CS224N Project Proposal

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2. Mentor: N/A

3. **Problem Description:** Many websites with user-submitted content must deal with toxic or abusive comments, but require increasingly fine-grained comment classifications to maintain civility without interfering with normal discourse. In this project, we will improve the identification and fine-grained classification of toxic online comments (Kaggle challenge suggested on course website, online at:

https://www.kaggle.com/c/jigsaw-toxic-comment-classification-challenge)

- 4. Data: We will be using a dataset of 159,571 comments from Wikipedia's talk page edits which have been labeled by human raters for toxic behavior. The types of toxicity are: toxic, severe_toxic, obscene, threat, insult, and identity_hate.
- 5. Methodology/Algorithm: We will first establish the classification baseline with a simple multi-label logistic regression and averaging all word embeddings of a comment. We will then move on to more elaborate methods like: (1) multilayer perceptrons with word embeddings, (2) RNNs/LSTMs with word embeddings, (3) convolutional neural networks (CNN) with word embeddings and/or character embeddings, and (4) attention-based networks [1]. We plan to utilize existing implementation when available, and to implement our own custom network architectures using TensorFlow.
- 6. **Related Work:** The earliest work on machine learning-based detection of online harassment can be traced back to Yin *et al.* [2], where they applied SVM on content, sentiment and contextual features to classify harassment. More recently, a paper by the group that has published the Kaggle challenge [3] focused on binary identification of toxic comments (no fine-grained classification). They found success with relatively simple n-gram NLP methods, and left more complex methods like LSTMs as future work. Last year, a group of students looked at the same dataset and found CNNs with character embeddings to be the most successful algorithm [4].
- 7. Evaluation Plan: We will plot ROC curves for classification of each class of toxic comment, and will run on a test set with true labels withheld by Kaggle, scored according to mean column-wise ROC AUC, a statistic which can be computed automatically.
- 8. Minimal Requirements:
 - (a) Number of examples $> 10000 \checkmark$
 - (b) Data collection already done? ✓
 - (c) Feasible task? ✓
 - (d) Automatic evaluation metric? <
 - (e) NLP required? ✓

References

- [1] A. Vaswani, et al. Attention Is All You Need. arXiv:1706.03762.
- [2] D. Yin, Z. Xue, L. Hong, B. Davison, A. Kontostathis, L. Edwards. Detection of Harassment on Web 2.0. In Proceedings of the Content Analysis in the WEB 2.0 (CAW2.0) Workshop at WWW2009, 2009

- [3] E. Wulczyn, N Thain, L Dixon. Ex Machina: Personal Attacks Seen at Scale. Proceedings of the 26th International Conference on World Wide Web, 2017.
- [4] T. Chu, K. Jue, and M. Wang. Comment Abuse Classification with Deep Learning. In CS224n Final Project Report, 2017.