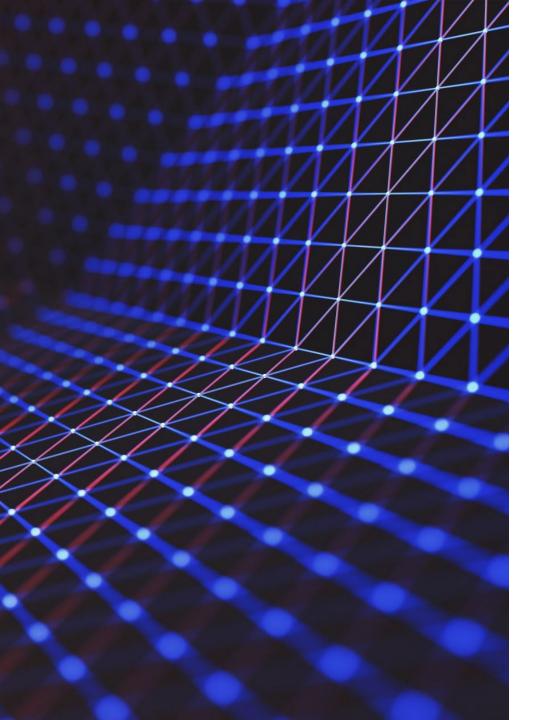
Battle of the Sexes – Blogger Edition

Saint Gau



The Problem

- Incorrect stereotypes: Studies have found that women do not use significantly more words than men, yet these stereotypes persist
- Developing research: Not many studies have been done on gender differences in vocabulary
- NLP for business: Accurate predictors of gender based on text used could be very useful for marketing or other businessrelated classifications



Who might care?





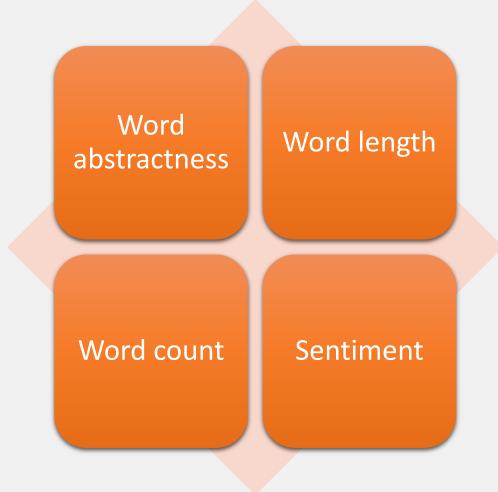


Advertisers

Researchers

General public

How might men and women blog differently?





Collected August 2004

Data Information



681,288 blogs written by 19,320 bloggers



Number of fields: 7

is on PBS tonight. Check your local listings for the time.

	id	gender	age	industry	sign	date	blog
0	1000331	female	37	indUnk	Leo	31,May,2004	Well, everyone got up and going this morning
1	1000331	female	37	indUnk	Leo	29,May,2004	My four-year old never stops talking. She'll
2	1000331	female	37	indUnk	Leo	28,May,2004	Actually it's not raining yet, but I bought 15
3	1000331	female	37	indUnk	Leo	28,May,2004	Ha! Just set up my RSS feed - that is so easy!
4	1000331	female	37	indUnk	Leo	28,May,2004	Oh, which just reminded me, we were talking ab

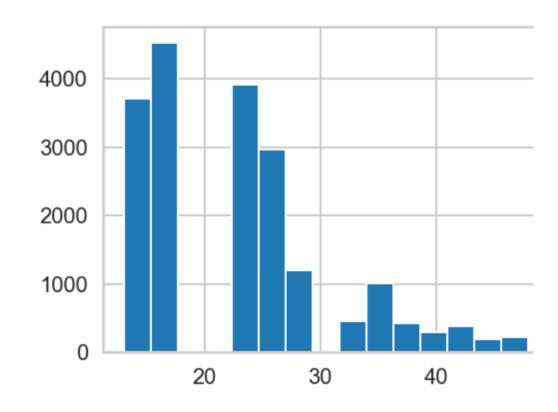
DataFrame Creation

Data Exploration

- Age
- Gender
- Industry
- Date
- Languages
- Sentiment analysis
- Word clouds

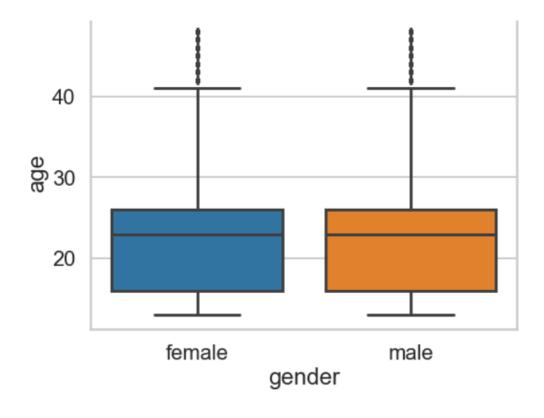
Age

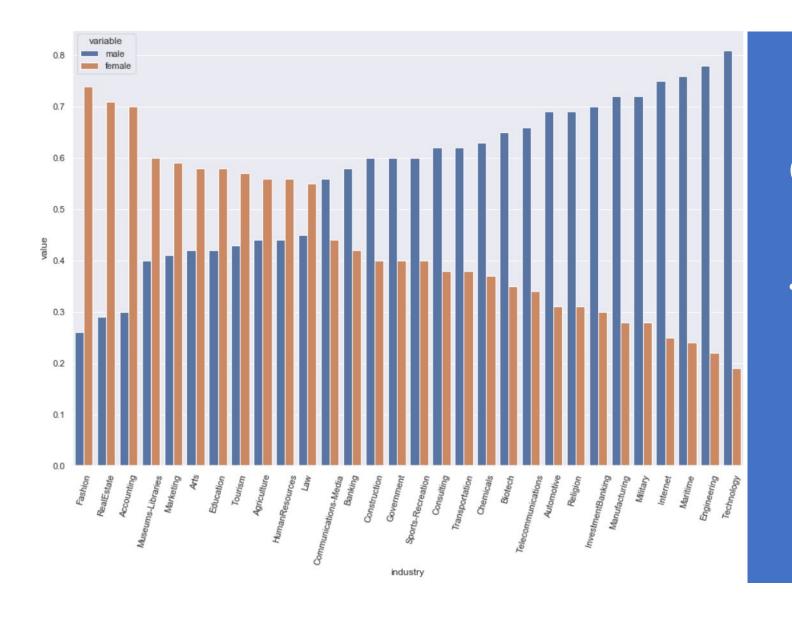
• Gaps in age distribution



Age vs. Gender

• No gender differences in age



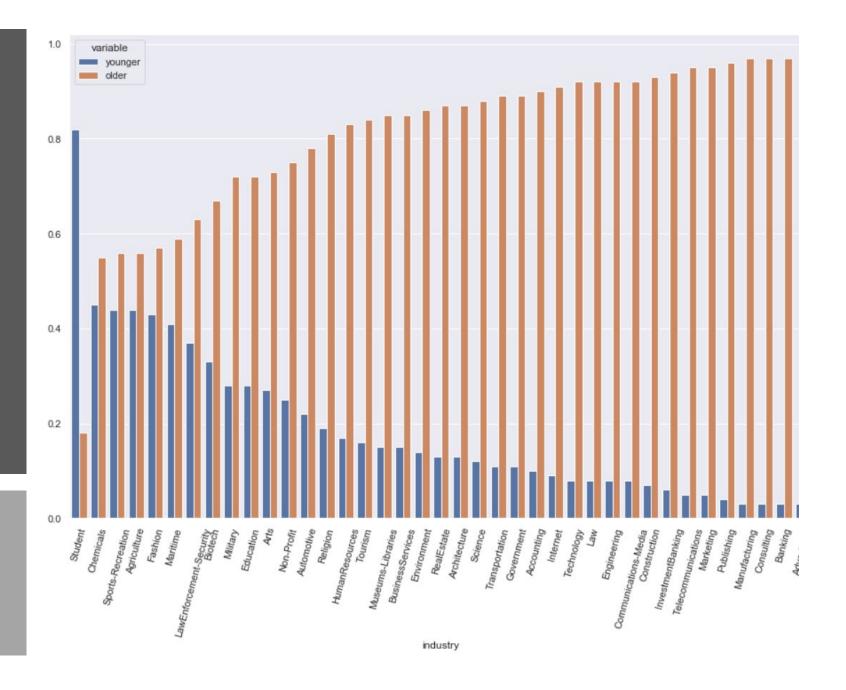


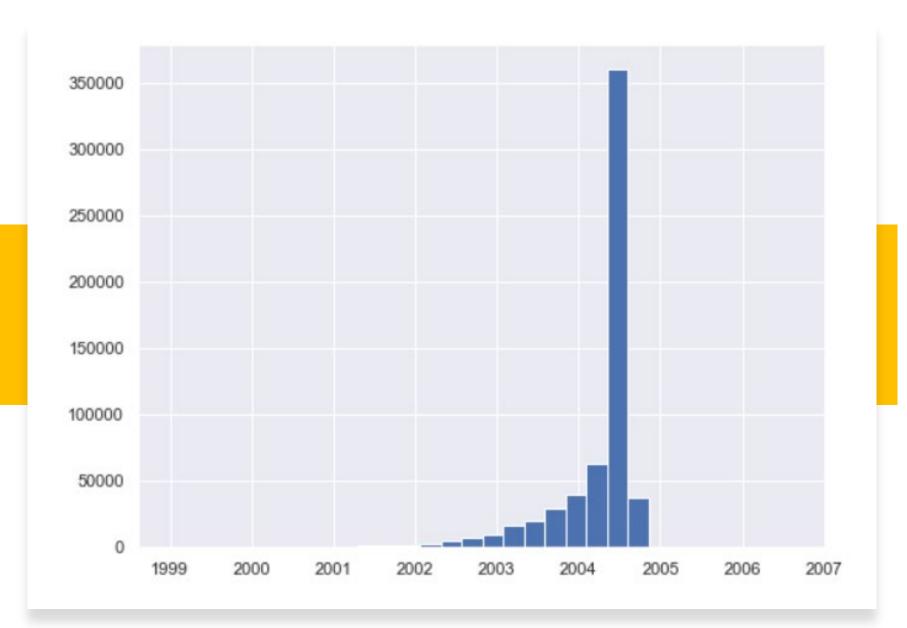
Gender vs. Industry

 Gender differences ranged widely across different industries

Age vs. Industry

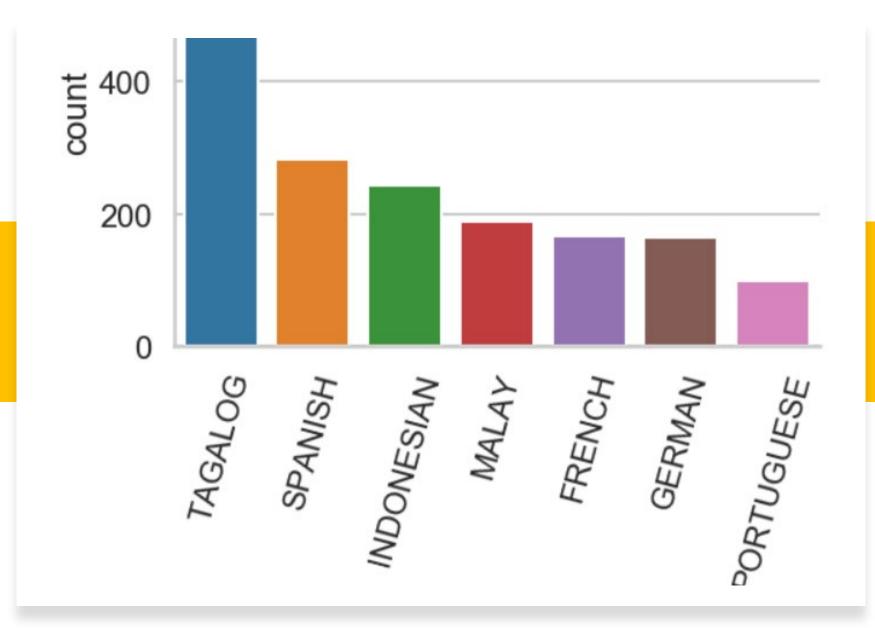
Most industries were made up of mostly older bloggers





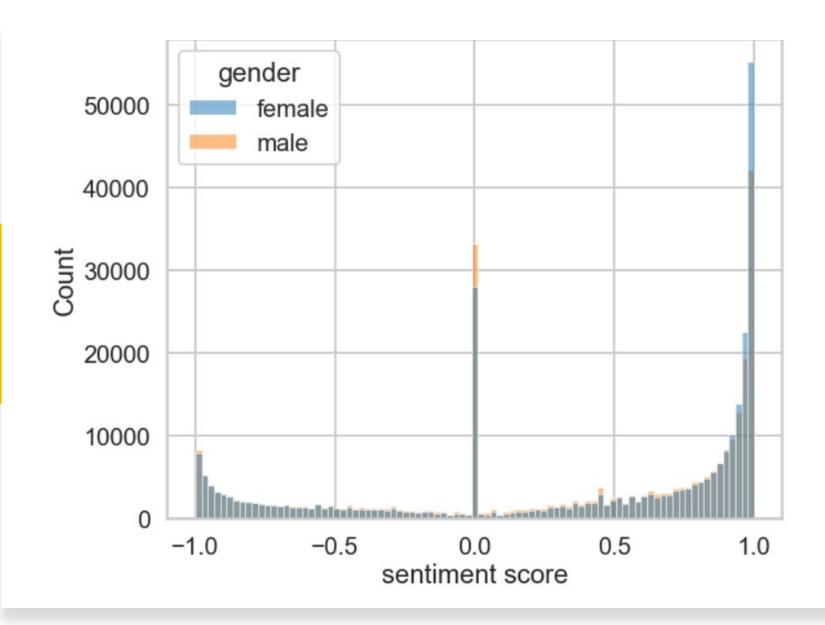
Date

Most blogs were written in 2004



Languages

Most common language of foreign blogs was Tagalog

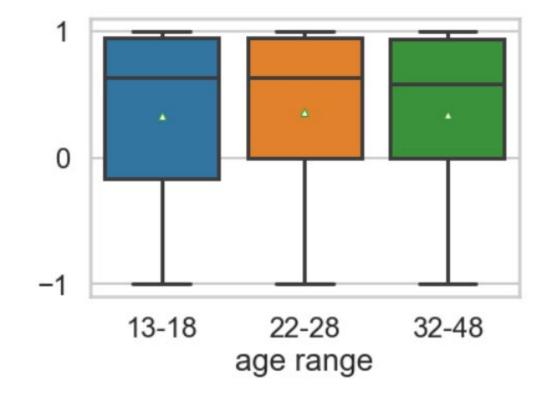


Sentiment Analysis by Gender

Female bloggers were significantly more positive than male bloggers

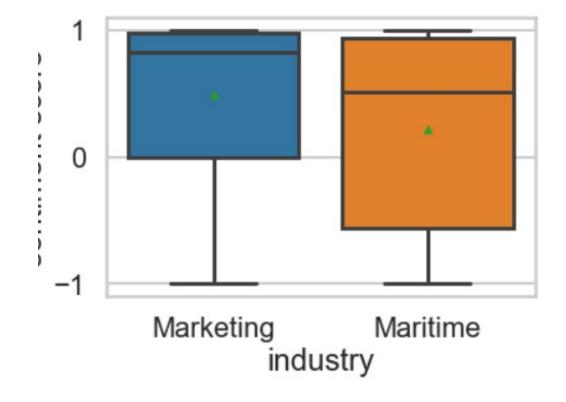
Sentiment Analysis by Age

• 22-28-year-old bloggers were significantly more positive than 13-17-year-old bloggers



Sentiment Analysis by Industry

 Marketing bloggers were significantly the most positive, while maritime bloggers were significantly the most negative



Word Clouds

Word clouds between different groups were mostly similar

Teen word cloud



Young word cloud

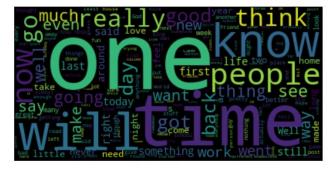


Old word cloud



Male word cloud

Female word cloud







Machine Learning Modeling







Modeling Steps

- Pipeline
 - Data Pre-Processing
 - 1. One-Hot Encoding
 - 2. Data splitting into training and test sets (80%-20%)
 - 3. Scaling
 - Cross-Validation (CV) for Hyperparameter tuning
 - 1.5-fold CV
 - 2. Using scikit-learn's grid search method
 - 3. Evaluation metric: AUC



Performance evaluation using holdout dataset (20% of data used)

Classifier Algorithms Used



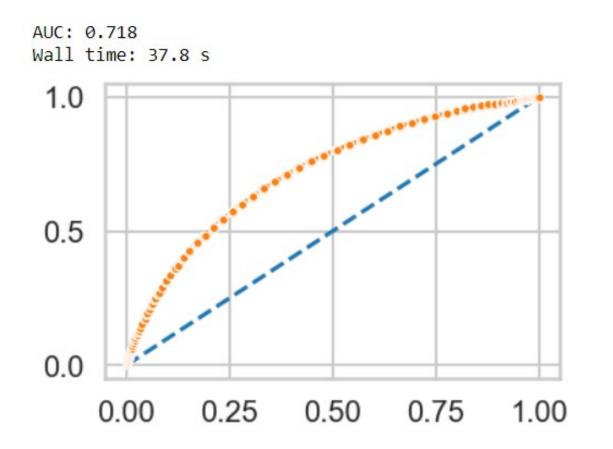
1. Random Forest Classifier

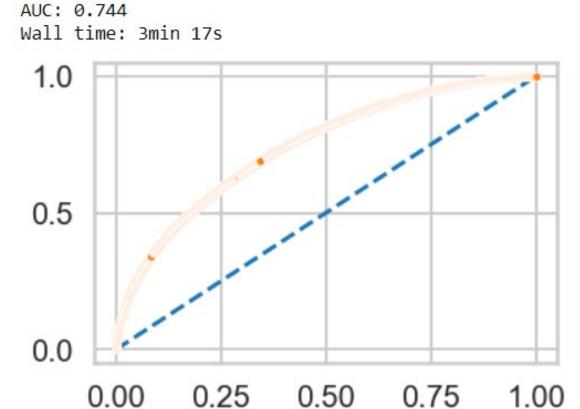


2. XGBoost Classifier

Model Comparisons

XGBoost performed better





Some Details on the Best Model

Tfidfvectorized

Standard scaling

Using the Model

- Input blog into model
- Use model pipeline on new data and predict gender of blogger

Conclusion

• Men and women do use different words when blogging

Assumptions, Limitations, and Disclaimers



We assume that all bloggers are independent



Blogs were mostly written in 2004, more than 15 years ago



Model would be more accurate if entire dataset were used

More Ideas to Improve the Model in the Future



Diversify information with more features



Extract information from more recent blogs



Include location of bloggers to account for regional differences in language

Thank you!

Saint Gau

Email: transaintgau@gmail.com

https://www.linkedin.com/in/saintgau/

https://github.com/transaint/Professional-Portfolio

Final project report:

https://github.com/transaint/Springboard-Projects/blob/master/Springboard%20Projects/Predicting %20a%20Table's%20Tips/Final%20Project%20Report.ipyn b

