

Computational linguistic analysis of dehumanization of LGBTQ groups in the media



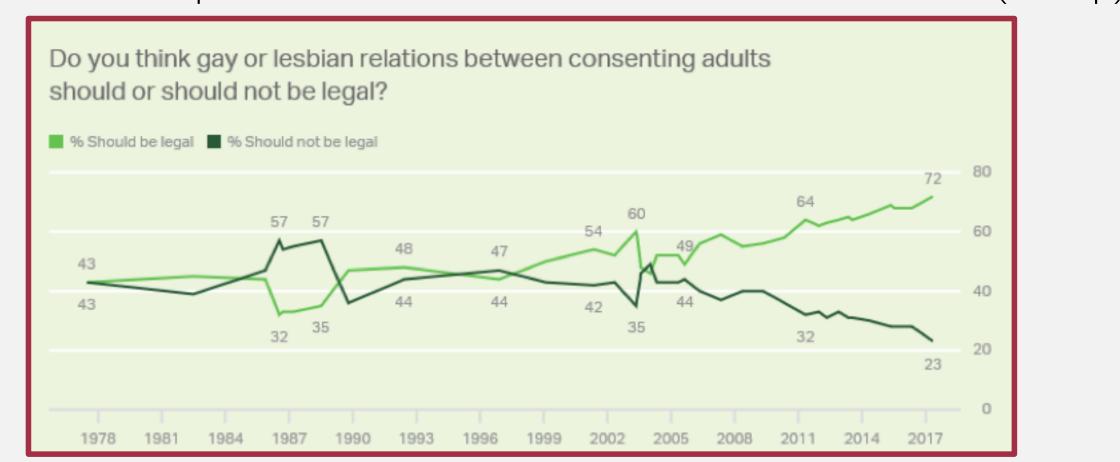
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Goals

- Propose & evaluate techniques to quantify dehumanization
- Understand how LGBTQ representation has changed over time
- Explore different meanings of terms gay and homosexual

Background

- **Dehumanization**: 'the act of perceiving or treating people as less than human'' (Haslem & Stratemeyer, 2016).
- Typically consists of multiple component psychological processes
- LGBTQ acceptance in USA has increased in recent decades (Gallup)



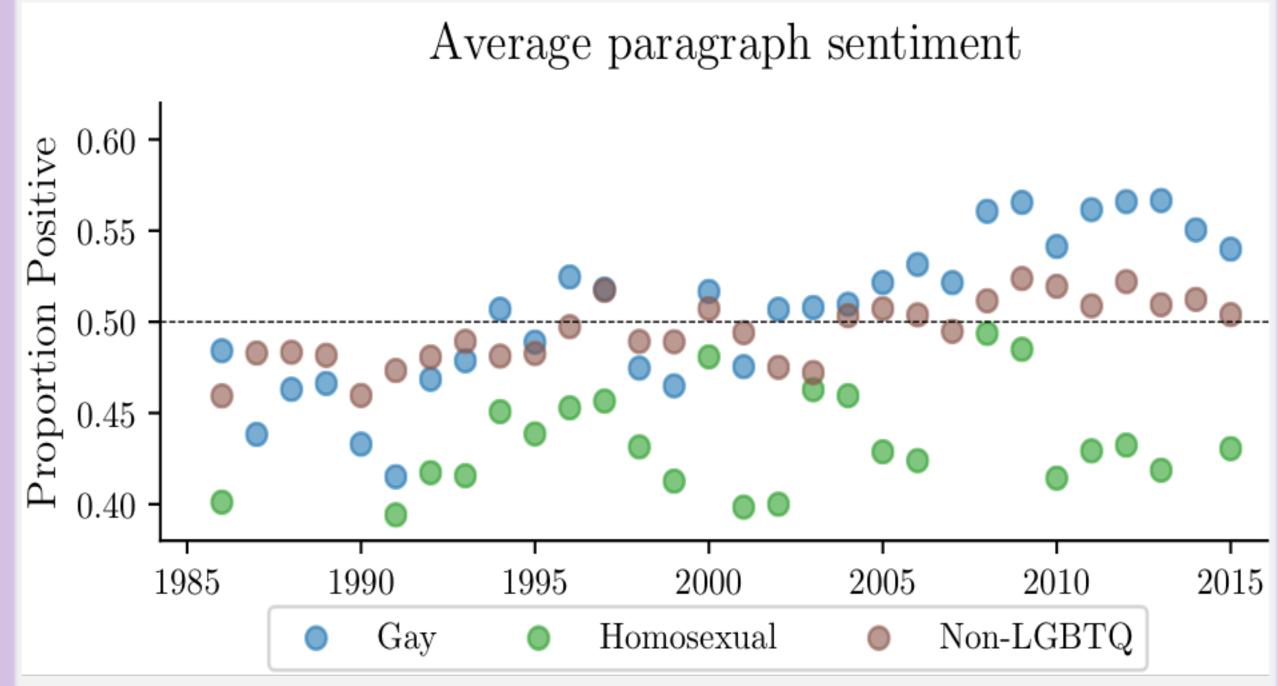
• Term homosexual considered outdated, clinical, derogatory (GLAAD)

Data

- Paragraphs from New York Times (NYT) containing LGBTQ terms
- Jan. 1986 Apr. 2016; over 110,000 paragraphs and 7 million words

Negative Evaluation of Out-Group

- Approximated by sentiment analysis using Bing Liu's sentiment lexicon
- Proportion of positive words (relative to negative) averaged by year



Negative language is not always sign of dehumanization. More negative words are used in debates and reporting on LGBTQ-related issues

Future Work

- Improve methods for precise, robust, and interpretable measurements
- Incorporate other dehumanization components (e.g. denial of agency)
- Analyze more groups, different newspapers, and social media

Experiment Overview

Dehumanization Component	Linguistic/Quantitative Correlate
Lexical association exploration	Log-odds & word embeddings
Disgust	Vector-similarity disgust analysis
Vermin Metaphor	Vector-similarity metaphor analysis
Negative evaluation of out-group	Sentiment analysis
Psychological distance	Definite plurals
Essentialization	Pejorative nominalization

Lexical Associations

Log-odds: find words overrepresented in one corpus relative to another

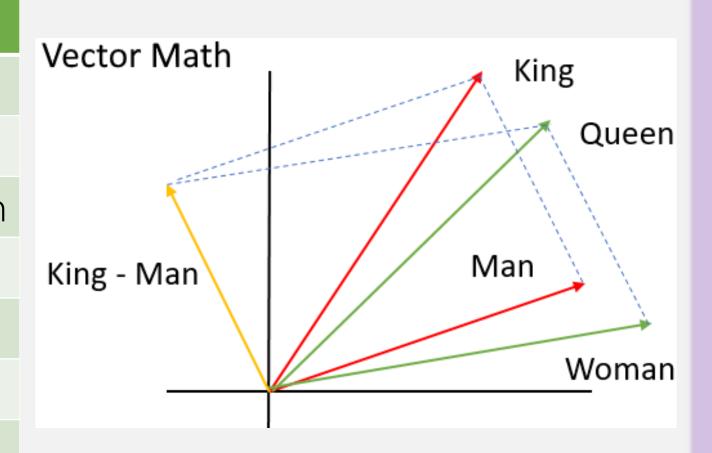
Calculated log-odds ratios of words in earliest decade relative to latest decade

Most personal pronouns (you, I, my, her, etc.) in top 50 words associated with later years. Personal stories & experiences in later years \rightarrow less denial of subjectivity

Early Years	Later Years
homosexuals	marriage
homosexual	samesex
aids	couples
drug	to
don't	you
clinton	on
parade	marriages
disease	supreme
virus	unions
military	state
groups	married

Word embeddings: represent words as points in vector space based on word co-occurrences. Geometric relationships between word vectors capture aspects of meaning, semantic change, and stereotypes.

Gay	Homosexual
sex	cohabiting
queers	indissolubility
marriage	excommunication
interracial	indissoluble
couples	bestiality
homophobia	abomination
closeted	mormons
openly	immoral
ordain	religious
mormons	sin
polygamy	sex
cohabiting	irja
equality	adultery



(Above) Vector arithmetic reveals relationships between words, such as man: king:: woman: queen analogy (L. Shure; blogs.mathworks.com)

(Left) words closest to gay & homosexual in Word2Vec model trained on 2015 New York Times data

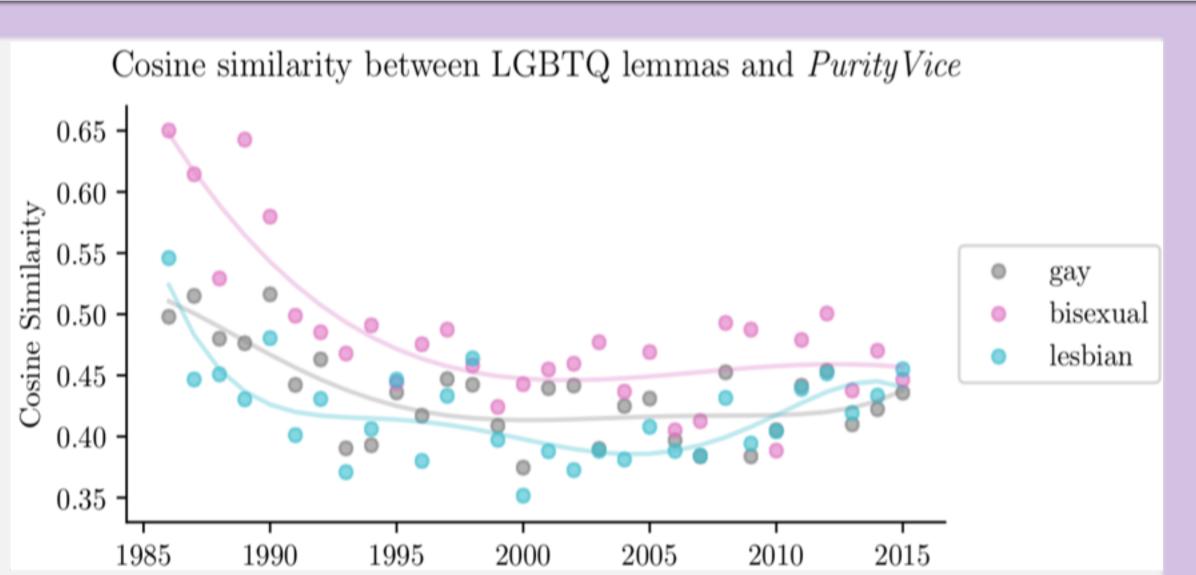
Conclusions

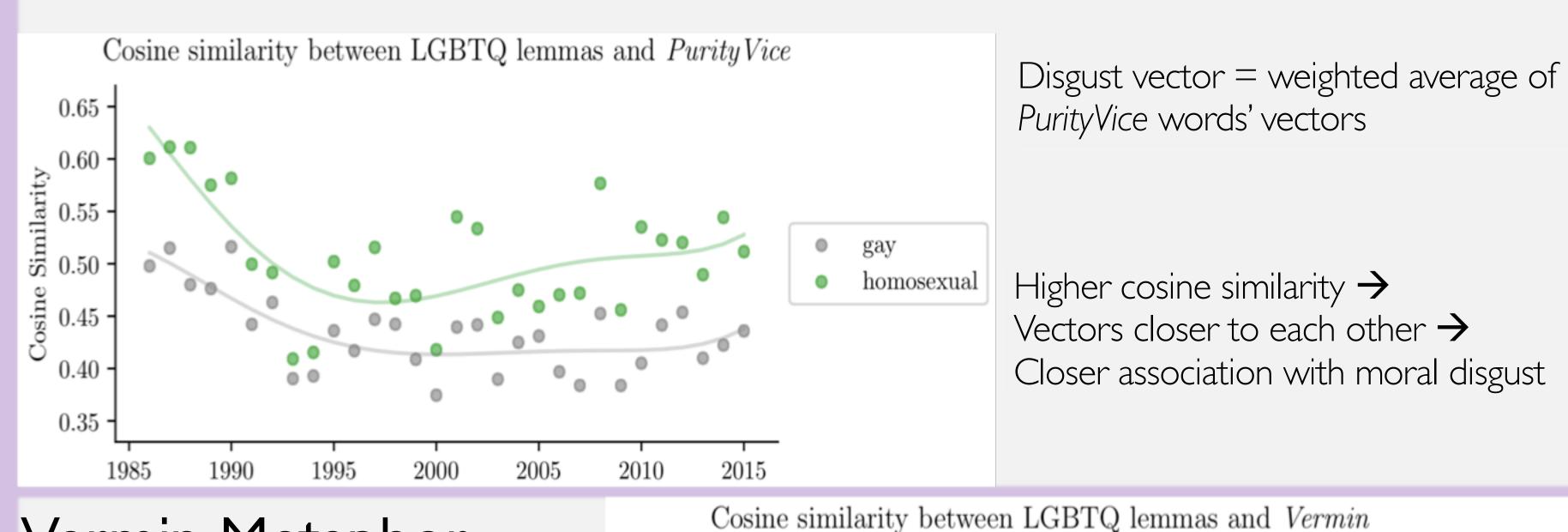
- LGBTQ individuals have been increasingly humanized in the NYT
- Homosexual is used in more dehumanizing contexts than gay
- Implications for detecting harmful bias and abusive language online
- NLP tools for understanding aspects of language variation and change

Moral Disgust

Similarities between LGBTQ vectors and moral disgust vector in Word2Vec captures extent of association

Disgust words from Moral Foundation Theory *PurityVice* dictionary

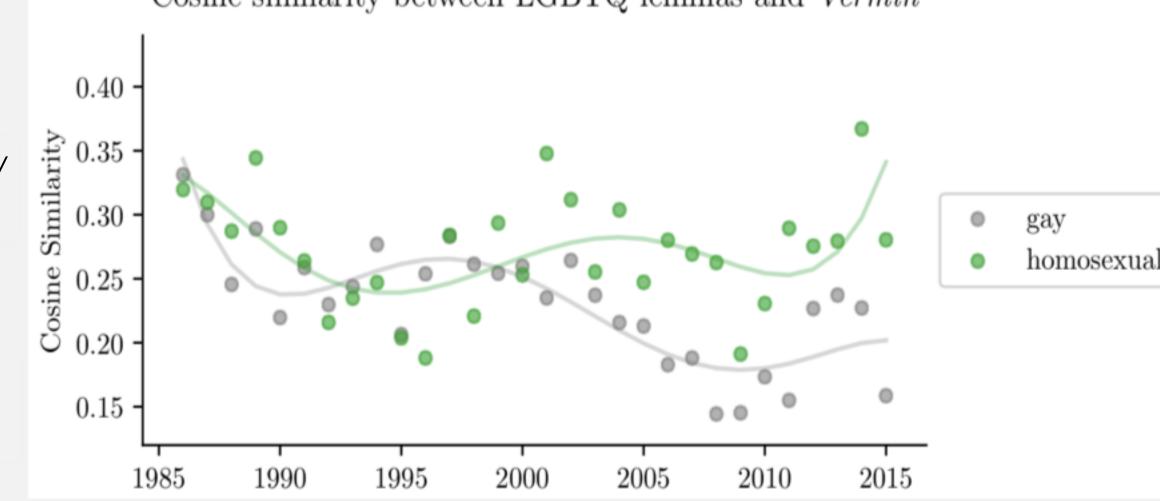


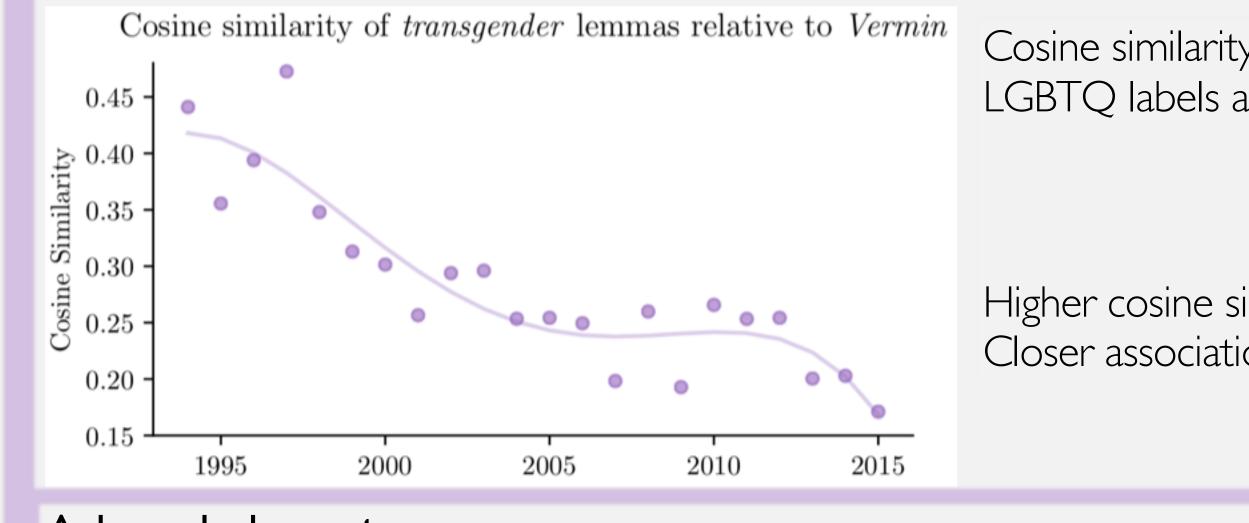




Dangerous, extreme, and rarely directly invoked, but exists in subtle forms

Vermin vector = Weighted average of vermin words (rats, cockroaches, etc.)





Cosine similarity to measure associations between LGBTQ labels and vermin representation.

Higher cosine similarity -> Closer association with vermin

Acknowledgments

Thanks to Dan Jurafsky, Yulia Tsvetkov, Chris Potts, and Stanford Linguistics Department for advice and feedback Thanks to Ethan Fast for supplying the New York Times data

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