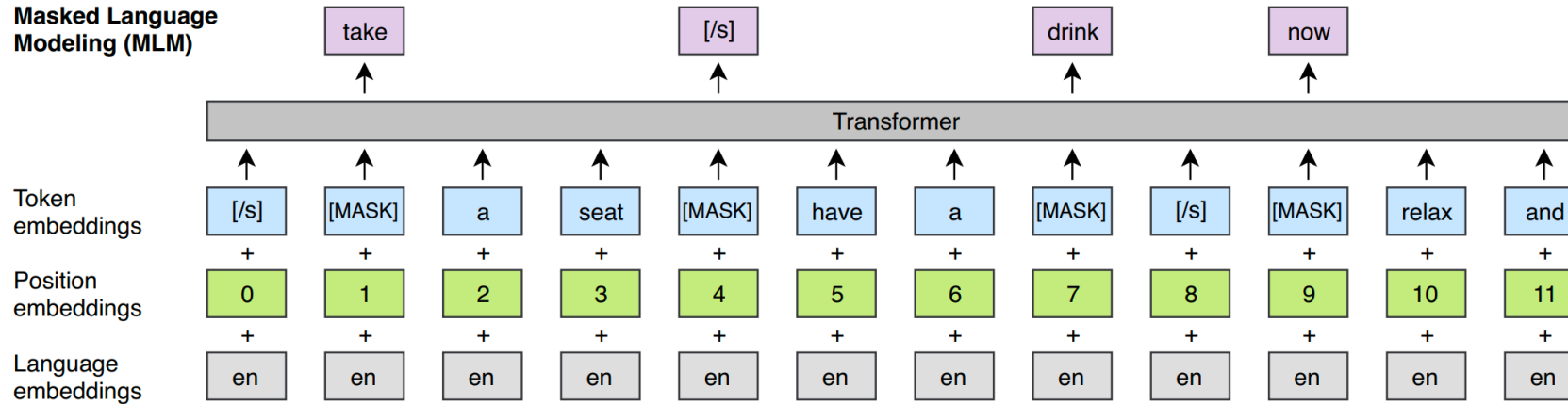


1. Causal Language Modeling (CLM)
 2. Masked Language Modeling (MLM)
- Unsupervised + Monolingual data

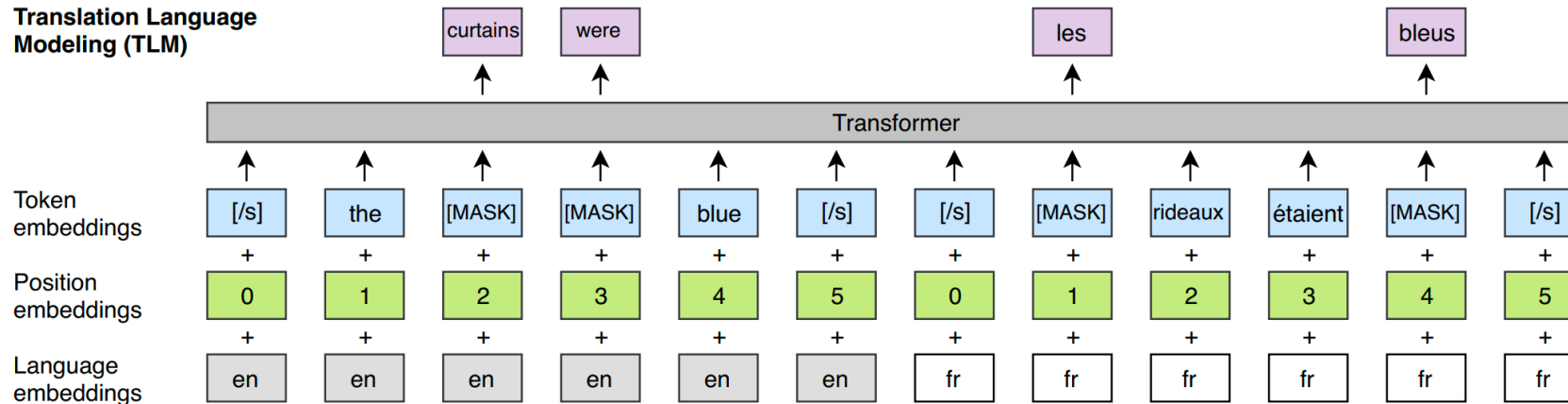


1. Translation Language Modeling (TLM)
- Supervised + parallel data

Masked Language Modeling (MLM)



Translation Language Modeling (TLM)



RESULTS ON CROSS-LINGUAL CLASSIFICATION ACCURACY

	en	fr	es	de	el	bg	ru	tr	ar	vi	th	zh	hi	sw	ur	Δ
<i>Machine translation baselines (TRANSLATE-TRAIN)</i>																
Devlin et al. (2018)	81.9	-	77.8	75.9	-	-	-	-	70.7	-	-	76.6	-	-	61.6	-
XLM (MLM+TLM)	<u>85.0</u>	<u>80.2</u>	<u>80.8</u>	<u>80.3</u>	<u>78.1</u>	<u>79.3</u>	<u>78.1</u>	<u>74.7</u>	<u>76.5</u>	<u>76.6</u>	<u>75.5</u>	<u>78.6</u>	<u>72.3</u>	<u>70.9</u>	63.2	<u>76.7</u>
<i>Machine translation baselines (TRANSLATE-TEST)</i>																
Devlin et al. (2018)	81.4	-	74.9	74.4	-	-	-	-	70.4	-	-	70.1	-	-	62.1	-
XLM (MLM+TLM)	<u>85.0</u>	79.0	79.5	78.1	77.8	77.6	75.5	73.7	73.7	70.8	70.4	73.6	69.0	64.7	65.1	74.2
<i>Evaluation of cross-lingual sentence encoders</i>																
Conneau et al. (2018b)	73.7	67.7	68.7	67.7	68.9	67.9	65.4	64.2	64.8	66.4	64.1	65.8	64.1	55.7	58.4	65.6
Devlin et al. (2018)	81.4	-	74.3	70.5	-	-	-	-	62.1	-	-	63.8	-	-	58.3	-
Artetxe and Schwenk (2018)	73.9	71.9	72.9	72.6	73.1	74.2	71.5	69.7	71.4	72.0	69.2	71.4	65.5	62.2	61.0	70.2
XLM (MLM)	83.2	76.5	76.3	74.2	73.1	74.0	73.1	67.8	68.5	71.2	69.2	71.9	65.7	64.6	63.4	71.5
XLM (MLM+TLM)	<u>85.0</u>	<u>78.7</u>	<u>78.9</u>	<u>77.8</u>	<u>76.6</u>	<u>77.4</u>	<u>75.3</u>	<u>72.5</u>	<u>73.1</u>	<u>76.1</u>	<u>73.2</u>	<u>76.5</u>	<u>69.6</u>	<u>68.4</u>	<u>67.3</u>	<u>75.1</u>

		en-fr	fr-en	en-de	de-en	en-ro	ro-en
<i>Previous state-of-the-art - Lample et al. (2018b)</i>							
NMT		25.1	24.2	17.2	21.0	21.2	19.4
PBSMT		28.1	27.2	17.8	22.7	21.3	23.0
PBSMT + NMT		27.6	27.7	20.2	25.2	25.1	23.9
<i>Our results for different encoder and decoder initializations</i>							
EMB	EMB	29.4	29.4	21.3	27.3	27.5	26.6
-	-	13.0	15.8	6.7	15.3	18.9	18.3
-	CLM	25.3	26.4	19.2	26.0	25.7	24.6
-	MLM	29.2	29.1	21.6	28.6	28.2	27.3
CLM	-	28.7	28.2	24.4	30.3	29.2	28.0
CLM	CLM	30.4	30.0	22.7	30.5	29.0	27.8
CLM	MLM	32.3	31.6	24.3	32.5	31.6	29.8
MLM	-	31.6	32.1	27.0	33.2	31.8	30.5
MLM	CLM	33.4	32.3	24.9	32.9	31.7	30.4
MLM	MLM	33.4	33.3	26.4	34.3	33.3	31.8

Results on unsupervised MT

Pretraining	-	CLM	MLM
Sennrich et al. (2016)	33.9	-	-
ro \rightarrow en	28.4	31.5	35.3
ro \leftrightarrow en	28.5	31.5	35.6
ro \leftrightarrow en + BT	34.4	37.0	38.5

Results on supervised MT

LOW-
RESOURCE
LANGUAGE
MODEL

Training languages	Nepali perplexity
Nepali	157.2
Nepali + English	140.1
Nepali + Hindi	115.6
Nepali + English + Hindi	109.3

BEYOND
XLM:
XLM-R

Model	D	#M	#lg	en	fr	es	de	el	bg	ru	tr	ar	vi	th	zh	hi	sw	ur	Avg
<i>Fine-tune multilingual model on English training set (Cross-lingual Transfer)</i>																			
Lample and Conneau (2019)	Wiki+MT	N	15	85.0	78.7	78.9	77.8	76.6	77.4	75.3	72.5	73.1	76.1	73.2	76.5	69.6	68.4	67.3	75.1
Huang et al. (2019)	Wiki+MT	N	15	85.1	79.0	79.4	77.8	77.2	77.2	76.3	72.8	73.5	76.4	73.6	76.2	69.4	69.7	66.7	75.4
Devlin et al. (2018)	Wiki	N	102	82.1	73.8	74.3	71.1	66.4	68.9	69.0	61.6	64.9	69.5	55.8	69.3	60.0	50.4	58.0	66.3
Lample and Conneau (2019)	Wiki	N	100	83.7	76.2	76.6	73.7	72.4	73.0	72.1	68.1	68.4	72.0	68.2	71.5	64.5	58.0	62.4	71.3
Lample and Conneau (2019)	Wiki	1	100	83.2	76.7	77.7	74.0	72.7	74.1	72.7	68.7	68.6	72.9	68.9	72.5	65.6	58.2	62.4	70.7
XLM-R_{Base}	CC	1	100	85.8	79.7	80.7	78.7	77.5	79.6	78.1	74.2	73.8	76.5	74.6	76.7	72.4	66.5	68.3	76.2
XLM-R	CC	1	100	89.1	84.1	85.1	83.9	82.9	84.0	81.2	79.6	79.8	80.8	78.1	80.2	76.9	73.9	73.8	80.9
<i>Translate everything to English and use English-only model (TRANSLATE-TEST)</i>																			
BERT-en	Wiki	1	1	88.8	81.4	82.3	80.1	80.3	80.9	76.2	76.0	75.4	72.0	71.9	75.6	70.0	65.8	65.8	76.2
RoBERTa	Wiki+CC	1	1	91.3	82.9	84.3	81.2	81.7	83.1	78.3	76.8	76.6	74.2	74.1	77.5	70.9	66.7	66.8	77.8
<i>Fine-tune multilingual model on each training set (TRANSLATE-TRAIN)</i>																			
Lample and Conneau (2019)	Wiki	N	100	82.9	77.6	77.9	77.9	77.1	75.7	75.5	72.6	71.2	75.8	73.1	76.2	70.4	66.5	62.4	74.2
<i>Fine-tune multilingual model on all training sets (TRANSLATE-TRAIN-ALL)</i>																			
Lample and Conneau (2019) [†]	Wiki+MT	1	15	85.0	80.8	81.3	80.3	79.1	80.9	78.3	75.6	77.6	78.5	76.0	79.5	72.9	72.8	68.5	77.8
Huang et al. (2019)	Wiki+MT	1	15	85.6	81.1	82.3	80.9	79.5	81.4	79.7	76.8	78.2	77.9	77.1	80.5	73.4	73.8	69.6	78.5
Lample and Conneau (2019)	Wiki	1	100	84.5	80.1	81.3	79.3	78.6	79.4	77.5	75.2	75.6	78.3	75.7	78.3	72.1	69.2	67.7	76.9
XLM-R_{Base}	CC	1	100	85.4	81.4	82.2	80.3	80.4	81.3	79.7	78.6	77.3	79.7	77.9	80.2	76.1	73.1	73.0	79.1
XLM-R	CC	1	100	89.1	85.1	86.6	85.7	85.3	85.9	83.5	83.2	83.1	83.7	81.5	83.7	81.6	78.0	78.1	83.6

[2] Conneau, Alexis, et al. "Unsupervised cross-lingual representation learning at scale." *arXiv preprint arXiv:1911.02116* (2019).

THANK YOU



QUESTIONS?