Project: Hate Speech Detection using Transformers

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https://github.com/syoungk7/Hate Speech Detection

Problem description

With the advent of social media, people can express their opinions at any time, regardless of time

and space. However, this freedom of expression has made it possible to use derogatory or

discriminatory language against individuals or groups. Racial and social discrimination continues

online, too. Detecting and mitigating such negative and hate-related speech will help create a safe

and inclusive online environment.

Business understanding

Social media platforms such as Twitter often cause social problems because online platforms are

highly contagious. Detecting hate speech and identifying and reviewing or removing potentially

harmful content can help alleviate these problems.

From a business perspective, these systems will be able to provide users with a safe social media

environment, which will not only increase user satisfaction and engagement, but will also

automatically improve the reputation of the platform.

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Project Plan

- 1. **Understand the problem**: We understand the nature of hate speech, its impact, and the importance of addressing it on online platforms. We define the problem as a sentiment classification task based on labeled Twitter data.
- 2. **Data cleaning and normalization**: Preprocess the dataset by cleaning and normalizing the text data. This includes handling noise, removing irrelevant information, and ensuring consistency in the representation of tweets.
- 3. **Expression learning**: We leverage Transformer-based architecture for representation learning. Transformers have achieved significant success in natural language processing tasks and are well-suited to capturing contextual information in text data.
- 4. Model building and training: We develop a deep learning model for hate speech detection using preprocessed data. The model is trained on the training data set to optimize accuracy and minimize false positives and false negatives.
- 5. **Performance evaluation and reporting**: Evaluate model performance on the test dataset using relevant metrics such as accuracy, precision, recall, and F1 score. Provides a comprehensive report on the strengths and limitations of the model.
- 6. **Model Deployment**: Deploy the trained model into a production environment, ready to integrate with online platforms to actively detect and filter hate speech.
- 7. **Model inference**: By implementing the model's inference mechanism, we can classify new tweets in real time to identify and flag potential instances of hate speech.