Identifying toxic content online

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Background and aim

The internet can be a hostile place, and it's a challenge to monitor and regulate. Some degree of automation might help - you could, for example, keep an eye out for certain offensive terms, although toxic behaviour is often much more subtle than this.

So the question is: can you train a machine learning classifier to recognise toxic content?

Can it pick up patterns in the words people use, and the way they use them, to predict whether a post is toxic or not?

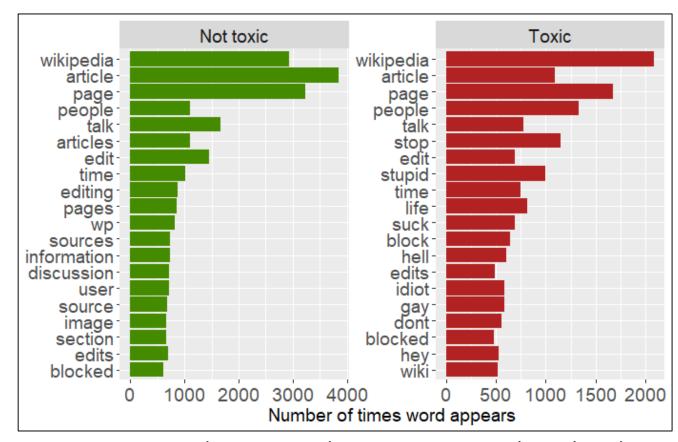


Fig. 1: Top 20 words in toxic and non-toxic posts in the Wikipedia Detox dataset (n = 30,000)

The process

This is a classic **text classification** problem. I am taking a **supervised learning** approach. There's also a bit of **sentiment analysis** involved.

GET TRAINING DATA

- The Wikipedia Detox Project
- Discussions from talk pages
- Annotated for toxicity by crowdsourcing

EXTRACT FEATURES

- Bag of words: unigrams, bigrams, more?
- Word embeddings
- Sentiment scores
 - Punctuation, capitalisation, repeated words and letters

BUILD CLASSIFIER

- Logistic regression
- Random forest
- Naïve Bayes
- Support Vector Machines

EVALUATE

- How does the model perform on unseen data?
- Accuracy, precision, recall
- Does it work on data from a different website?

All data manipulation and analysis carried out in R, using RStudio and key packages such as tidytext and caret.

Preliminary results

Looking for specific terms is certainly a good starting point – as Fig. 1 shows, there are clear differences in the most used words, although they also have several in common.

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It might also be useful to look at which words are *proportionally* more likely to appear in toxic posts (Fig. 2).

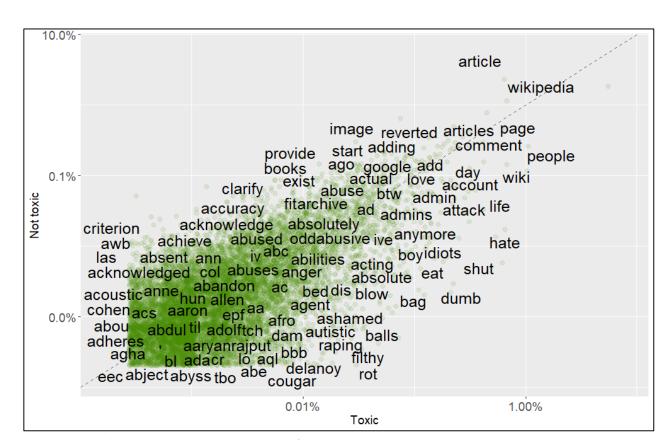


Fig. 2: Relative proportions of words in toxic and non-toxic posts

Next steps

Build and test classifiers, experimenting with different models and feature sets, and see which perform the best.

