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# Methods: Mind the Gap Webinar Series

## Geospatial Data for Healthy Places: Building Environments for Active Living Through Opportunistic GIScience

Presented by

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The Ohio State University





# Geospatial Data for Healthy Places:

Building Environments for Active Living through  
Opportunistic GIScience

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# Introduction

## Healthy places

- Encourage physical, mental and social well-being
- Facilitate physical activity and access to healthy food
- Minimizing exposure to risks such as poor air quality, heat stress and injuries
- Inclusive, legible and convivial environments that maximize social interaction



## Introduction

### Many places are unhealthy!

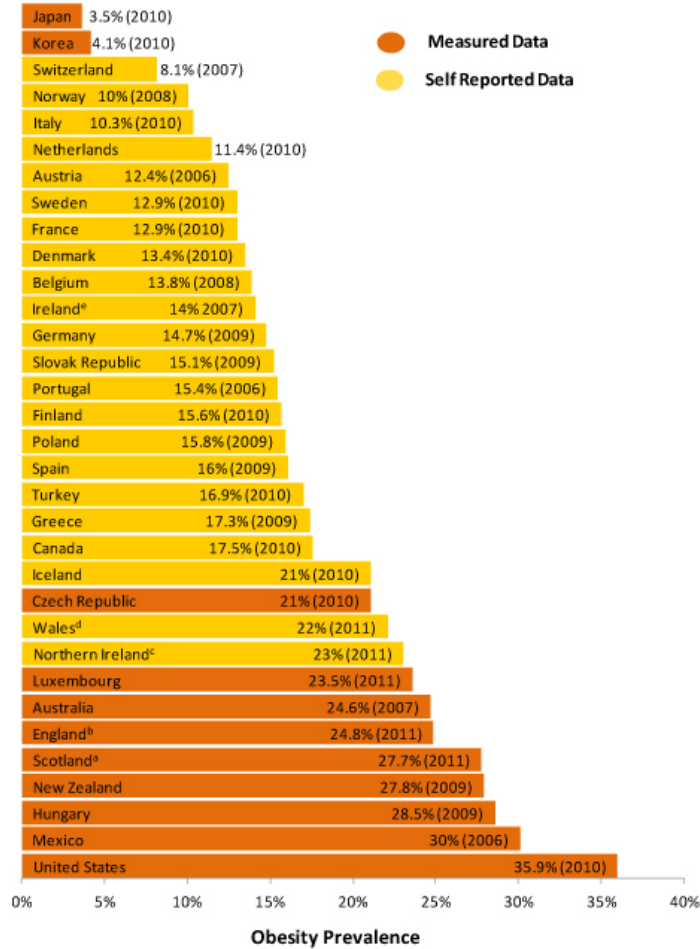
- Discourage physical activity  
Physical activity is unnecessary  
And often unpleasant, unsafe, unhealthy
- Health consequences  
Sedentary lifestyle diseases such as obesity, diabetes, heart disease, cancer  
Leading risk factor for mortality in US  
Fourth leading risk factor globally



<http://www.newurbanism.org/>

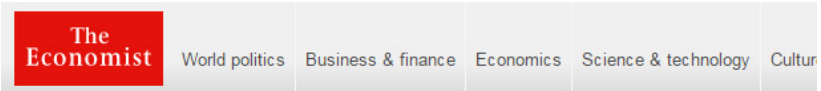


Figure 1: Adult obesity prevalence, latest available data



Public Health England <https://www.noo.org.uk>

# Introduction



Chronic diseases in developing countries

## Growing pains

Poor countries are developing the diseases of the rich, with lethal consequences

Sep 24th 2011 | KAMPALA AND NEW YORK | From the print edition

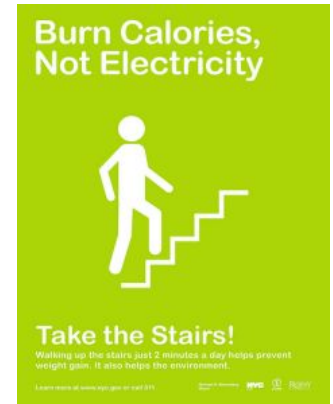




## Introduction

### Active living and healthy cities

- Integrate planning/design with evidence-based health policy
  - Planning and design interventions
  - Data collection and evaluation
- Science and policy initiatives
  - Healthy Cities
  - Active Living Research
  - Complete Streets
  - Active Design



# Introduction

## Problems

- Human systems are complex
- Policy interventions can have unclear and counterintuitive outcomes

## Solutions

- Ecological approach
- Evidence-based policy

## Strategies

- Opportunistic GIScience
- Geographic Information Observatories





## Remainder of talk

### Healthy environments

- Design interventions

- Ecological approach

### Real-world experiments

- Moving Across Places Study (MAPS)

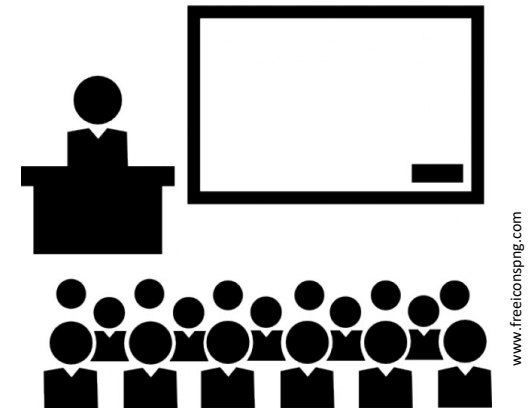
- Does public transit generate *new* physical activity?

### Opportunistic GIScience

- Geographic information observatories

- Leverage real world events

### Conclusion





## Healthy environments

### Physical activity by design

#### Five D's of walkability

1. Diversity of activities
2. Density
3. Distance to public transit
4. Destination accessibility
5. Design

### Complete Streets

Roads as habitats for all users, including pedestrians, cyclists and transit users



*First Avenue Improvements, New York City*

Complete Streets intervention in NYC



# Healthy environments

## Active design

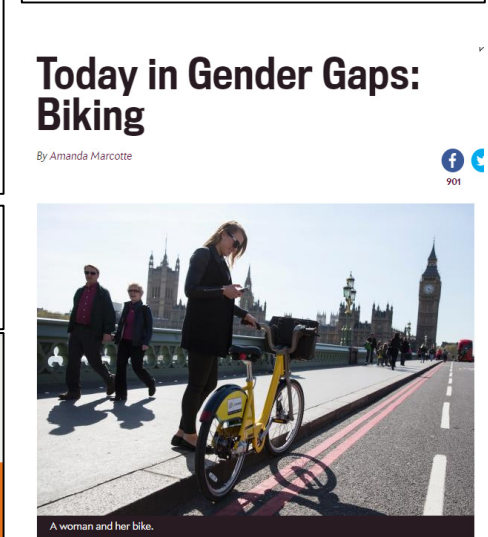
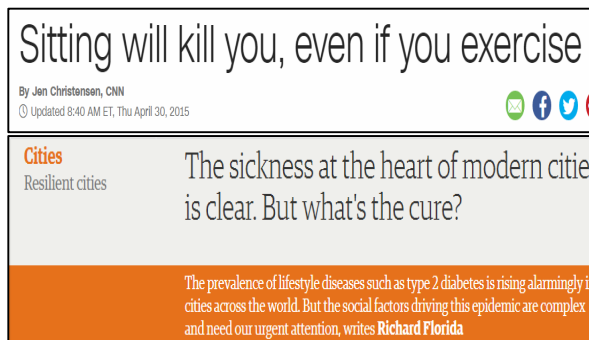
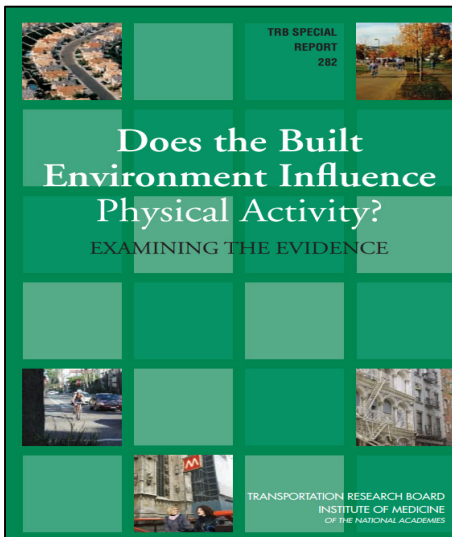
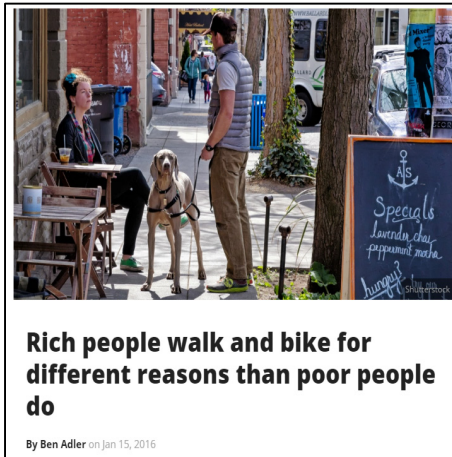
- **Urban design strategies** that encourage walking, bicycling, and active transportation and recreation
- **Building design strategies** for promoting active living where we work and live and play, through the placement and design of stairs, elevators, and indoor and outdoor spaces



Active design in Hangzhou



# Unclear outcomes from design interventions





## Healthy environments

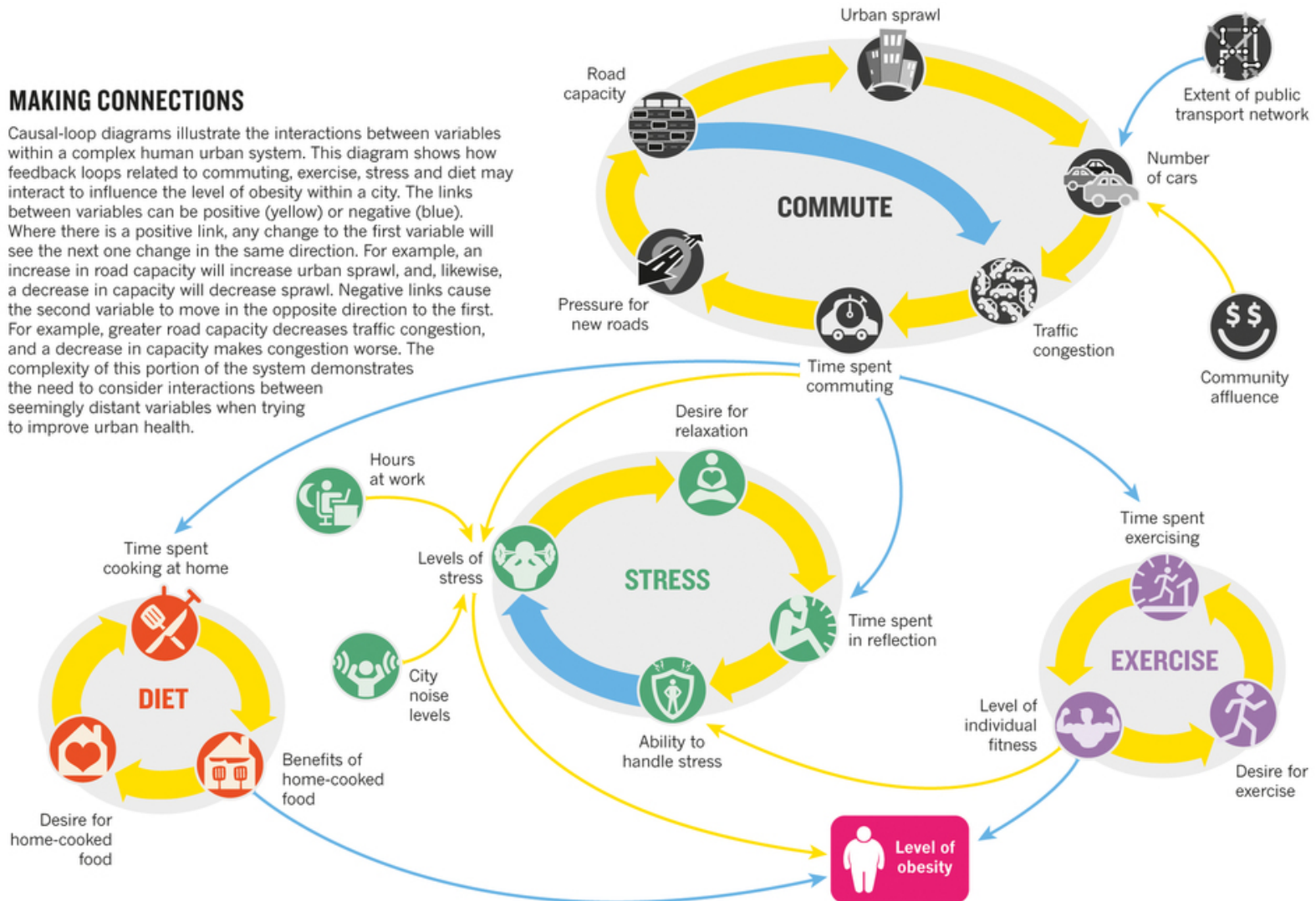
### Why unclear outcomes?

- **Endogeneity:** e.g., people self-select for walkable neighborhoods
- **Context-dependence:** e.g., walkability near home versus work
- **Social factors:** socioeconomic status, age, gender roles
- **Compensation effects:** e.g., standing at work leads sitting at home
- **Unhealthy environments:** e.g., air quality in urban canyons, heat stress, sick buildings

# Cities and obesity – multiple layers of feedbacks

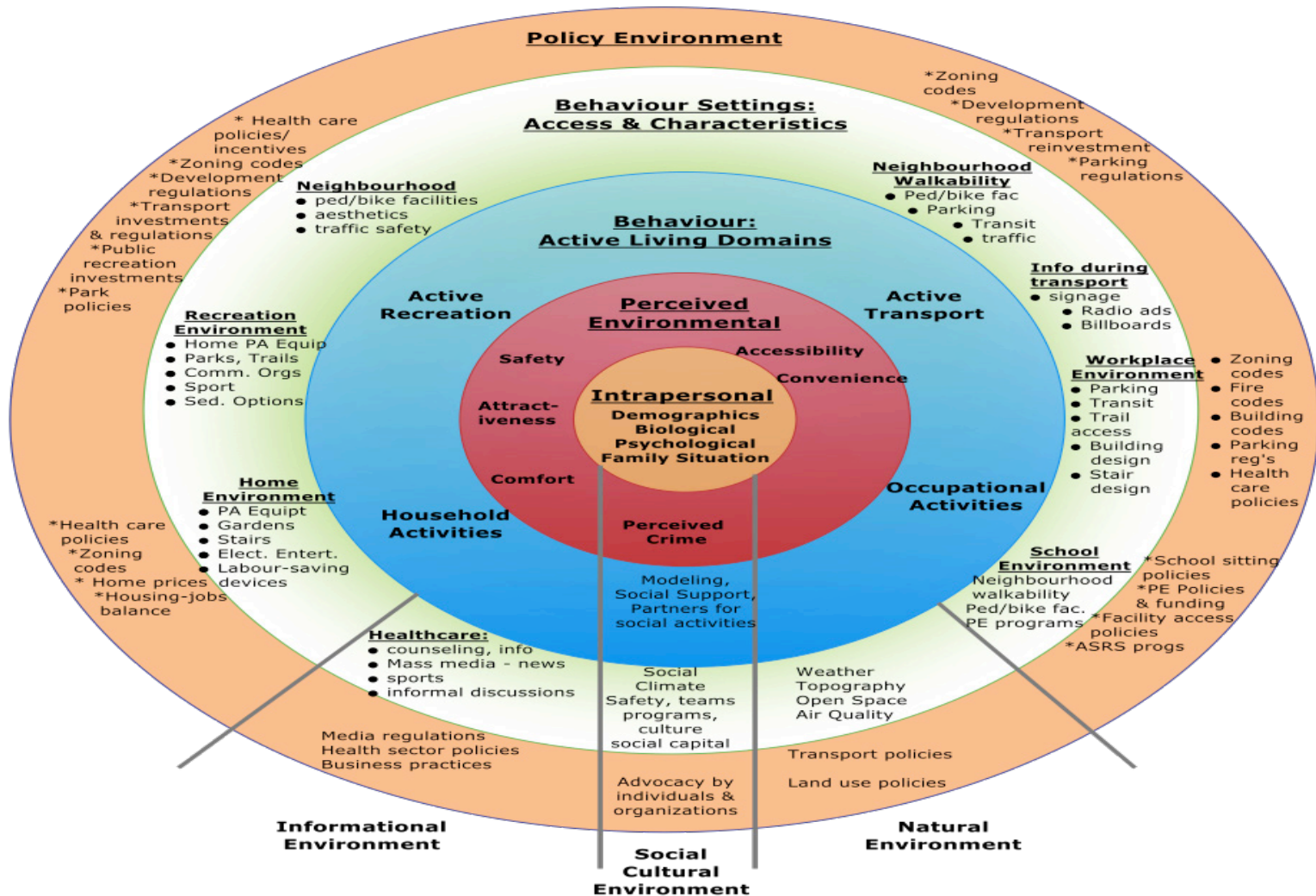
## MAKING CONNECTIONS

Causal-loop diagrams illustrate the interactions between variables within a complex human urban system. This diagram shows how feedback loops related to commuting, exercise, stress and diet may interact to influence the level of obesity within a city. The links between variables can be positive (yellow) or negative (blue). Where there is a positive link, any change to the first variable will see the next one change in the same direction. For example, an increase in road capacity will increase urban sprawl, and, likewise, a decrease in capacity will decrease sprawl. Negative links cause the second variable to move in the opposite direction to the first. For example, greater road capacity decreases traffic congestion, and a decrease in capacity makes congestion worse. The complexity of this portion of the system demonstrates the need to consider interactions between seemingly distant variables when trying to improve urban health.





# Ecological approach to active living research



Sallis J.F. et al. (2006). An ecological approach to creating active living communities. *Annual Review of Public Health*, 27 297-322.





# Healthy environments

## Ecological approach to healthy design

### Multiple levels of influence

Personal characteristics, behavioral domains, settings and policy

### Holistic

Transportation, recreation, work, household activities  
All major facets of active living

### Evidence-based

- Objective measures > Self-reports
- Longitudinal studies > Cross-sectional
- *Experiments: Gold standard for causality*



# Real-world experiments

## Treatment vs. control?

Lab: Random sorting

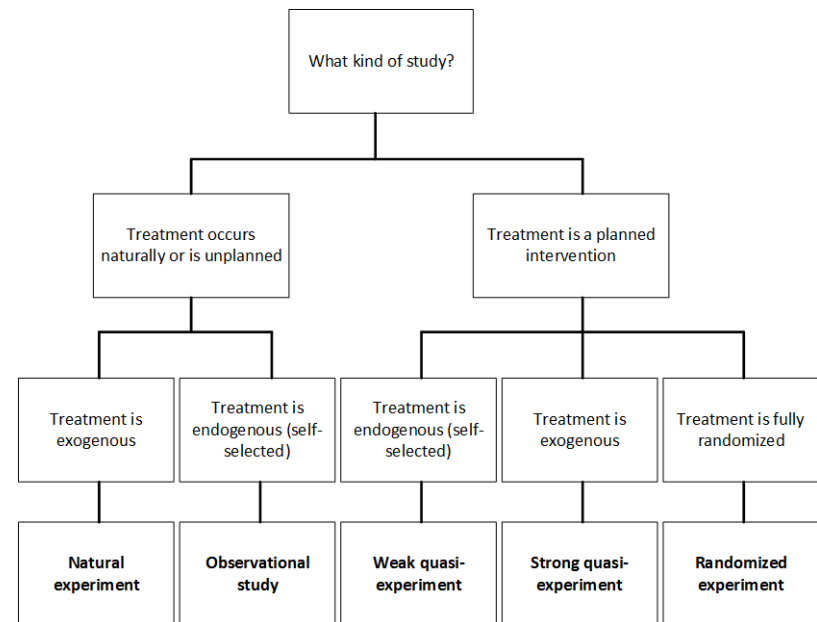
Real world: Infeasible,  
unethical

## Quasi-experiments

**Planned** events or  
interventions in the real world

Natural experiments

**Unplanned or natural events**  
in the real world



## Natural versus quasi-experiments

Remler D K and Van Ryzin GG (2015). *Research Methods in Practice: Strategies for Description and Causation*, second edition.

## Real-world experiments

### Moving Across Places Study (MAPS)

- Impacts of Light Rail Transit and Complete Streets on physical activity (PA)
- Salt Lake City, Utah, USA

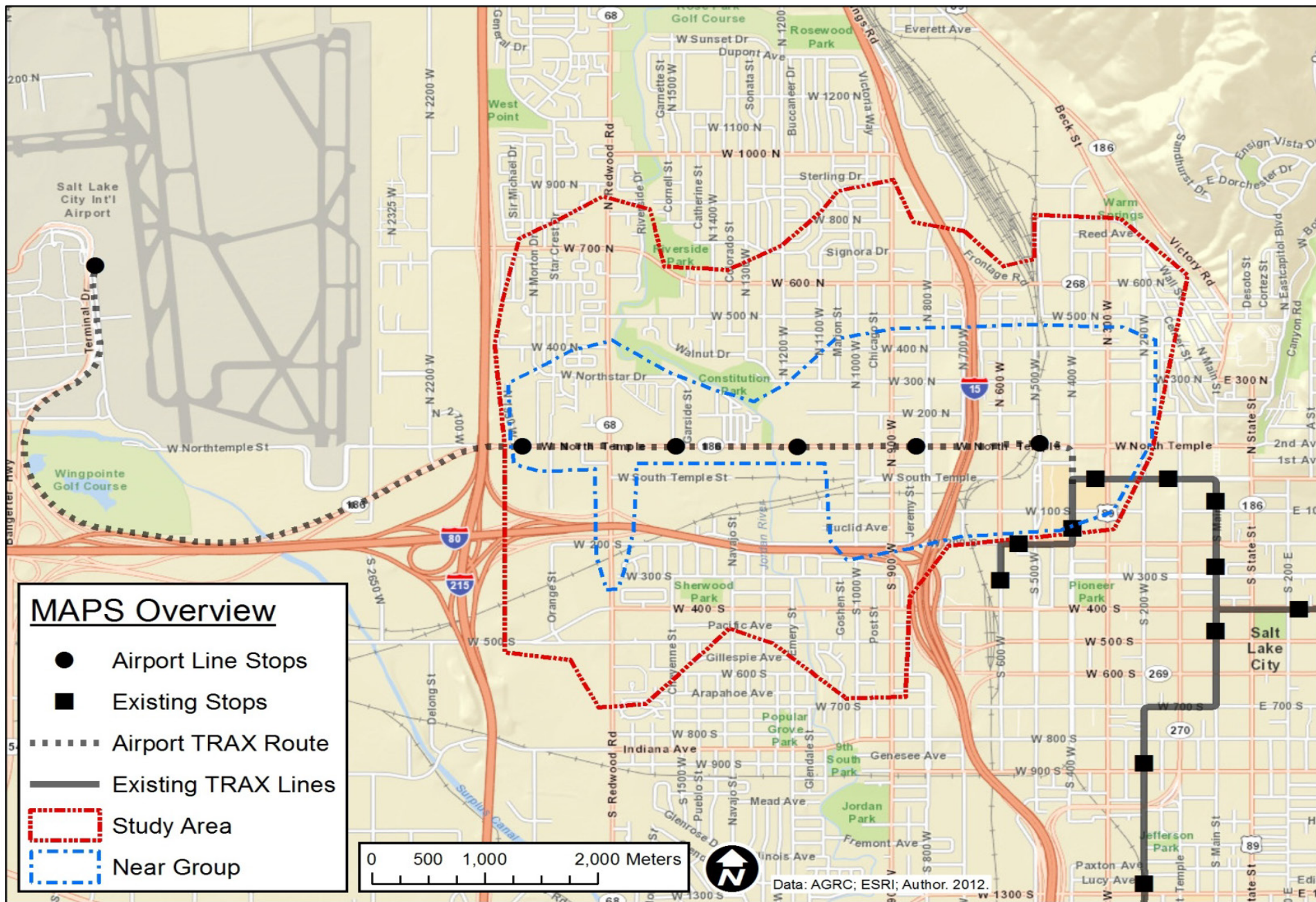


### A quasi-experiment

- Measurements of same participants before and after planned intervention
- Case (near) and control (far) groups







## Real-world experiments

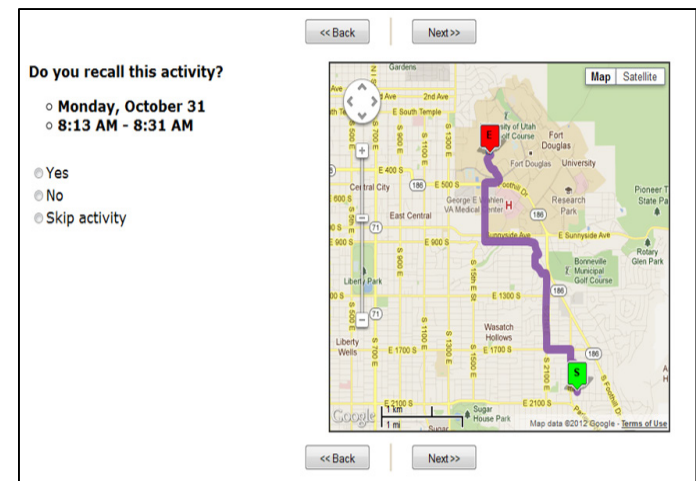
### Data collection

- Height, weight measurements; attitudinal surveys
- GPS + accelerometer wear for one week
- 2012 (before) and 2013 (after)
- Complete sample:  $n = 536$

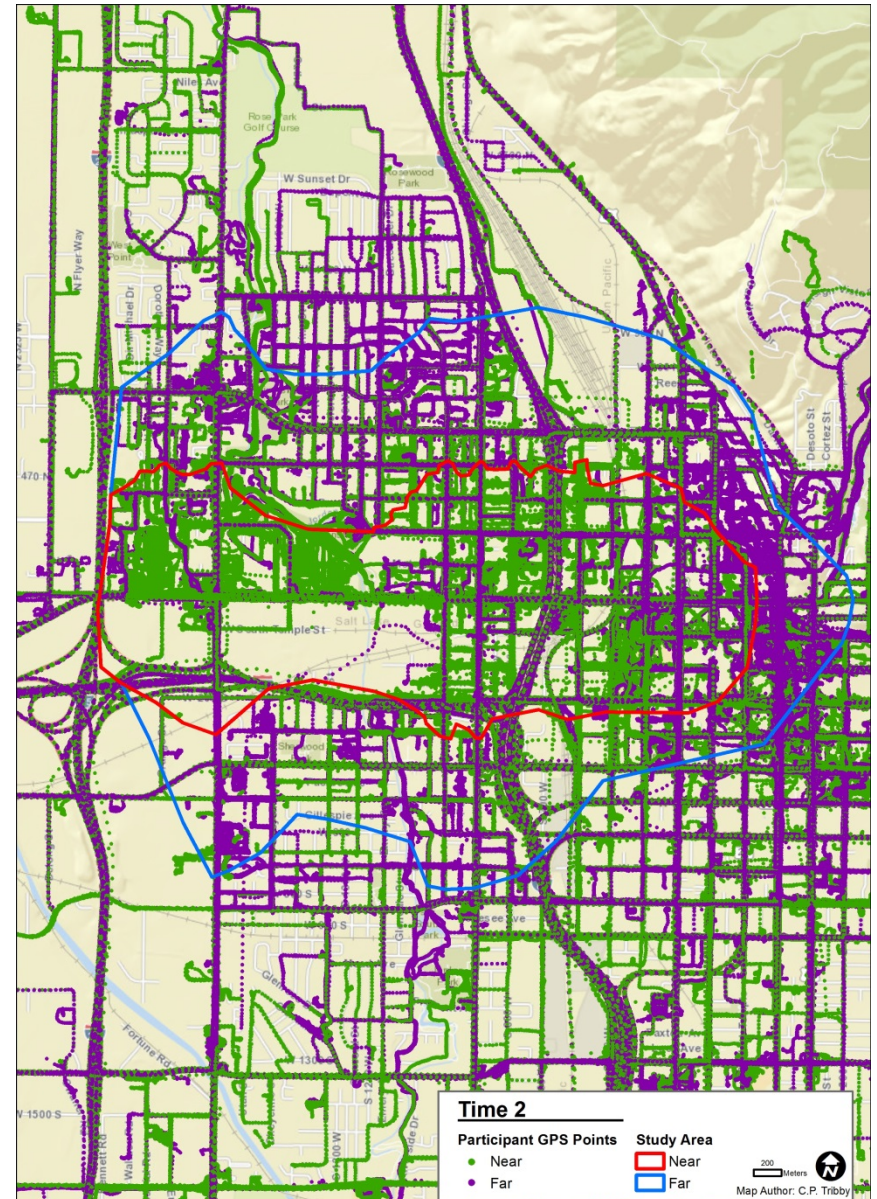
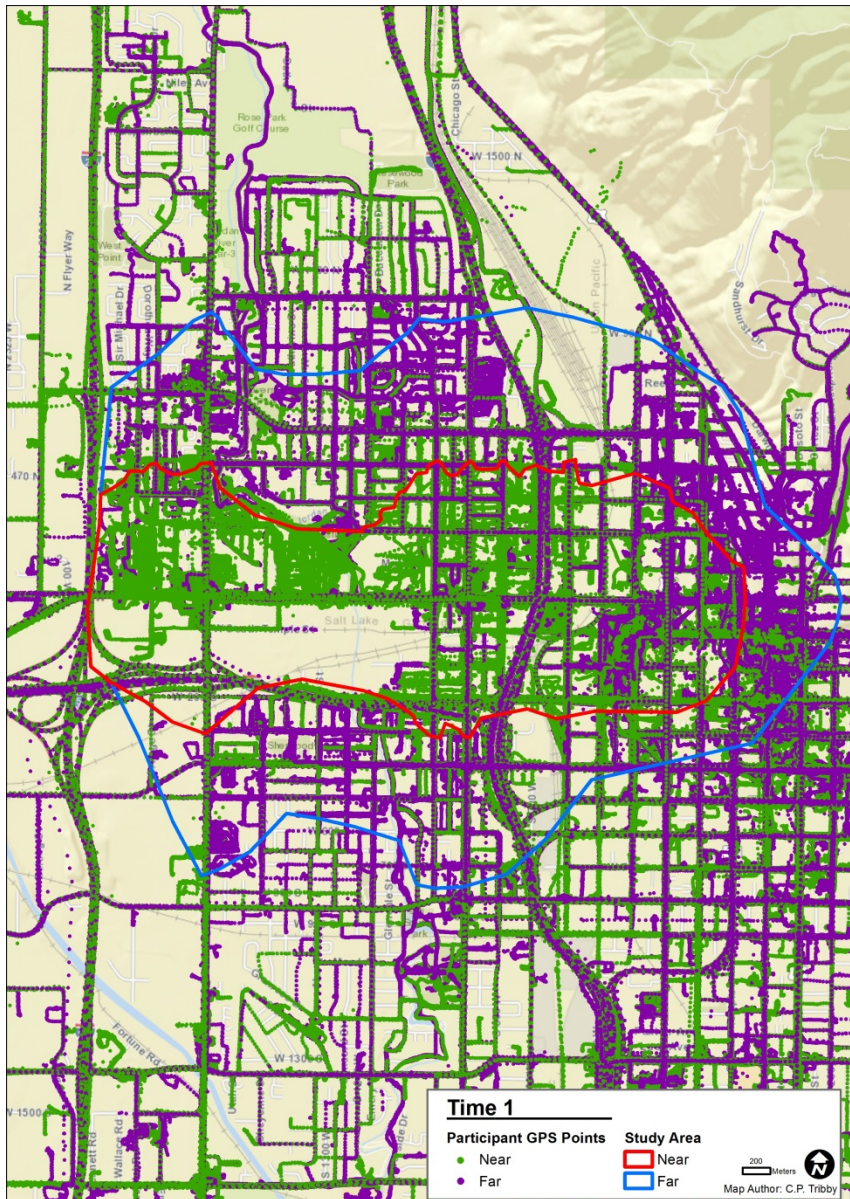


### Data pre-processing (Westat)

- Uploaded, fused and map-matched
- Download for participant review
- Mode detection: Walk, bike, car, bus, LRT







Big geographic data? Approximately 4 million GPS points for each time period 20





## Real-world experiments

### Walkability audit

#### Irvine Minnesota Inventory (IMI)

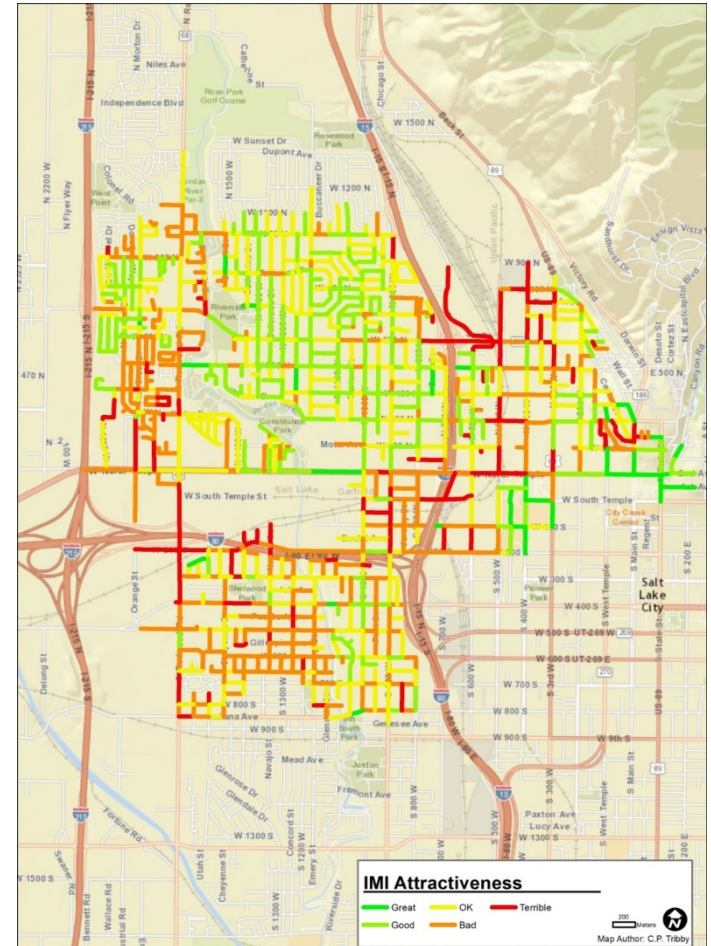
160 attributes / 6 dimensions

1. Accessibility
2. Attractiveness
3. Traffic safety
4. Crime safety
5. Density
6. Pedestrian access

#### Block-level survey

1000+ block faces

2012 and 2013





## Real-world experiments

### Major project aims

1. Increases in transit riders, cyclists and pedestrians?  
Brown BB, Tharp D, Tribby CP, Smith KR, Miller HJ and Werner CM, *Journal of Transport & Health*.
2. Living proximal → larger activity increases? Brown BB, Smith KR, Tharp DS, Werner CM, Tribby CP and Miller HJ (in press) *Journal of Physical Activity & Health*.
3. Perceived walkability, pro-neighborhood → greater use?  
Brown BB, Werner CM, Smith KR, Tribby CP, and Miller HJ (2014) *Preventive Medicine*, 66, 140–144.
4. Complete Streets users → healthy PA + BMI changes?  
Brown BB, Werner CM, Tribby CP, Miller HJ and Smith KR (2015) *American Journal of Public Health*.



# Real-world experiments

## Other studies

### **Walking and built environment** (Calvin Tribby)

- **Walkability summary and visualization.** Tribby CP, Miller HJ, Brown BB, Werner CM and Smith KR (2016) *Journal of Transport and Land Use*, 9, 1-21.
- **Geographic regions for assessing walking trips.** Tribby CP, Miller HJ, Brown BB, Smith KR and Werner CM. (2017) *Health and Place*, 45, 1-9.
- **Walking route choice through built environments.** Tribby CP, Miller HJ, Brown BB, Werner CM, and Smith KR (in press) *Environment and Planning B: Planning and Design*.

### **Public transit**

- **LRT did not cannibalize bus ridership.** Werner CM, Brown BB, Tribby CP, Tharp D, Flick K, Miller HJ, Smith KR and Jensen W (2016) *Transport Policy*, 45, 5–23.
- **Does public transit generate new PA?**



## Real-world experiments

### Does public transit generate *new* PA?

We know transit users are more active  
Self-reports, cross sectional analysis

But this may not be *new* PA  
Undermines claimed health benefits

### Why not new PA?

Confounding

Other, non-transit factors such as density, mix

Substitution

More time for transit → Less time for recreation





## Real-world experiments

### Physical activity bouts

Min 5 minute with a min of 1000  
accelerometer counts per minute  
(Saelens et al. 2014, *AJPH*, 854-859.)



### Types of PA

Transit-related PA (**PA-Transit**): PA within a trip that contains bus or LRT

Non-transit PA (**PA-Other**): PA that does not occur within a trip with bus or LRT

Overall PA (**PA-Total**): PA-Transit + PA-Other



## Real-world experiments

### Public transit user

Participant who rode either bus or LRT at least once during data collection week

### Transit groups (below)

Transit group	N	Public transit user in:	
		2012?	2013?
<b>Never</b>	391	No	No
<b>Continued</b>	51	Yes	Yes
<b>Former</b>	42	Yes	No
<b>New</b>	52	No	Yes



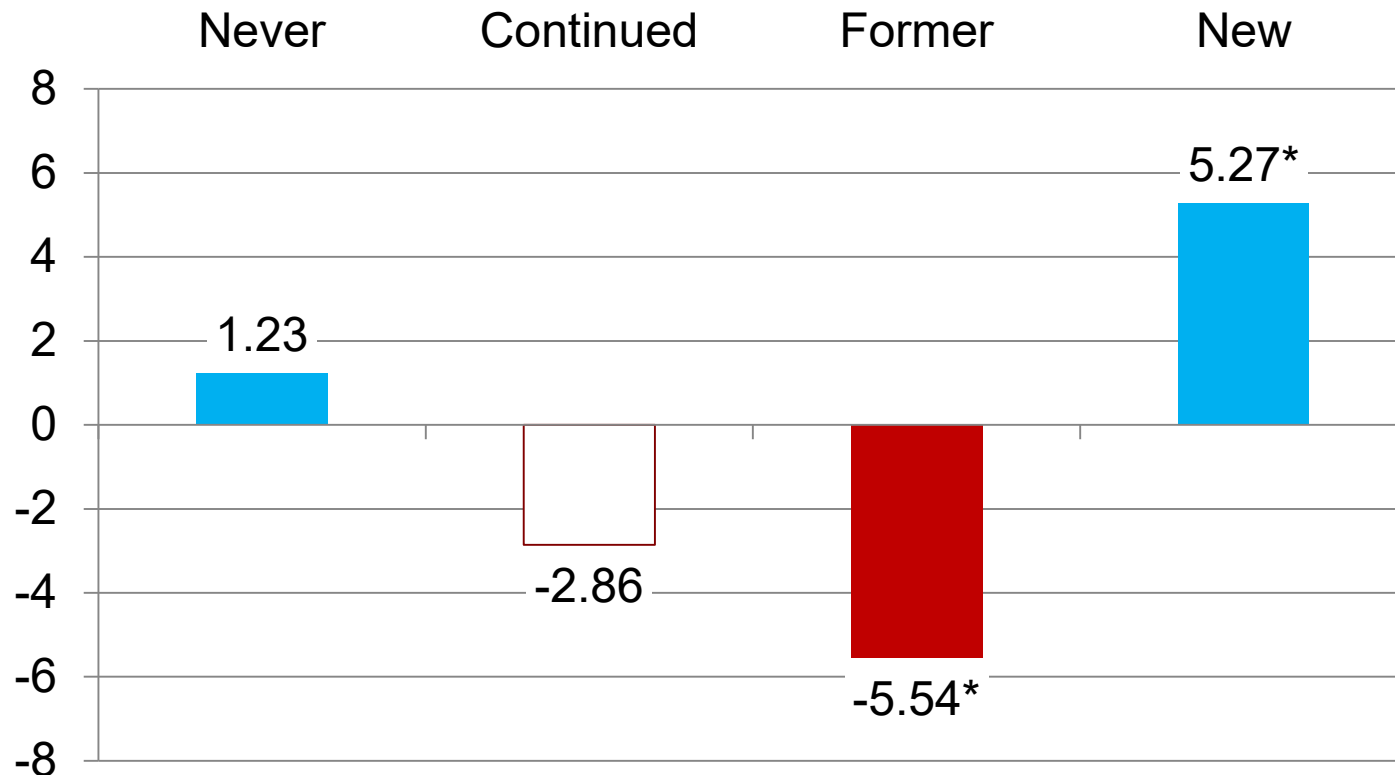
## Real-world experiments

## Hypotheses

New PA implies:	Public transit user group:			
	Never	Continued	Former	New
i) no confounding	No change in PA-Other	No change in PA-Other	Decrease in PA-Transit	Increase in PA-Transit
ii) no substitution			No increase in PA-Other	No decrease in PA-Other
Net change in PA-Total	No change	Any change	Decrease	Increase



## Changes in **PA-Total** time by group (within-person differences, 2013 - 2012)

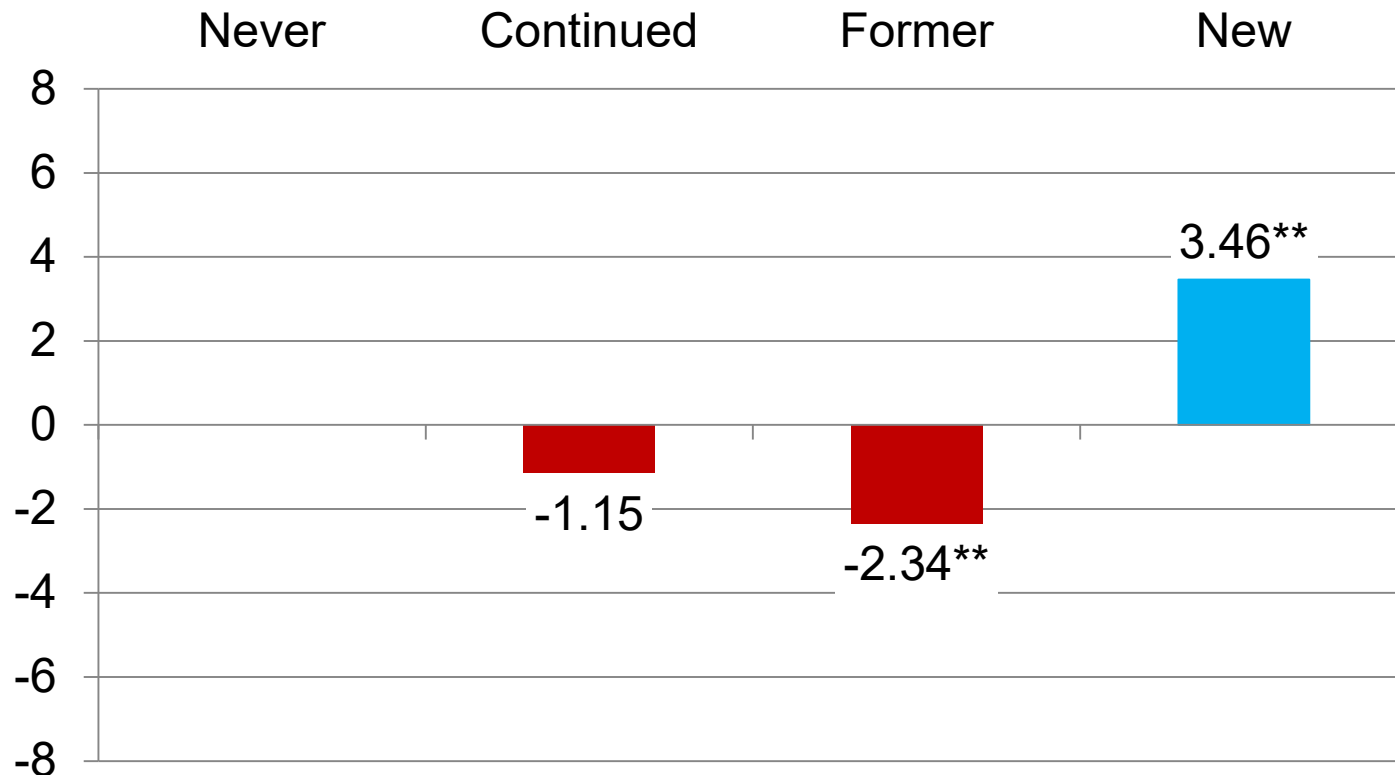


Average time: Minutes per 10 hr. wear period  
PA: Min 1000 cpm in min 5 minute bout  
Within group differences: \*  $p < 0.1$





## Changes in **PA-Transit** time by group (within-person differences, 2013 - 2012)



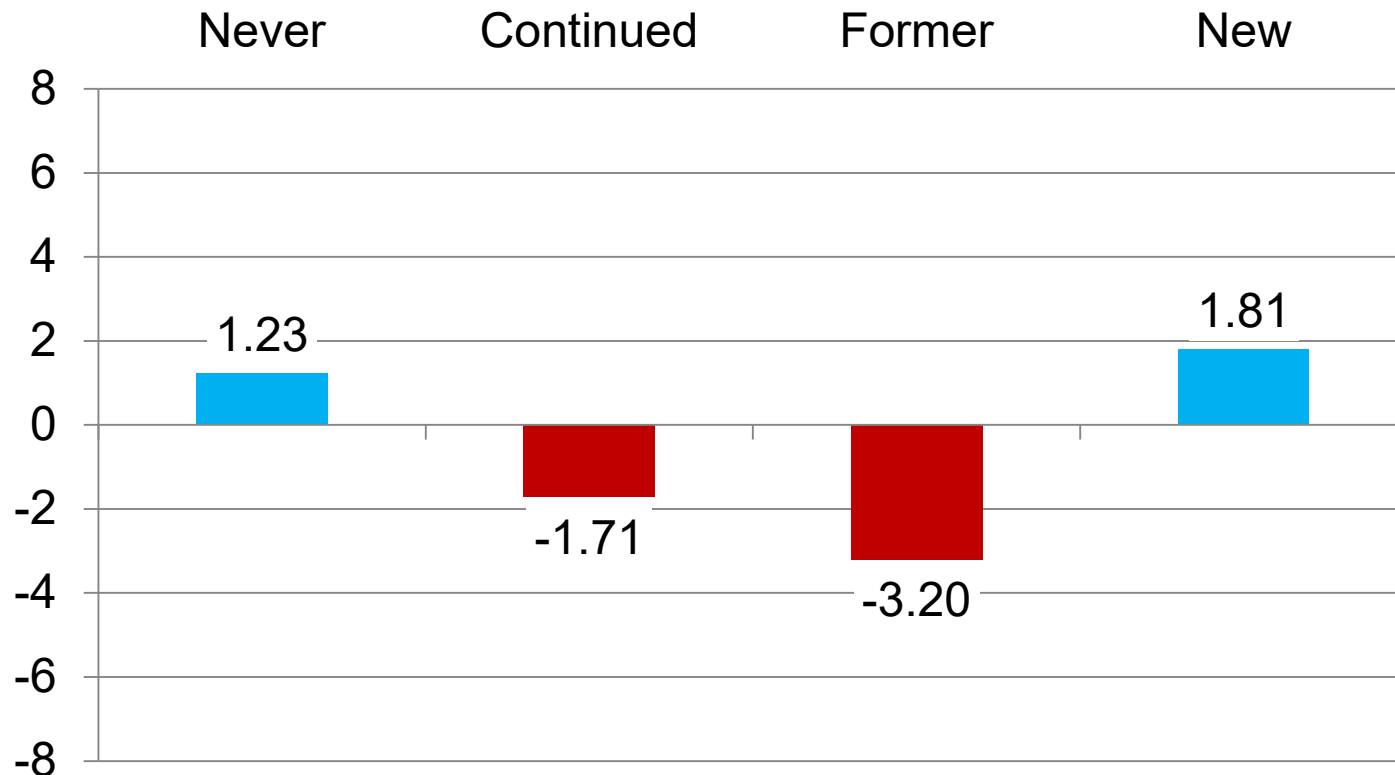
Average time: Minutes per 10 hr. wear period

PA: Min 1000 cpm in min 5 minute bout

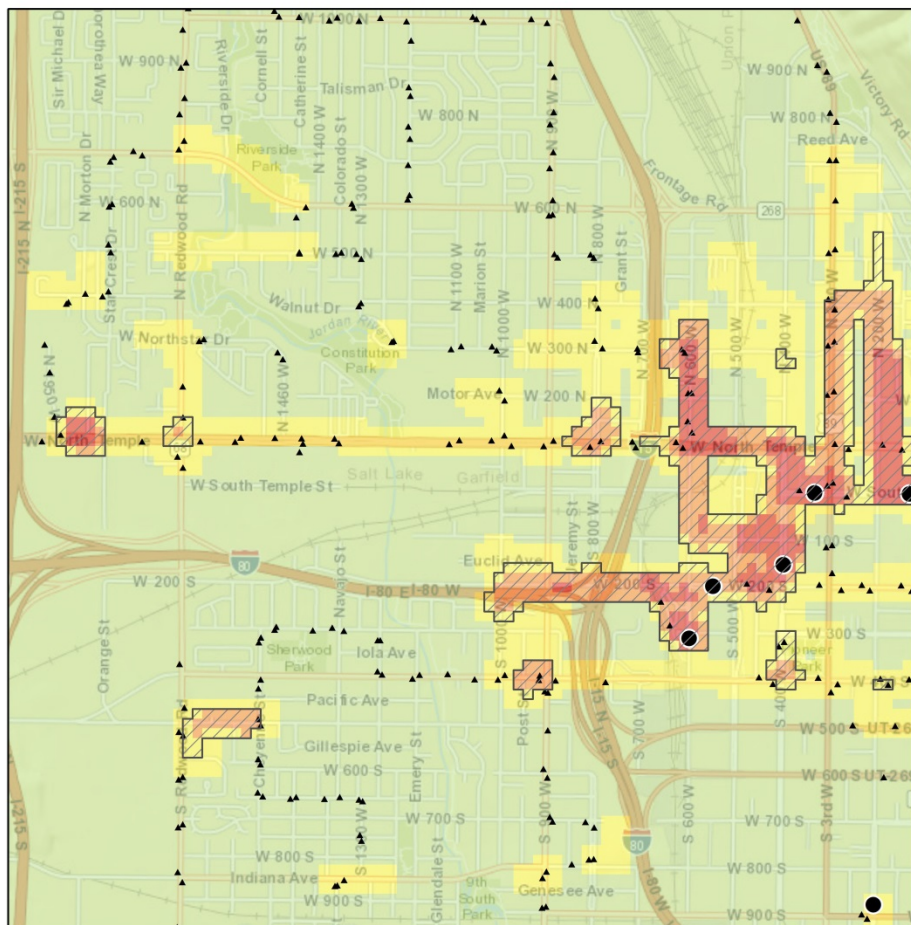
Within group differences: \*\*  $p < 0.05$



## Changes in **PA-Other** time by group (within-person differences, 2013 - 2012)

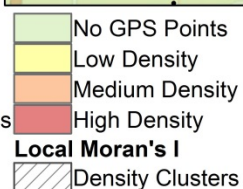


Average time (minutes per 10 hr. wear period);  
PA: Min 1000 cpm in min 5 minute bout  
Within group differences: **None significant**

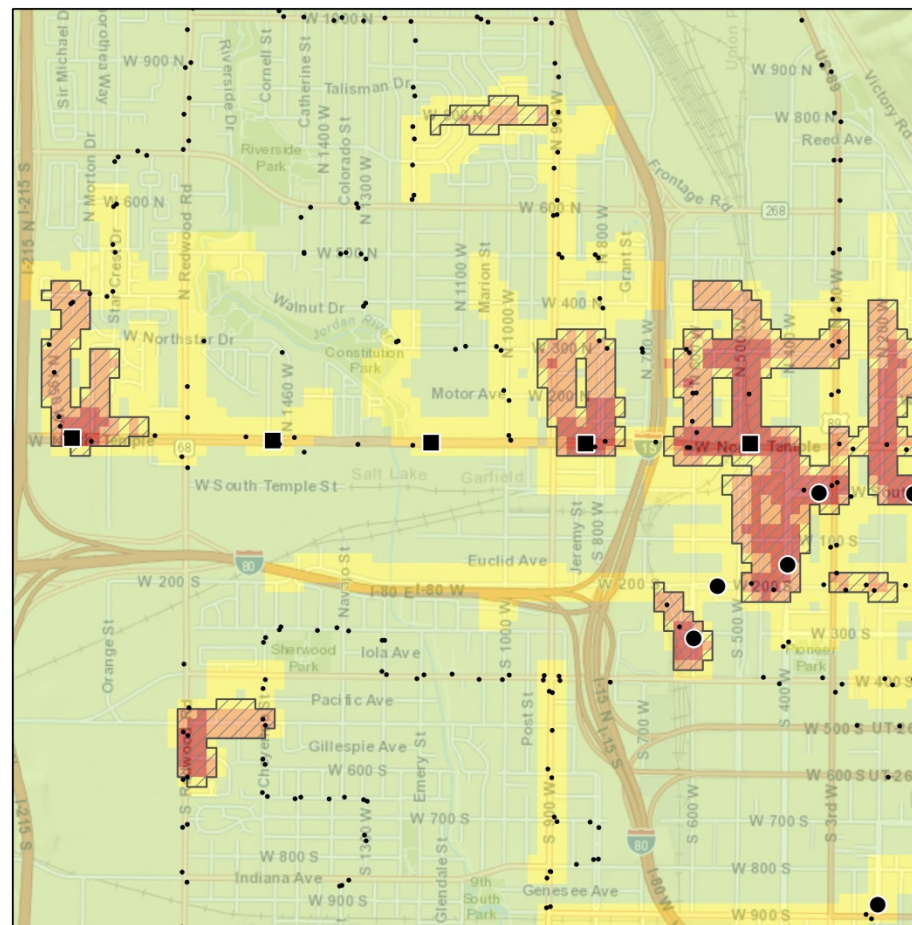


2012 Transit-related  
Physical Activity

- ▲ 2012 Bus Stops
- 2013 Bus Stops
- New TRAX Stations
- Existing TRAX Stations

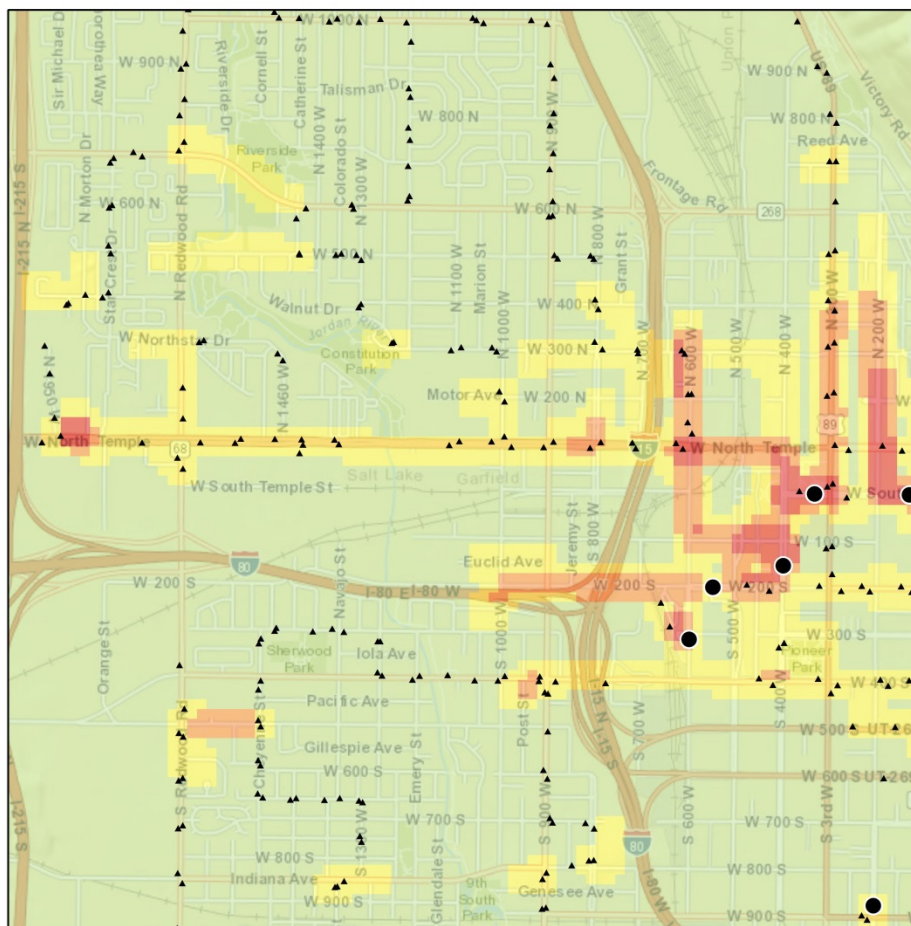


Basemap: Esri; Transit stops: AGRC. 1 Kilometers



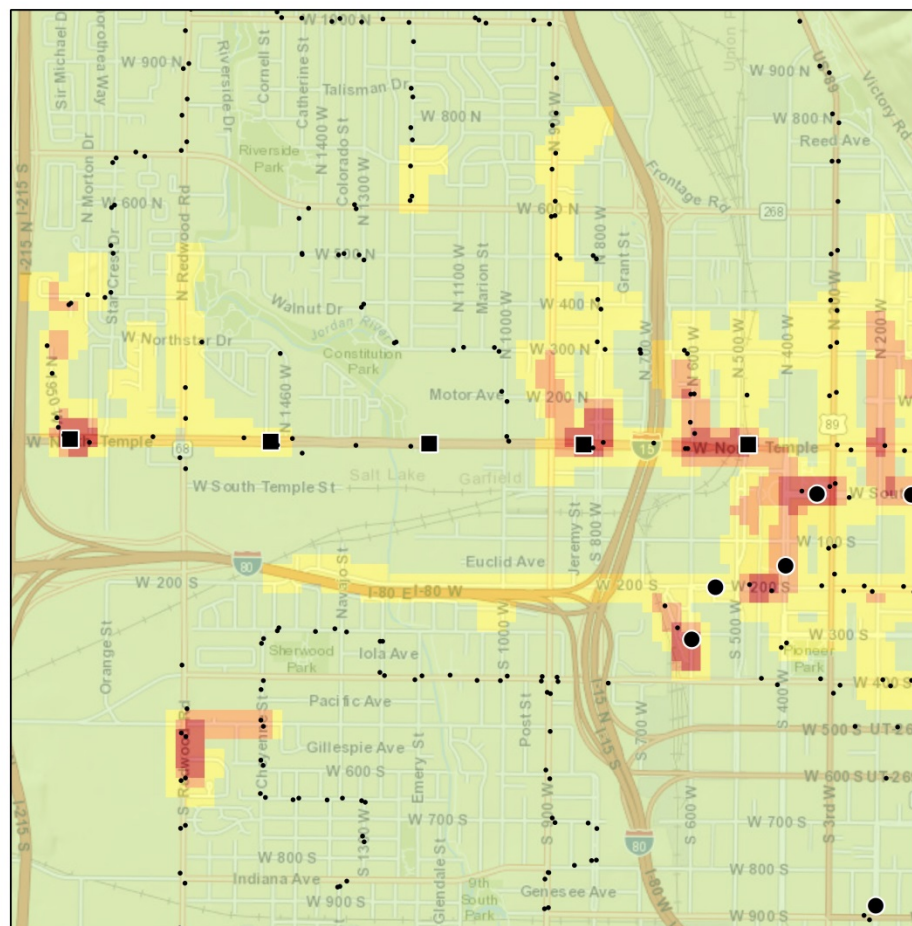
2013 Transit-related  
Physical Activity





2012 Transit-related  
Physical Activity  
CONTINUED

- ▲ 2012 Bus Stops
- 2013 Bus Stops
- New TRAX Stations
- Existing TRAX Stations

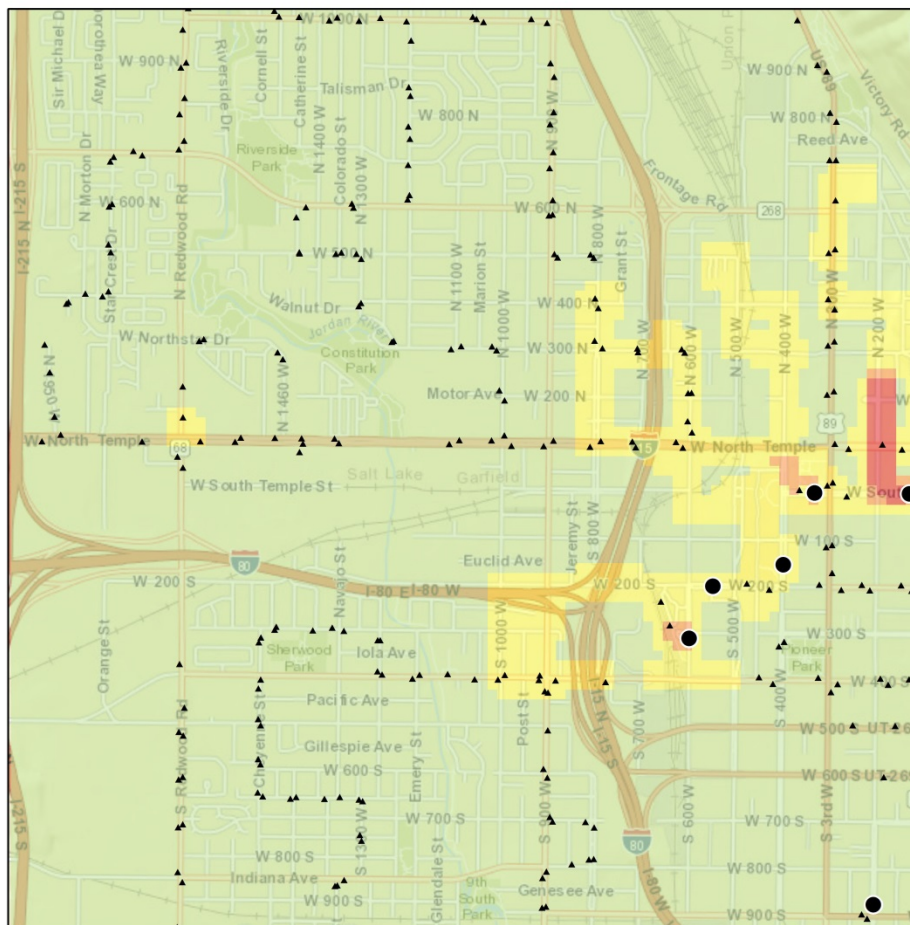


2013 Transit-related  
Physical Activity  
CONTINUED

- No GPS Points
- Low Density
- Medium Density
- High Density

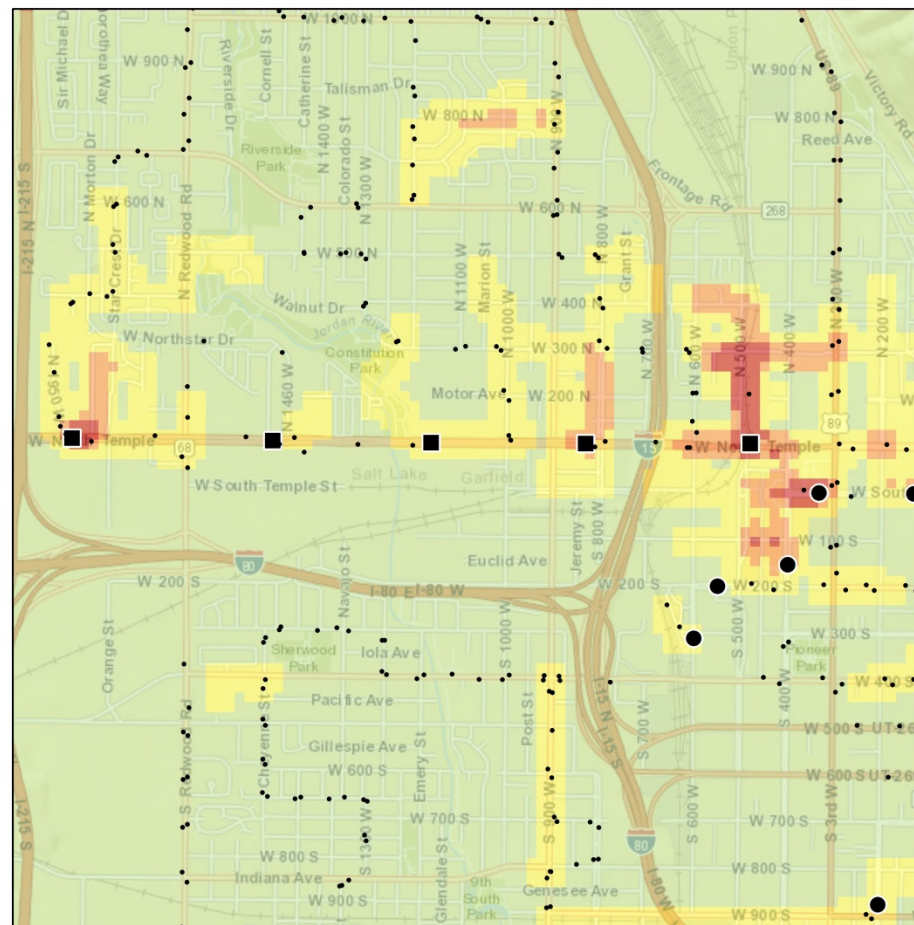
Basemap: Esri; Transit stops: AGRC.  1 Kilometers





2012 Transit-related  
Physical Activity  
FORMER

- ▲ 2012 Bus Stops
- 2013 Bus Stops
- New TRAX Stations
- Existing TRAX Stations



2013 Transit-related  
Physical Activity  
NEW

- No GPS Points
- Low Density
- Medium Density
- High Density

Basemap: Esri; Transit stops: AGRC.   1 Kilometers



## Summary

User behavior (2013 vs. 2012)	PA-Total	PA-Transit	PA-Other
<b>Did not change</b> (Never; Continuing)	No change	No change	No change
<b>Stopped using transit</b> (Former)	Decrease	Decrease	No change
<b>Started using transit</b> (New)	Increase	Increase	No change

No confounding factors

No substitution for non-transit PA

→ Transit PA is *new* PA!

## Opportunistic GIScience

### MAPS: a “one-off” experiment with a limited study horizon

Anticipated event

New LRT and street rehabilitation

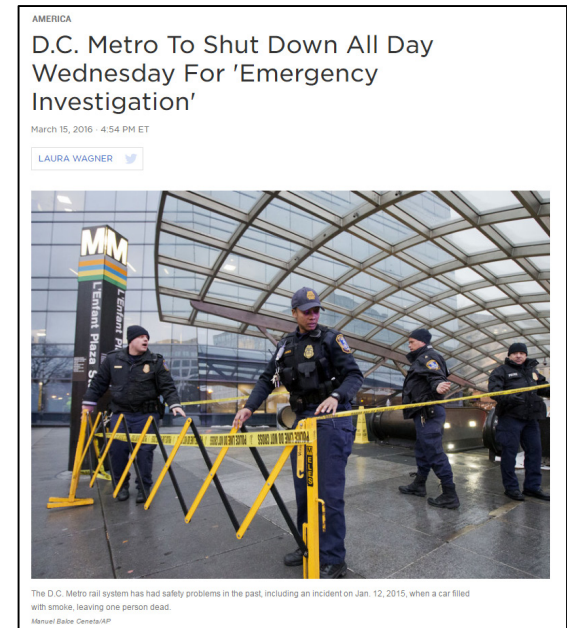
Planned for years

Five-year study

No follow-up activity measurement

What about **unanticipated** events?

What about policies with **complicated**  
**outcomes and time horizons?**





# Opportunistic GIScience

## Observatory science

### What?

**Ongoing** data collection based on a favored view, supported by technology and organizational processes

### Why?

**Discovery:** Generate new, surprising hypotheses

**Dynamics:** Complex multi-scale dynamics

**Monitoring and policy:** e.g., volcano observatories

**Ready when something happens!**



<http://www.brown.edu/>





# Opportunistic GIScience

## Human observatories

## Social observatories

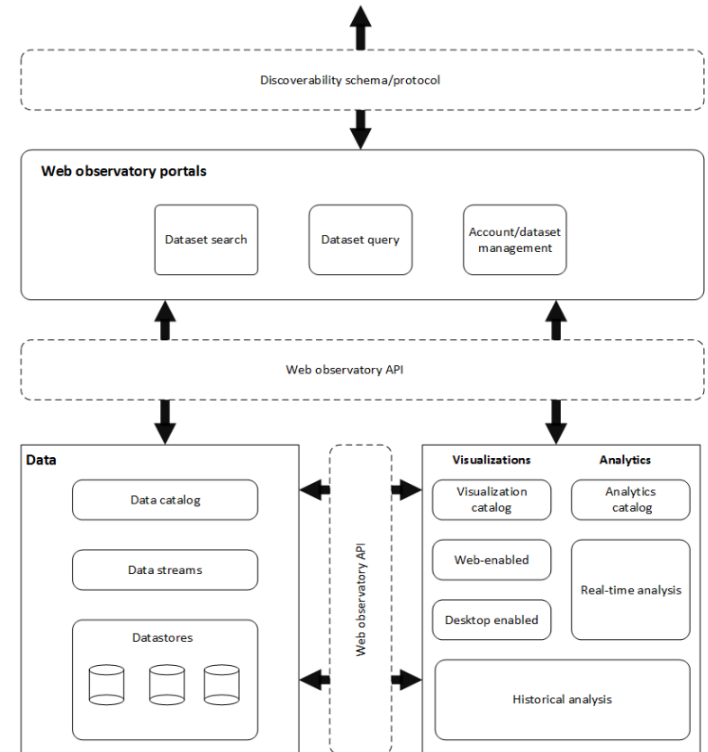
Collect and integrate data about social phenomena

Often focusing on a place or region

## Web observatories

Collecting, sharing, querying and analysis of Web data

Middleware for **broad data** – complex data from diverse sources



Architecture for a real-time web observatory (Tinati et al. 2015)

# Opportunistic GIScience

## Geographic Information Observatories (GIOs)

### What?

Integrate and analyze **broad geographic data** associated with a place or region

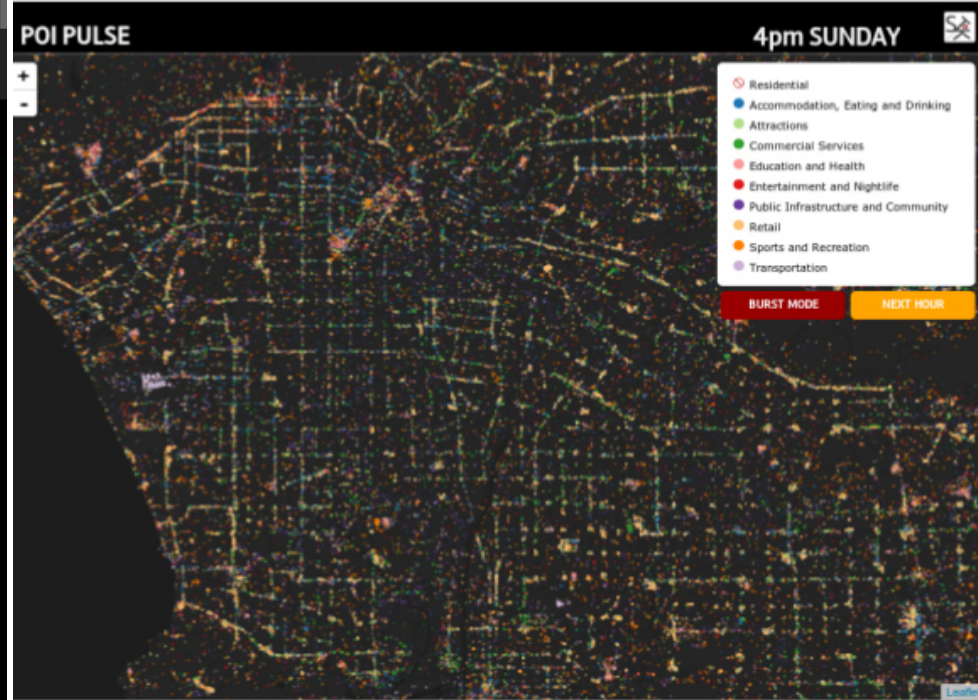
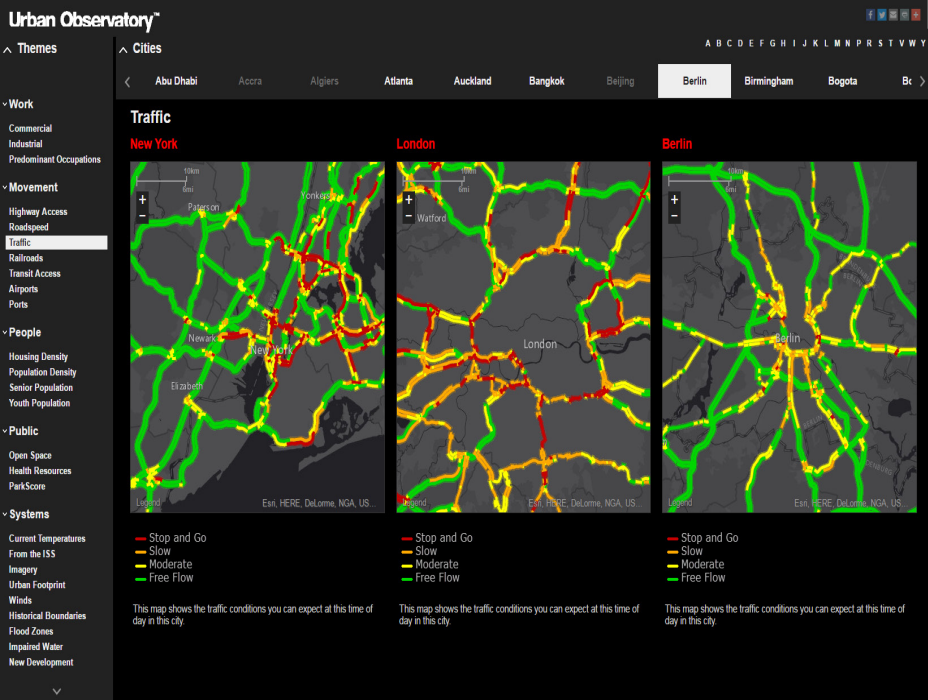
### Why?

**Observatory of geographic information (OGI):** Explore and analyze the poorly understood geographic information universe

**Geographic observatory (GO):** Study the real world through the geographic information it produces and consumes



POI Pulse (McKenzie et al. 2015)



## GIO examples

Upper left: Urban Observatory Project  
Upper right: POI Pulse (McKenzie et al. 2015)

Lower left: Columbus Urban and Regional Information Observatory (CURIO)



# Opportunistic GIScience

## Opportunistic GIScience

Geographic knowledge discovery, construction and use that **leverages planned and unplanned events in the real world**

Facilitated by GIO capabilities for **persistent data collection and analysis**

Three facets

1. Opportunistic observation
2. Opportunistic experiments
3. Opportunistic decision-making



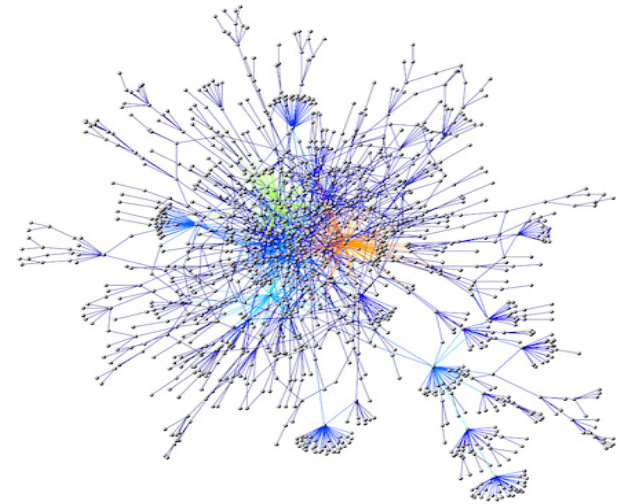
# Opportunistic GIScience

## Opportunistic observation

Human and coupled human-natural systems are complex

→ Geographic context matters!

→ History matters!



Ongoing, parallel observation is crucial

- Different geographic contexts
- Time periods sufficient to elucidate temporal dynamics
- Ready when surprises occur



# Opportunistic GIScience

## Opportunistic experiments

Leverage broad geographic data

## Mundane events

Baseline measures

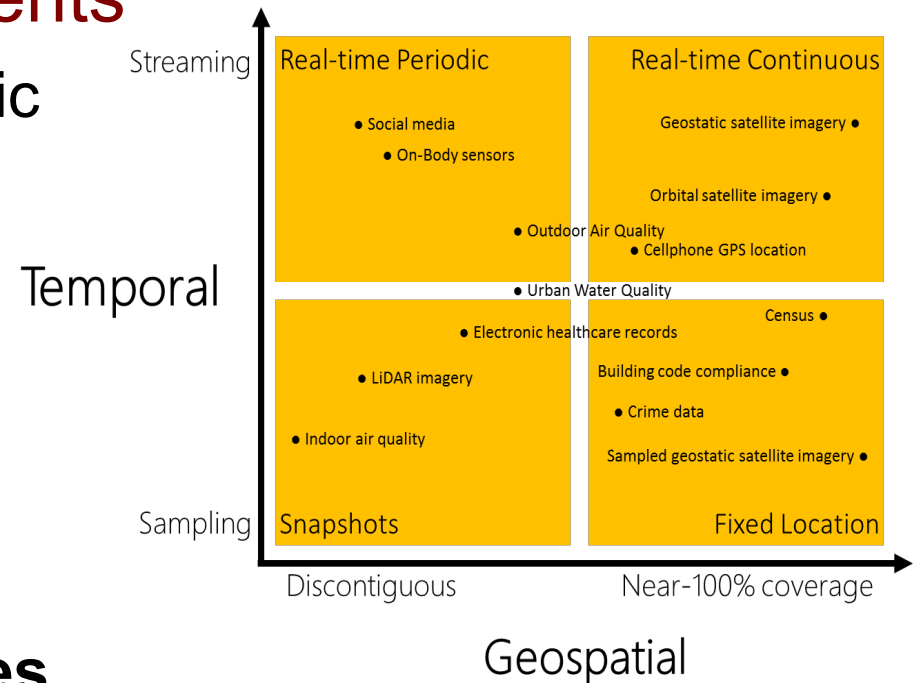
## Unplanned events

Natural experiments

## Complex policy outcomes

Non-binary outcomes

Multiple time horizons



Miller HJ and Tolle K (2016) "Big Data for healthy cities," *Built Environment*, 42, 441-456.

# Opportunistic GIScience

## Opportunistic decision-making

Learning from evidence can be slow with complex systems

- Despite lots of data!
- Feedback loops, delays, nonlinearities, self-organization and path dependence

## GIOs as virtual laboratories

- Agent-based modeling and geosimulation
- Integrate empirical and simulated data
- Perform experiments infeasible in the real world



www.futuretimeline.net

# Opportunistic GIScience

## GIOs as Mirror Worlds (Gelernter 1993)

Virtual reality **tightly coupled to the real-world** via streaming data

- A real-time, comprehensive, detailed, interactive and discoverable portrayal of a real-world system

Tool for **investigating** and **managing** reality

- Help scientists, leaders and citizens work together to understand and manage complex real world systems



MC Escher - "Print Gallery" (c) 2004 The M. C. Escher Company





## Conclusion

### Human systems are complex

- Policy interventions have unintended consequences
- Designing healthy places is a good example

### Geographic data collection is much easier

- Allows researchers to design and execute real-world experiments, with stronger support for causality
- Opportunities for natural experiments are happening all the time!

### Next step: Geographic information observatories

- Broad geographic data associated with a place or region
- Persistent observation and analysis supports opportunistic observation, experimentation and shared decision-making



## Conclusion

### Issues and concerns

#### Experimental ethics

Innocent real-world experiments?

Ask Facebook!

#### Privacy

Differential privacy & indistinguishability

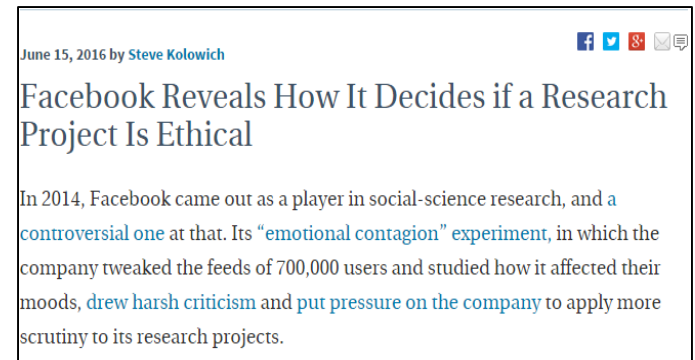
3D urban models as a metaphor and mechanism (Miller and Tolle 2016)

#### Data hubris

The parable of Google Flu (Lazer et al. 2014, *Science*, 343, 1203-1205)

Understand processes that generate data

Respect theory and administrative data





# Thank you!

## Papers (available at: [u.osu.edu/miller.81](http://u.osu.edu/miller.81))

Miller HJ (2017) “Geographic information observatories and opportunistic GIScience,” *Progress in Human Geography*, 41, 489-500.

Miller HJ and Tolle K (2016) “Big data for healthy cities: Using location-aware technologies, open data and 3D urban models to design healthier built environments,” *Built Environment*, 42, 441-456.

Miller HJ, Tribby CP, Brown BB, Smith KR, Werner CM, Wolf J, Wilson L and Oliveira MGS (2015) “Public transit generates new physical activity: Evidence from individual GPS and accelerometer data before and after light rail construction in a neighborhood of Salt Lake City, Utah, USA,” *Health and Place*, 36, 8-17.

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