Self-Attention Sentimental Sentence Embedding For Sentiment Analysis

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Outline

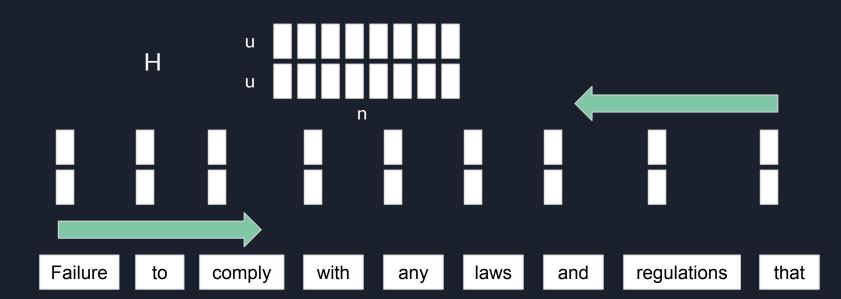
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 - Datasets
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- Conclusion

Introduction

- Pre-trained word embedding has shown success in NLP
- Some fields, such as Finance, have very few labeled data and many domain specific words, which make the field-related NLP tasks more challenging.
- There are many domain specific dictionaries; can we integrate the human knowledge into NLP tasks?
- Yes. In this paper, we integrate dictionary into a self-attention model for sentiment analysis

Our Methodology - Self Attention

Word

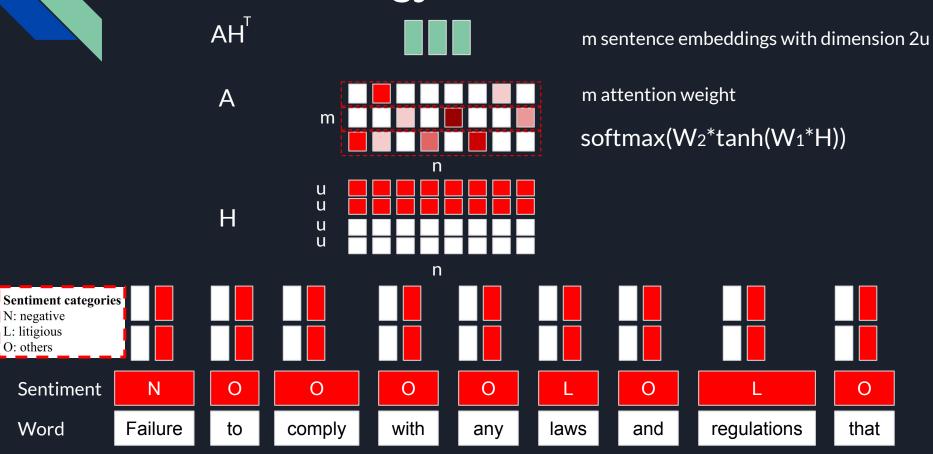


Our Methodology - Self Attention



Word Failure to comply with any laws and regulations that

Our Methodology - Self Attention + Sentiment



Experiment

- Datasets
 - 10K-Sentence (Label: Risky / Non-risky)
 - Yelp (Polarity 2-way, Full 5-way)
 - Amazon (Polarity 2-way, Full 5-way)
- Dictionary
 - General dictionary
 - Domain specific dictionary

Experiment

| Accuracy | Dictionary | Full (2-class) | Yelp | | Amazon | |
|-----------------|-----------------|----------------|--------------------|----------------|--------------------|----------------|
| | | | Polarity (2-class) | Full (5-class) | Polarity (2-class) | Full (5-class) |
| BOW | | 87.86% | 90.41% | 46.50% | 86.31% | 41.00% |
| LSTM | | 85.40% | 91.10% | 52.40% | 89.10% | 51.70% |
| CNN | | 85.37% | 87.22% | 58.78% | 83.02% | 51.36% |
| fastText | | 87.00% | 92.40% | 58.10% | 89.10% | 52.10% |
| Self-Attn | | 85.80% | 94.41% | 62.38% | 91.77% | 56.63% |
| Senti-Self-Attn | General-purpose | 87.24% | 94.66% | 62.95% | 91.78% | 56.25% |
| Senti-Self-Attn | Domain-specific | 88.13% | 94.23% | 62.07% | 91.83% | 55.58% |

 Table 1: Performance comparison

Case Study

while we believe our general operating and financial characteristics including a significant amount of cash on our balance sheet would enable us to ultimately respond effectively to an interruption in the availability of bonding capacity such an interruption would likely cause our revenues and profits to decline in the near term

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(a) Self-Attn

(b) Senti-Self-Attn (domain-specific lexicon)

Fig. 2: Heatmap of a risky sentence

Conclusion

- Demonstrate how to incorporate human knowledge, such as dictionary, into machine learning models.
- The empirical results show the effectiveness of our proposed model, especially for finance specific dataset.
- The attention heat map analysis shows how dictionary facilitates model to learn domain specific words.