

Bilingual Sentiment Embeddings: Joint Projection of Sentiment Across Languages

Motivation

Goal: How can you create bilingual word representations for cross-lingual sentiment analysis when you have little parallel data?

Problem: There is no labeled sentiment data or large amounts of parallel data available in most languages in the world.

Approach: We jointly train the projection and sentiment classifier.

Experimental Setup

- Sentence-level Cross-lingual Sentiment Analysis on hotel reviews: the OpeNER and MultiBooked Corpora. This avoids additional domain problems.
- Three language pairs:
 - English \rightarrow Spanish (ES)
 - English \rightarrow Catalan (CA)
 - English \rightarrow Basque (EU)
- Binary and 4-class.
- Train on the English annotated sentiment corpus only. Test on the target language test set.
- Compare to two bilingual embedding methods with similar data requirements (ARTETXE, BARISTA) and Machine Translation (MT).

Code Available

https://github.com/jbarnesspain/blse



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Bilingual Sentiment Embedding Model



Results for Cross-lingual Sentiment Classification

- BLSE outperforms other projection methods on macro F_1 (p < 0.05 highlighted) and is close to MT.
- ENSEMBLE of BLSE and MT performed better than other combinations (not shown here).

| | | Artetxe | Barista | Blse | MT | Ensemble |
|---------|----|---------|---------|-------------|------|----------|
| Binary | ES | 67.1 | 61.2 | 74.1 | 79.0 | 80.3 |
| | CA | 60.7 | 60.1 | 72.9 | 77.2 | 85.0 |
| | EU | 45.6 | 54.8 | 69.3 | 69.4 | 73.5 |
| 4-class | ES | 34.9 | 39.5 | 41.2 | 48.8 | 50.3 |
| | CA | 23.0 | 36.2 | 35.9 | 53.7 | 53.9 |
| | EU | 21.3 | 33.8 | 30.0 | 43.6 | 50.5 |



BLSE learns that "sentiment synonyms" should be close in vector space and "sentiment antonyms" should be farther away.

BLSE performs better on binary task. It (1) assigns too much sentiment to functional words and (2) does not preserve the original semantics of the embeddings.



Future Work



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Projection Lexicon

Model Analysis

• Move our approach to aspect-level.

 Include multi-word phrases into projection to improve negation and modifiers.