Fake news: an algorithmic approach

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2 Research question

The following research question is defined: *what is the performance of combinations of pre-trained embedding techniques with machine learning algorithms?*

For this research question, the following subquestions will be answered:

- RQ1: How well do Transformer architectures capture relevant information from text compared to non-Transformer text embeddings?

- RQ2: How well do neural network architecture classify fake news compared to non-neural classification algorithms?

3 Related Work

3.1 RQ1

Fake news as a term only caught public attention starting from the end of 2016, during the Presidential Elections of the United States [7].

3.2 RQ2

In the last couple of years, using transfer learning for natural language processing has given promisable results. The following sentence embeddings will be used to detect fake news:

- Bag of Words as a baseline for performance of non-pretrained embeddings;

- Facebook's InferSent [1];

- ELMo from the Allen Institute for Artificial Intelligence [5];

- OpenAI's GPT-2 [6];

- Transformer-XL [2];

- Microsoft's MT-DNN [4];

- and Google's BERT [3].

3.3 RQ3

Aligned with the original research on this dataset by Wang [8], the following machine learning algorithms will be used to test the applicability of the abovementioned embedding techniques:

- SVMs;

- Logistic regression;

- Bi-LSTMs;

- CNNs.

References

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