## Contextual Token Representations

ULMfit, OpenAl GPT, ELMo, BERT, XLM

Noe Casas



#### Background: Language Modeling

- Data: Monolingual Corpus
- Task: predict next token given previous tokens (causal):

 $P(T_i \mid T_1 \dots T_{i-1})$ 

Usual models: LSTM, Transformer.



#### Contextual embeddings: intuition

- Same word can have different meaning depending on the context. Example:
  - Please, **type** everything in lowercase.
  - What **type** of flowers do you like most?
- Classic word embeddings offer the same vector representation regardless of the context.
- Solution: create word representations that depend on the context.

#### Articles

	Model Alias	Org.	Article Reference	
	ULMfit	fast.ai	Universal Language Model Fine-tuning for Text Classification Howard and Ruder	
	ELMo	AllenNLP	Deep contextualized word representations Peters et al.	
	OpenAl GPT	OpenAl	Improving Language Understanding by Generative Pre-Training Radford et al.	
	BERT	Google	BERT: Pre-training of Deep Bidirectional Transformers for Language Understanding Devlin et al.	
	XLM	Facebook	Cross-lingual Language Model Pretraining Lample and Conneau	

#### Overview

- Train model in one of multiple tasks that lead to word representations.
- Release pre-trained models.
- Use pre-trained models, options:
  - A. Fine-tune model on final task.
  - B. Directly encode token representations with model.

## Overview (graphical)



#### Differences

Alias	Alias Model		Tasks	Language
ULMfit	LSTM	word	Causal LM	English
ELMo	LSTM	word	Bidirectional LM	English
OpenAl GPT	Transformer	subword	Causal LM + Classification	English
BERT	Transformer	subword	Masked LM + Next sentence prediction	Multilingual
XLM	Transformer	subword	Causal LM +Masked LM + Translation LM	Multilingual

#### ULMFiT

- Task: causal LM
- Model: 3-layer LSTM
- Tokens: words



#### ELMO





## OpenAl GPT

- Task: causal LM
- Model: self-attention layers
- Tokens: subwords



#### **エエイ**



• **Tasks**: masked LM + next sentence prediction

15% of tokens get masked

**Model:** self-attention layers



**Tokens:** subwords

## XLM



- **Tasks**: LM + masked LM + Translation LM
- **Model**: self-attention layers

Masked LM with parallel sentences

• Tokens: subwords

Projection

and softmax

are omitted

\*figure from "Cross-lingual Language Model Pretraining"

#### Downstream Tasks

- Natural Language Inference (NLI) or Cross-lingual NLI.
- Text classification (e.g. sentiment analysis).
- Next sentence prediction.
- Supervised and Unsupervised Neural Machine Translation (NMT).
- Question Answering (QA).
- Named Entity Recognition (NER).

#### Further reading

- "Looking for ELMo's friends: Sentence-Level Pretraining Beyond Language Modeling", Bowman et al., 2018
- "What do you learn from context? Probing for sentence structure in contextualized word representations", Tenney et al., 2018.
- "Assessing BERT's Syntactic Abilities", Goldberg, 2018
- "Learning and Evaluating General Linguistic Intelligence", Yogatama et al., 2019.

# Differences with other representations

Note the differences of contextual token representations with:

- Non-word representations like in (CoVe): Learned in Translation: Contextualized Word Vectors by McCann et al. 2017 [salesforce].
- Fixed-size sentence representations like in Massively Multilingual Sentence Embeddings for Zero-Shot Cross-Lingual Transfer and Beyond by Artetxe and Schewnk, 2018 [facebook].

#### Other resources

- <u>https://nlp.stanford.edu/seminar/details/jdevlin.pdf</u>
- <u>http://jalammar.github.io/illustrated-bert/</u>
- <u>https://medium.com/dissecting-bert/dissecting-bert-part2-335ff2ed9c73</u>
- https://github.com/huggingface/pytorch-pretrained-BERT

## Summary



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#### **Bonus slides**

#### Are these really token representations?

- They are a linear projection away from token space.
- Word-level nearest neighbours in corpus finds same word with same usage.

