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## An Empirical Study of Multilingual Representations from Language Modeling and Translation

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## Work in progress

- a principled standpoint and train comparable MT and LM systems to contrast their cross-lingual and monolingual downstream performances;
- an empirical study on publicly available pretrained LM and MT systems and study whether continued training on MT helps or hinders the emergence of cross-lingual capabilities.
- Data
  - UNPC (Ziemski et al., 2016)
  - OpenSubtitles (Tiedemann, 2012)
- Languages: Arabic, Chinese, English, French, Russian, and Spanish
- Models
  - Masked Language Modeling (MLM) with the BERT architecture (Devlin et al., 2019);
  - Causal Language Modeling (CLM) with the GPT-2 architecture (Radford et al., 2019);
  - Translation Language Modeling (TLM) with the GPT-2 architecture, where the input is the concatenation of a language pair following a setup similar to Conneau and Lample (2019);
  - Oenoising Sequenece-to-Sequence Langauge Modeling with BART architecture (Lewis et al., 2020);
  - Machine Translation (MT) with the classic encoder-decoder transformer architecture (Vaswani et al., 2017) and the BART architecture (Lewis et al., 2020).



## Preliminary Results

	Model	NC	XNLI	PAWS-X	Tas QAM		WPR	NER	POS
LM	mBERT XLM-R mBART	82.1	65.2 <b>73.5</b> 67.6	86.6 88.9 <b>89.2</b>	64.6 67.4 <b>67.8</b>	63.1 <b>66.9</b> 65.5	<b>75.3</b>	77.5 <b>78.7</b> 77.7	79.7
МТ	NLLB 600M	76.0	68.3	73.4	61.5	63.9	73.7	54.2	71.4
СР	mBART m2o mBART o2m mBART m2m	65.4	48.1	85.6 81.7 87.2	63.9 58.4 63.2	63.9 62.7 62.8	73.2	61.5 55.1 71.9	55.7

Table: Average performance on cross-lingual tasks. We use the base architecture for mBERT and XLM-R. mBART scores are derived from the 12-layer encoder.

