Lyrical Analysis of Rap Misogyny

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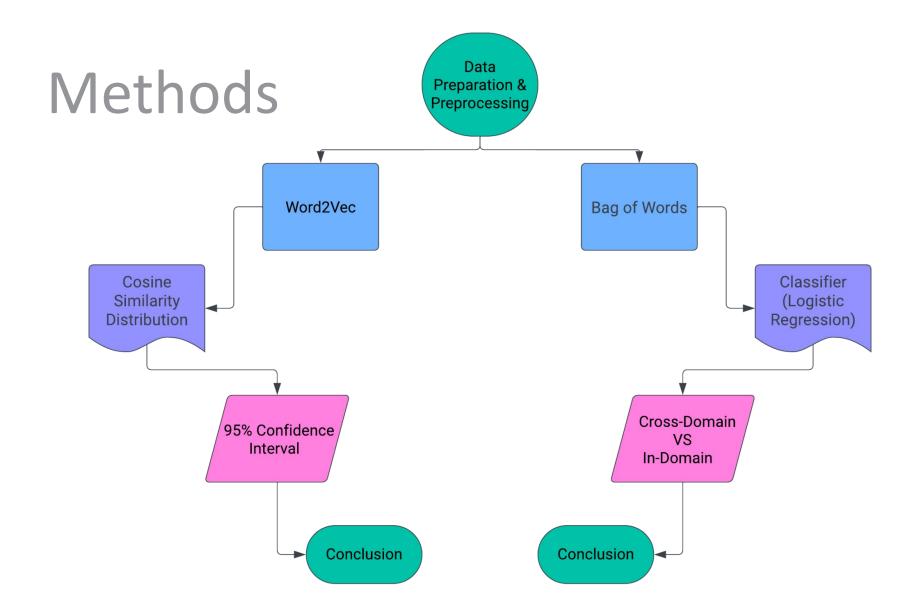
Background

The influence of American rap can be felt around the world. This **cultural export** raises important concerns about the role of rap & hip hop culture on the spread of misogyny in pop culture. We want to evaluate the truth of some assumptions that are commonly held, using textual data:

- Stereotypes: Is rap music really as misogynistic as is commonly believed?
- Diverse Interpretations: Can the prevalence of misogyny in the music industry be attributed to

Data

- 1. Lyrics: Database of scraped lyrics Rap-genred songs on Kaggle.
 - Rap Song Dataset: 4,578 lyrics; Rappers, Verses.
 - General Lyrics Dataset: 1,840 lyrics; Artists, Genres, Lyrics.
- 2. Misogynistic Words: Dataset of Reddit posts from traditionally misogynistic communities. Previous researchers have conducted labeling and annotation tasks to determine if the posts are misogynistic or not.
 - Labeled Dataset: 6,567 Reddit posts; "Misogynistic" or "Non-Misogynistic".



Cosine Similarity Scoring

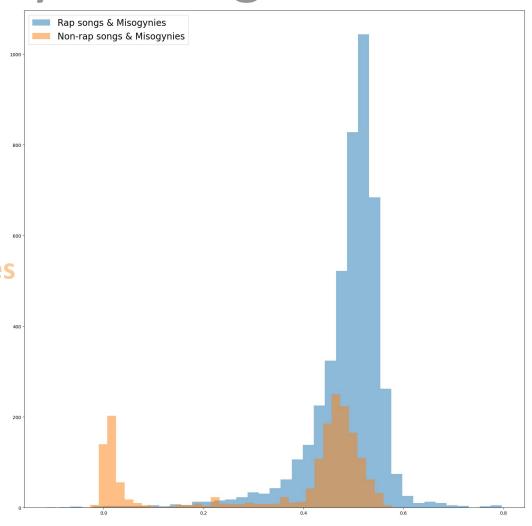
Confidence Intervals:

Rap Song & Misogynies

(0.4871, 0.4915)

Non-Rap Songs & Misogynies

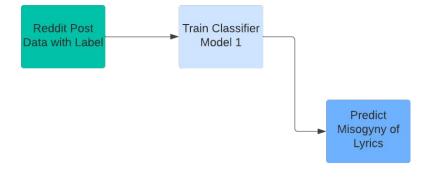
(0.3362, 0.3548)



Domain-Based Predictions

We were able to apply our dataset of misogynistic words to our rap lyrics via **cross-domain** and **in-domain** prediction to obtain a broader scope:

	precision	recall	f1-score	support
Misogynistic Nonmisogynistic	0.79 0.94	0.54 0.98	0.64 0.96	147 1167
accuracy macro avg weighted avg	0.87 0.93	0.76 0.93	0.93 0.80 0.93	1314 1314 1314



Train Classifier Model 2

> Predict Misogyny of

Cross-Domain Prediction Results

	precision	recall	f1-score	support
Misogynistic Nonmisogynistic	0.57 0.77	0.57 0.77	0.57 0.77	7 13
accuracy macro avg weighted avg	0.67 0.70	0.67 0.70	0.70 0.67 0.70	20 20 20

In-Domain Prediction Results

Limitations

Some examples of technical limitations which should be considered for further analysis are as follows:

- Word2Vec Limitations with Slang: Word2Vec is unable to fully understand slang and culturally-specific language.
- Sole Use of Logistic Regression Classifier: Reliance only on Logistic Regression Classifier for analysis.
- Subjectivity in Misogynistic Interpretation: Variability in interpreting and labeling lyrics as misogynistic.
- Data Constraints: Limited size of the dataset + lack of comprehensive labeled lyric data.

Complexities to Consider

There are several intriguing social issues and aspects of language that we should consider:

- Cultural Impact: Rap music's broad influence sparks important discussions about misogyny in pop culture.
- Language Nuances: Rap's rich linguistic structure, utilizing slang, metaphors, and filler words, makes identifying misogyny challenging.
- Context Sensitivity: Misogynistic interpretations vary with context.

Next Steps

Enhancements Enhance Word2Vec Model to mprove accuracy and contextual

capabilities for precision
Try to use models other than
logistic regression classifer

Next Steps

Expanding and Deepening the Analysis

Expansion

Incorporate a wider range of songs, convering more genres, artists and time periods.

Cross-Genre Comparison

conduct a comparison of misogyny across different music genres

Thank You! Q&A

Appendix

Rap song lyrics data:

https://github.com/fpaupier/RapLyrics-Scraper/tree/master/lyrics_US

General lyrics data:

https://www.kaggle.com/datasets/neisse/scrapped-lyrics-from-6-genres

Labeling data:

Coates, Joshua, and Danushka Bollegala. "Frustratingly easy meta-embedding – computing meta-embeddings by averaging source word embeddings." *Proceedings of the 2018 Conference of the North American Chapter of the Association for Computational Linguistics: Human Language Technologies, Volume 2 (Short Papers)*, 2018, https://doi.org/10.18653/v1/n18-2031.