

# Leveraging Natural Language Processing and Thematic Analysis to Understand Personal Sensing Acceptability in Opioid Use Disorder Patients

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# Opioid Use Disorder (OUD)

- The chronic use of opioids for non-medical reasons that causes **clinically significant** distress or impairment.
- Affects over **2.1 million people** in the United States.
- In 2021, **over 80,000 people** in the United States died from an opioid overdose.
- Very **few people** seek treatment.



# Individual

- Stigma
- Lack of social support
- Cost of care





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## Provider/Institutional

- Low treatment availability in rural areas
- Low variety of treatment options
- System navigation





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## Societal/Systematic

- Laws surrounding treatment
- Flawed justice system
- Lack of education about addiction

# Digital Therapeutics

- Accessible **24/7** and can be used in conjunction with traditional care or by itself.
- These **personalized** treatments use personal sensing and machine learning to **predict and recommend**.
- While the potential for reaching a largely underserved population is high, it requires individuals find personal sensing **acceptable**.



# Acceptability of Personal Sensing

- People with alcohol use disorder using sensing methods for three months found methods to be **acceptable**.
- There are reasons to suspect this could be different among those with other substance use disorders, particularly **opioid use disorder**.



# Demographic Differences

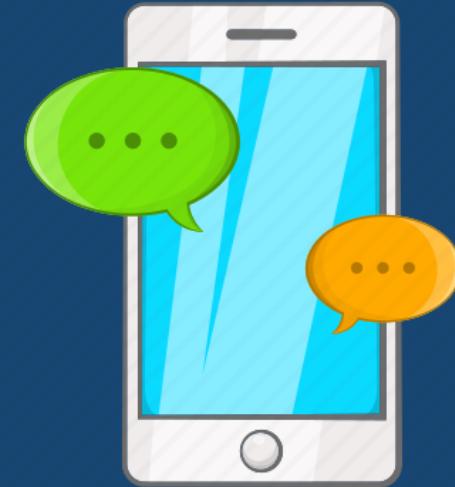
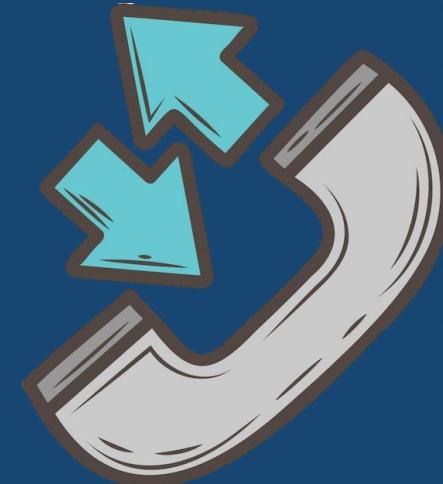
- In urban populations, such as Philadelphia, PA, distrust of the police may result in bystanders **not intervening when someone is experiencing an overdose**.
- In rural areas, limited availability of medication treatment has led to overdose-related deaths **surpassing urban overdose-related deaths** in 2015.
- African Americans and other historically marginalized groups are **less likely** than White people to receive medication, live near a provider of such medications, or continue treatment for opioid use disorder.
- Women have **higher rates** of comorbid depression and anxiety, poorer quality of life, and face more stigma.
- Individuals of a lower SES are **less likely** to be treated or be referred for treatment for opioid use disorder.

# Aims of Study

- Assess if individuals with opioid use disorder from a national sample find personal sensing methods to be **acceptable**.
- Identify important themes from qualitative data to identify **strengths, challenges**, and **barriers** for applying sensing-based treatments to people with OUD.



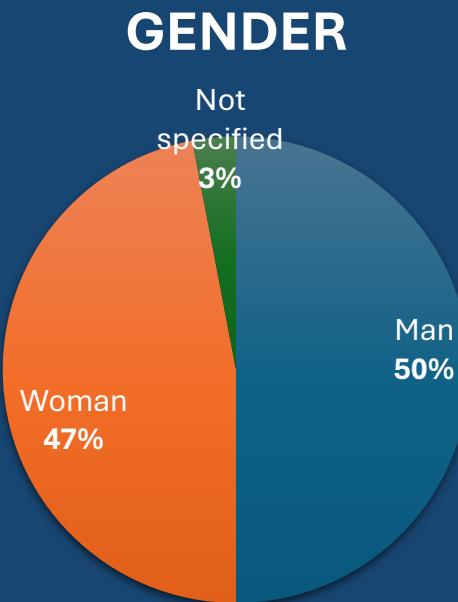
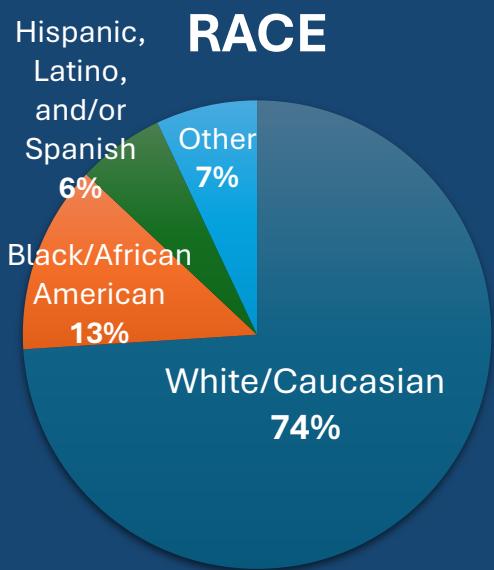
# Procedure



*“Please share any positive or negative comments you have about [Sensing Method].”*



# Study Demographics



Age ranged from 22-65+

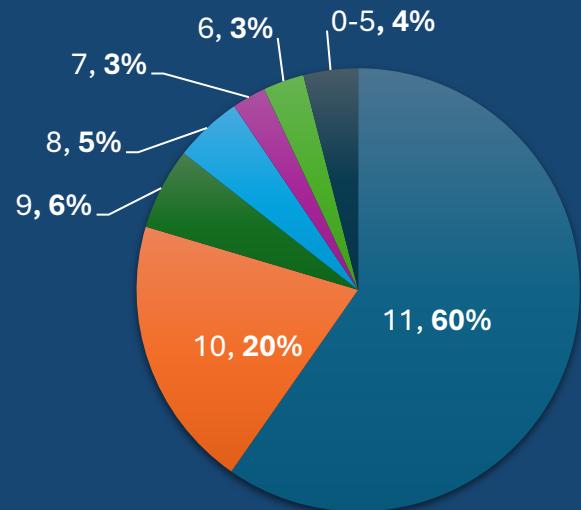
90% of our sample fell between 26 – 55 years old

Participants reside in 38 states across the US

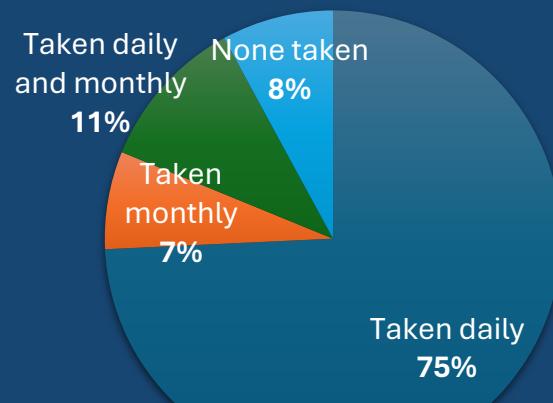
**N = 200**

# Clinical Characteristics

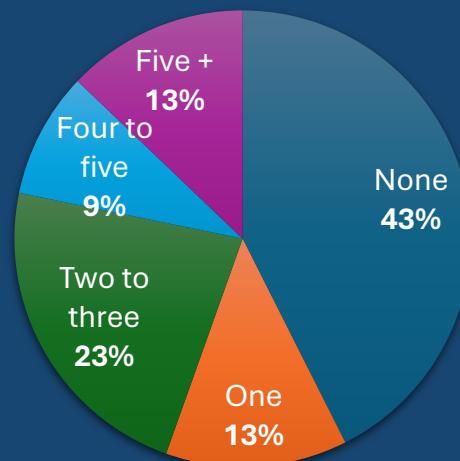
**DSM 5 SYMPTOM SCORE**



**MEDICATION STATUS**

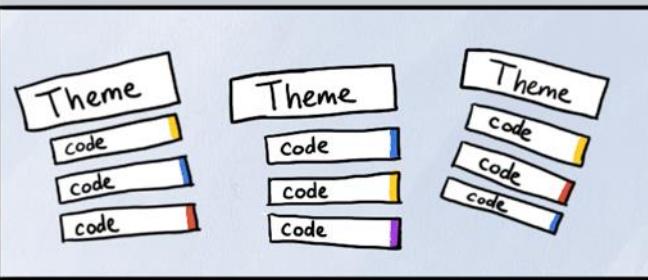


**LIFETIME OVERDOSES**



N = 200

# Methods

Thematic Analysis	Topic Modeling
<ul style="list-style-type: none"><li>• Top-down approach</li><li>• Driven by research aims and questions</li><li>• Deductive codes informed by prior research</li><li>• Inductive codes iteratively created</li></ul> 	<ul style="list-style-type: none"><li>• Bottom-up approach</li><li>• Driven by data</li><li>• Identifies clusters of words that co-occur frequently together.</li><li>• Allows new and unexpected categories to emerge as being important</li></ul> 

# Thematic Analysis

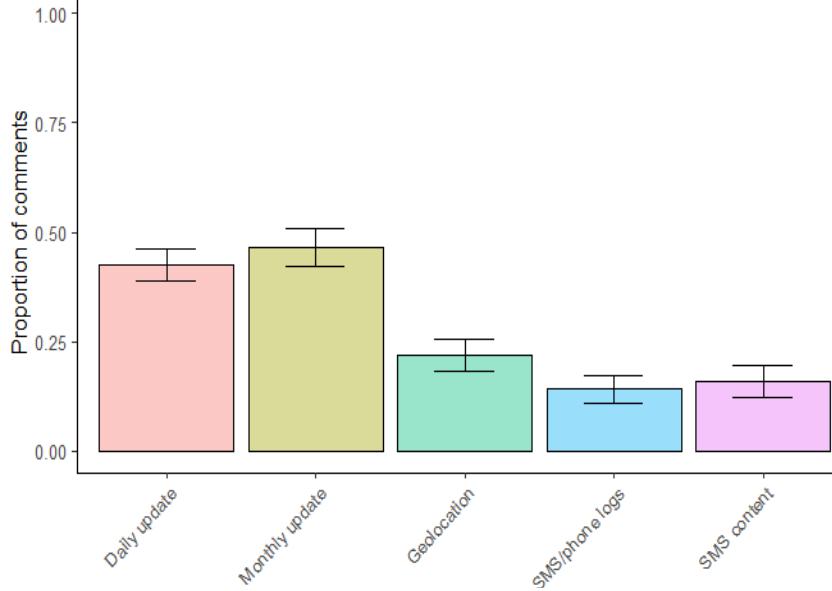
- Total comments: 1,356
- Total comments coded: 647
- Discrepancies were found using R and resolved by coder #1
- Report the overall percentage of comments in each theme and separately by data type
- We then report percentage of comments by data type and demographic characteristic (race/ethnicity, gender, and income)

# Thematic Analysis - Themes

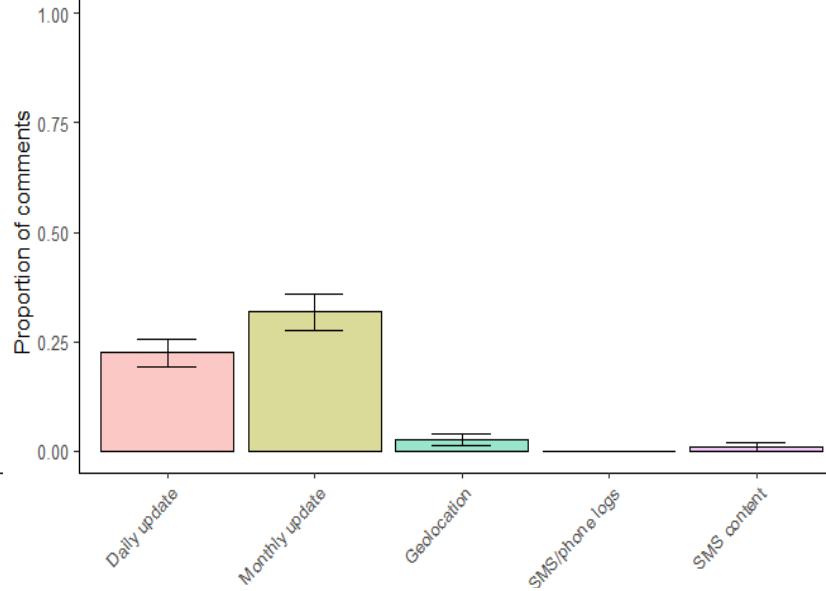
Theme	Comment	% of comments, (n)
Acceptability 	“It’s a good tool that makes me look back and reflect on how I’m doing in my recovery.”	30.14%, (195)
Sustainability	“I have worked this daily update into my routine.”	13.14%, (85)
Benefits 	“It helps me stay sober.”	22.87%, (148)
Trust/Privacy	“It’s a bit of an invasion of privacy but I understand why it’s needed for the study.”	15.45%, (100)
Usability 	“I think it’s great and very precise about the information you asked.”	15.76%, (102)
Feedback	“It would be nice to explain certain things that caused me to use.”	51.93%, (336)

# Proportion of Comments by Data Type - Themes

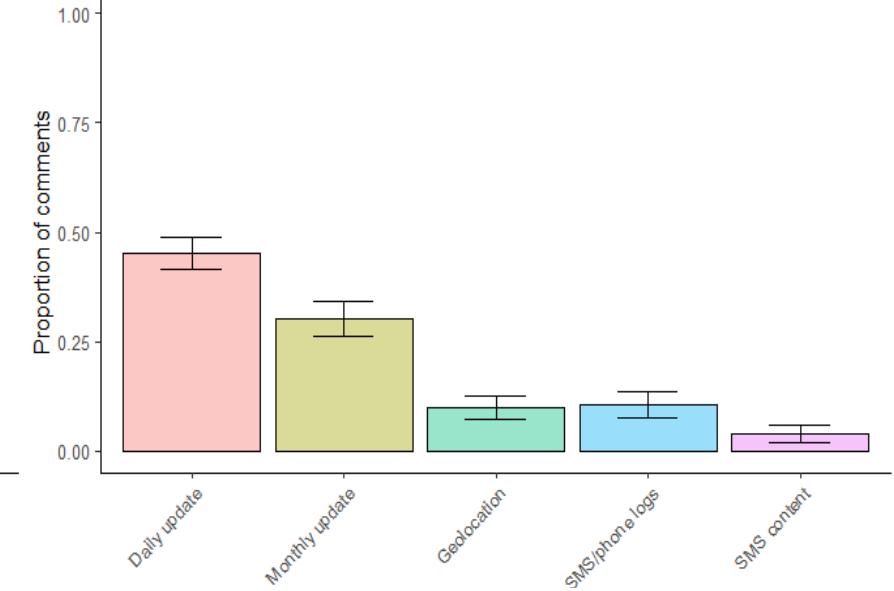
Proportion of Acceptability Comments by Data Type



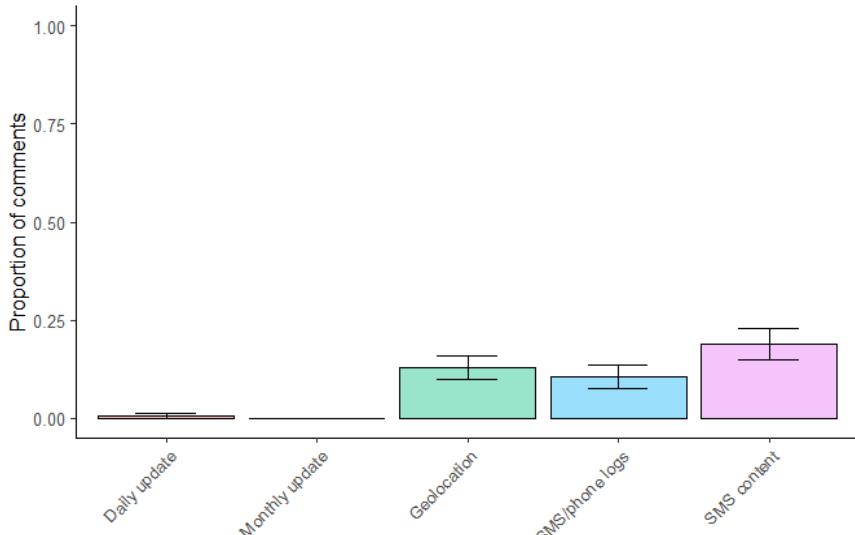
Proportion of Sustainability Comments by Data Type



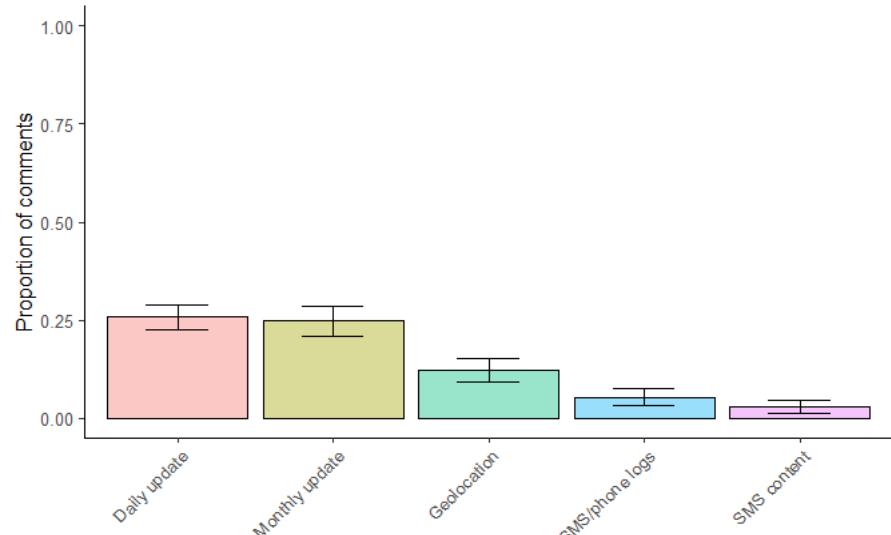
Proportion of Benefits Comments by Data Type



Proportion of Trust Comments by Data Type



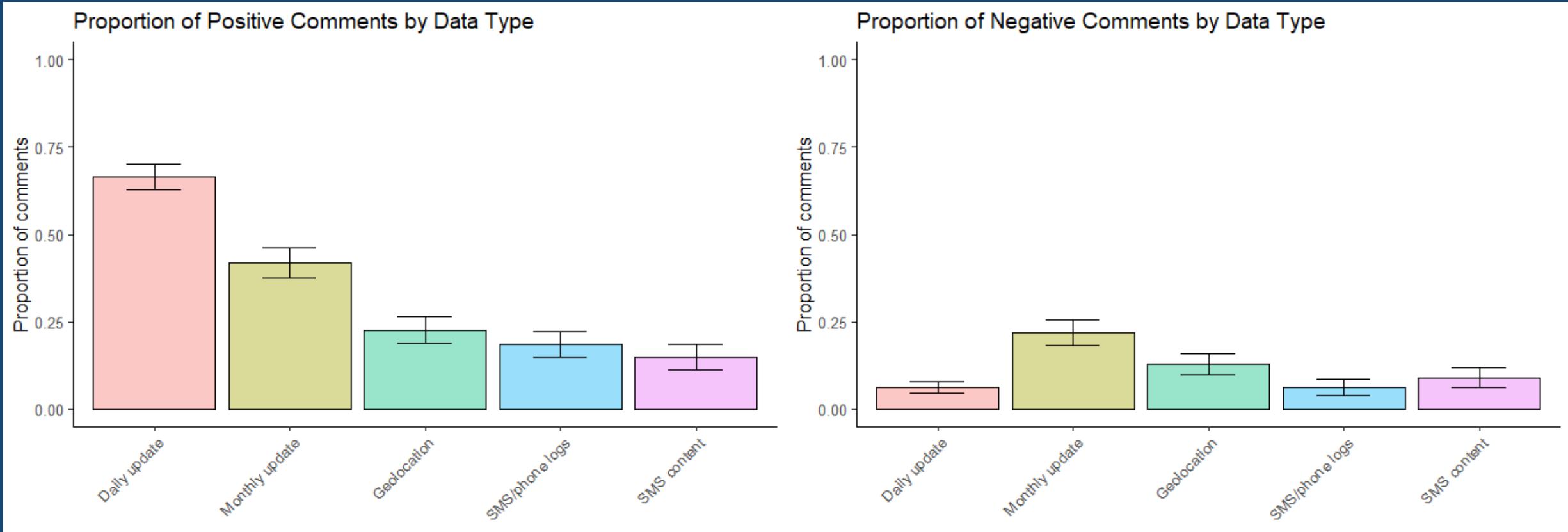
Proportion of Usability Comments by Data Type



# Thematic Analysis - Affect

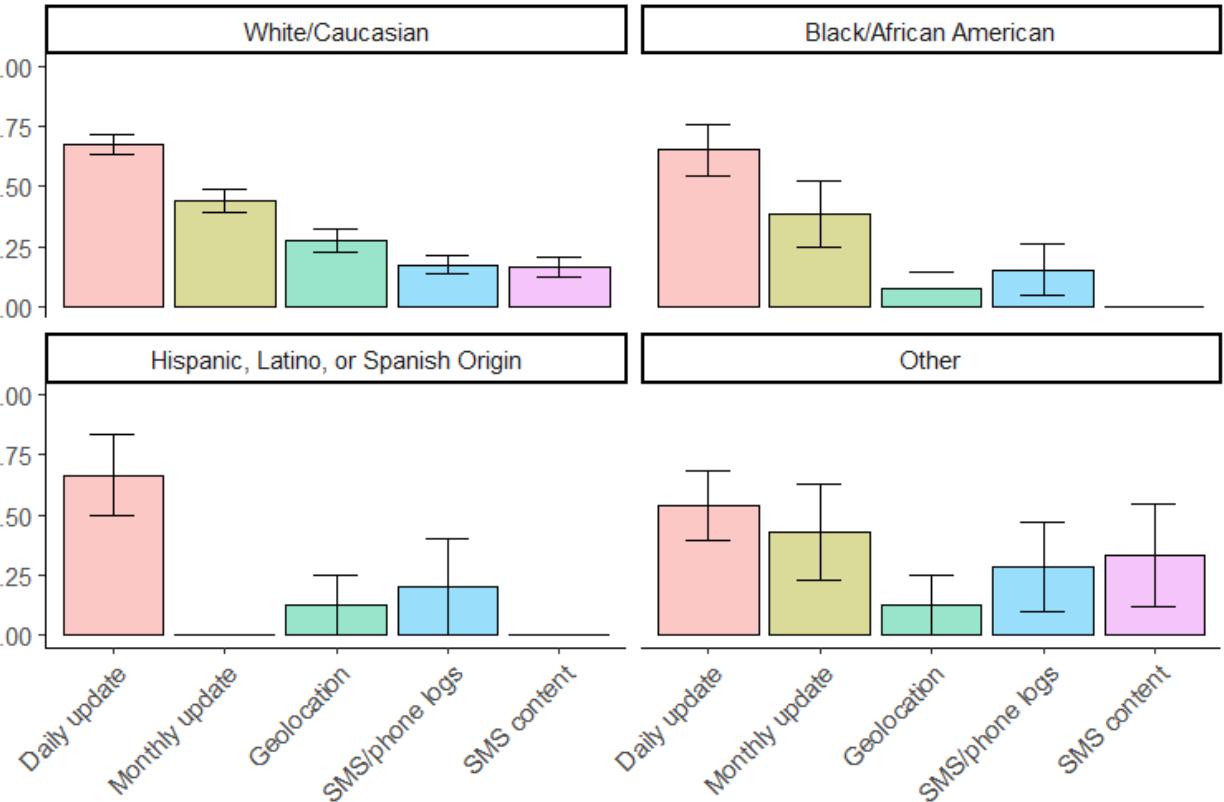
Affect	Comment	% of comments, (n)
Positive	 “The program and daily updates have helped my recovery and helped me to evaluate my emotions and actions.”	36.63%, (237)
Negative	“I just disliked how it drained the battery on my phone so quickly.”	10.97%, (71)
Neutral/mixed	“I don’t mind providing this information.”	52.24%, (338)

# Proportion of Comments by Data Type - Affect

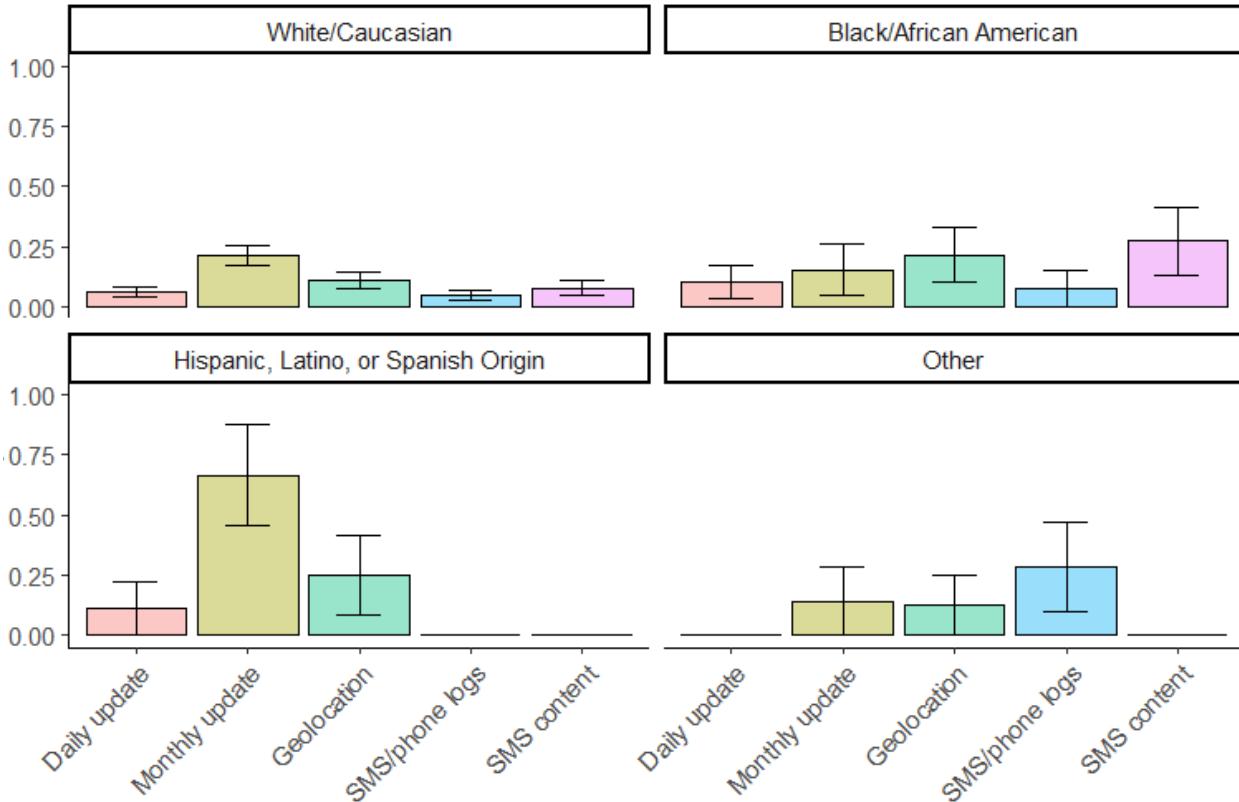


# Demographic Differences in Themes - Race

Proportion of Positive Comments by Race/Ethnicity

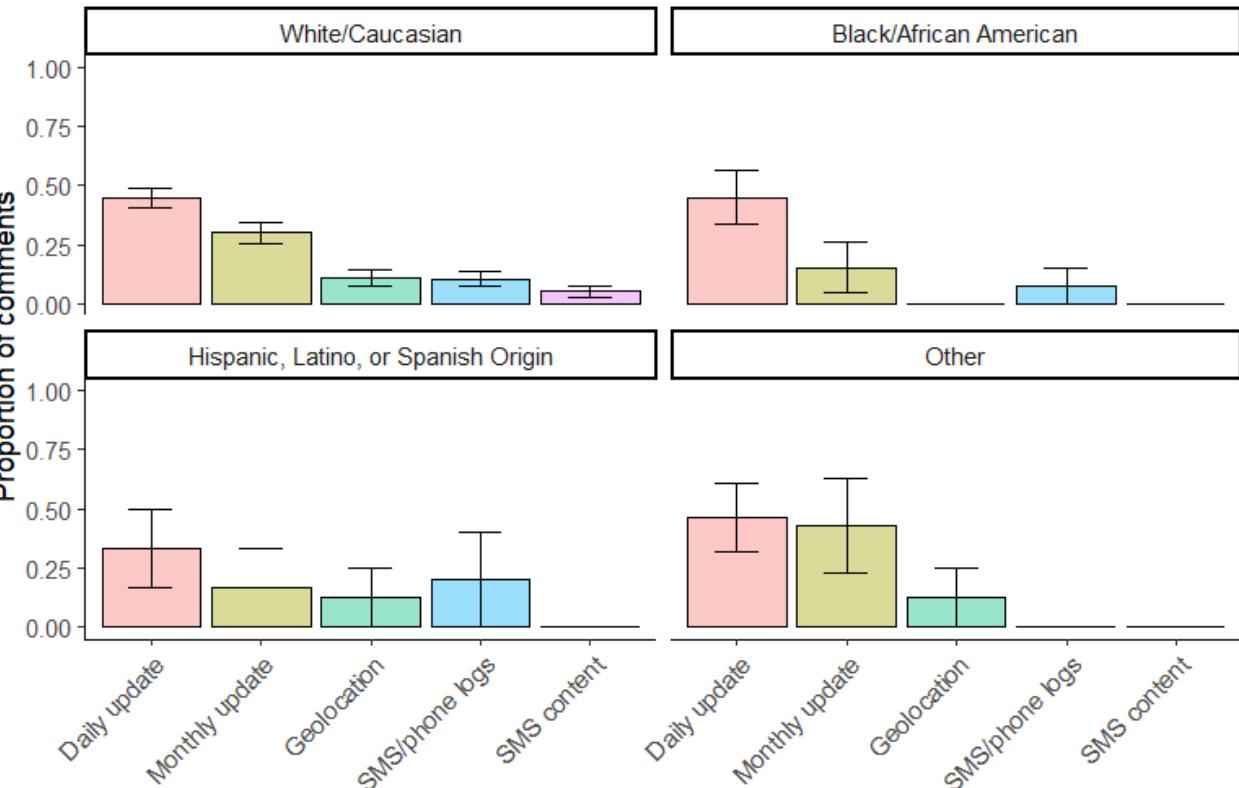


Proportion of Negative Comments by Race/Ethnicity

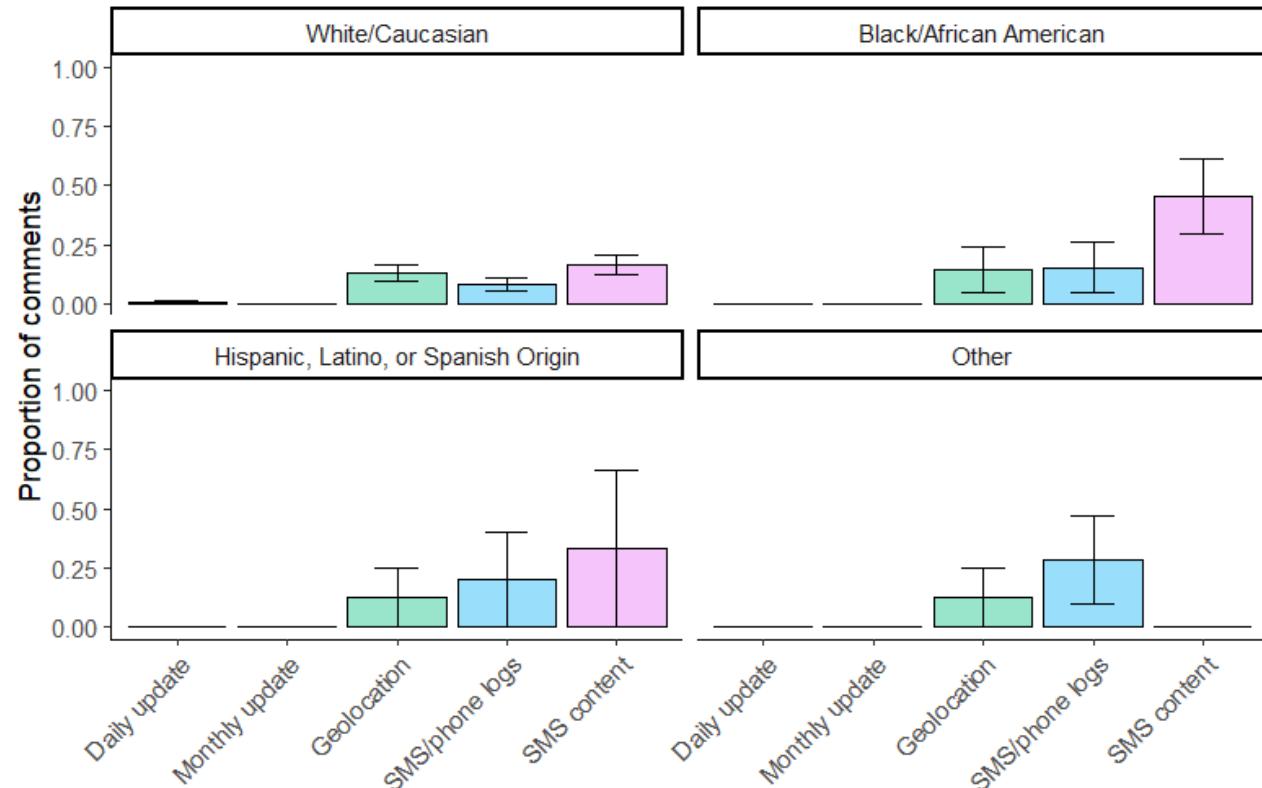


# Demographic Differences in Themes - Race

Proportion of Benefits Comments by Race/Ethnicity

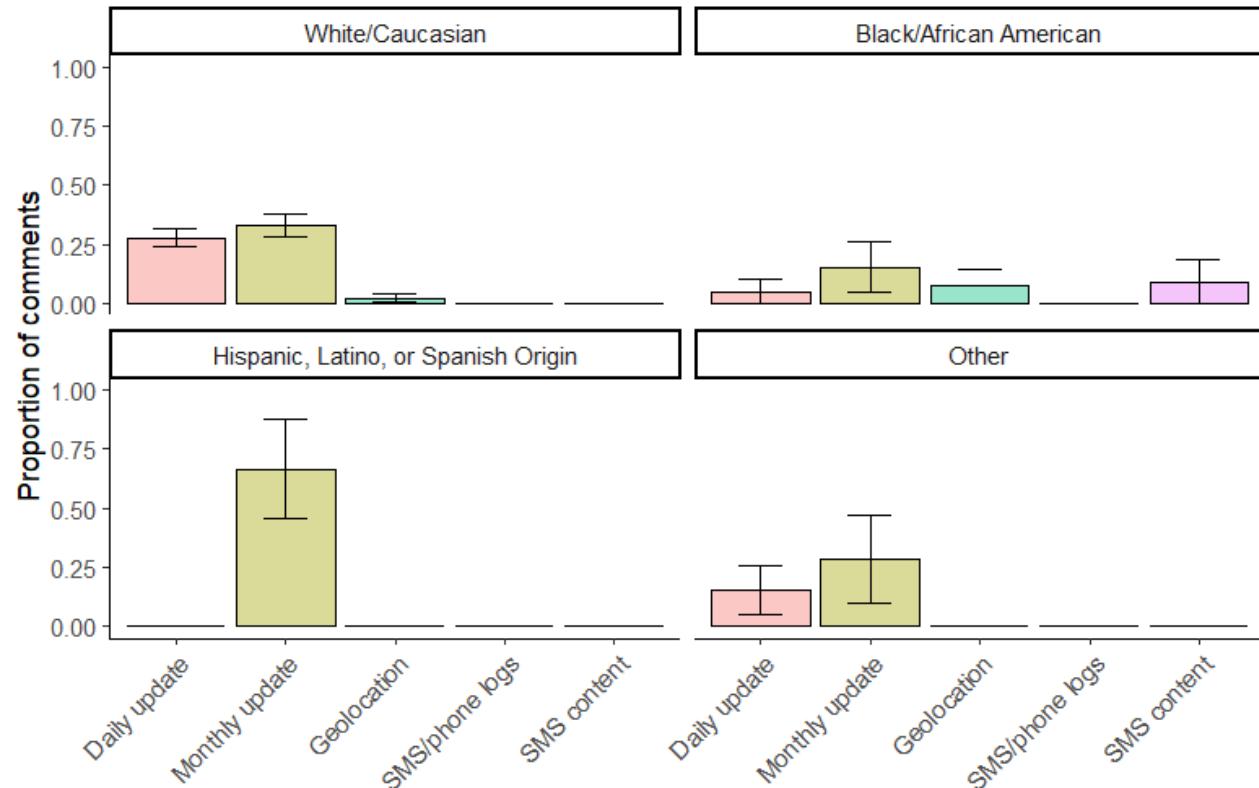


Proportion of Trust Comments by Race/Ethnicity

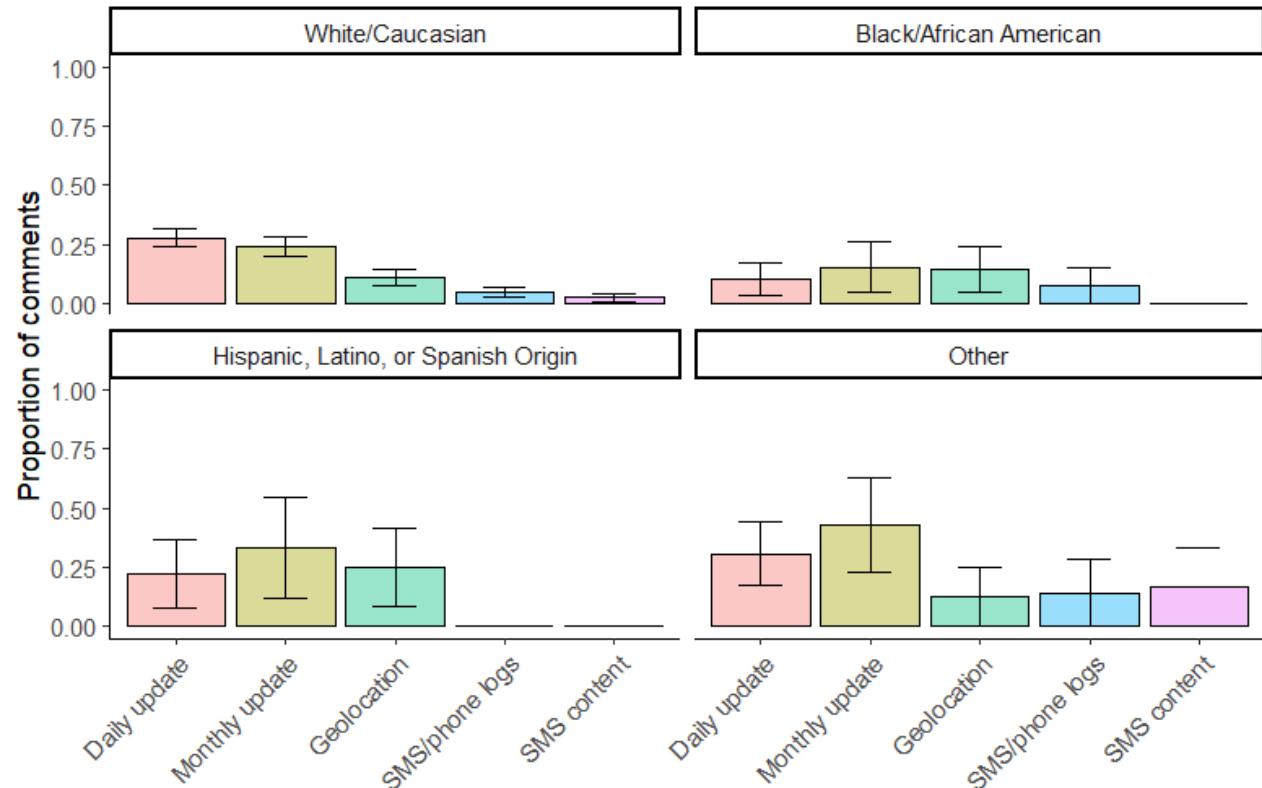


# Demographic Differences in Themes - Race

Proportion of Sustainability Comments by Race/Ethnicity

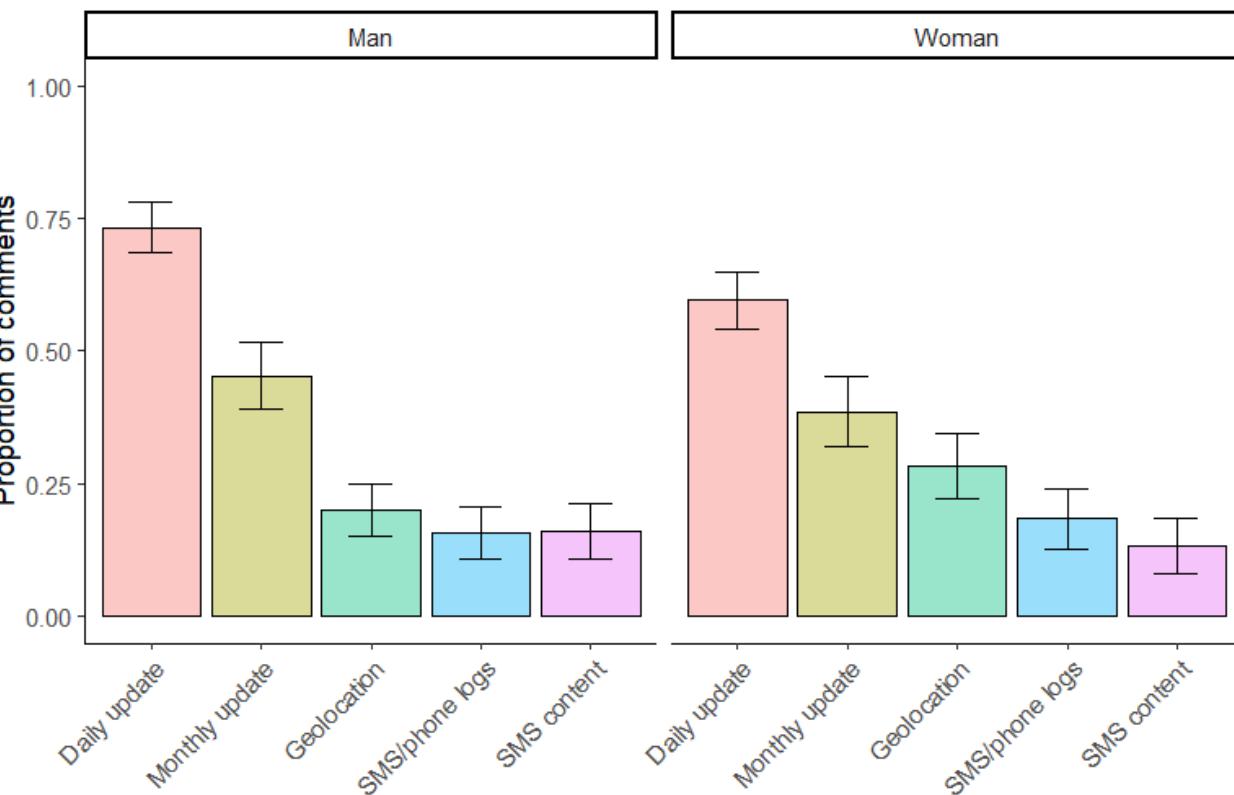


Proportion of Usability Comments by Race/Ethnicity

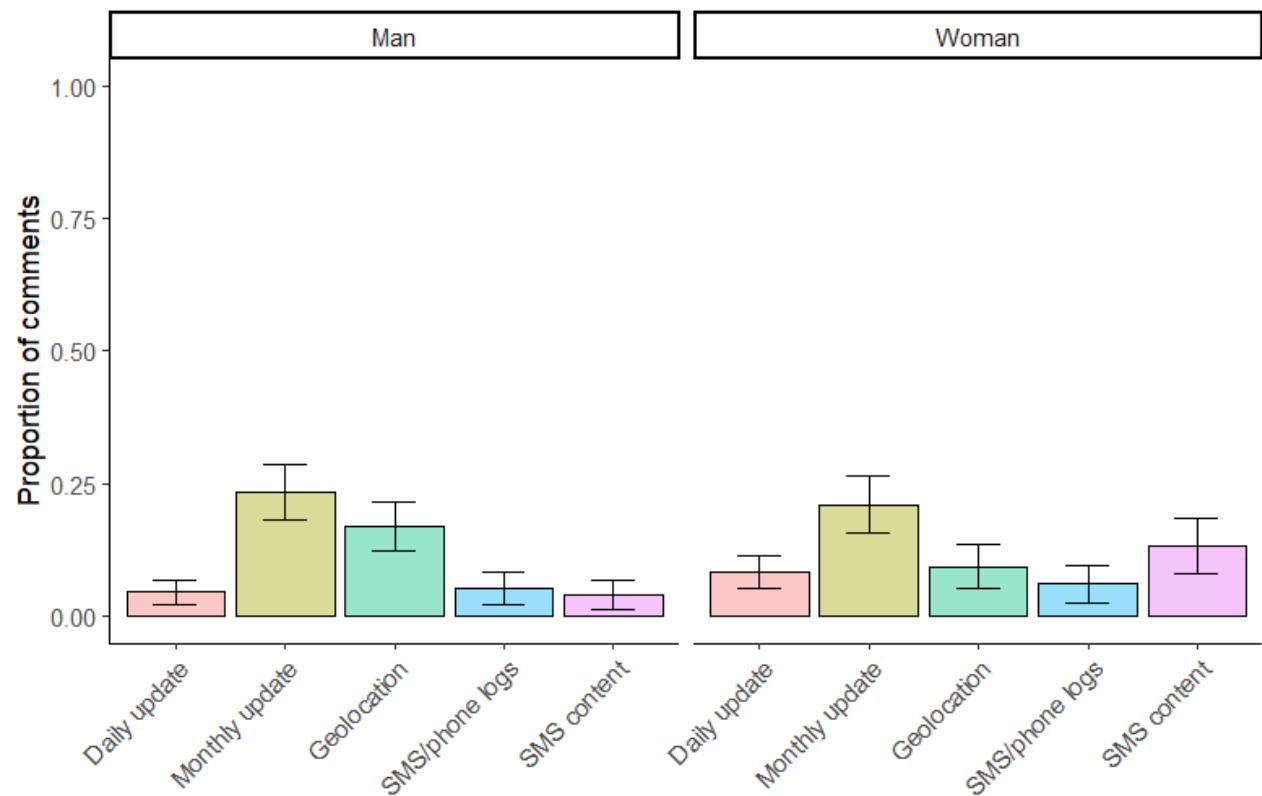


# Demographic Differences in Themes - Gender

Proportion of Positive Comments by Gender

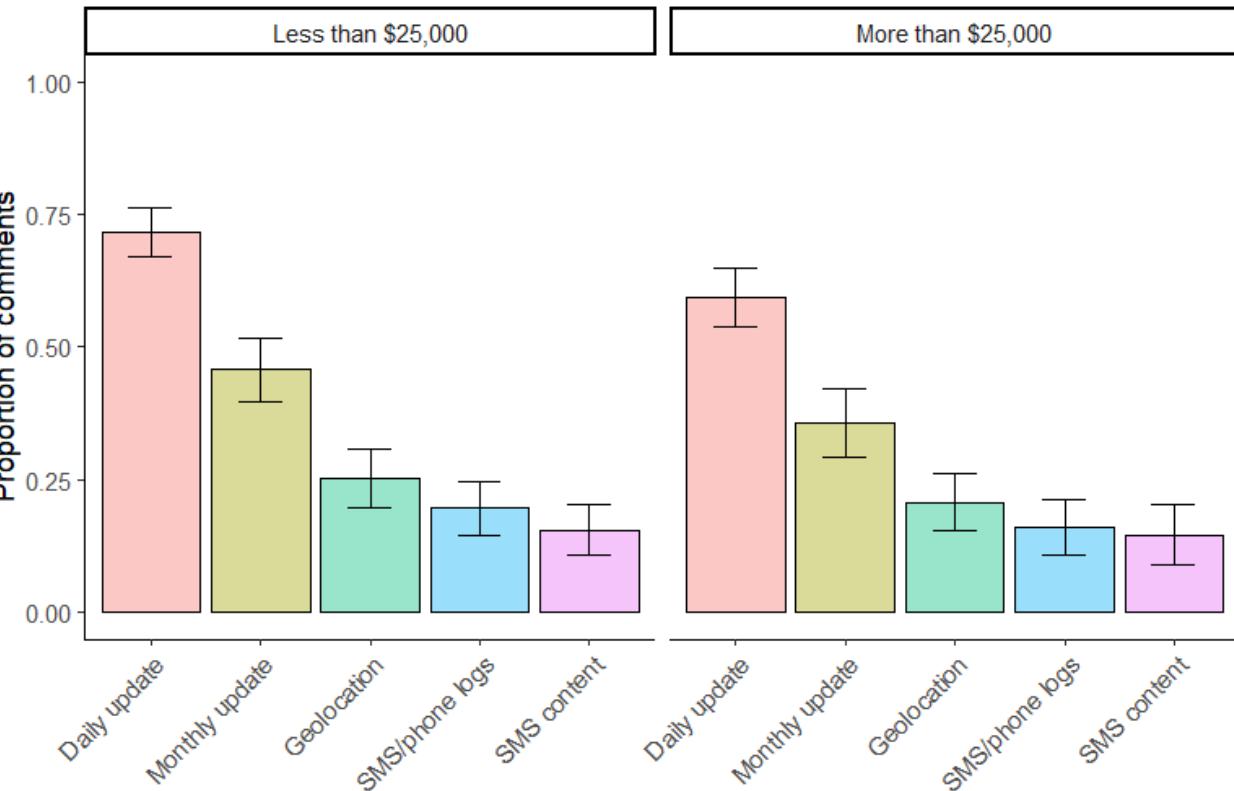


Proportion of Negative Comments by Gender

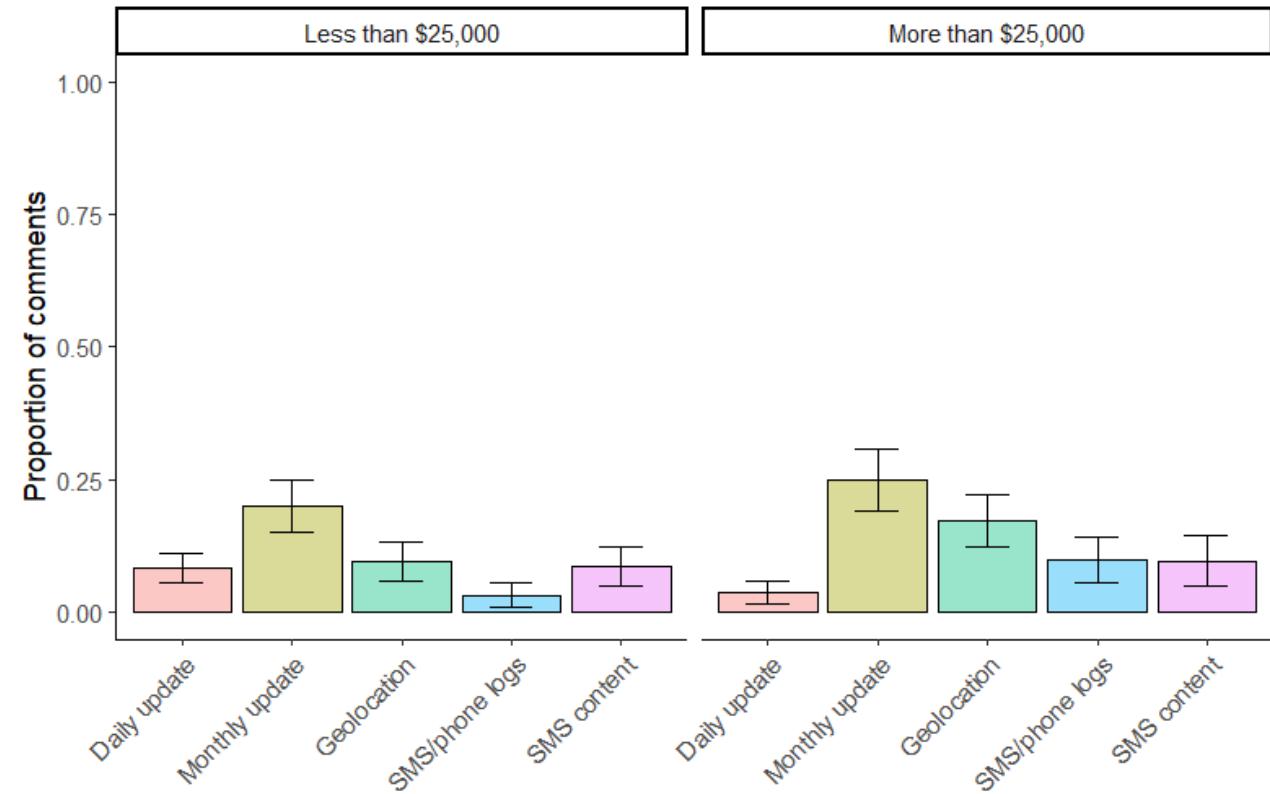


# Demographic Differences in Themes - Income

Proportion of Positive Comments by Income



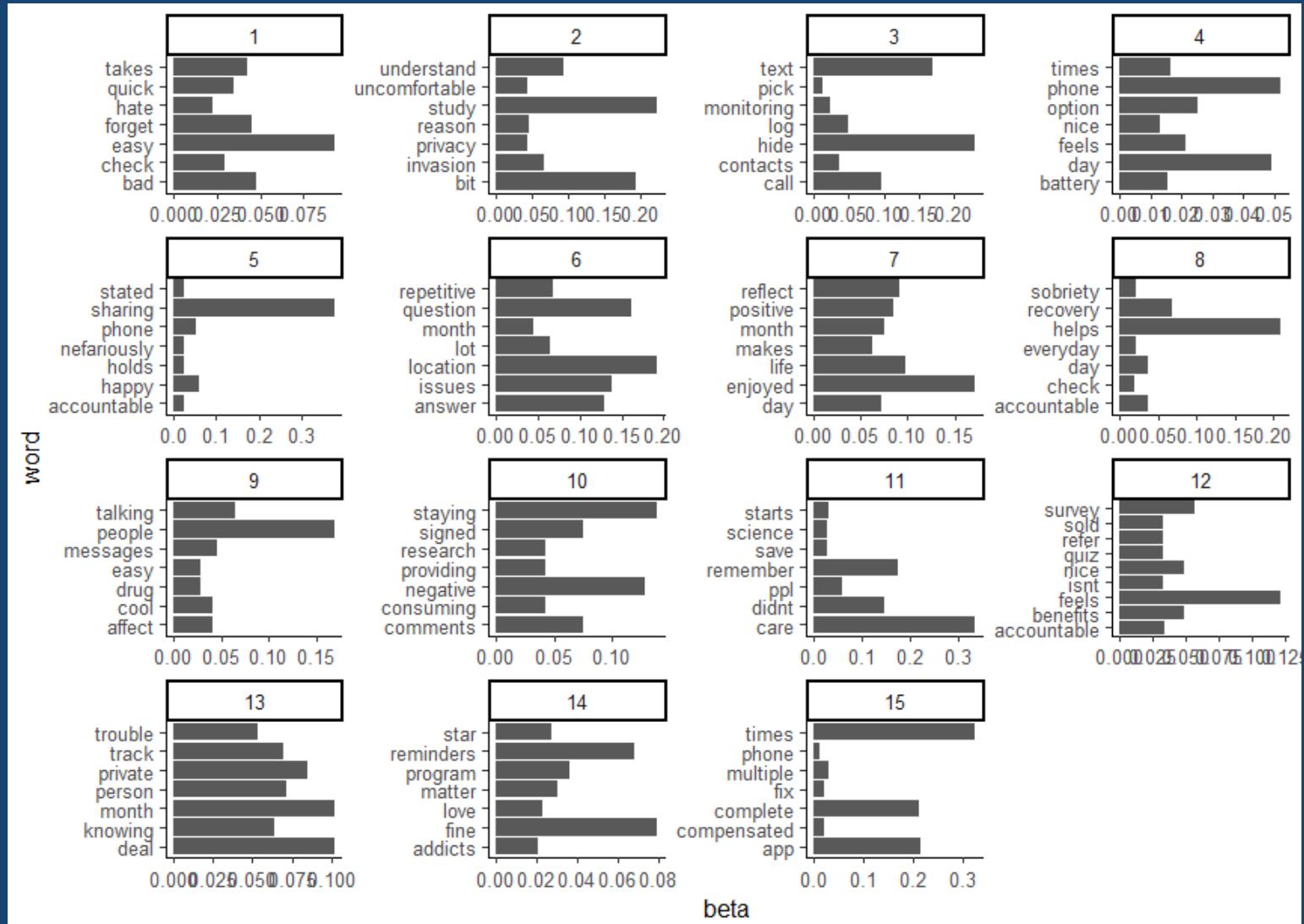
Proportion of Negative Comments by Income



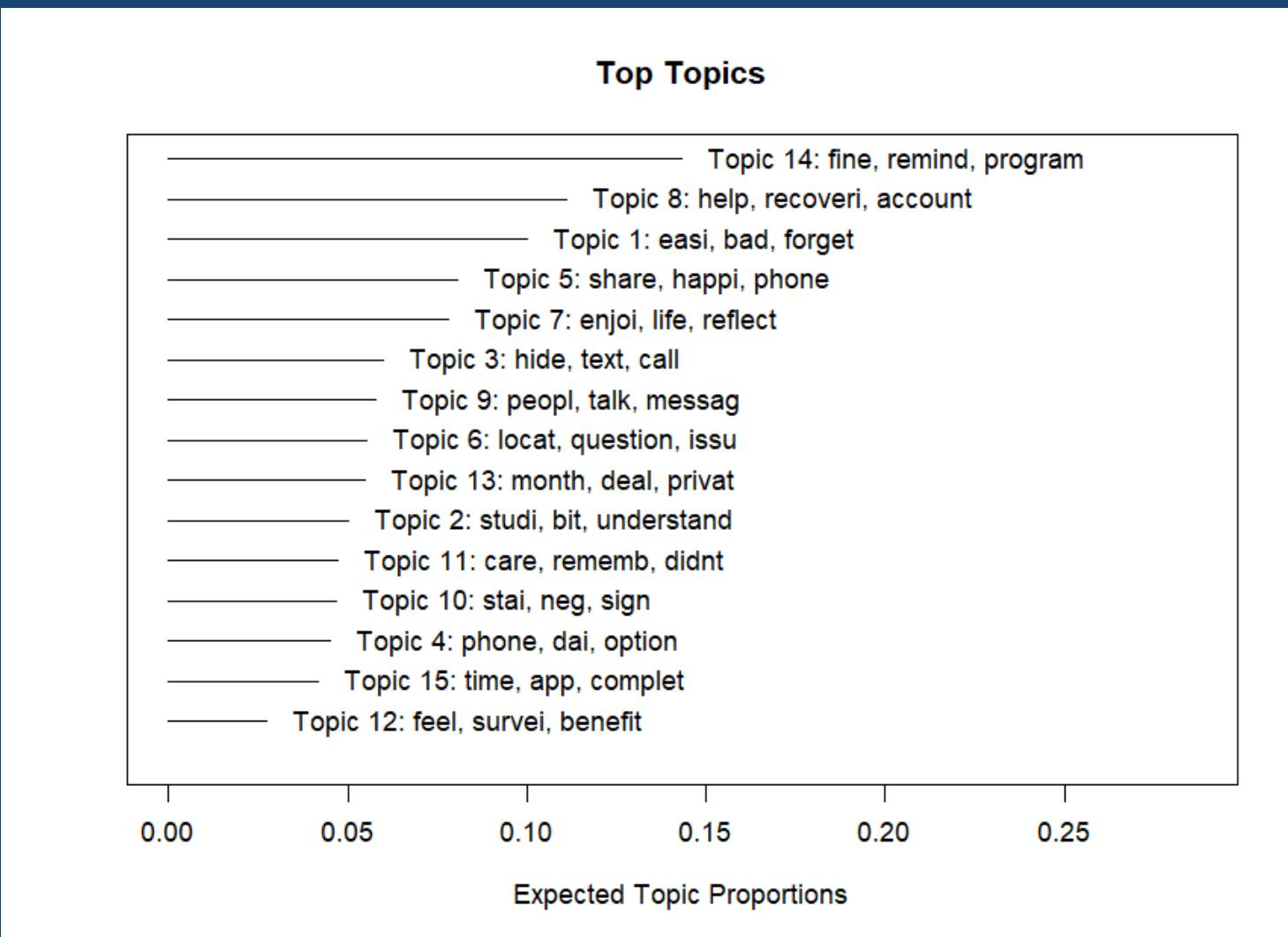
# Topic Modeling

- Fit a structural topic modeling controlling for covariates of race and data type
- We looked at the top words defined by FREX, a metric that evaluates word frequency and exclusivity to a topic
- We found that there were 15 unique topics

# Identified Topics

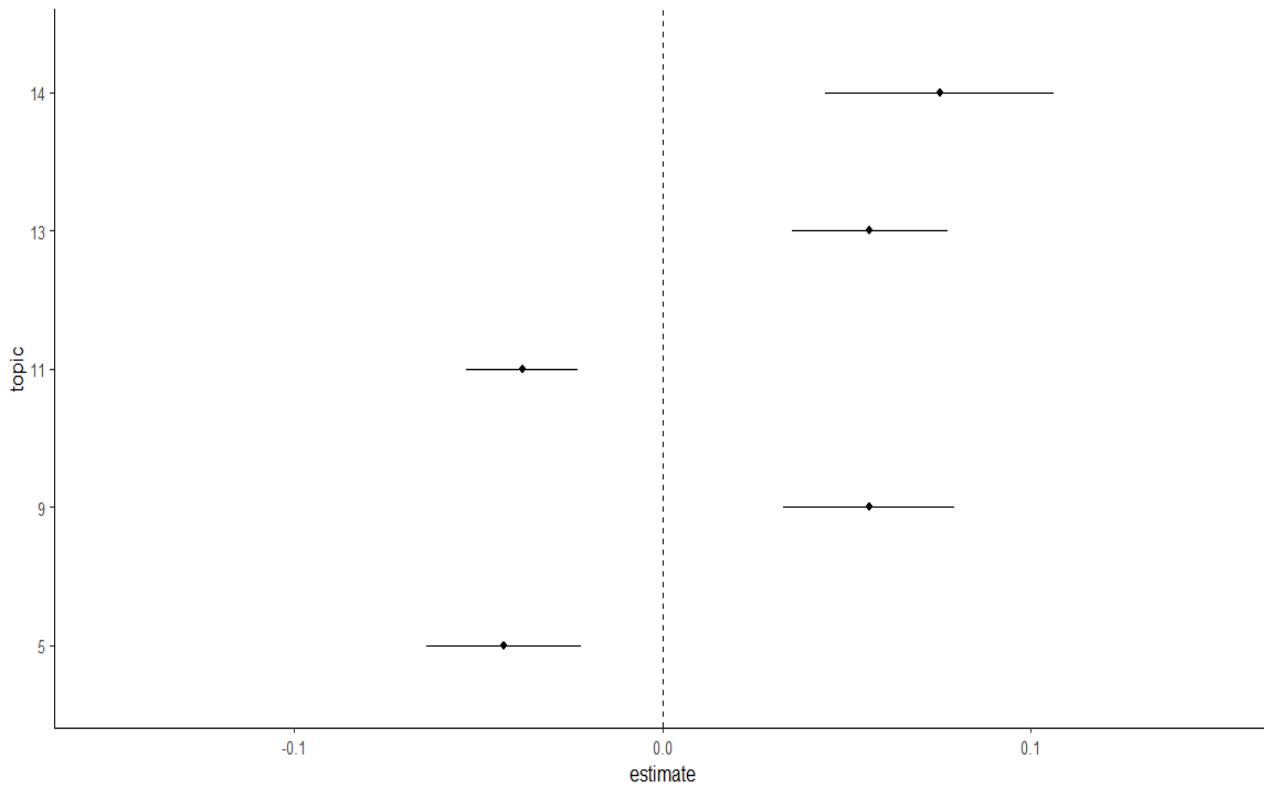


# Proportion of Topics

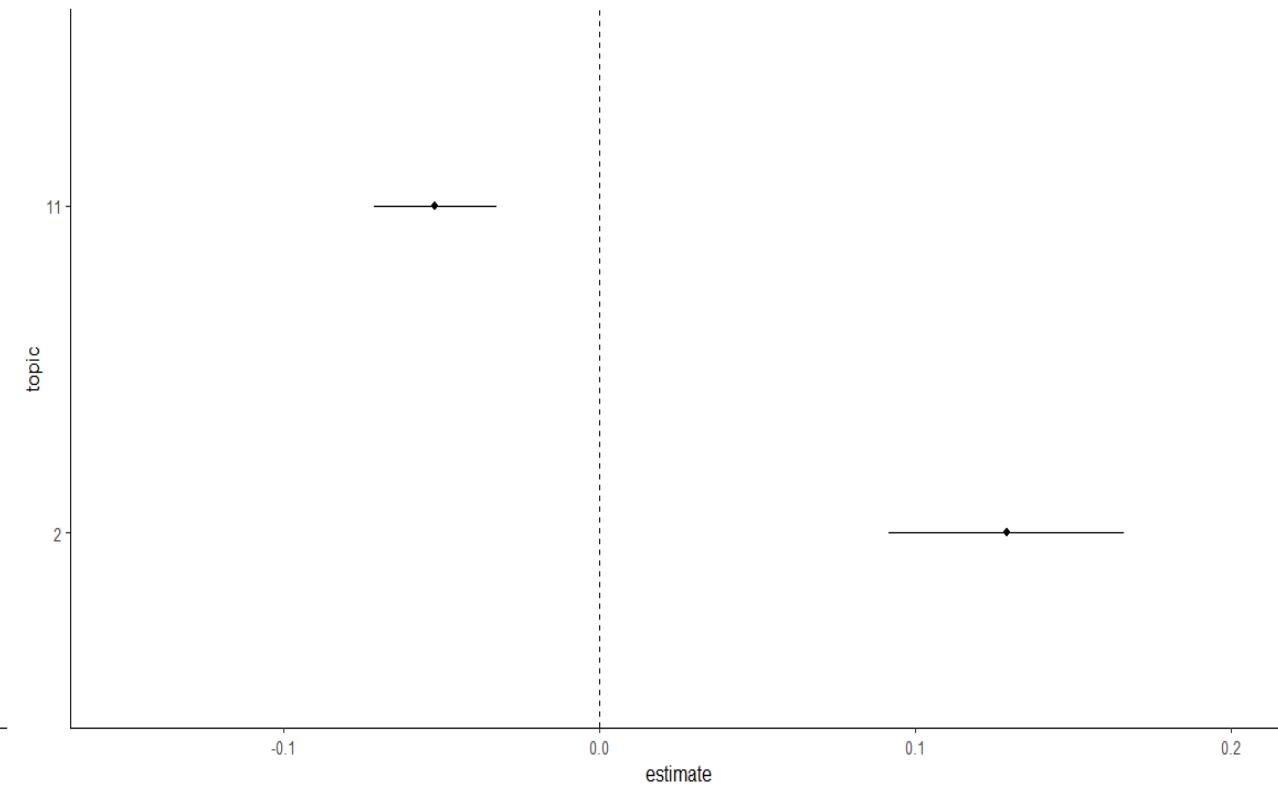


# Demographic Differences in Themes - Race

Difference in proportion of topics among Black compared to White participants



Difference in proportion of topics among Hispanic compared to White participants



# Implications

- Participants found these methods to be **acceptable** and saw **benefits** from using them
- Participants largely reported **positively** about these methods
- Acceptability of personal sensing types **varies** across **racial/ethnic groups**, but is relatively the same across gender and income
- Digital therapeutics could be a **viable treatment method** for those with opioid use disorder

# Future Directions

- End-of-study survey
  - Gauging participant reactions after using personal sensing methods for a full year
- Location data
  - Looking at if participants live in a rural, suburban, or urban setting
- Linguistic Inventory and Word Count (LIWC)
  - Top-down approach that analyzes text based on predefined linguistic categories

# Thank You!

John Curtin and Kendra Wyant  
Addiction Research Center

Brad Postle and Sarah Sant'ana  
PREP Faculty and Mentors

# Questions?

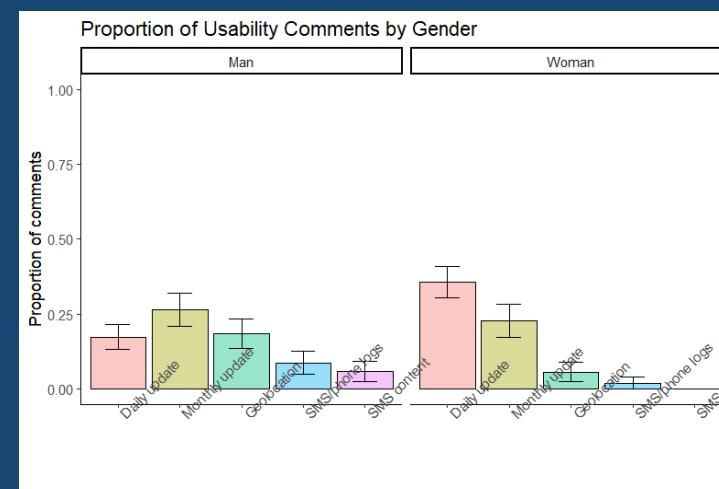
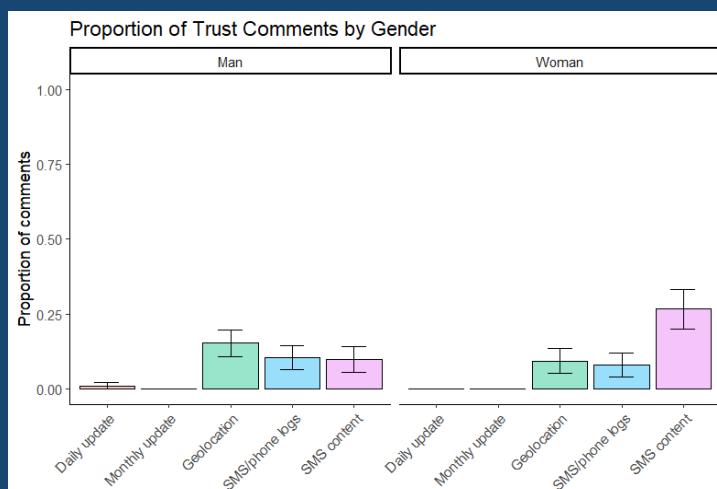
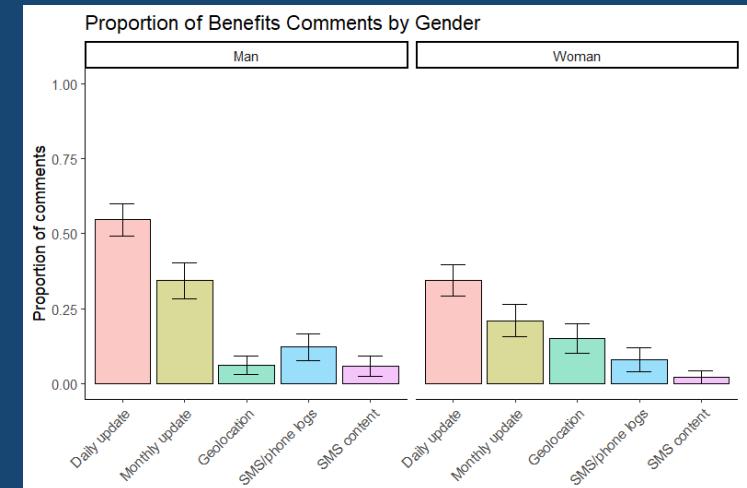
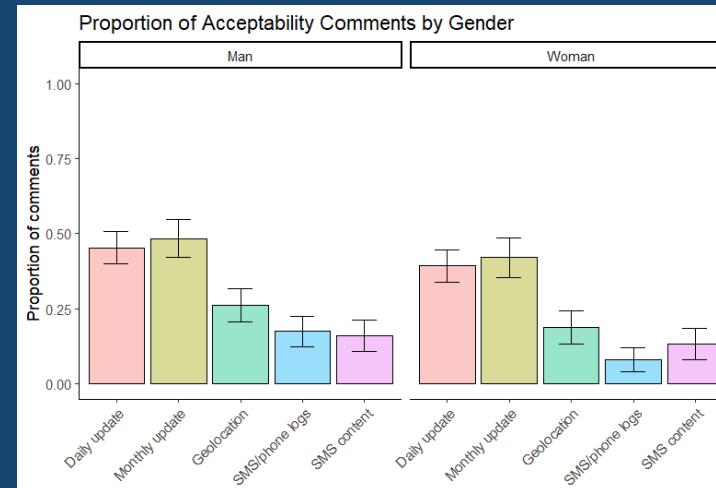
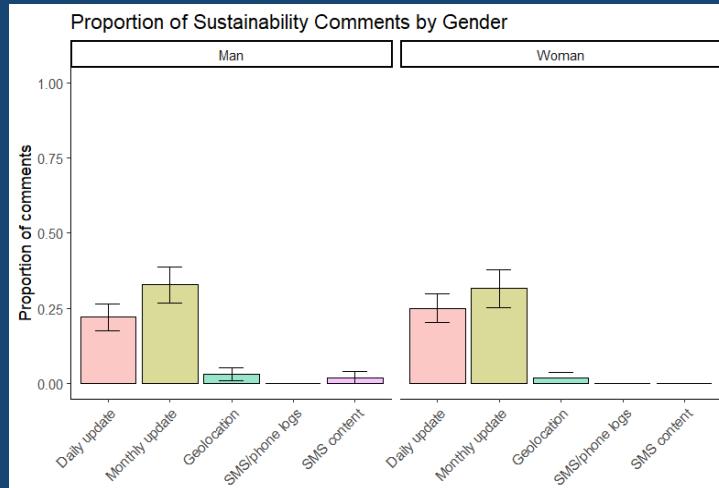


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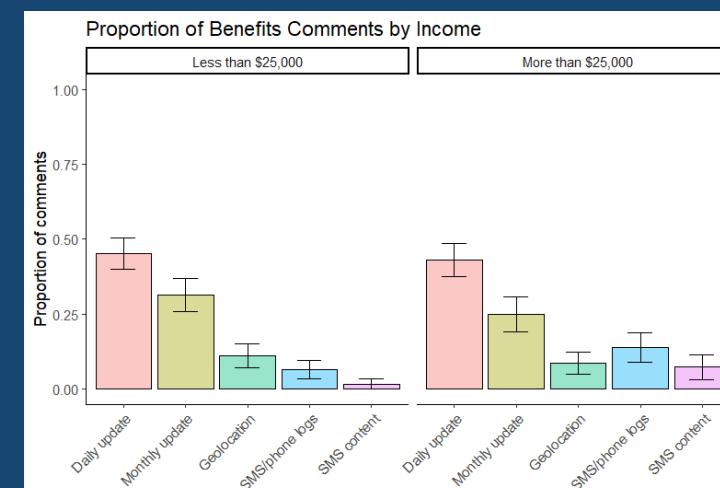
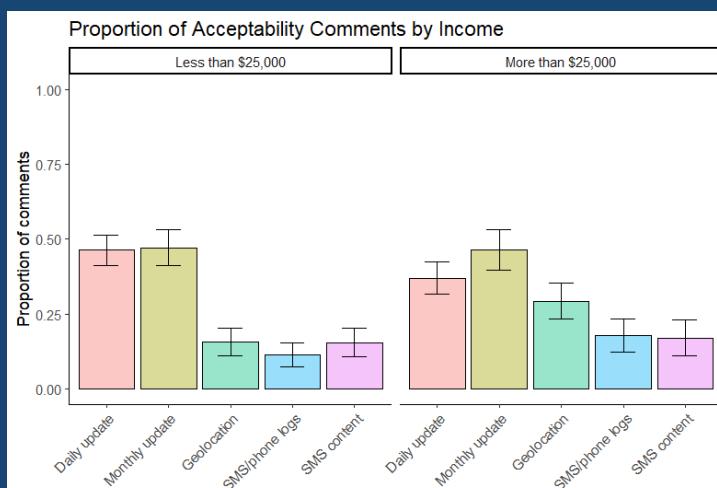
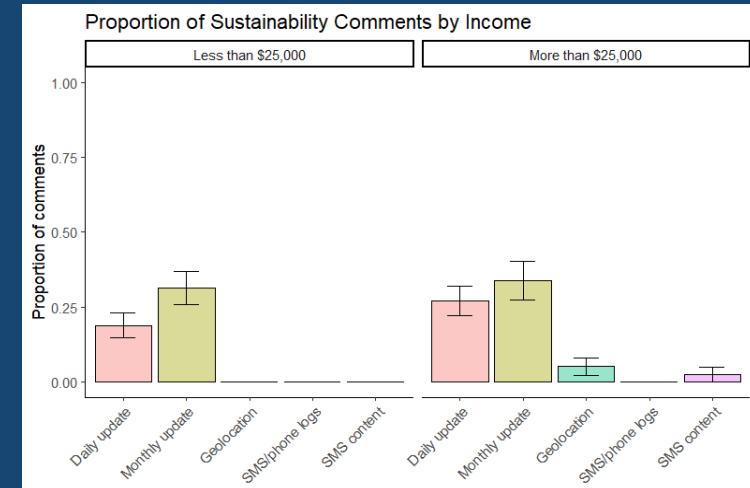
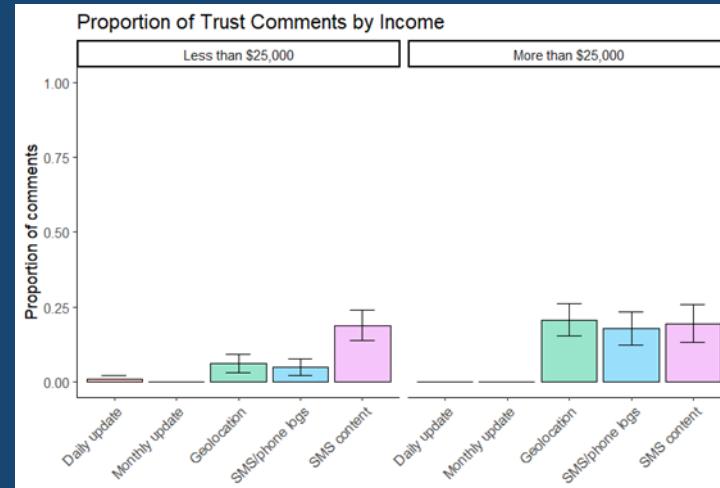
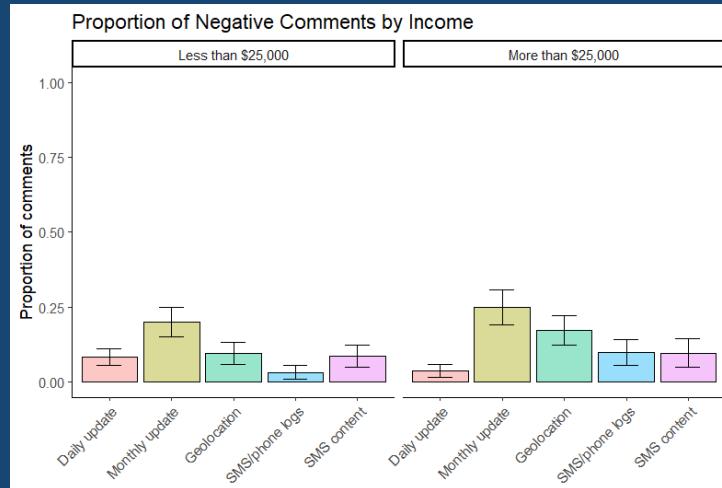
# Thematic Analysis – Additional Themes

Theme	Comment	% of comments, (n)
Getting in trouble	“Idk. Don't wish to get anyone in trouble.”	0.77%, (5)
Data already tracked	“Everyone is being tracked.”	1.08%, (7)
Research/Study/Science	“Nothing to say its part of the program its not hard to do.”	5.87%, (38)
Honesty	“Nothing to hide here.”	4.79%, (31)

# Demographic Differences in Themes - Gender



# Demographic Differences in Themes - Income



# Feedback Notes