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# Evolving views of transit service: Inter-temporal analysis of safety interventions and satisfaction

Joint work with **Spencer Aeschliman** (PhD student, NSF GRFP fellow). Collaboration with the Regional Transportation Authority, Amy Hofstra and Peter Fahrenwald



Graduate Research Fellowship Program

## Context-changes

### \* Schedule/activities

Accelerated e-commerce  
Lasting telework/hybrid

## Medium-Long term dynamics

Public transport ridership

Safety & Health

Residential location

Business locations

Mobility business models

Public transport service & funding

Social equity & DEI

Polarization & distrust in authorities

Urban decentralization / depopulation / renewal

## Pandemic shock

Stress & Isolation



Economic decline

Active mobility

Social distancing

Transit abandonment

### \* People

Values and priorities  
Equity and climate  
Mobility changes  
Localization preferences

### \* Employers/Businesses

Hybrid policies

### \* Technology

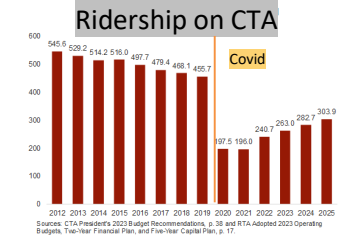
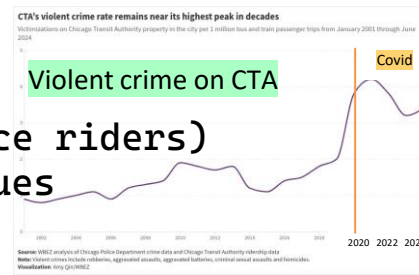
Tele-everything  
Hybrid substitution

### \* Society/Governance/Politics

Low trust in gov & 'experts'  
Economic stress  
Electrification, Climate policy

# Setting the scene

- Transit fiscal cliff
- Lapsed ridership (choice riders)
- WFH and Safety key issues

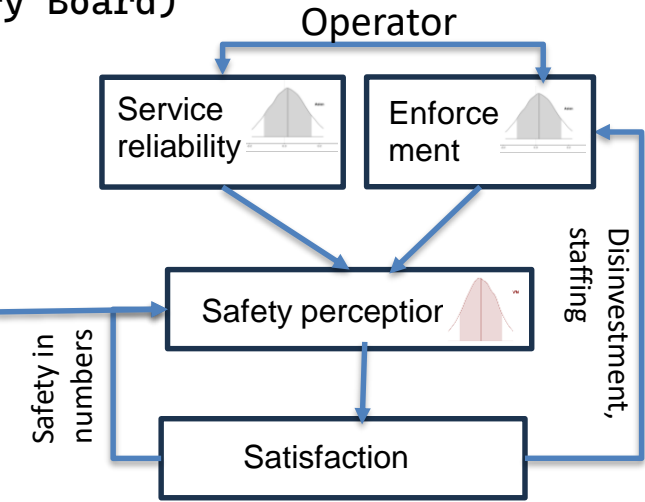
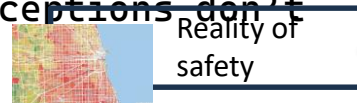


# Safety & transit

- “Safety is kind of a hard issue because how safe people feel is just as important, sometimes more important, than actual safety,” (CTA’s Citizens Advisory Board)

# A complex issue

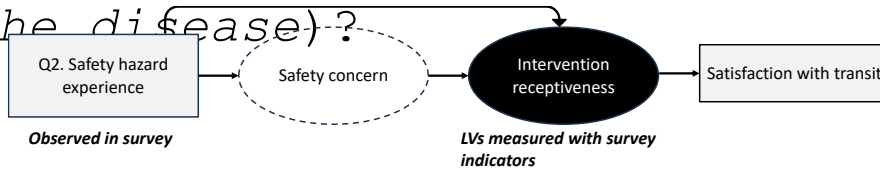
- Different views of safety
- Different views of safety measures
- Balance interventions
- Reality and Perceptions don't match



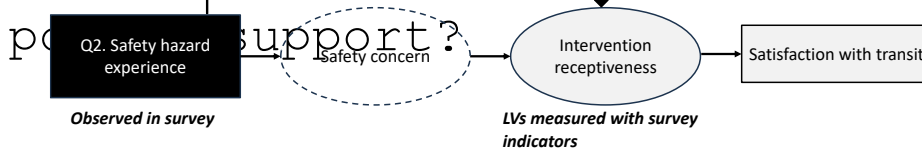
# Modeling evolving transit satisfaction dynamics? 3 investigations

## • 3 avenues to capture evolving behavior

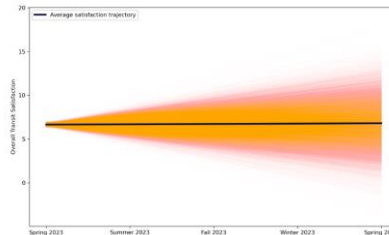
Q1: How can we model policy trade-offs in a divided society (*Is the cure worse than the disease*)?



Q2: Do safety hazard experiences shape



Q3: How do we capture at trajectories (Variation noise)?



## Data

- Data for Q1 & Q2
- Shared by the RTA
- Spring '23
- #2292 transit riders
- Focus on service, safety & enforcement
- Service Satisfaction
- Responder profile

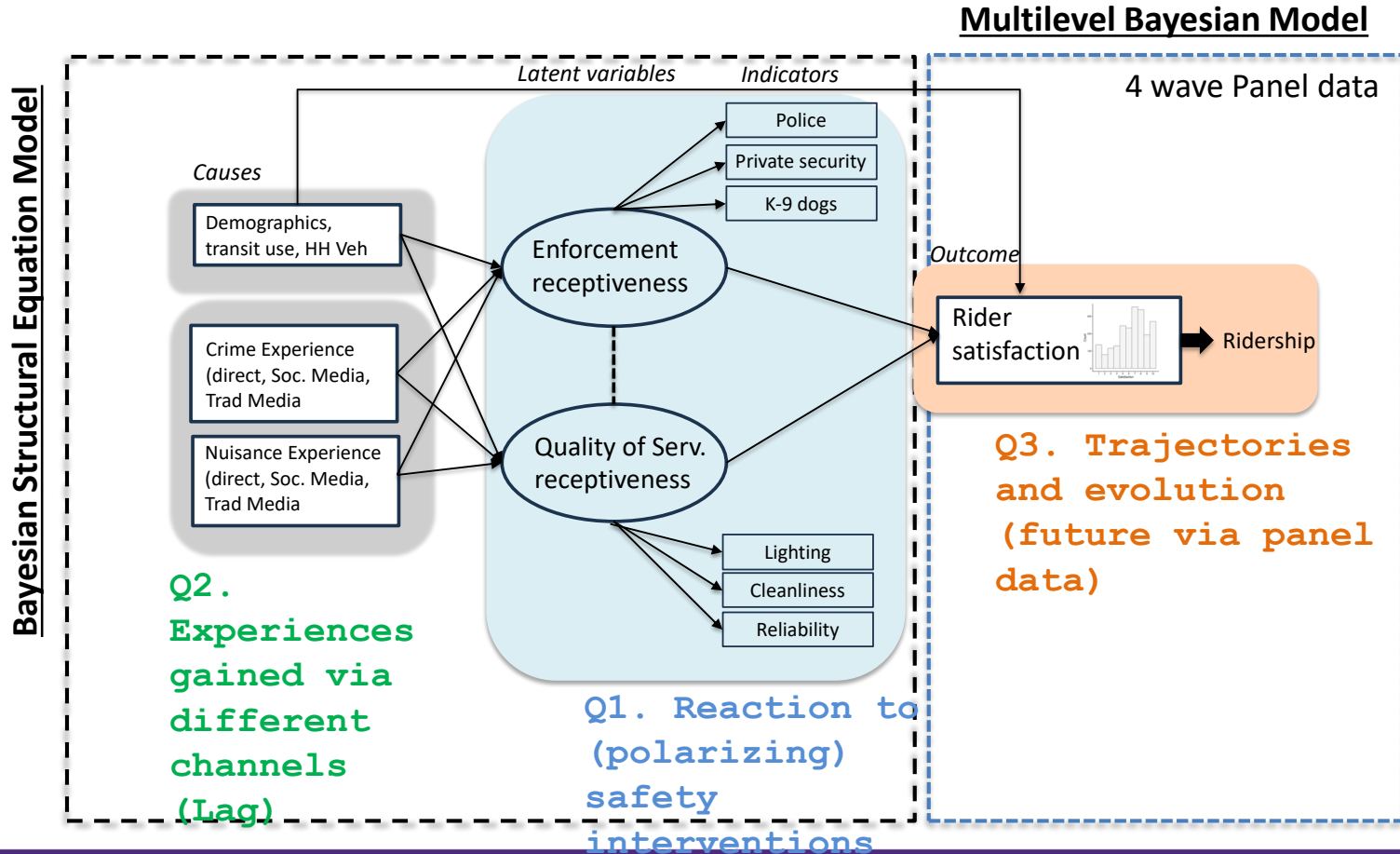
- Panel Data for Q3
- 4 waves of panel data spring '23 - spring '24
- 460 respondents in all 4 waves
- Satisfaction focus

# Q1. Is the cure worse than the disease?



Figure 1: Rider responses to potential safety interventions

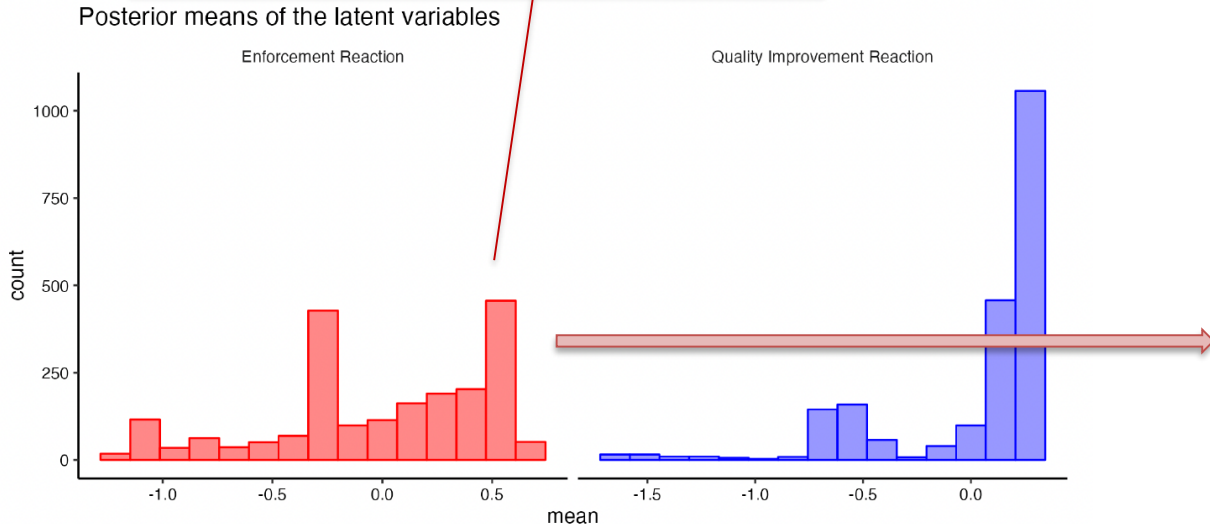
# Research Methods



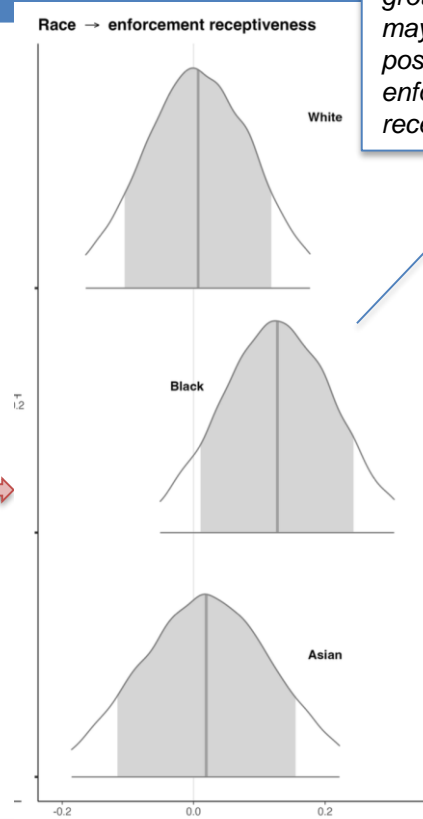
# Results Q1: Safety measures divisiveness (estimation)

*Difference by ethnic groups: Black resp. may have a slightly positive enforcement receptiveness*

*'Enforcement' has wider spread of values*



Histogram of individual means of sampled latent variable values, with 3000 draws for each ind. in the sample. Higher values → more positive reactions to transit safety measures.

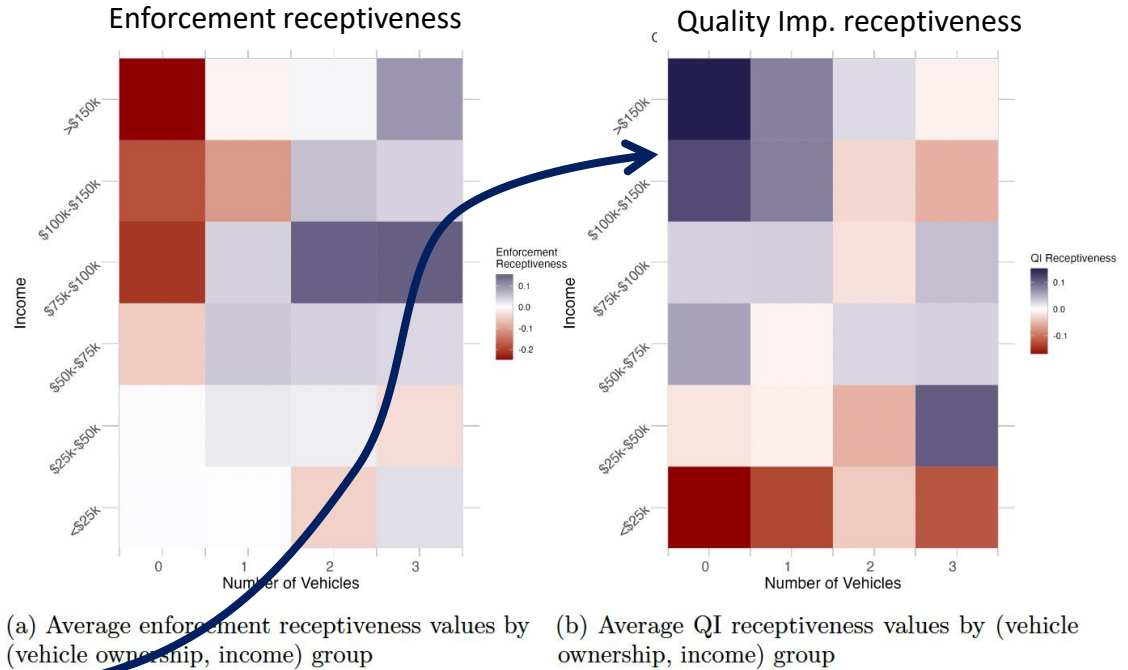


# Take-aways: Divided reaction to issues and policies

**Q1:** Enforcement-related policies have stronger association with satisfaction, *but* also come with the downside of 10-20% of riders feeling less safe.

**Metric of note:** show nuanced policy trade-off for different rider segments, with implications for equity across groups.

E.g. “choice riders”







IMPROVED LIGHTING AND  
SIGNAGE PROVIDE

## 2. Experiential shaping

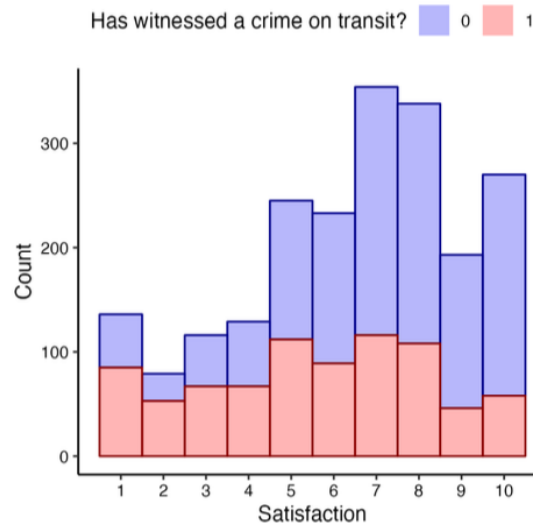
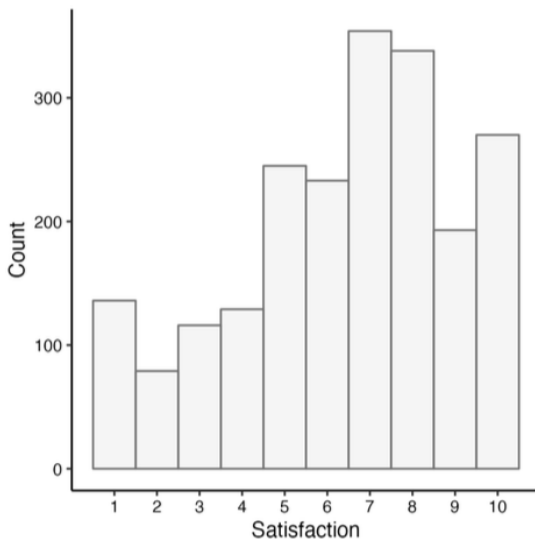


Figure 2: Distributions of overall transit satisfaction in the Chicago region. Left: the full sample used in the model. Right: the full sample, but colored by if those riders have witnessed a crime (assault, robbery, harassment) on transit or not.

ATION  
ENGER  
URITY

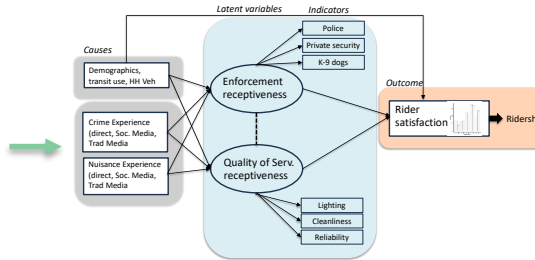


D WALL FINISHES  
EASE OF  
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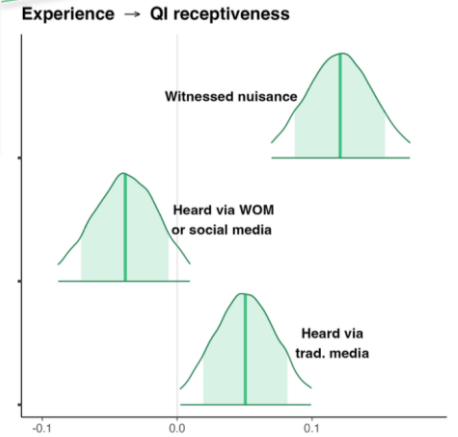
# Results Q2: experience of hazards

**Q2:** We examined experience channels and impacts

- 2 types of experiences (Crime, Nuisance) via 3 channels



*Impact differs  
Traditional media  
leads to higher  
enforcement support*



Experience	Description	Channel
Crime Experience	Experience of assault, robbery or harassment on transit	-Witness 1 <sup>st</sup> hand -Social media /WOM -Traditional news
Nuisance Experience	Experience of dirtiness, loud noise, damage, substance abuse, or people using transit as shelter	Same as above

## Q2 Take-aways: Divided reaction to issues and policies

### Q2. How experiences are gained matters

#### Finding:

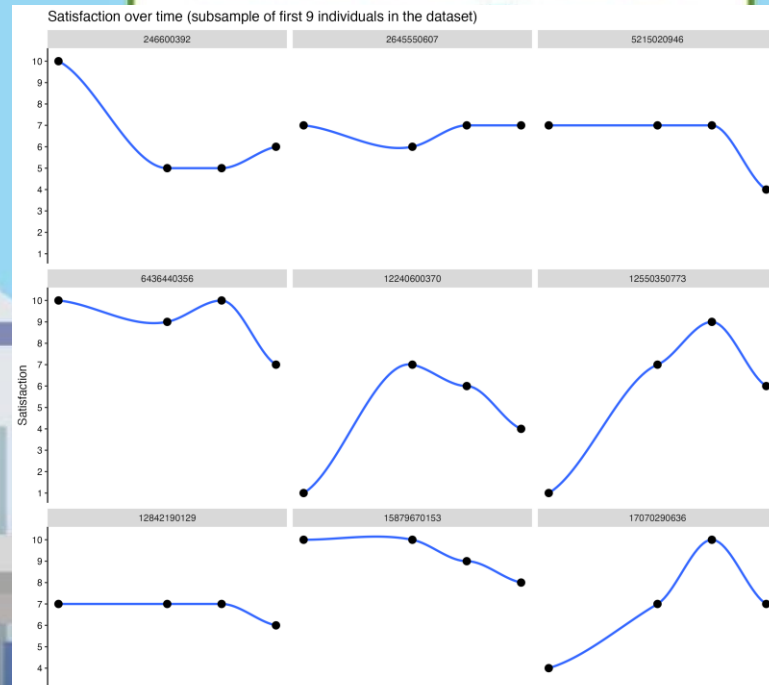
- Poor experiences lead to higher acceptance of safety policies, and lower satisfaction (*crime:witnessed is most impactful*)
- We learn that own experience > social media, but the ordering is different for crime vs nuisance.

#### Impact & Metrics.

- Visualize how people acquire information about safety hazards.
- Suggests different strategies by experience channel, e.g. social media lowers support for policy measures.

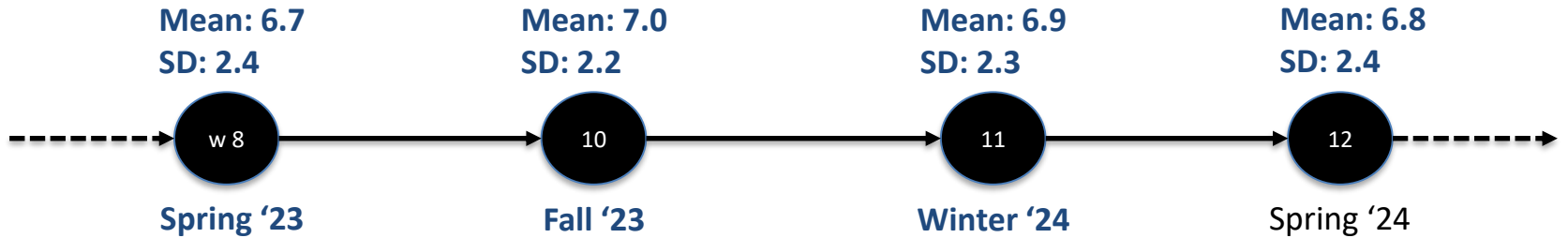
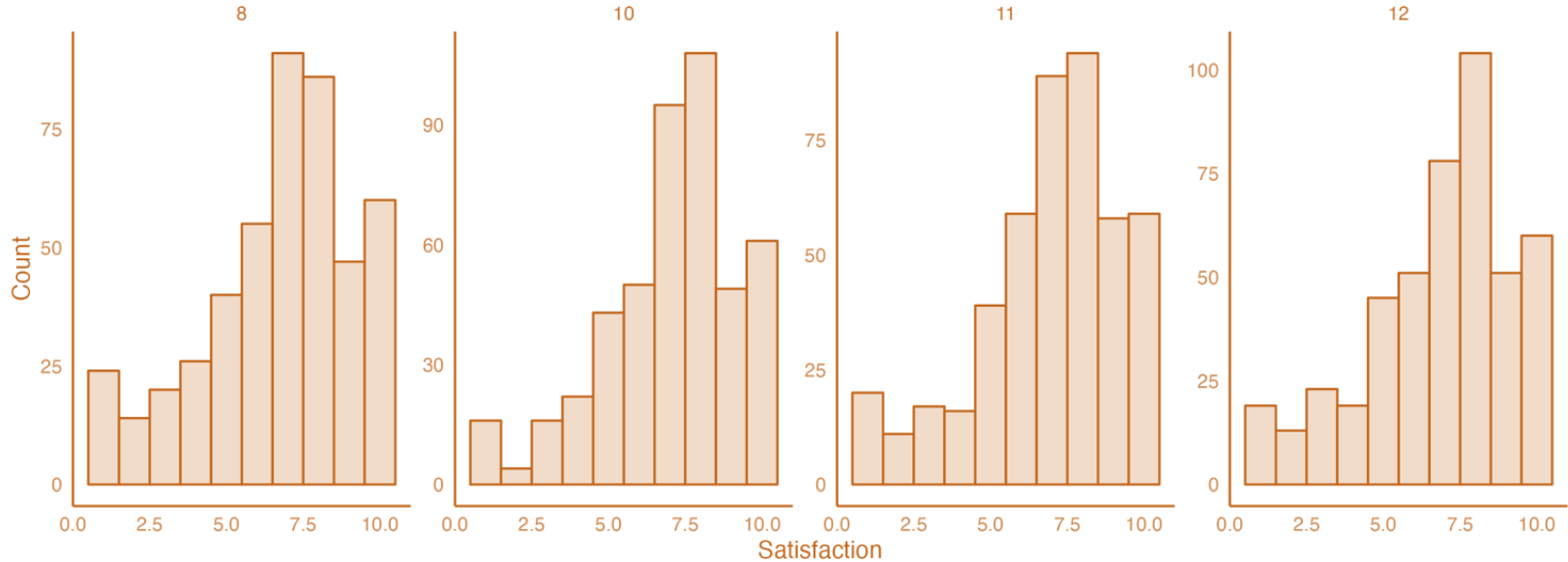
# How satisfied were commuters with Public Transport Services in 2023?

## 3. Capture satisfaction trajectories

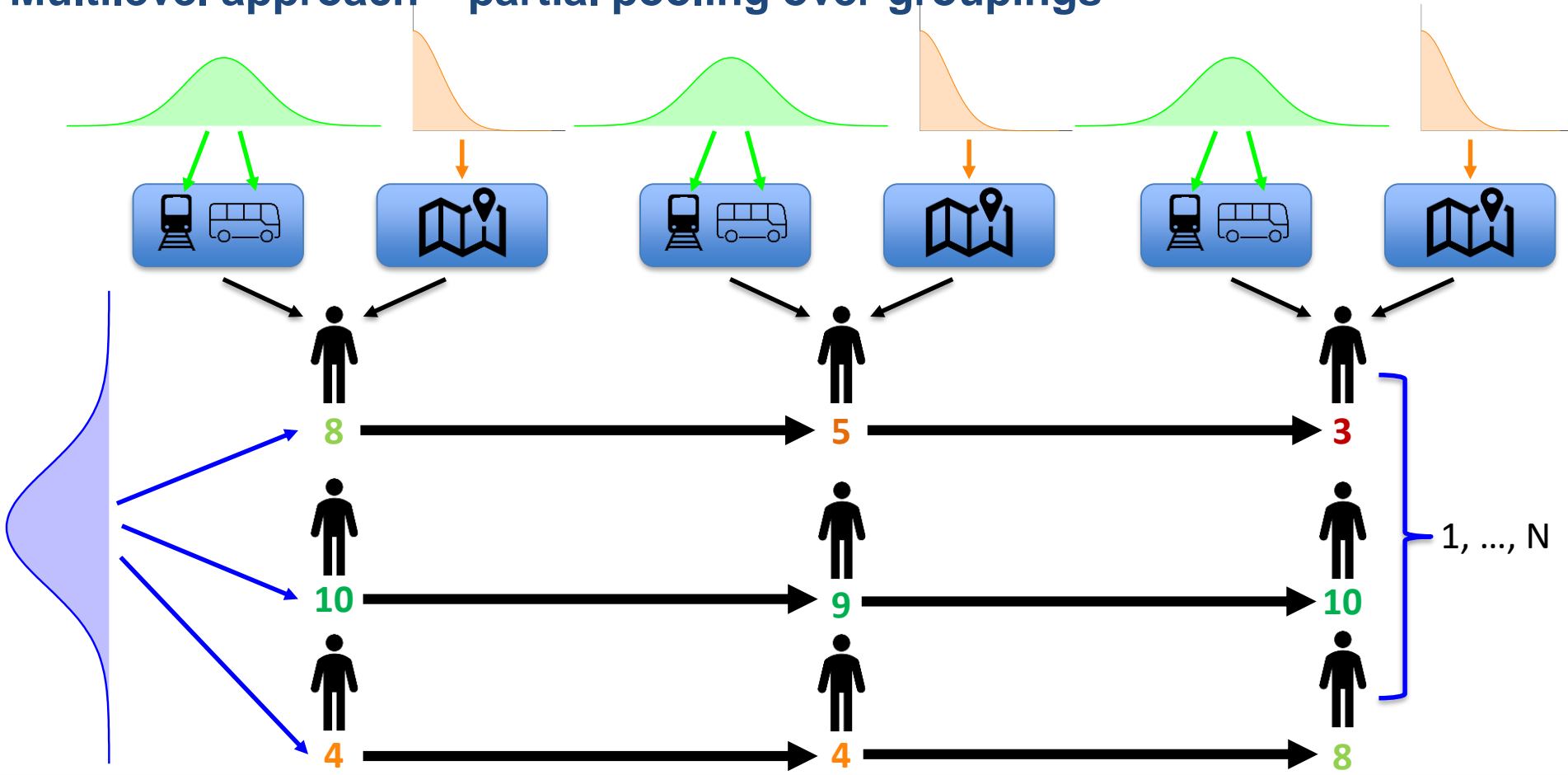


# Panel data summary

Satisfaction distribution across waves

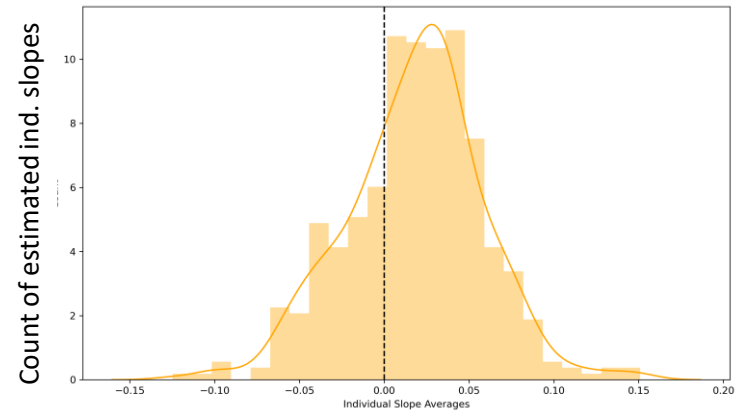
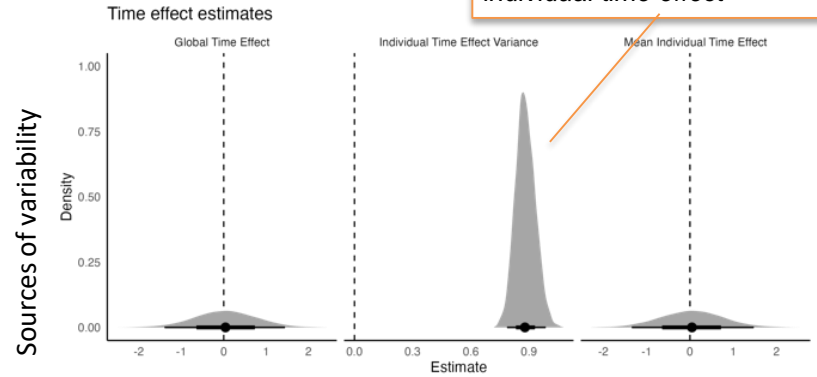
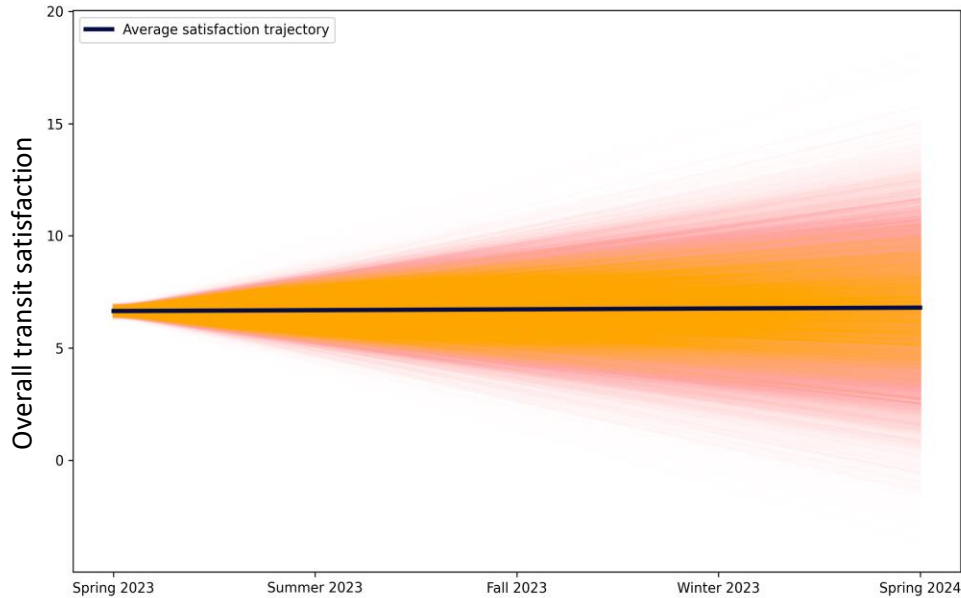


# Multilevel approach – partial pooling over groupings



# Multilevel predictions (simple structures)

*Most variance in partial pooling model owed to individual time effect*

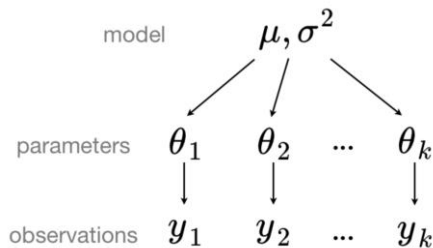


# Results Q3: Trajectory analysis

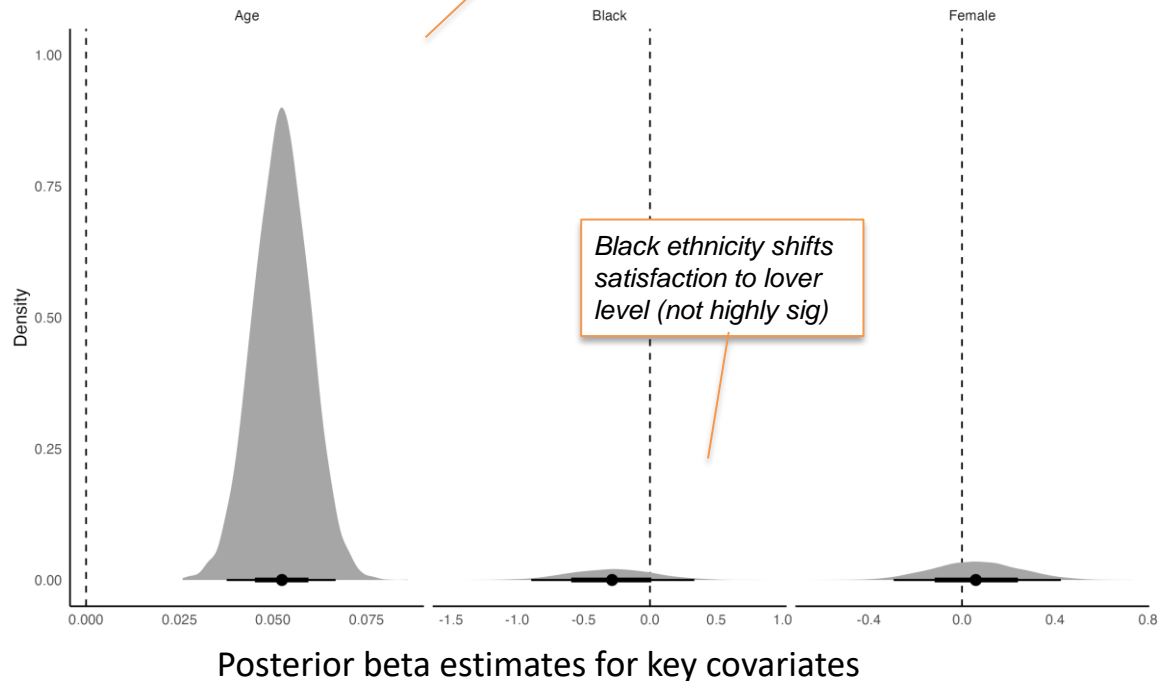
**Age** has largest impact on satisfaction (older more satisfied)

**Q3:** In what way is transit satisfaction changing, how rapid is that change, and how dispersed are opinions at each time period?

- Find home to variance
- Explore sources of variability (individual, mode, location, time)

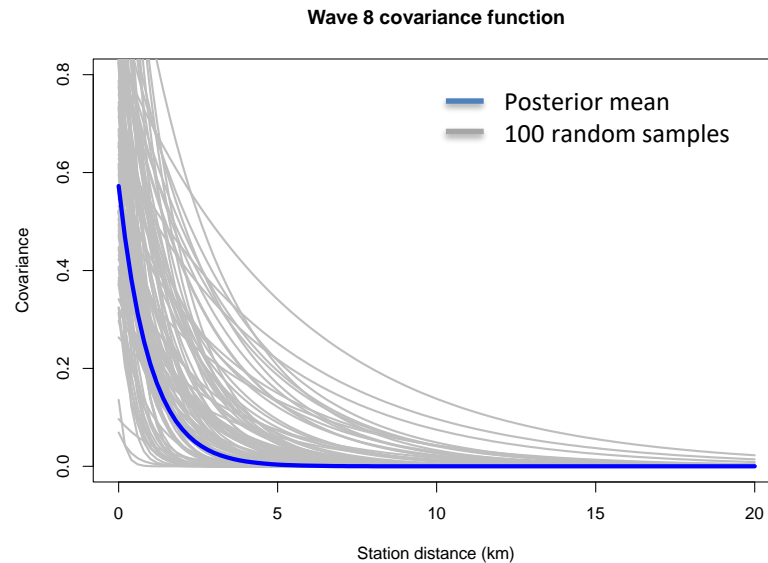
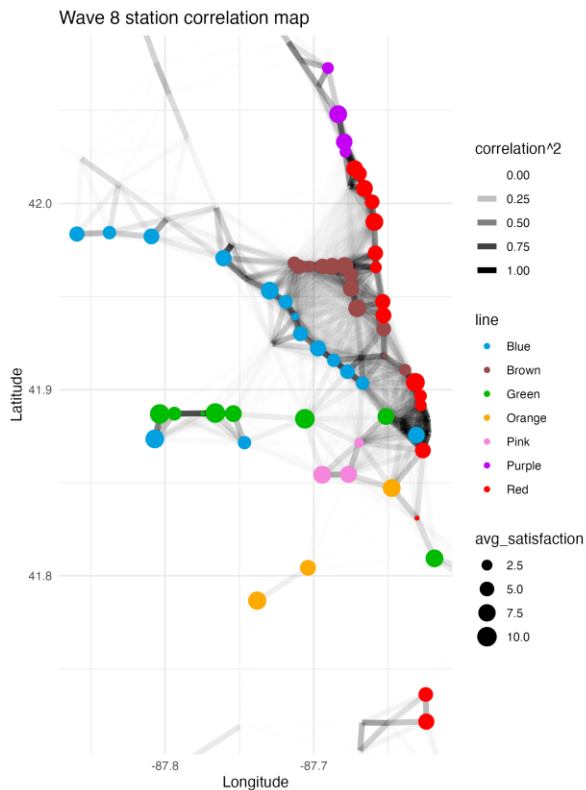


Coefficient estimates - individual characteristics





# Peak into spatial spillover (ongoing work)

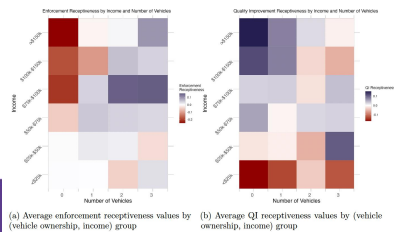


# Summary of insights

## Q1. Divided reaction to safety measures

**Finding:** Pro enforcement ⇔ higher overall transit satisfaction, but is also divisive

**Impact/Metric :** Account for heterogeneity & tailor safety programs by knowing rider profiles



## Q2. Experience sources matter

**Finding:** safety hazard experience → enforcement acceptance → satisfaction (stronger for Enforcement).

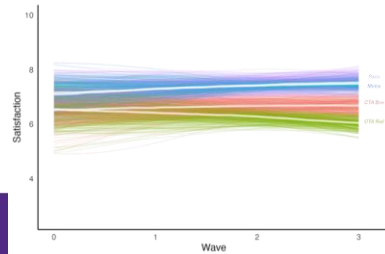
**Impact/Metric:** Account for experience channel. Quantify impact



## Q3. Trajectory MLM Bayesian modeling

**Finding:** Partial pooling reveals source of variance: slopes mainly vary by individual time 'drift'

**Impact/Metric:** Account for both main trend and variability. Allocates sources of variance



# Questions

- As modelers how should we contend with divisive “cure is worse than the problem” scenarios (e.g. social distancing, climate policy, road pricing, enforcement and safety)
  - Bayesian marginal posterior distributions give new insight
- How can we account for politicization and ideological aspects in transportation decisions.
  - Can the ‘winners’ really compensate the ‘losers’: should we apply a different weight for a vulnerable group, e.g. transit-reliant household?
  - Does illustrating divisiveness help make practical decisions?

# Question about modeling learning/experiences

- How can we account for different **learning** channels in a model setting (e.g. direct experience or WOM)?
  - Own experience > Word-of-mouth → satisfaction
  - Traditional media vs. social media have different ranking by safety measure
- Can MLM partial pooling approach help model attitude evolution (esp. with small samples, complex hierarchy, longitudinal data)?

## Discussion question: Individual posterior estimates & Agent Based Models?

- Should agents be allowed a ‘discordant’ view of attributes and policies?
- Should we give agents prior experiences? Are they predisposed for updating preferences in some way?
- Should individual agent trajectories change endogenously (or be more/less deterministic informed by MLM)

# Credibility of research – some thoughts

**Method and data transparency:** availability of design details, analytic choices, and underlying data);

If a finding reported with lower levels of transparency is to be considered less credible → behavioral research may be penalized (possibly also some big-data sets)

**Analytic reproducibility:** ability of reported results to be reproduced by repeating the same data processing and statistical analyses on the original data);

Sophisticated modeling starts out with simple exploration of data, e.g., unexpected or nuanced relations traced through modeling

Worked with different data-cleaning and exclusion practices, rerun model

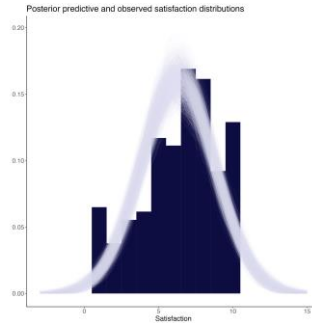
**Analytic robustness:** robustness of results to different data-processing and data-analytic decisions); and

Convergence and Model Checks as subsection in paper; compare posterior to data

Tested different priors: e.g. weakly-informative priors, diffuse priors

**Effect replicability:** ability of the effect to be consistently observed in new samples, at a magnitude similar to that originally reported, when methodologies and conditions similar to those of the original study are used

Advancement of machine learned models more difficult to replicate: Challenging to understand how transferability is verified in complex models developed using large (opaque) training data sets



# Thank you! Questions?

- Papers and Resources
- Aeshliman & Stathopoulos “Trade-offs in transit public safety interventions: balancing enforcement and service quality improvements” under review in Transportation Research Part A: Policy and Practice.
  - [https://papers.ssrn.com/sol3/papers.cfm?abstract\\_id=4989514](https://papers.ssrn.com/sol3/papers.cfm?abstract_id=4989514)
- Working paper: Aeshliman & Stathopoulos “Tracking public transit satisfaction trajectories with longitudinal multilevel models”
  - to be presented at TRB 2025



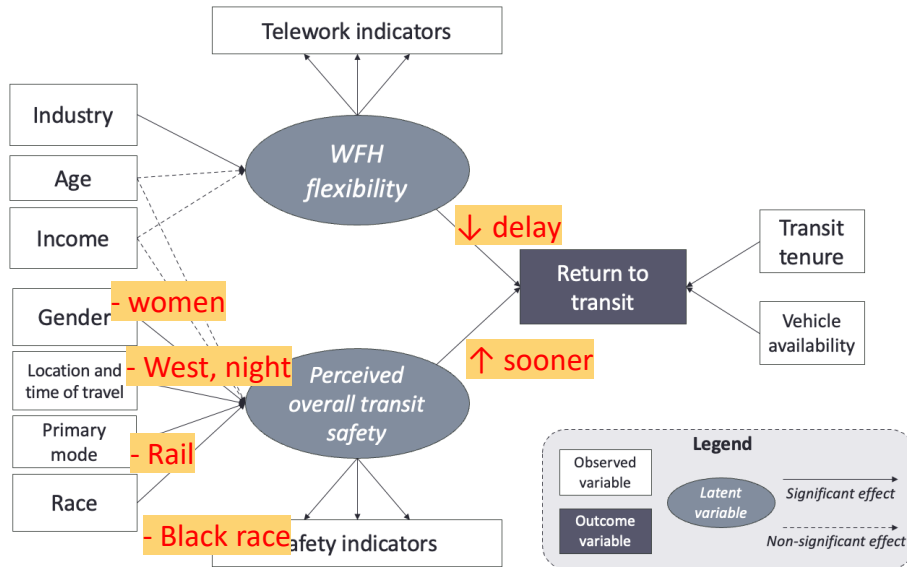
Graduate Research Fellowship Program

## Other references

- *Currie, G., Delbosc, A., and Mahmoud, S. (2013). Factors influencing young peoples' perceptions of personal safety on public transport. Journal of public transportation, 16(1):1–19.*
- Guerra, E. (2022). What the heck is a choice rider? A theoretical framework and empirical model. *Journal of Transport and Land Use, 15(1):165–182.* Publisher: Journal of Transport and Land Use.
- Klein, N. J., Ralph, K., Thigpen, C., and Brown, A. (2022). Political partisanship and transportation reform. *Journal of the American Planning Association, 88(2):163–178.*
- Barajas, J. M. (2021). Biking where Black: Connecting transportation planning and infrastructure to disproportionate policing. *Transportation Research Part D: Transport and Environment, 99:103027.*



# Additional information



## Primary Hypotheses

1. Increased sense of safety/security shortens RTT timelines
2. Increased WFH flexibility extends RTT timelines
3. Circumstances of regular transit use (mode, time of day, etc.) play a role in formation of perceived safety
4. Longer-tenured transit riders return sooner

