

# INTEGRATING RISK SCIENCE AND URBAN PLANNING

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Lecture of Civil Systems  
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# My background

- BE (Hons), **Natural Resources Engineering**, University of Canterbury, New Zealand
- BSc, **Mathematics**, University of Canterbury, New Zealand

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- BSc, **Mathematics**, University of Canterbury, New Zealand
- MS, **Geography and Environmental Engineering**, Johns Hopkins University, Baltimore, MD, USA
- PhD, **Industrial and Operations Engineering**, University of Michigan, Ann Arbor, MI, USA

# My background

- Returned to UC as a lecturer in civil systems engineering



# Why a systems approach

# Maladaptation

Maladaptation occurs when actions unintentionally **increase** the **vulnerability** of a community

# Systems approach: integration

- Address issues simultaneously
- Can identify maladaptation potential
- Solutions that create synergies

# Risk science

- Evolution of risk over time
- Capturing consequences temporally (resilience)
- Repeated hazards and human feedbacks



# Prioritising intervention

- How do we capture the co-benefits and trade-offs?
- How do we evaluate alternative strategies?
- How do we address deep uncertainty?

“Research is needed on the effect of, and the dynamics between, adaptation alternatives for .. cities. Complex and dynamic feedback systems can result in seemingly **intuitive infrastructure solutions resulting in maladaptation.** ...” (Prieur-Richard et al., 2018)

# Overview

**Agent-based  
modeling**

**Machine learning  
and spatial data**

**Urban resilience  
framework**

Integrating **risk analysis** and **data-driven planning** for  
**urban resilience**

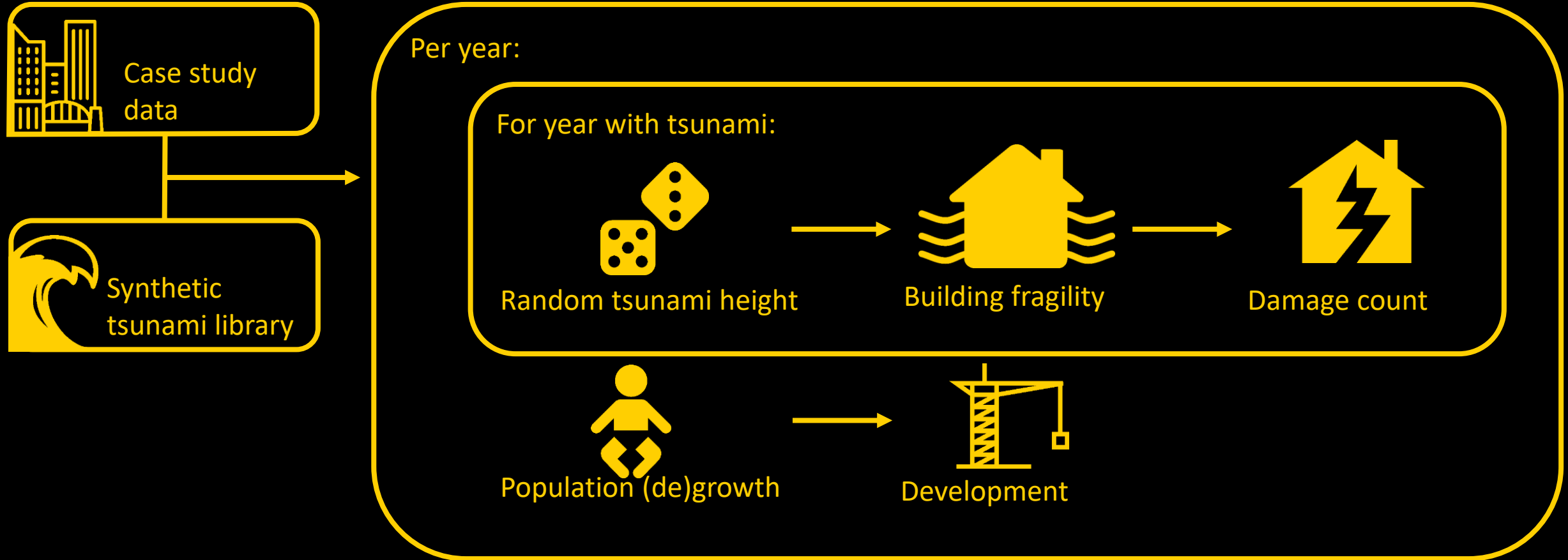
# Agent-based modeling and stochastic simulation

- Incorporating feedbacks
  - Behaviour
  - Time

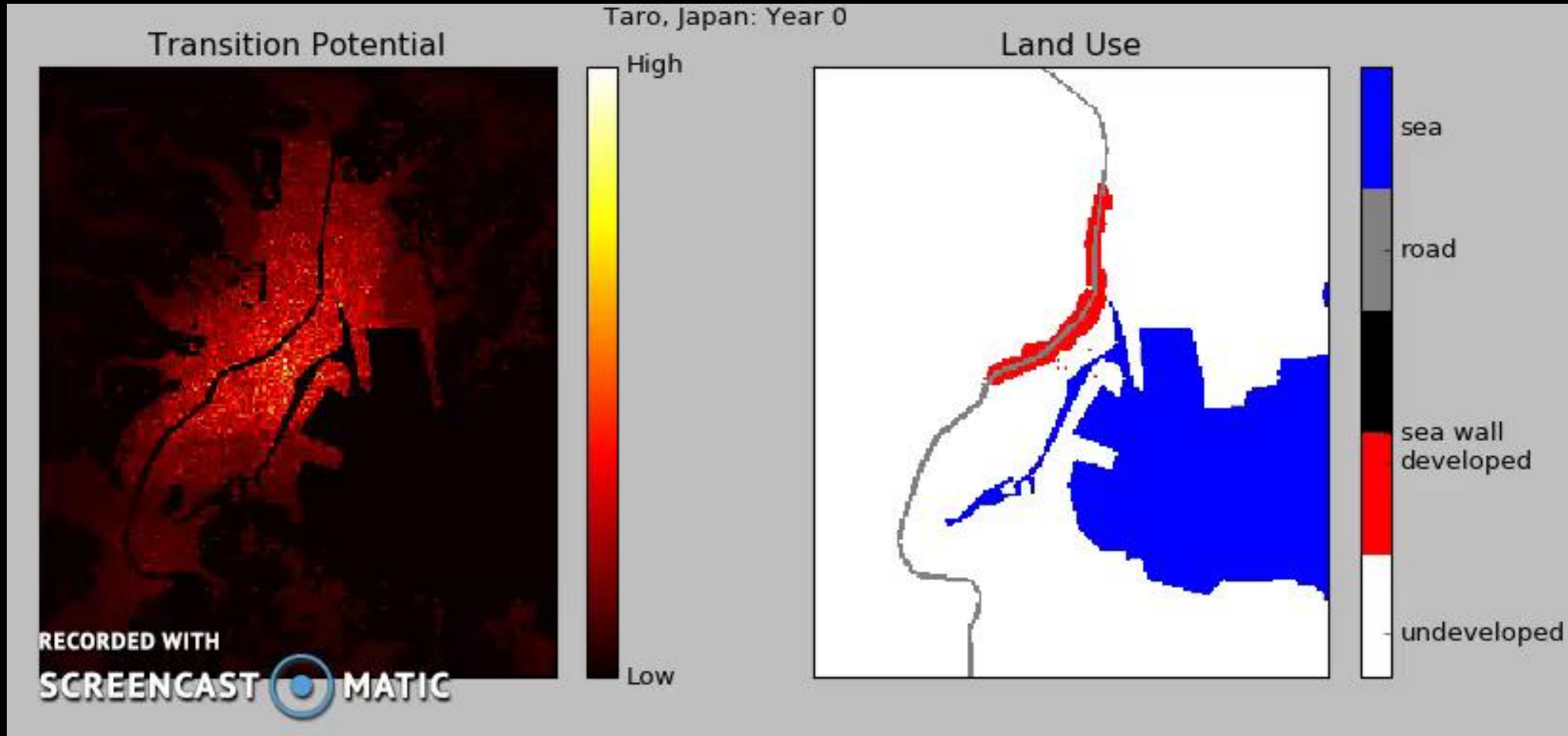
Logan, T. M., Guikema, S. D., & Bricker, J. D. (2018). Hard-adaptive measures can increase vulnerability to storm surge and tsunami hazards over time. *Nature Sustainability*



# Model



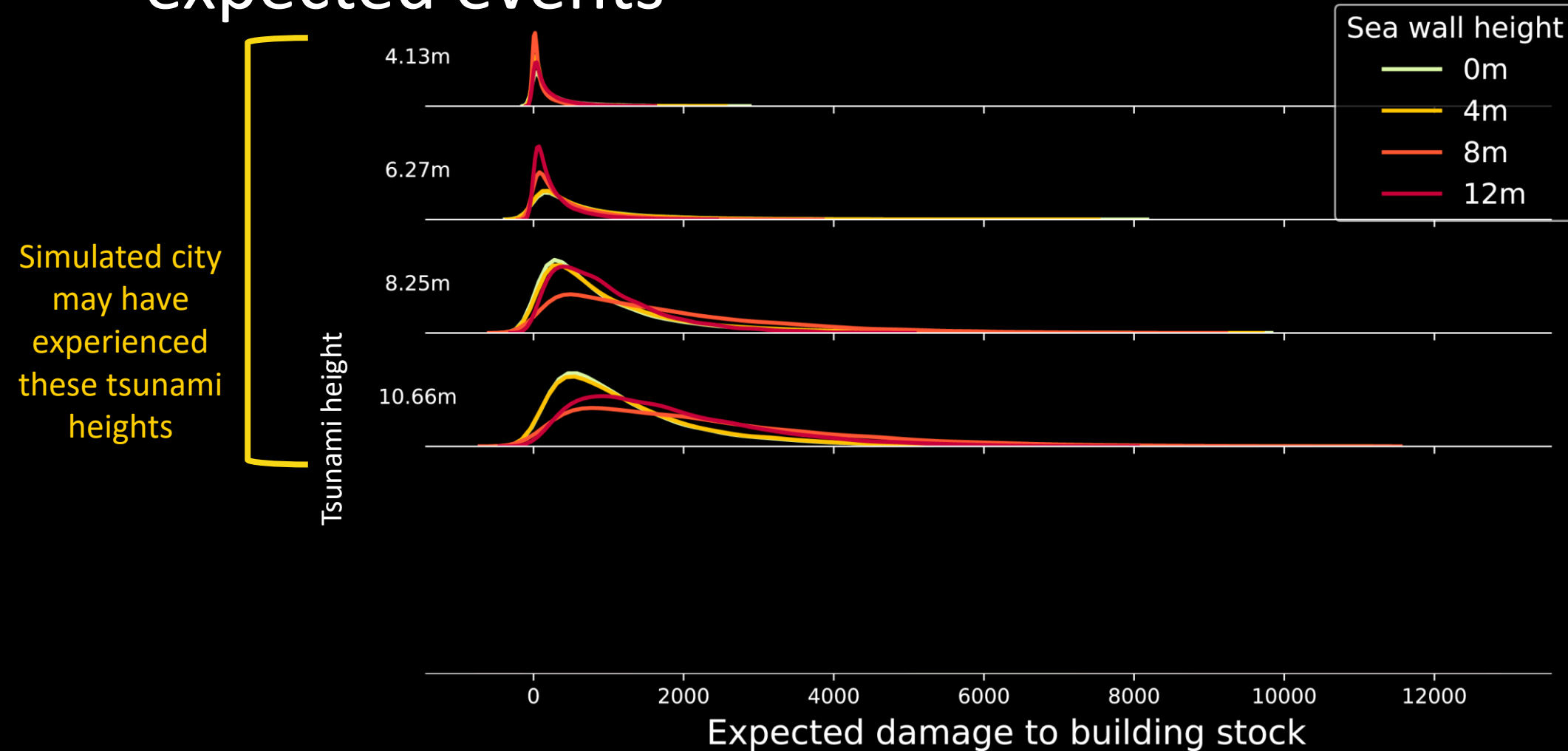
# Our model for land-use, tsunami, & human memory



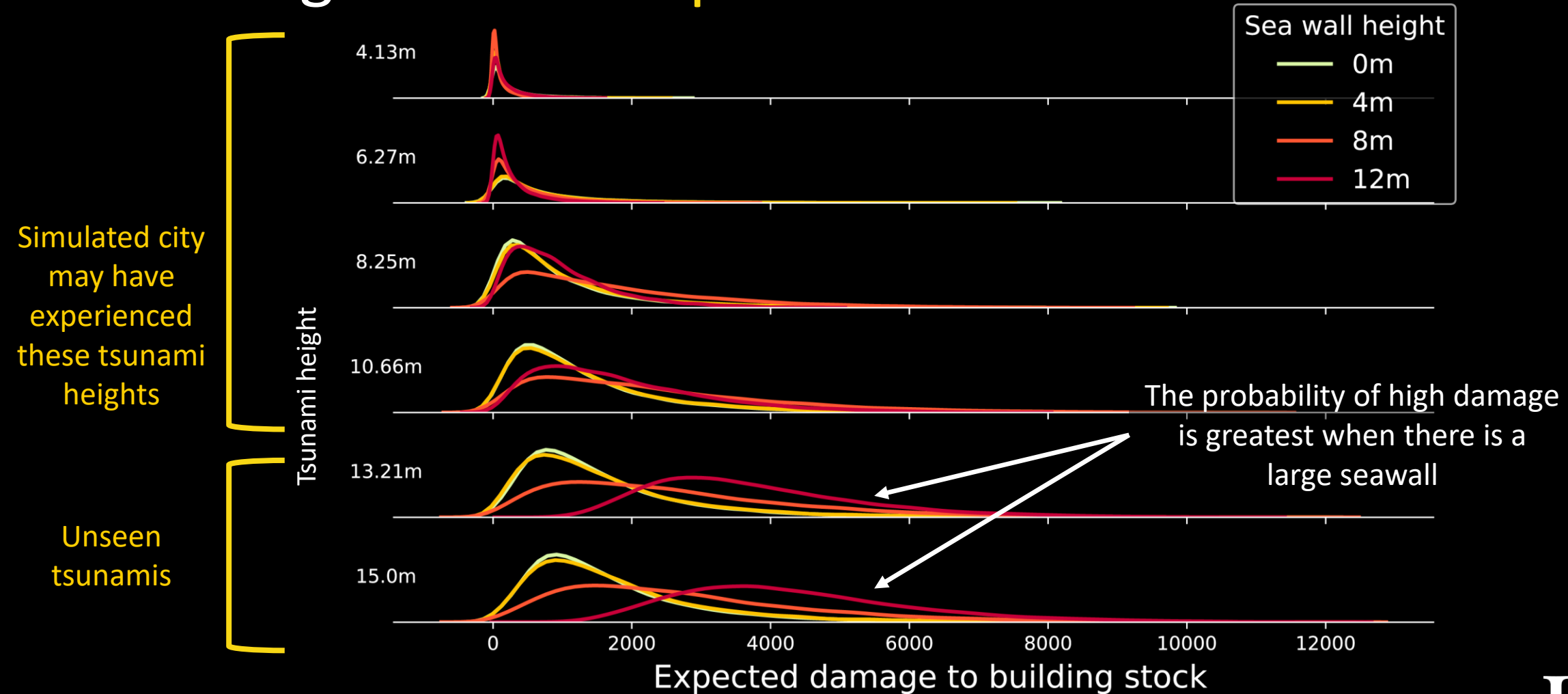




# Seawalls don't make much difference for expected events



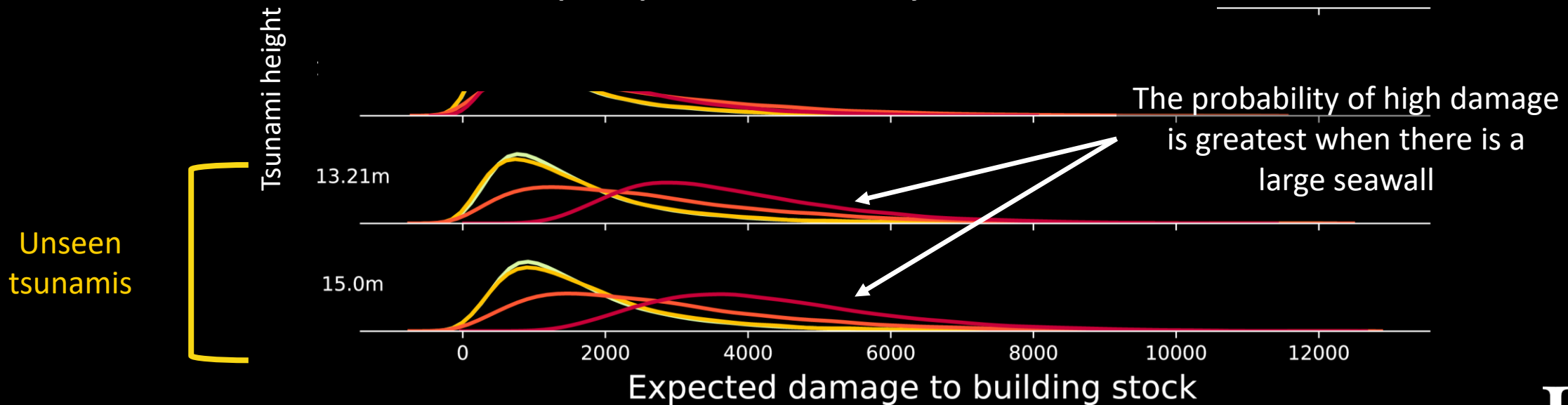
# Communities with large seawalls have more damage from **unexpected** events



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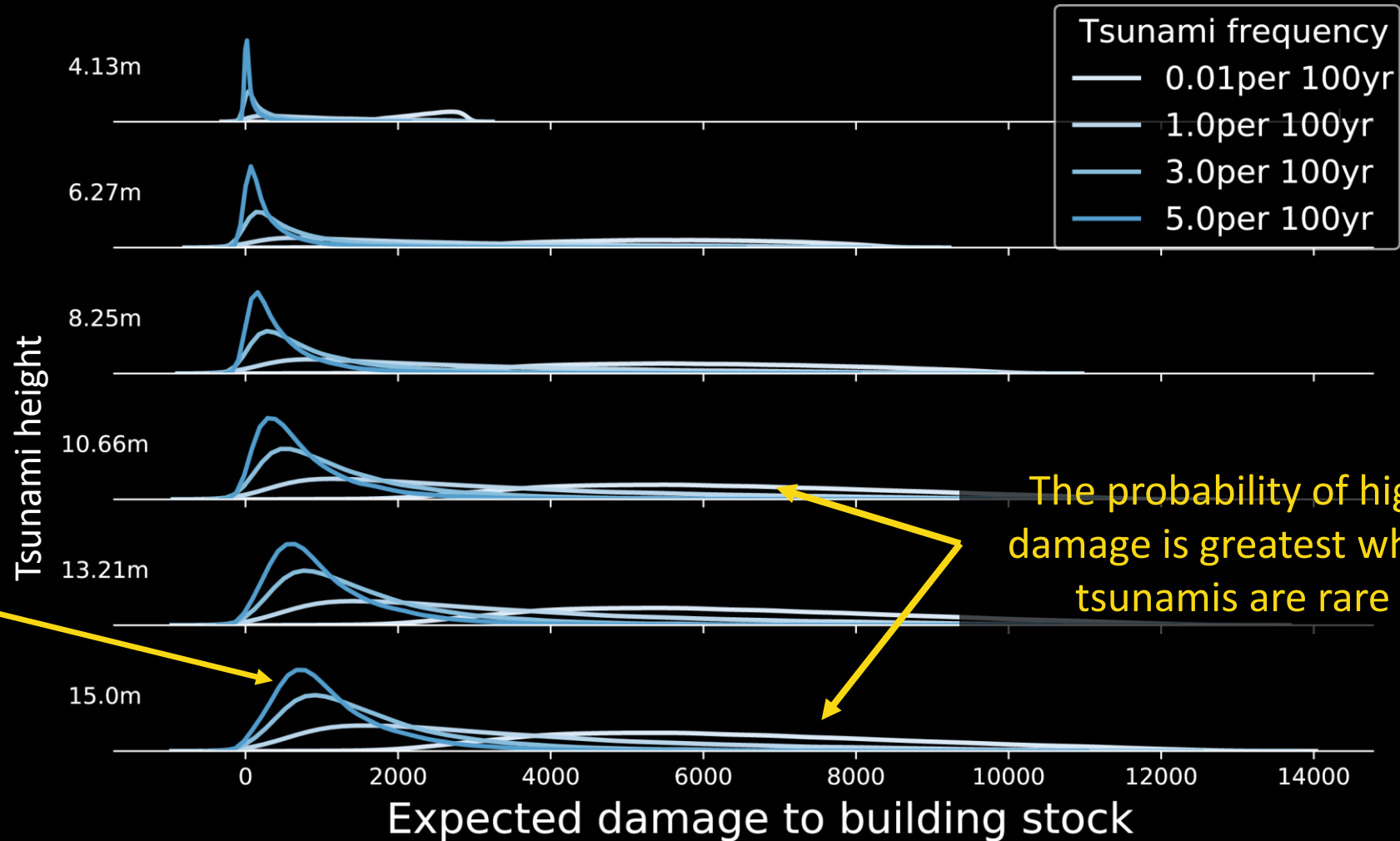
## Large walls

- Protect against smaller events
- Prevent community from **learning**
- It makes people **think** they're safe





# More frequent events **reduce** vulnerability





events that raise the salience of risk will make people and governments more likely to pay attention to that risk, act on that risk or demand action from others — **regardless of whether such actions actually reduce risk**

(Anderson et al. 2018)

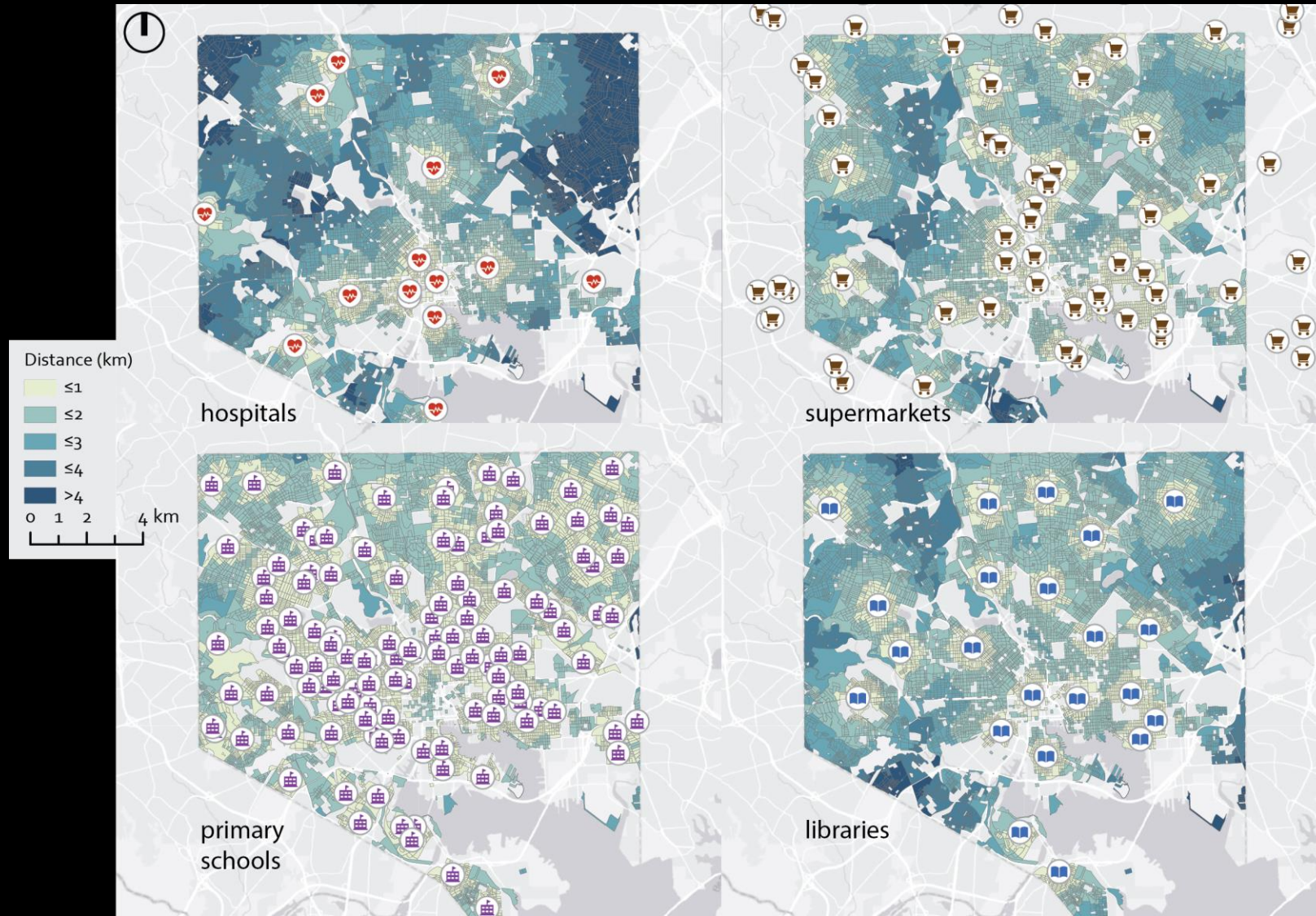




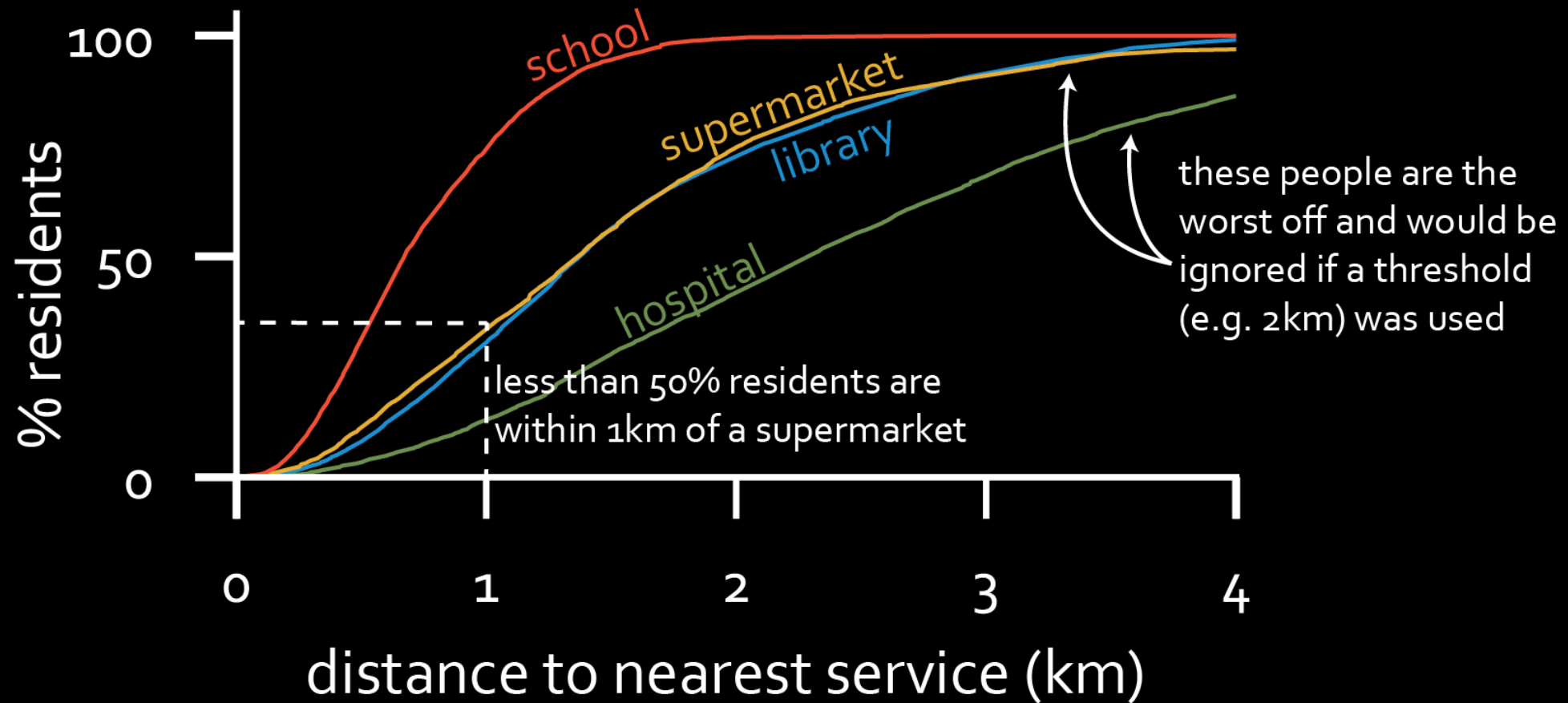
# Community resilience

- We need an approach that captures the impact of hazards and infrastructure disruptions on people
  - Logan, T. M., Williams, T. G., Nisbet, A. J., Liberman, K. D., Zuo, C. T., & Guikema, S. D. (2019). Evaluating urban accessibility: leveraging open-source data and analytics to overcome existing limitations. *Environment and Planning B: Urban Analytics and City Science*
  - Logan, T. M., & Guikema, S. D. (under review). Reframing Resilience: Equitable Access to Essential Services. *Risk Analysis*

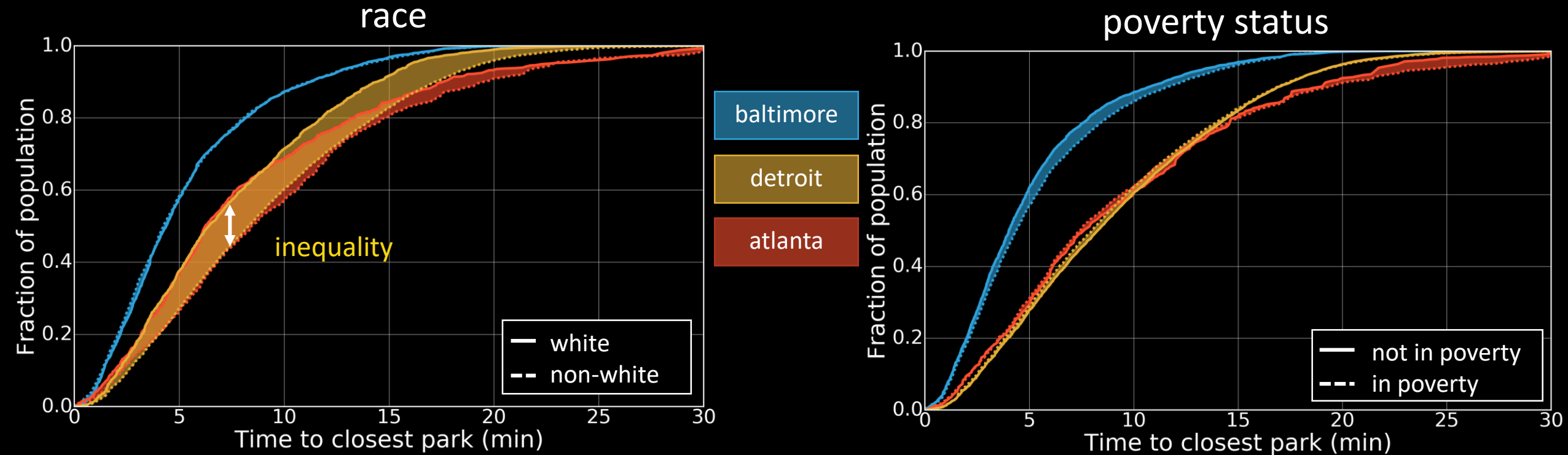
# We calculate the walking distance from every house to every service within a city



# Using census data we calculate the city-wide distribution for access to services



# Using census data we can compare the access to services for different demographic groups



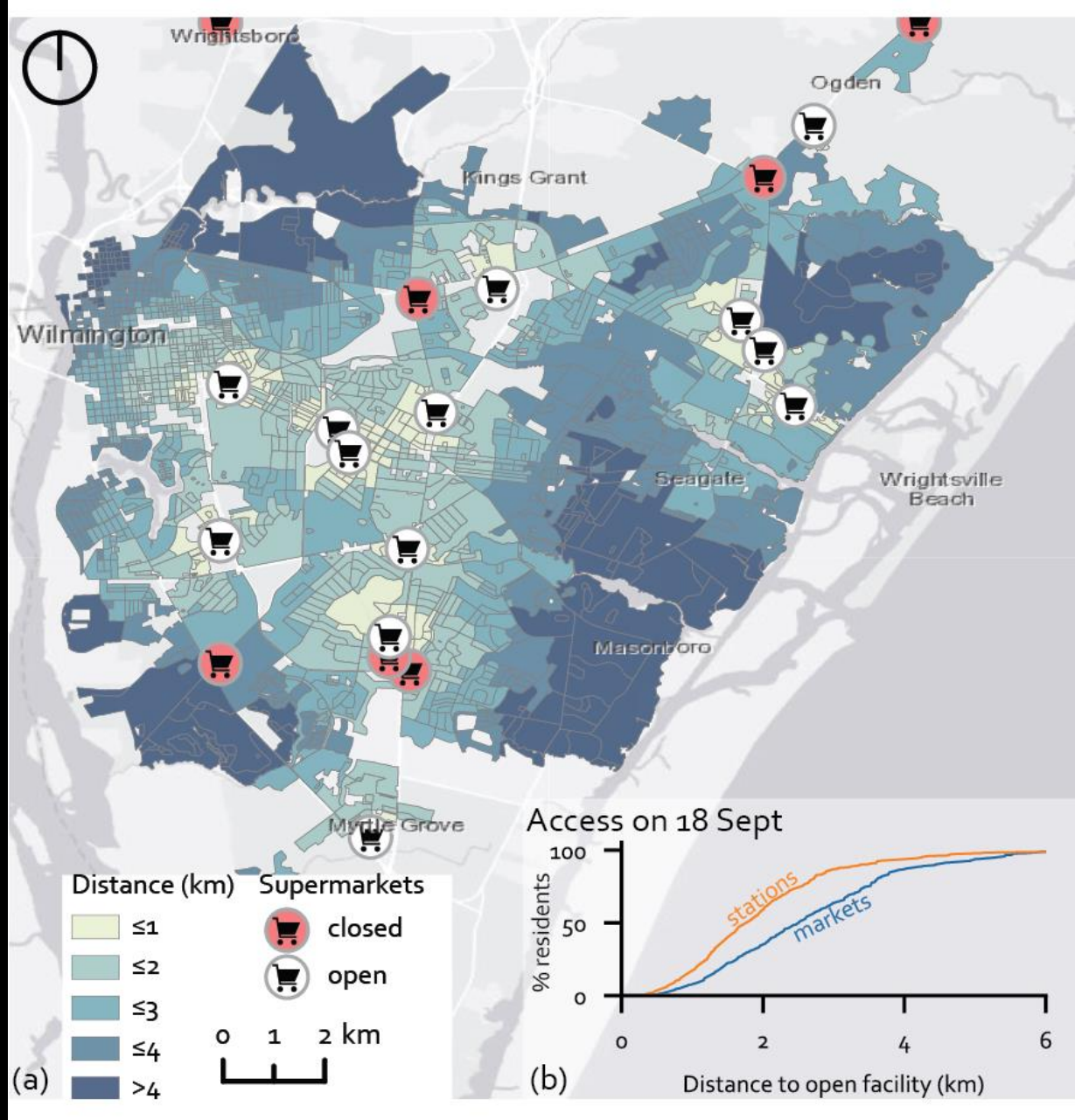


# We must use high spatial resolution otherwise access-poor people are overlooked



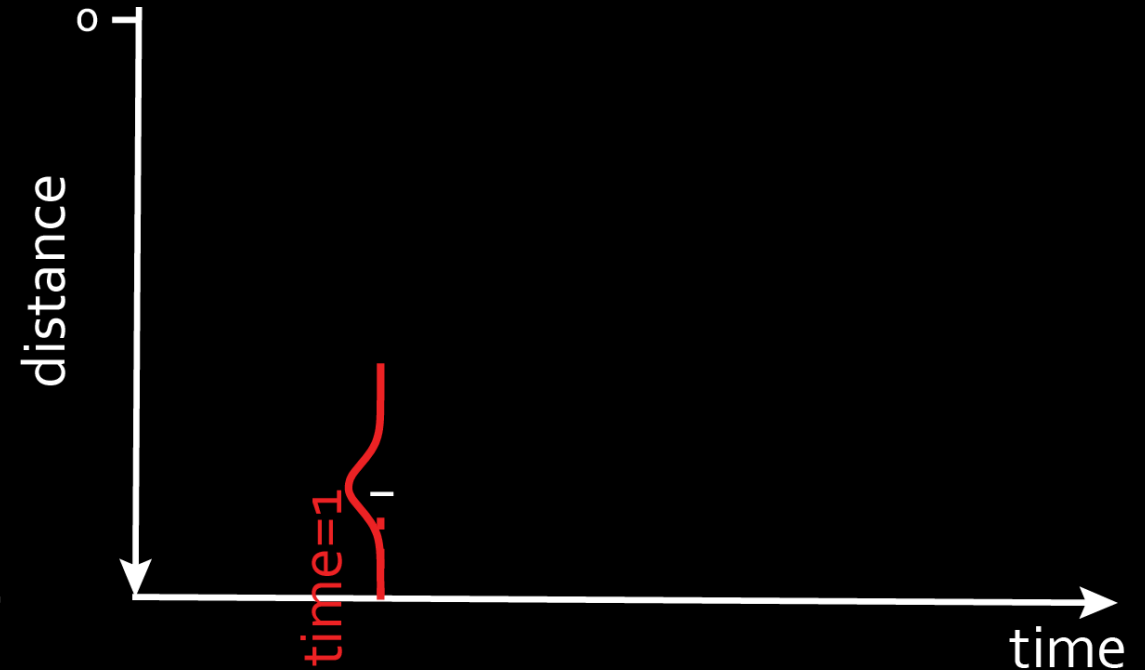
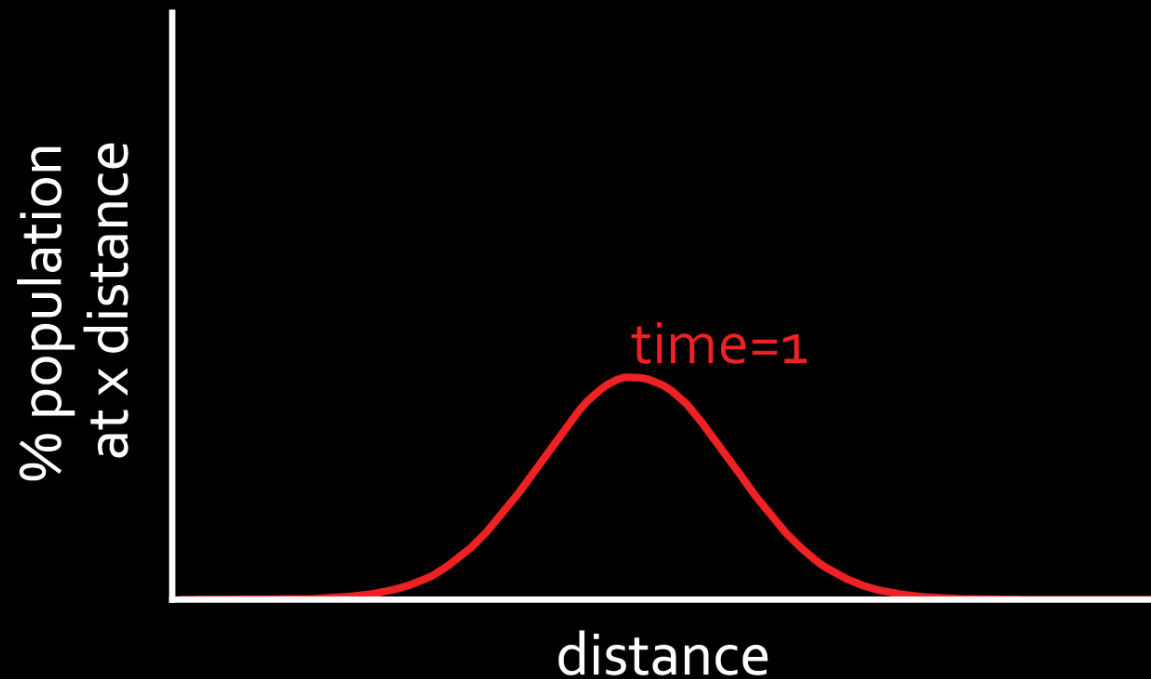




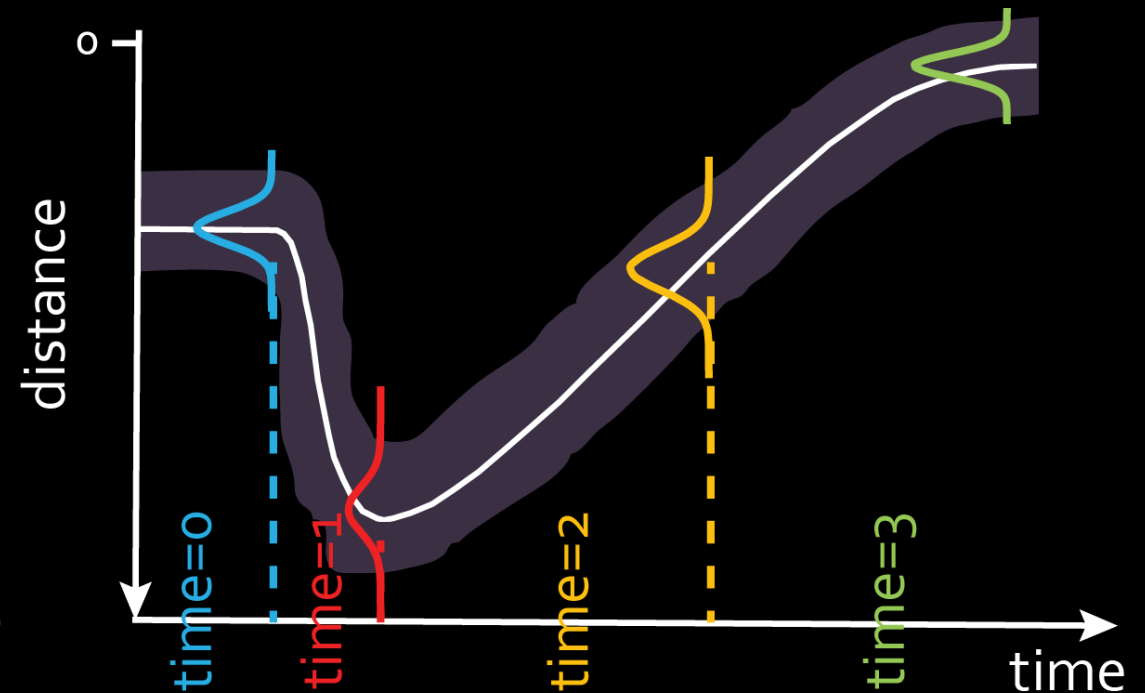
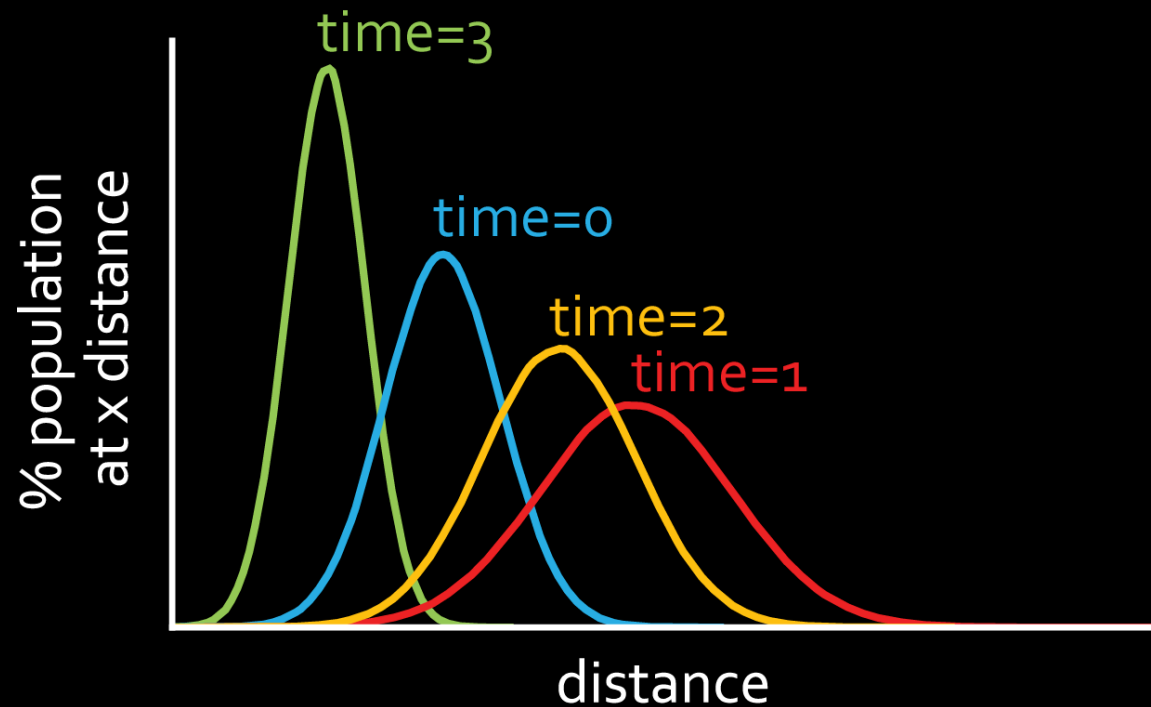


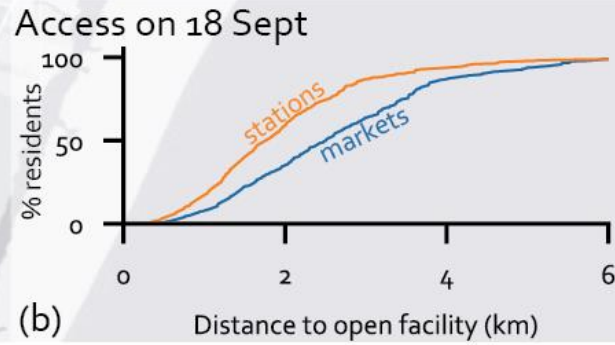
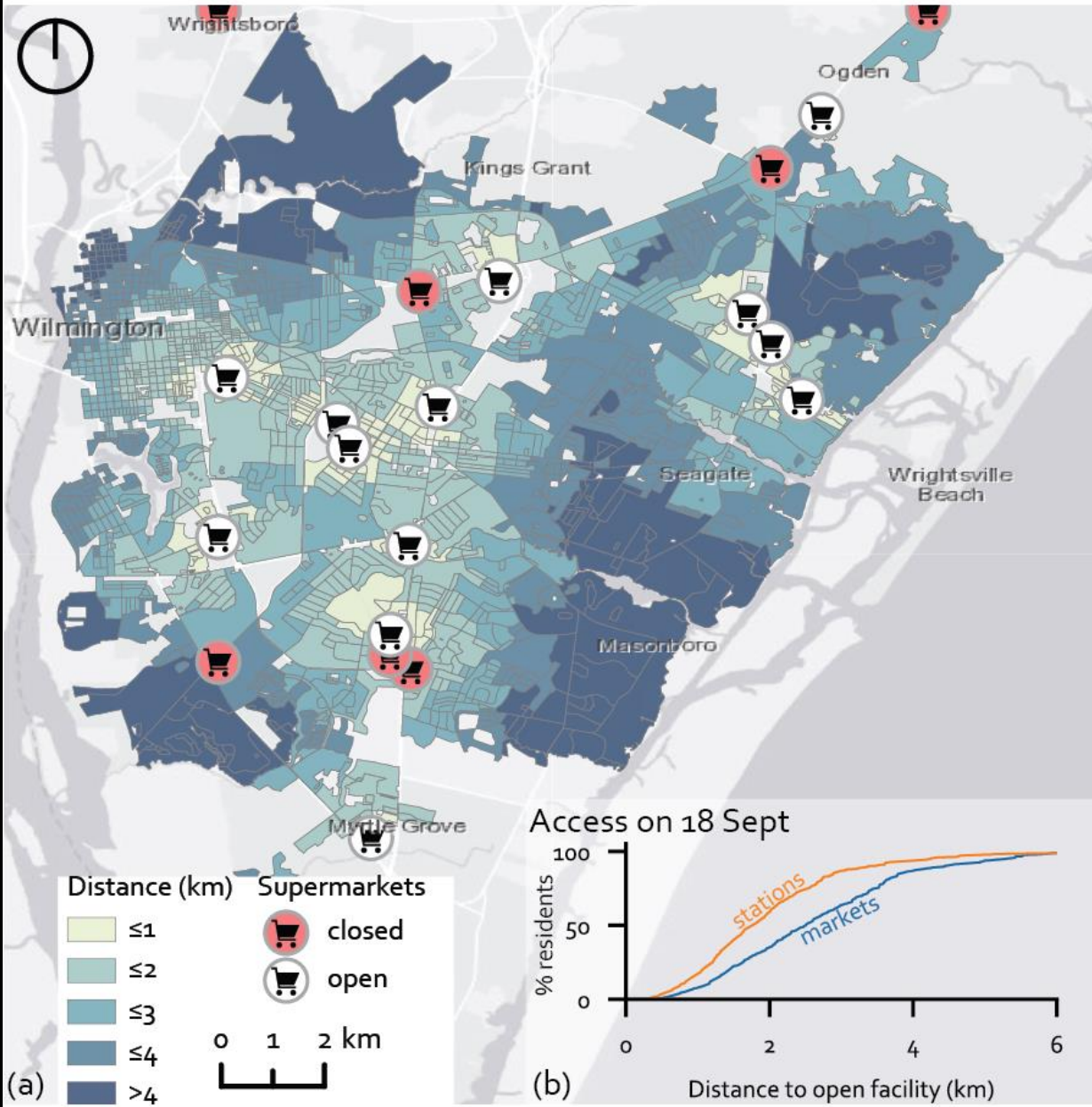


We map the distance distribution to a resilience curve by including time

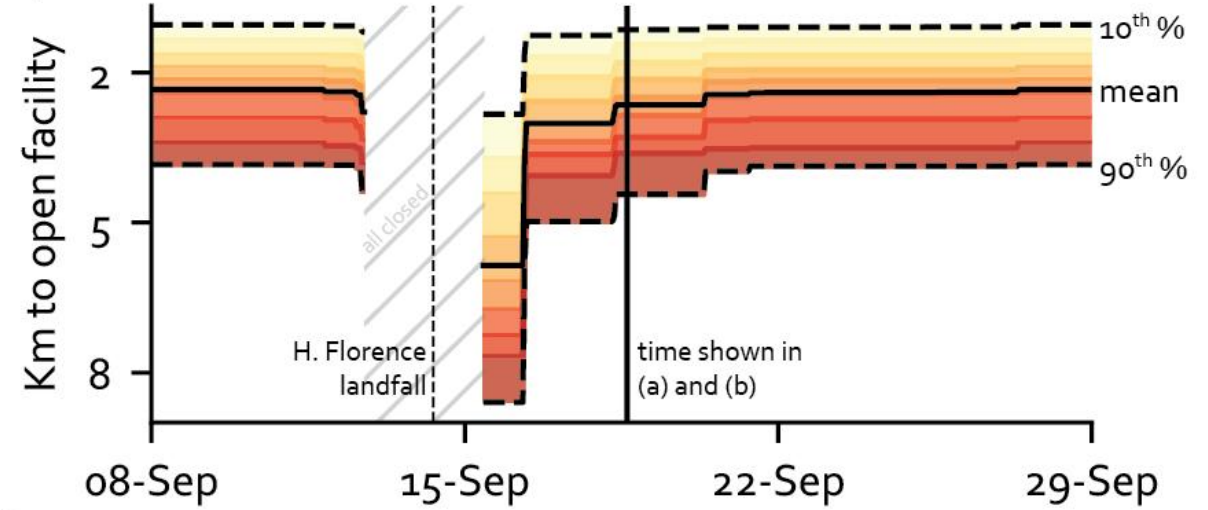


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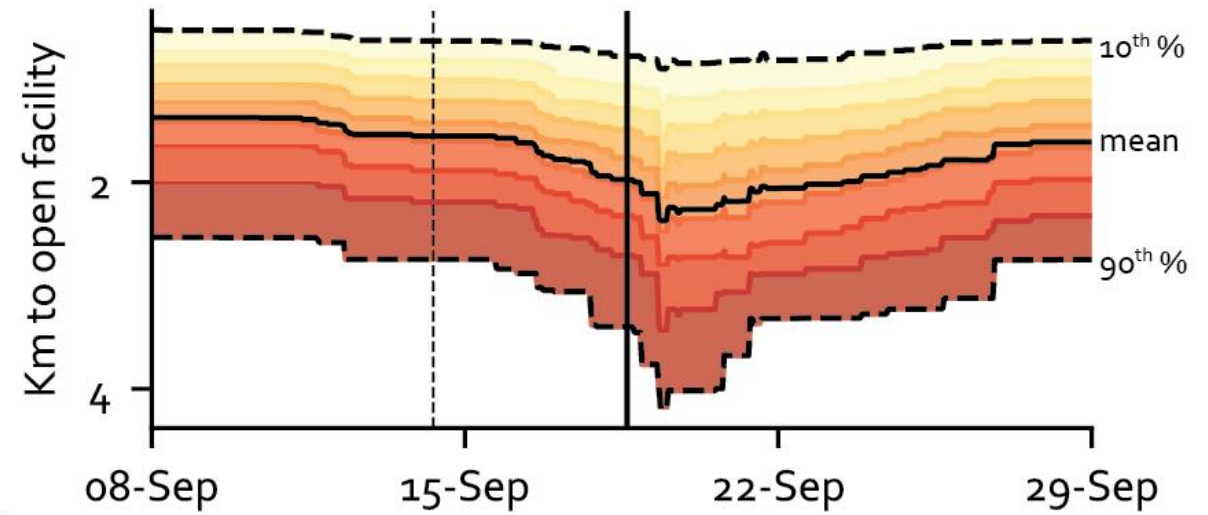




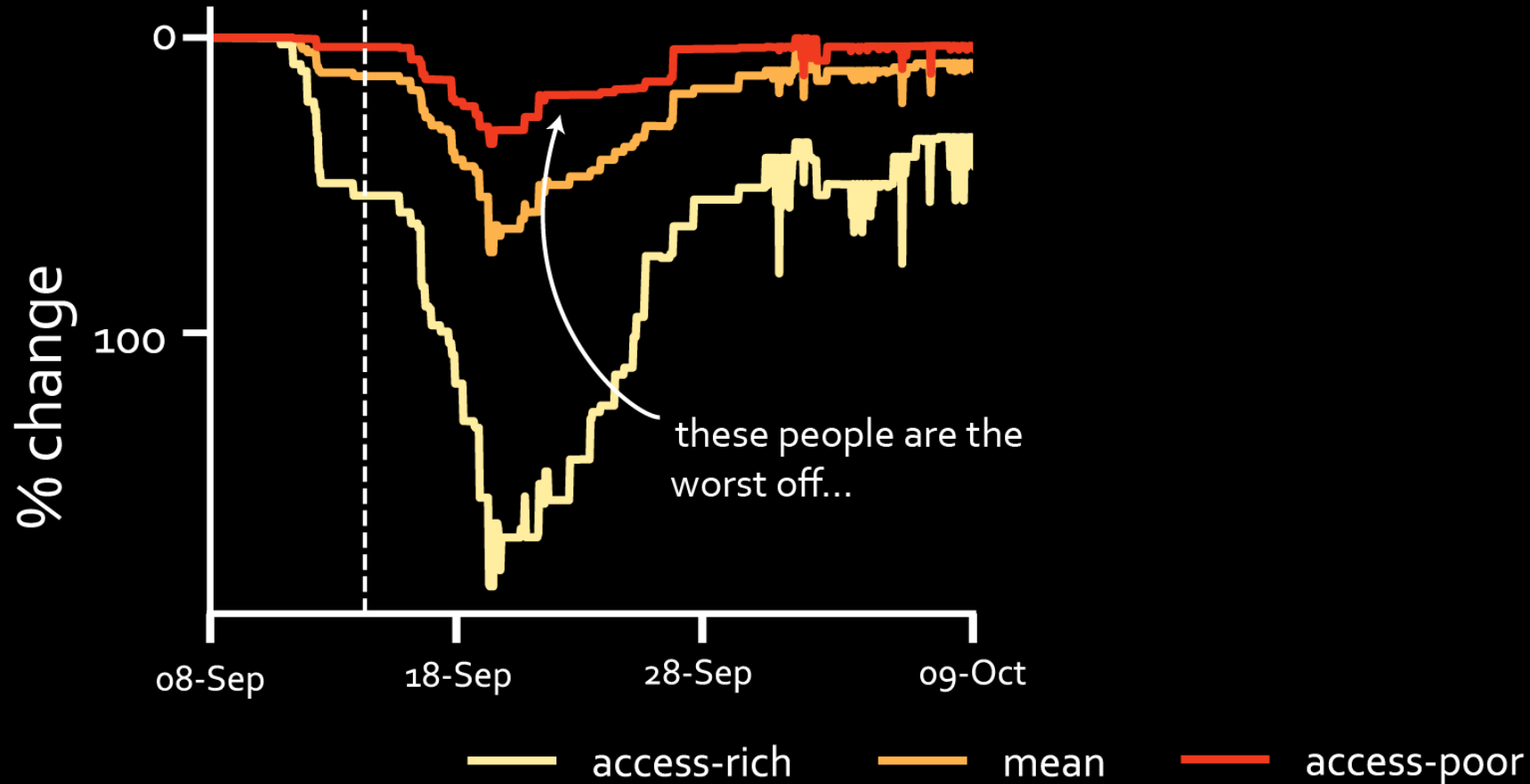
### Supermarkets



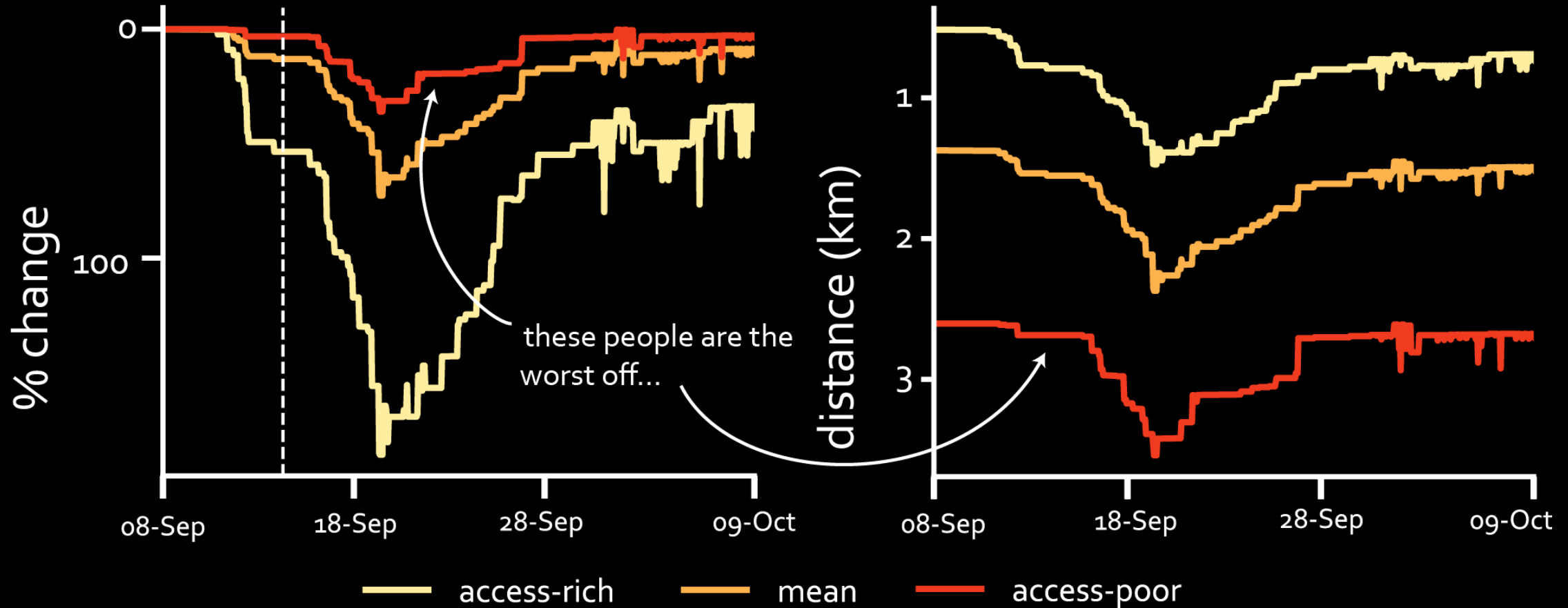
### Service stations



# How the resilience measure is calculated can perpetuate inequality

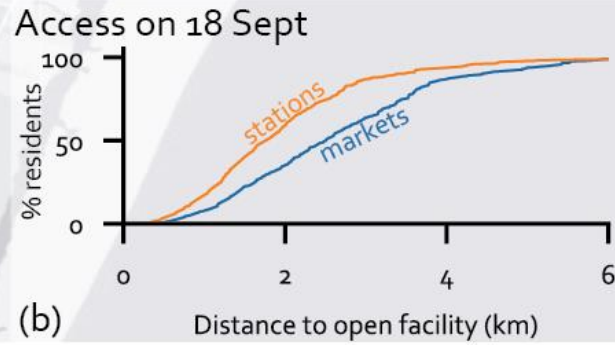
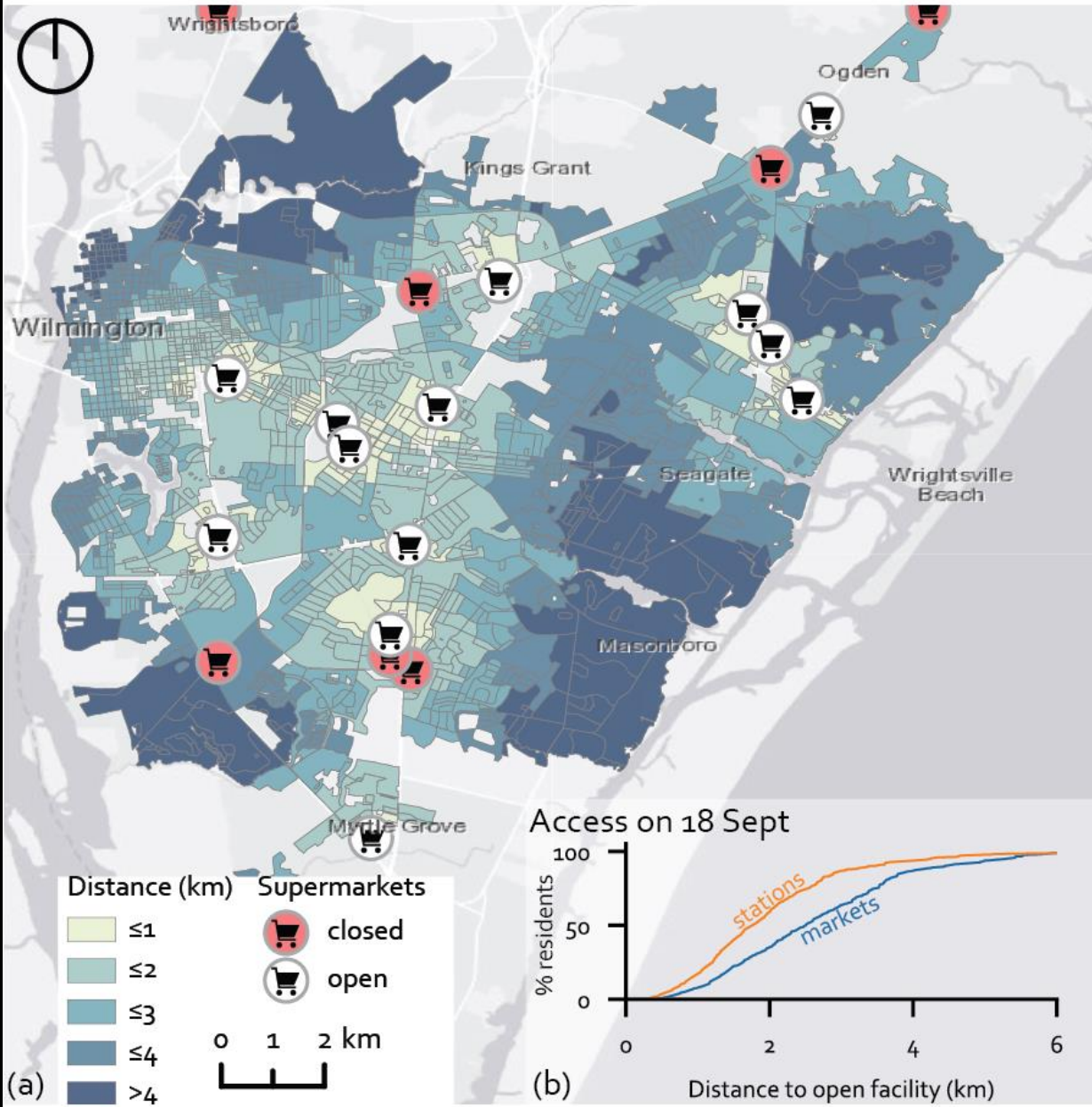


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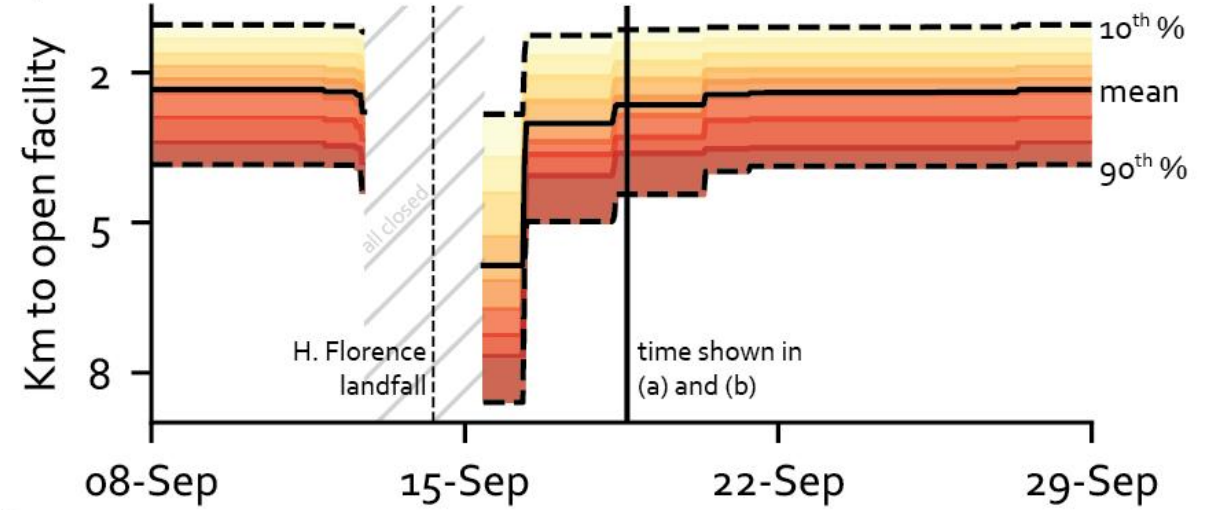




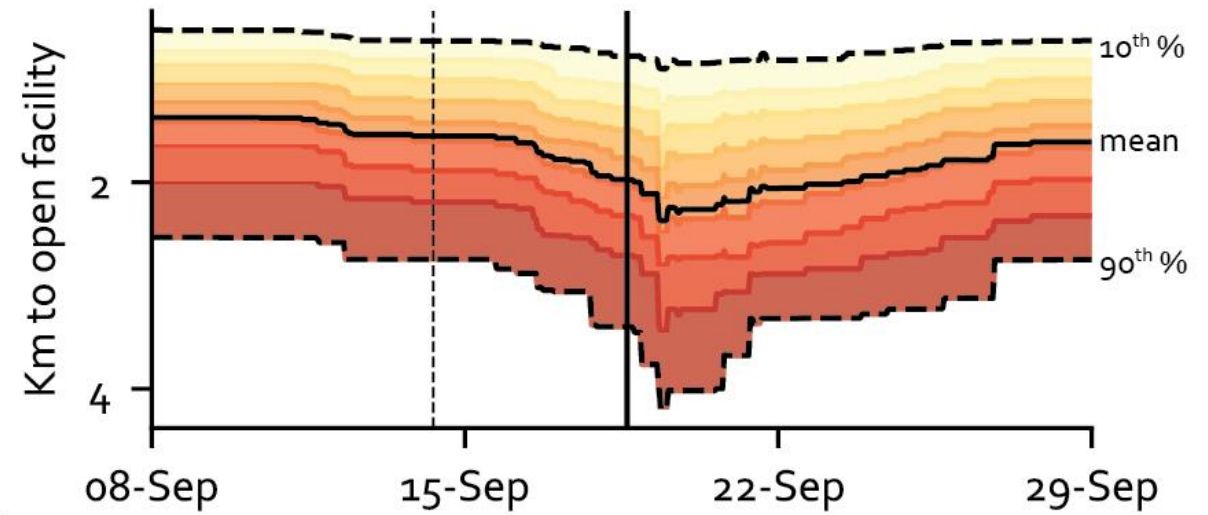




### Supermarkets



### Service stations



# Risk science

- Contemporary risk analysis tools can be used for resilience analysis
  - Yet there are moves to silo the two
- Working on papers that
  - Clarify how risk deals with time
  - Describe how resilience analysis fits within a risk framework
  - Identify how resilience metrics dictate the equity of interventions



# Machine learning and spatial data science

- Investigating the land characteristics that drive high urban land temperatures using a variety of AI models
  - Logan, T. M., Zaitchik, B., & Guikema, S. D. (under review). Night and Day: What is the influence and relative importance of urban characteristics on land surface temperature? *Remote Sensing of Environment*.
- Evaluating the effect of density on public health in US cities