Unsupervised Word Translation

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Yes we can!

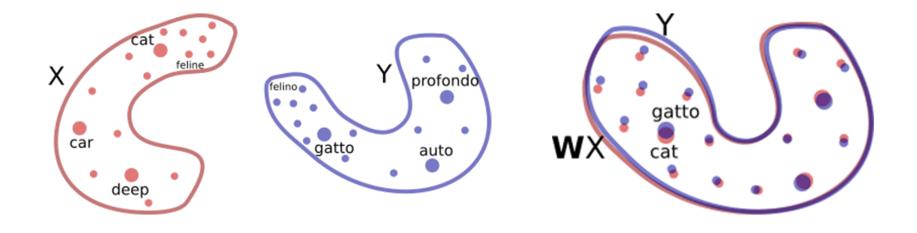
- Near the end of 2017, FAIR (Facebook AI Research) published a model called MUSE (Multilingual UnSupervised word Embeddings)
- MUSE can learn to translate between languages without any cross-lingual information!
- Achieves state of the art accuracy on hundreds of languages, even coming close to or surpassing supervised models!

Word Embeddings

- Word embeddings are models that map every word in a language to a fixed-size vector
- The idea is to map words in such a way that the resulting vector space somehow captures something about the relationships between words
- Most famous example: Word2Vec (Mikolov 2013)
- King Man + Woman = Queen

MUSE

- We start with a fixed set of word embeddings in each language, typically learned from a large corpus of text
- Given target vectors Y and source vectors X, we want to learn a mapping Y = XW between the two spaces
- We want to do this in such a way that the distribution of vectors in each of the two languages is the same



GANs

- MUSE does this by using a GAN (Generative Adversarial Network)
- We train a discriminator to try to tell whether two vectors are from the same language, and a generator to map the vectors from one language into each other
- The discriminator and the generator are adversaries – they each train to try to beat the other

MUSE

- MUSE has been incredibly successful, and set a new standard for word translation
- Many papers have been published following up on MUSE's techniques, but there are still open problems in the area
- One of the most important is to improve the performance on highly dissimilar languages and lowresource languages
- This is an area that could be an excellent opportunity for a research project

European	Languages	
En-Es	81.7	
Es-En	83.3	
En-Fr	82.3	
Fr-En	82.1	
En-De	74.0	
De-En	72.2	
En-Ru	44.0	
Ru-En	59.1	
En-It	76.9	
It-En	76.7	
Non-European Languages		
En-Fa	*	
Fa-En	*	
En-Hi	*	
Hi-En	*	
En-Bn	*	
Bn-En	*	
En-Ar	35.3	
Ar-En	49.7	
En-He	41.6	-
He-En	52.6	
En-Zh	32.5	-
Zh-En	31.4	_
En-Ja	*	-
Ja-En	4.2	_
En-Vi	*	-
Vi-En	*	
En-Id	67.8	-
Id-En	66.6	
427 11 1		

^{*}Failed to converge