

Cornell Sun Age Classification

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Project Overview

- ❖ Classify a Cornell Daily Sun article's likely readership
- ❖ Use data obtained by Google Analytics stored by the Cornell Daily Sun
- ❖ Naïve Bayes Classification model
- ❖ Create an iOS app that displays the classification information dynamically to the user

Model Overview

- ❖ Naïve Bayes Classification model
- ❖ Features: Words that appear in the content of an article
- ❖ Example:
 - ❖ { “hello”: 1, “world”: 2, “Cornell”: 3 }
 - ❖ [hello, world, world, Cornell, Cornell, Cornell]

The diagram shows the Naïve Bayes formula with arrows pointing from labels to the corresponding parts of the equation:

$$P(c | x) = \frac{P(x | c)P(c)}{P(x)}$$

Labels and their corresponding parts in the formula:

- Likelihood** points to $P(x | c)$
- Class Prior Probability** points to $P(c)$
- Posterior Probability** points to $P(c | x)$
- Predictor Prior Probability** points to $P(x)$

$$P(c | X) = P(x_1 | c) \times P(x_2 | c) \times \cdots \times P(x_n | c) \times P(c)$$

Obtaining Training Data

- ❖ Google Analytics only stores the title and sessions counts
- ❖ Strip out unnecessary unicode characters in the title
- ❖ Take the title and make an API request to the Cornell Sun's WordPress account
 - ❖ Only search results to took the best result and checked for a rendered title match
- ❖ Counted the words and wrote the word counts to a text file
- ❖ Manually classify the Google Analytics articles by counting the sessions in each age group

Training the Model in Swift

- ❖ Use the `.observe(...)` method in the Bayes module
- ❖ Takes in the classification and the list of observed words
 - ❖ Ex. “18-24” and [hello, world, world, Cornell, Cornell]
- ❖ Read in both the classifications text file and the word counts text file into the app
- ❖ Observe each of the articles when training with all the words associated with said article

Judging Model Accuracy

- ❖ Total of 819 articles found and classified
- ❖ 70% of data used to train the model (574 articles)
- ❖ 30% of data used to test the model (245 articles)
- ❖ Accuracy of 76% on the testing data after training
- ❖ Given more training data, the model generally performed better

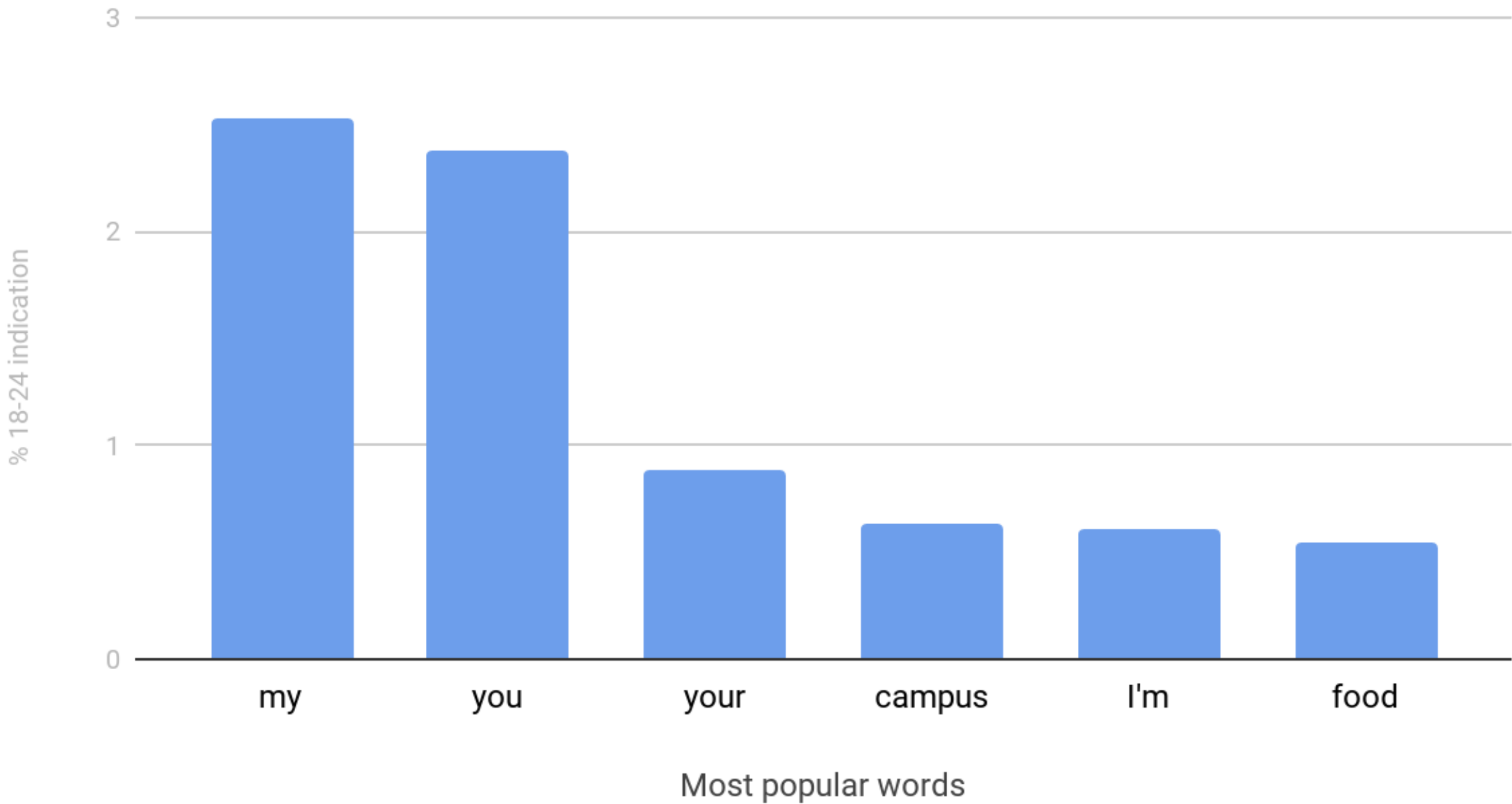
The Accuracy of our Naive Bayes Model



Analyzing Indicator Words

- ❖ Attempted to identify words that indicated the article's classification
- ❖ Found the probability a word appears given a certain classification
- ❖ Subtracted off the equivalent probabilities for the other classifications
- ❖ If this value was greater than 0.05, deemed this word as an **indicator**
- ❖ Majority of articles containing this word classified accordingly

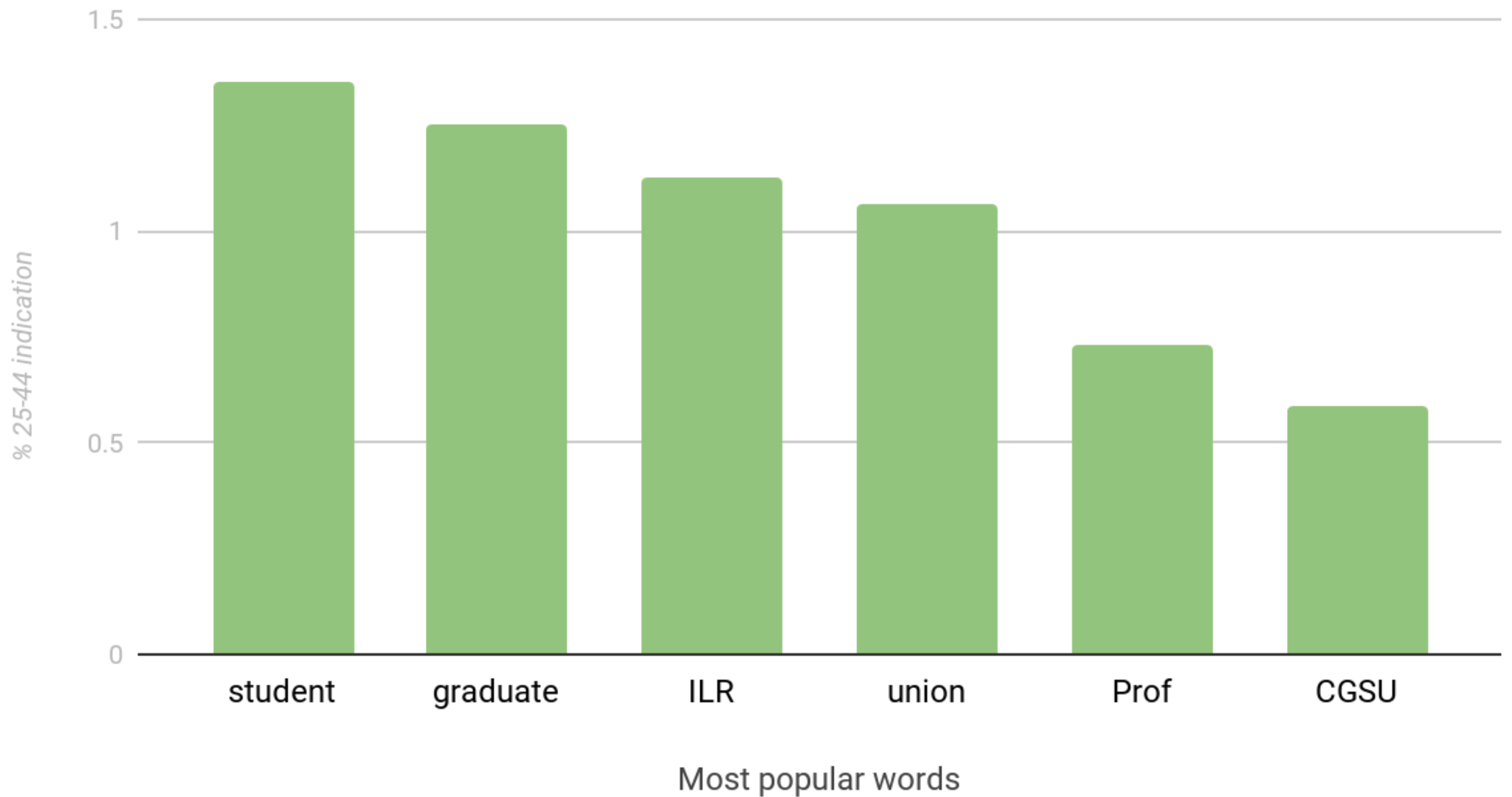
Most Popular Words Read By Ages 18-24



18-24 Classification Conclusions

- ❖ Articles containing casual language more likely to be consumed by younger audiences
 - ❖ “You”
 - ❖ “Your”
 - ❖ “My”
 - ❖ “Food”
- ❖ More close to campus and social-oriented

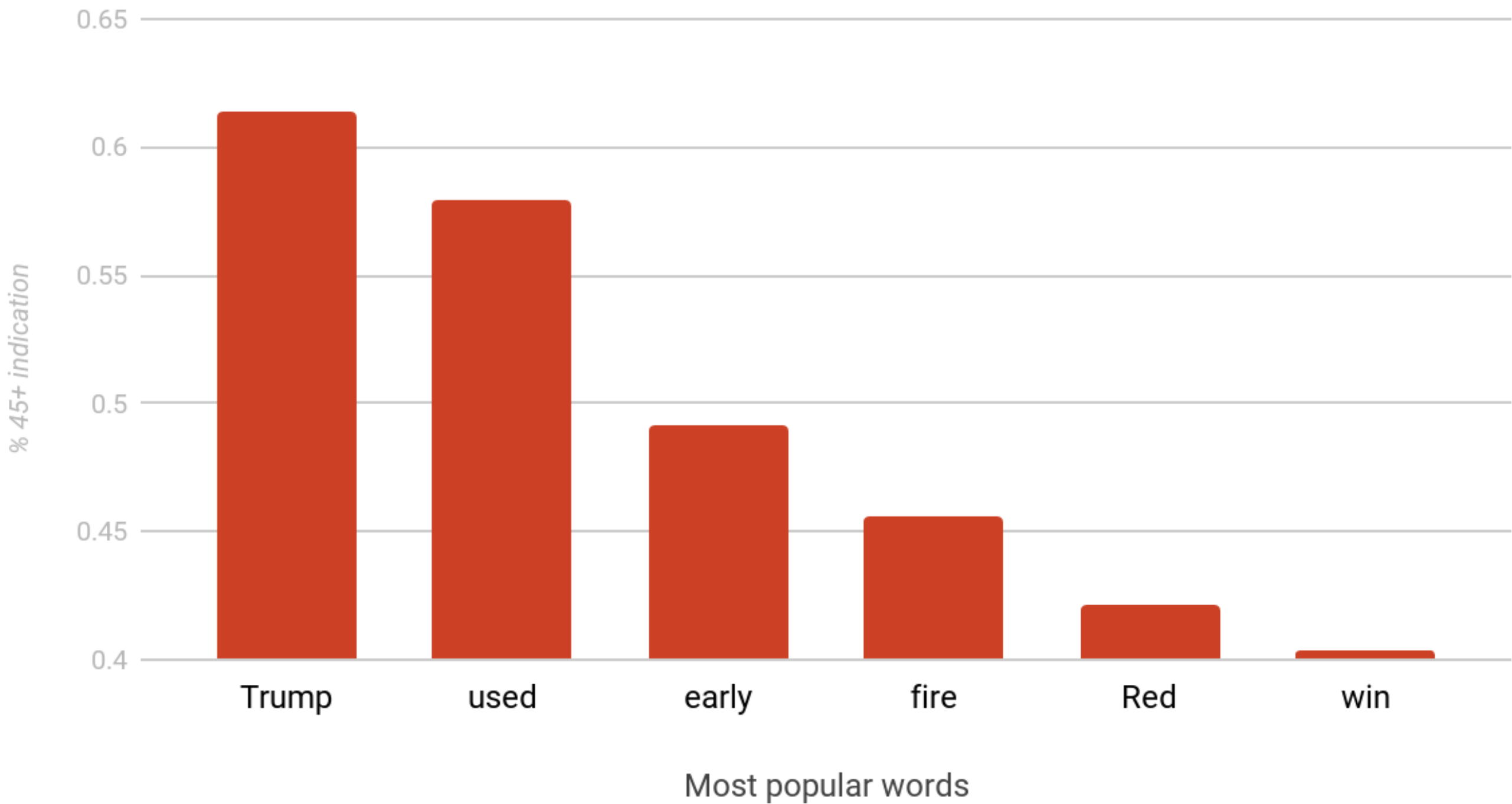
Most Popular Words Read by Ages 25-44



25-44 Classification Conclusions

- ❖ Articles containing language pertaining to economics, unionization, or academics
 - ❖ “Student”
 - ❖ “Graduate”
 - ❖ “Union”
 - ❖ “ILR”
- ❖ More academic or organization related

Most Popular Words Read by Ages 45+



45+ Classification Conclusions

- ❖ Articles containing language pertaining to politics or local news events
 - ❖ “Trump”
 - ❖ “Fire”
 - ❖ “Used”
- ❖ More oriented around politics or news stories

iOS Application and Demo

- ❖ Tap on article
- ❖ Shows statistics and classification for article
 - ❖ Most likely classification
 - ❖ Top 5 words
 - ❖ Corresponding weight for each word

