Using Paraphrases to Study Properties of Contextual Embeddings

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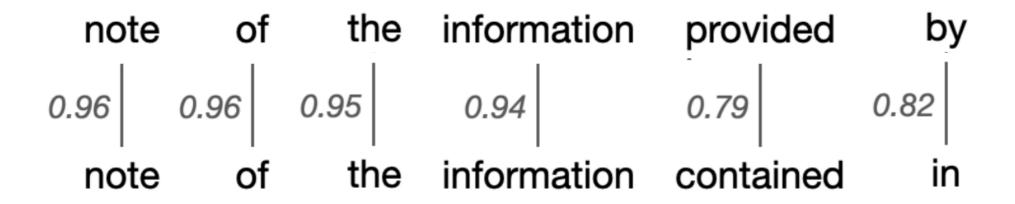


Big Idea

We use *paraphrases* to analyze contextualized embeddings

Why are paraphrases special?

- They naturally encode phrase semantics...
- ...and word semantics



- The Paraphrase Database 2.0 (PPDB)
 - Ganitkevitch et al., 2013; Pavlick et al., 2015
- Word alignment information, automatically generated quality rating, some human quality ratings

Outline

- Phrase-level Embeddings
- Word-level Embeddings
- 3 Punctuation

Phrase-level Embeddings

Can BERT distinguish between two phrases that are paraphrases and two phrases that are unrelated?

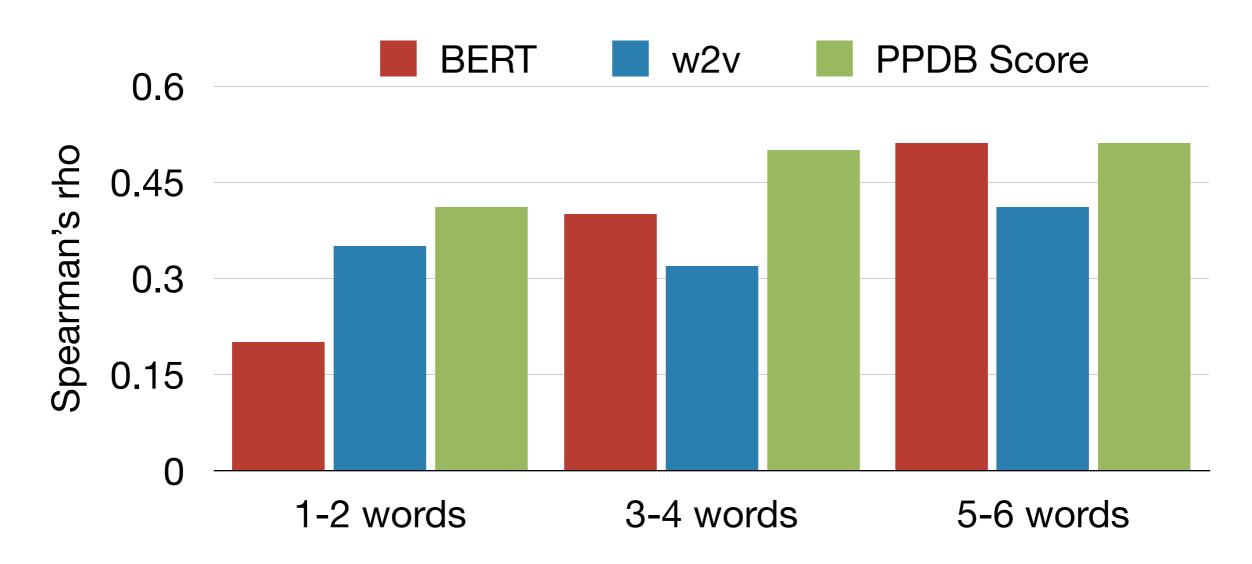
- Use phrase-level embeddings
 - Average together word embeddings to get a phrase embedding
 - Take cosine similarity between two phrase embeddings
 - Compare cosine similarities to human annotations (Spearman's correlation)

Phrase-level Embeddings

Can BERT distinguish between two phrases that are paraphrases and two phrases that are unrelated?

- Experiment details
 - Uncased base model of BERT
 - 25,736 phrase pairs with human annotations
 - Compare BERT with w2v trained on Wikipedia

Phrase-level Embeddings



- BERT does better with longer paraphrases
- With longest paraphrases, BERT is comparable to PPDB score

Outline

1 Phrase-level Embeddings

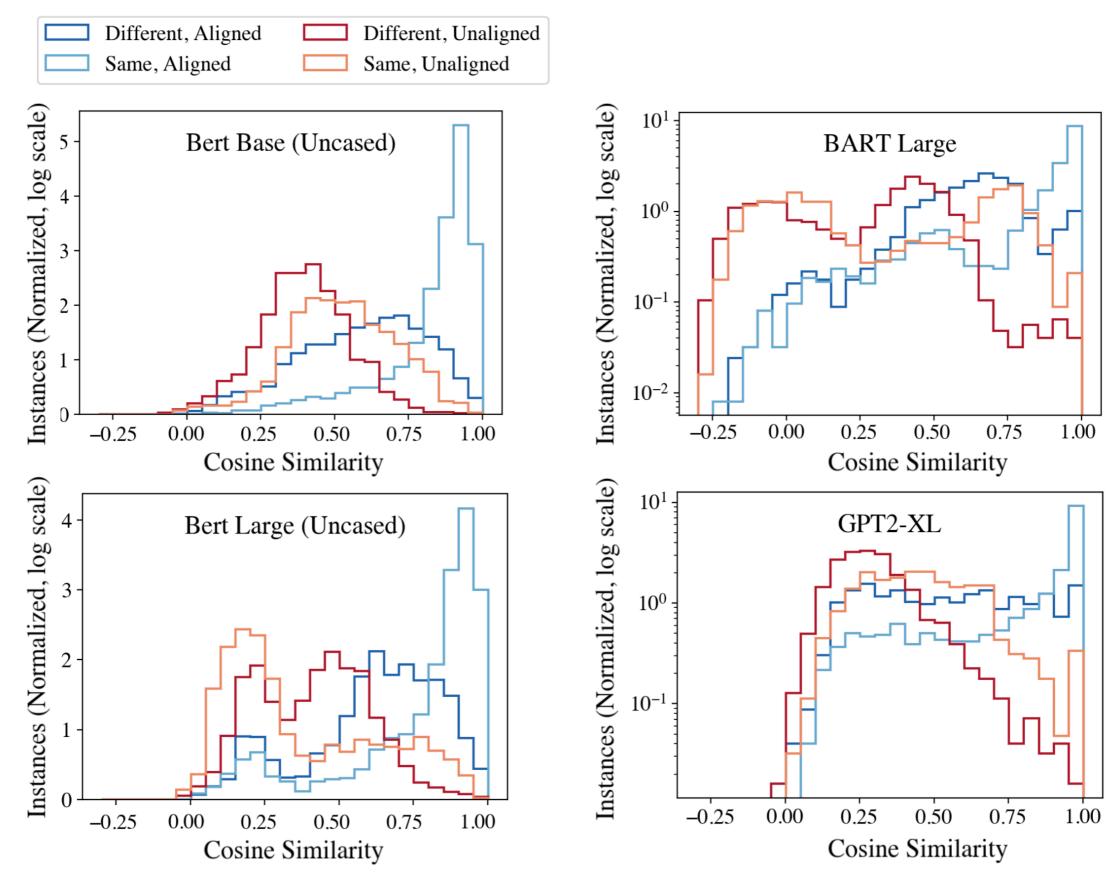
Word-level Embeddings

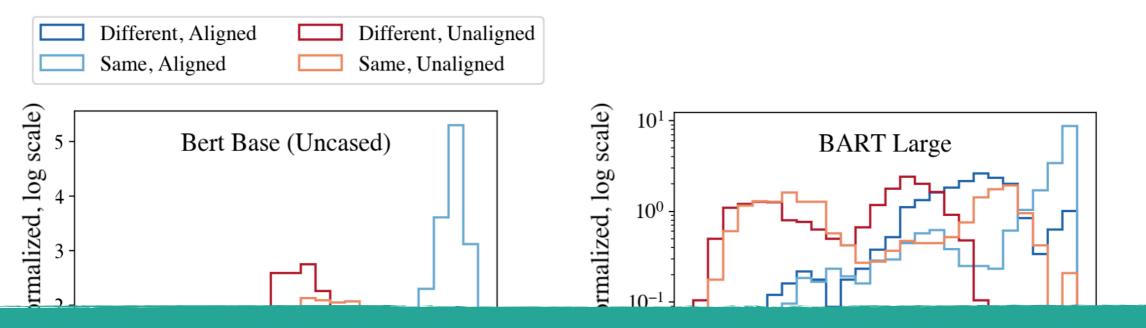
3 Punctuation

	Same	Different
Aligned	adopted by the general assembly at	, with a special focus on
	adopted by the assembly at	, with special emphasis on
Unaligned	okay, so everything 's fine	between the canadian government and
	you guys okay over there	between the government of canada and

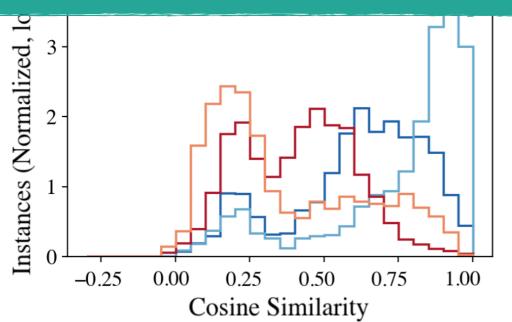
Does BERT recognize that aligned words are more similar than unaligned words?

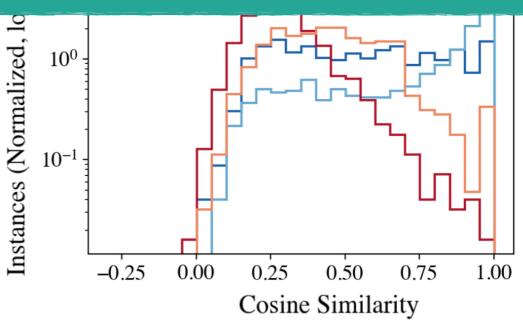
- Experiment details
 - Only use highest quality paraphrases in PPDB
 - Randomly sample 2,500 words from each category
 - For aligned words, only consider 1-1 alignment





Contextual embeddings consistently handle aligned words in paraphrases, but there are substantial variations across models in how peaked the distributions of same-aligned words are.





Outline

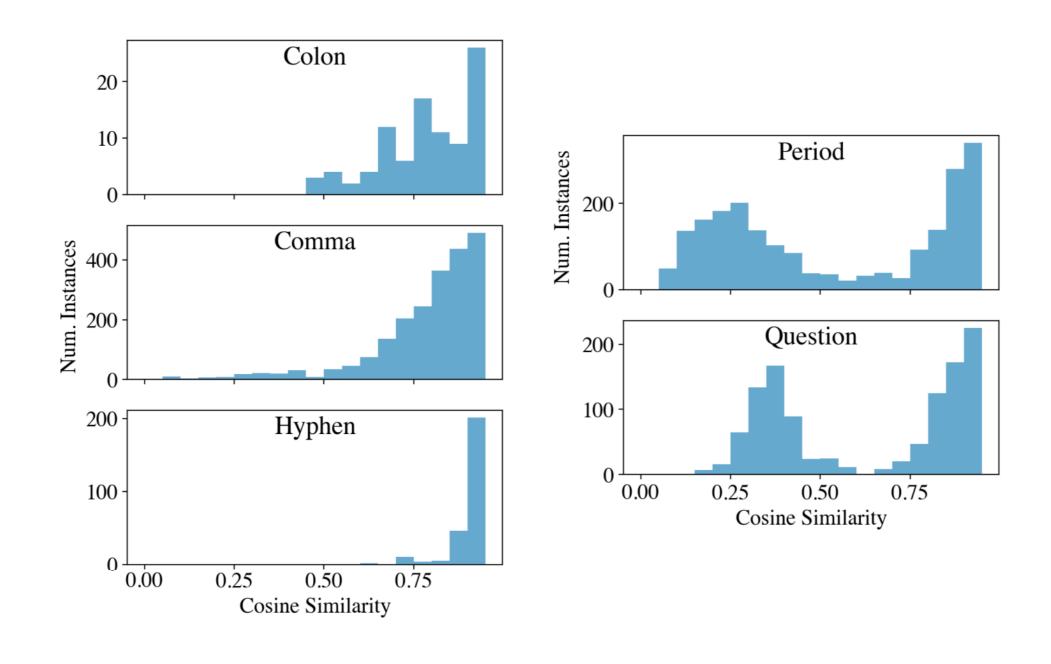
1 Phrase-level Embeddings

Word-level Embeddings

3 Punctuation

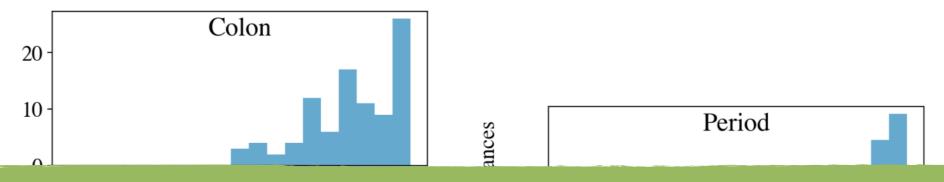
Punctuation

Is the distribution of embeddings for punctuation different than the distribution for other words?

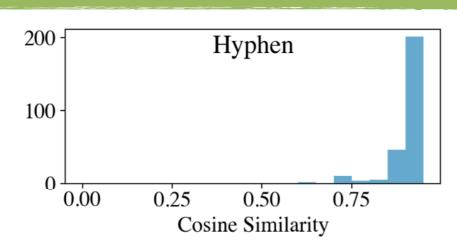


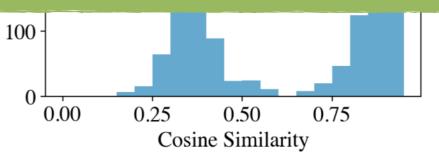
Punctuation

Is the distribution of embeddings for punctuation different than the distribution for other words?



BERT's representation of punctuation is surprisingly context sensitive, with substantial variation even when we control for meaning.





Takeaways

- BERT consistently represents paraphrases.
- We can use paraphrases to explore other representation methods!
- More in the paper...
 - Polysemous vs. non-polysemous words
 - One-word paraphrases (synonyms)
 - How the position of a word in a phrase affects its representation
 - How contextualized representations are at different BERT layers

Thank you!

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