

# MSDs and Urban Complexity

Significance of social science

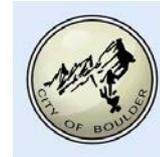
## Multi-Sector Dynamics Workshop

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# Significance of embedding science- and engineering- studies in behavioral science

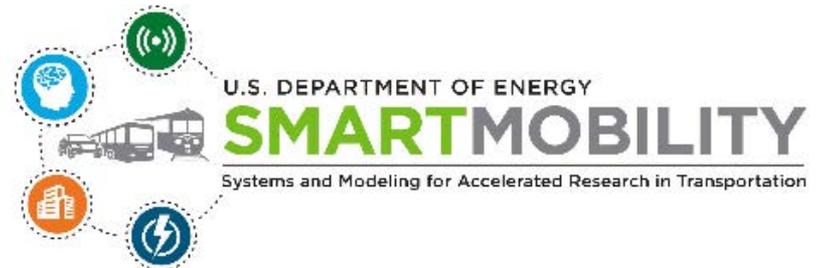
## I. Interdependencies and cascading effects



## II. Emerging technologies, ABMs and behavioral science



## III. Urban typology patterns and variations in outcomes



# I. Interdependencies mediating risks of cascading effects

An aerial photograph showing a two-story house with a brown roof and light-colored walls, partially submerged in a turbulent, greyish-brown river. The house is tilted and appears to be in the process of being washed away. The surrounding area is lush with green trees and vegetation, and a road is visible on the right side of the image.

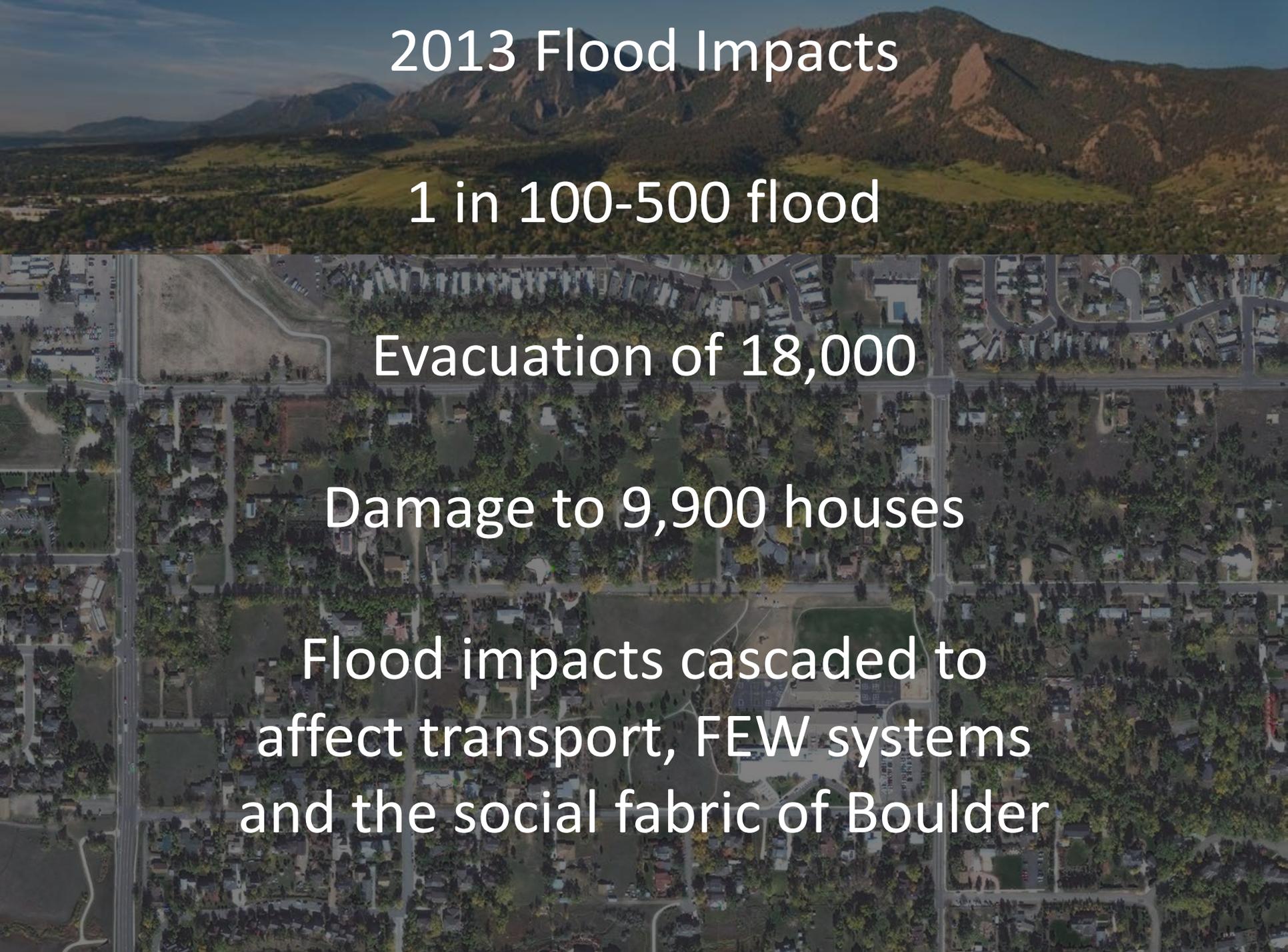
Participatory research based on

- Identification of gaps in engineering
- Integration of social science's approaches

I. Understanding of interdependencies that amplify or mitigate these effects

II. Options/barriers for learning & action

# 2013 Flood Impacts

An aerial photograph of Boulder, Colorado, showing the city and surrounding mountains. The image is used as a background for the text. The mountains are in the background, and the city is in the foreground. The text is overlaid on the image.

1 in 100-500 flood

Evacuation of 18,000

Damage to 9,900 houses

Flood impacts cascaded to  
affect transport, FEW systems  
and the social fabric of Boulder

# Methods

Interviews/Fuzzy Cognitive Maps co-created with 17 respondents

- Food-energy-water, 4 cities, local to state jurisdictions, mountains to plains

## Jurisdictional Level and Sectoral Representation of Respondents

Level	Sectors							
	<u>Other</u>	<u>Food</u>	<u>Energy</u>	<u>Water</u>	<u>Emergency Response</u>	<u>Infrastructure</u>	<u>Public Health</u>	<u>Flood Recovery</u>
<b>Local</b>	#5	#12, #16	#16	#6,#11 #16	#16	#10		#14
<b>County</b>	#8, #10	#3, #15		#2	#1, #4, #17	#9	#1	#7
<b>State</b>					#13			

# Infrastructural interdependencies

A photograph showing a severe flood event. A paved road with double yellow lines is partially submerged and broken. Turbulent, brown water flows over the road surface and through a gap where the road has been washed away. Debris, including a large fallen tree trunk and various rocks, is scattered in the water. The background shows a steep, rocky hillside with some sparse vegetation.

**Transportation was the foundation of interdependency amplifying flood risk**

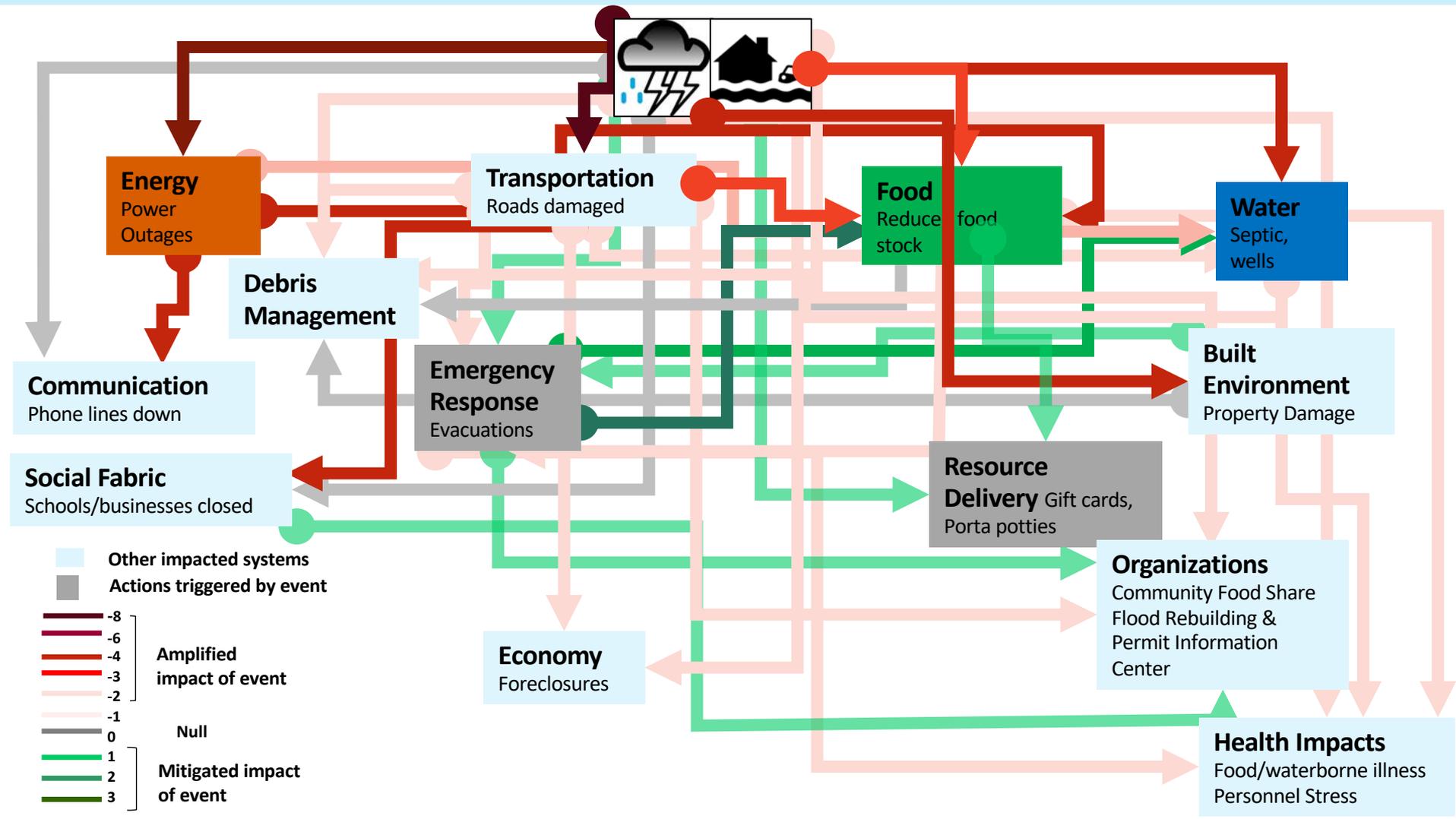


In many sectors, actions and policies mitigated the negative impacts of the floods

Short-term responses were effective, but longer-term recovery is harder



# Factors that amplified or mitigated cascading effects triggered by floods





**Through our framework, participants learned to see FEW systems connections, however,**

**Most actions occurred within organizations, which focused on one system at a time**



## II. Transformational Analysis through Behavioral Science: Symbiotic Autonomous Systems (SAS) in Cities

LDRD  
Patricia Romero Lankao and Bill Livingood (PIs)

# Symbiotic Autonomous Systems Augmented Minds

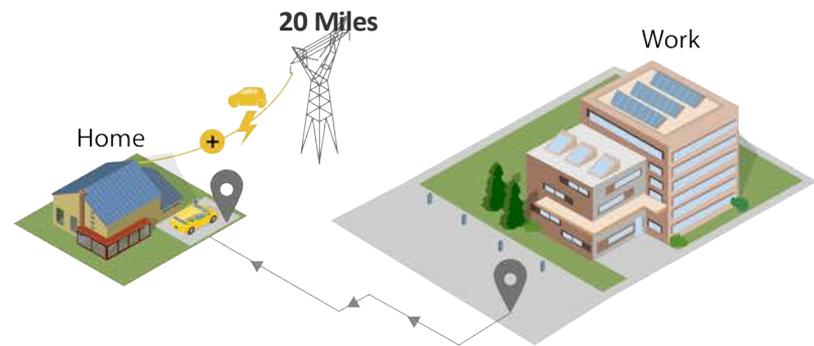
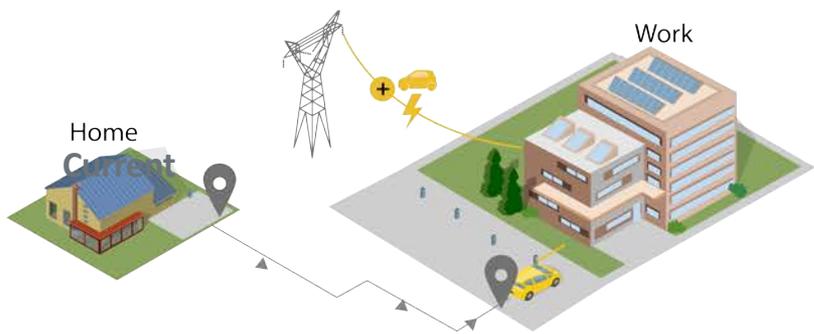


# Advance machines will interact with humans and cities to transform our lives and the fabric of our cities



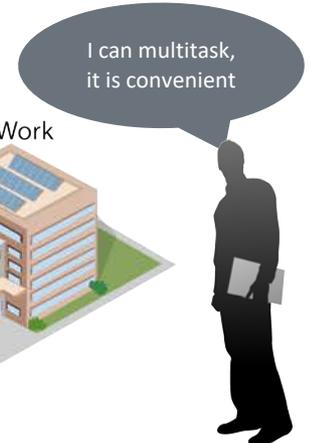
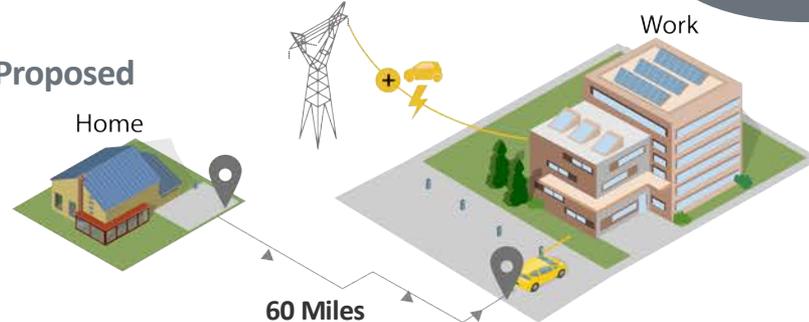
Autonomous Electric Vehicles:  
Transformed Urban Fabric

# Using current tools, NREL analyses assume no change in behavior, or at best, bake in static assumptions



Some strain on urban and suburban grids;  
no change to rural grids or building loads.

## Proposed

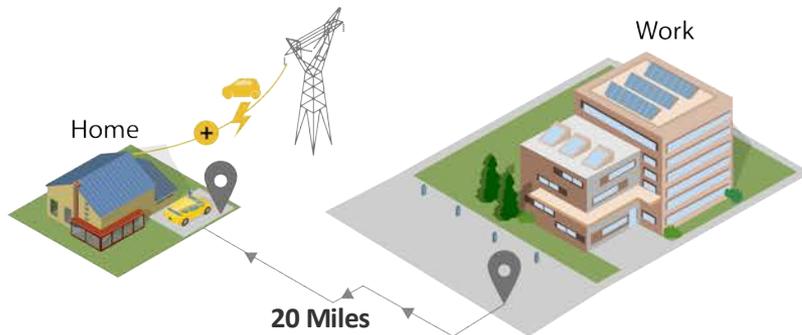
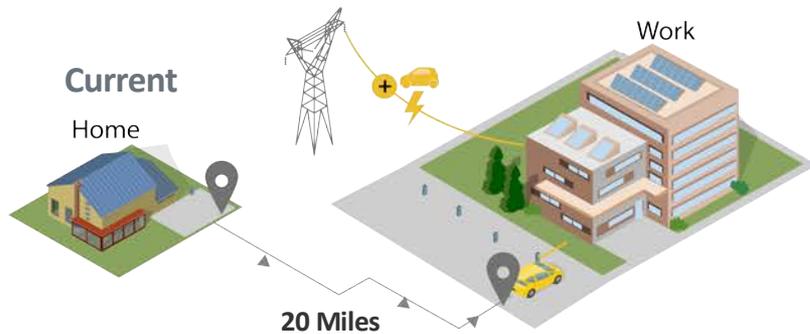


Large strain on urban and rural grids, but lower  
building load!

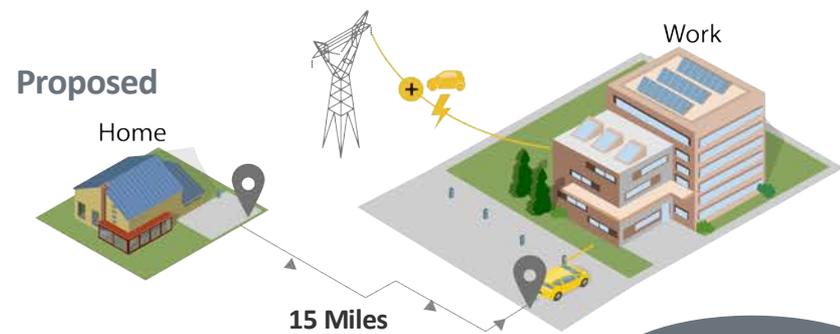
**Problem:** Accounting for autonomous EV adoption would look largely similar to current EV adoption

**Example:** Autonomous EV adoption increases:

# More, in our agent-based approach, our female would react differently



Some strain on urban and suburban grids;  
no change to rural grids or building loads.



Different strain on urban and rural grids, and on  
building load!

# Planned Approach

## *Proven Methods Accounting for Complexity of humans, robots and cities*

### M1: Inventory

- Toolsets and methods
- Delphi technique

### M2: Stakeholder Proving Ground

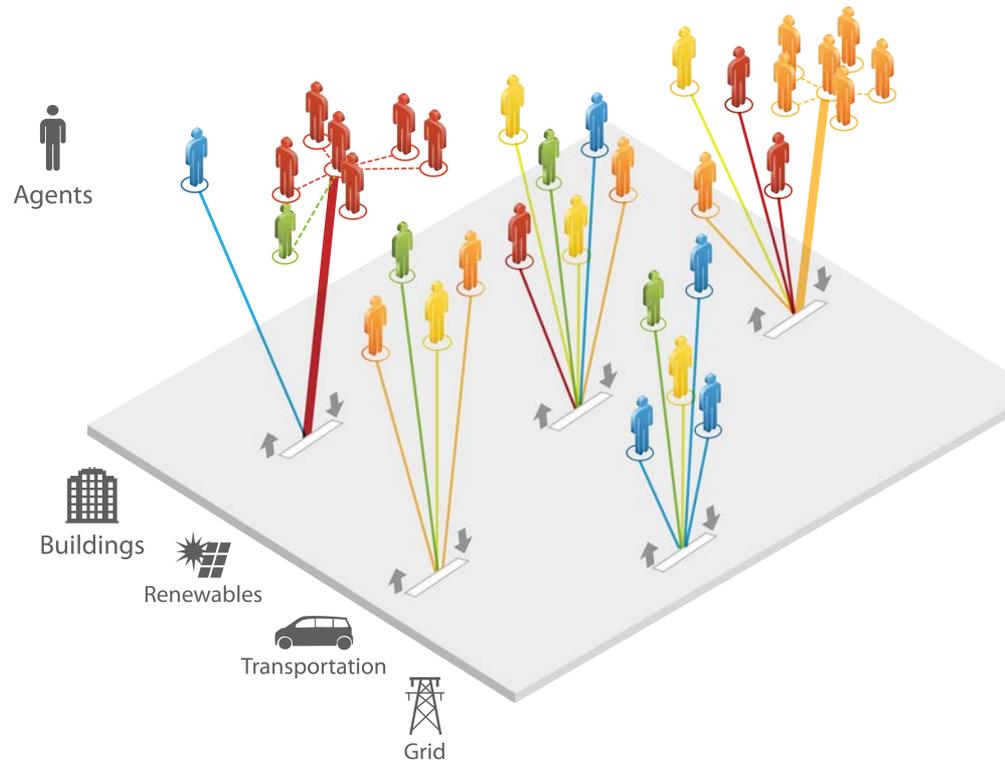
- Control scenario workshop
- Rank identified use cases

### M3: Implementation

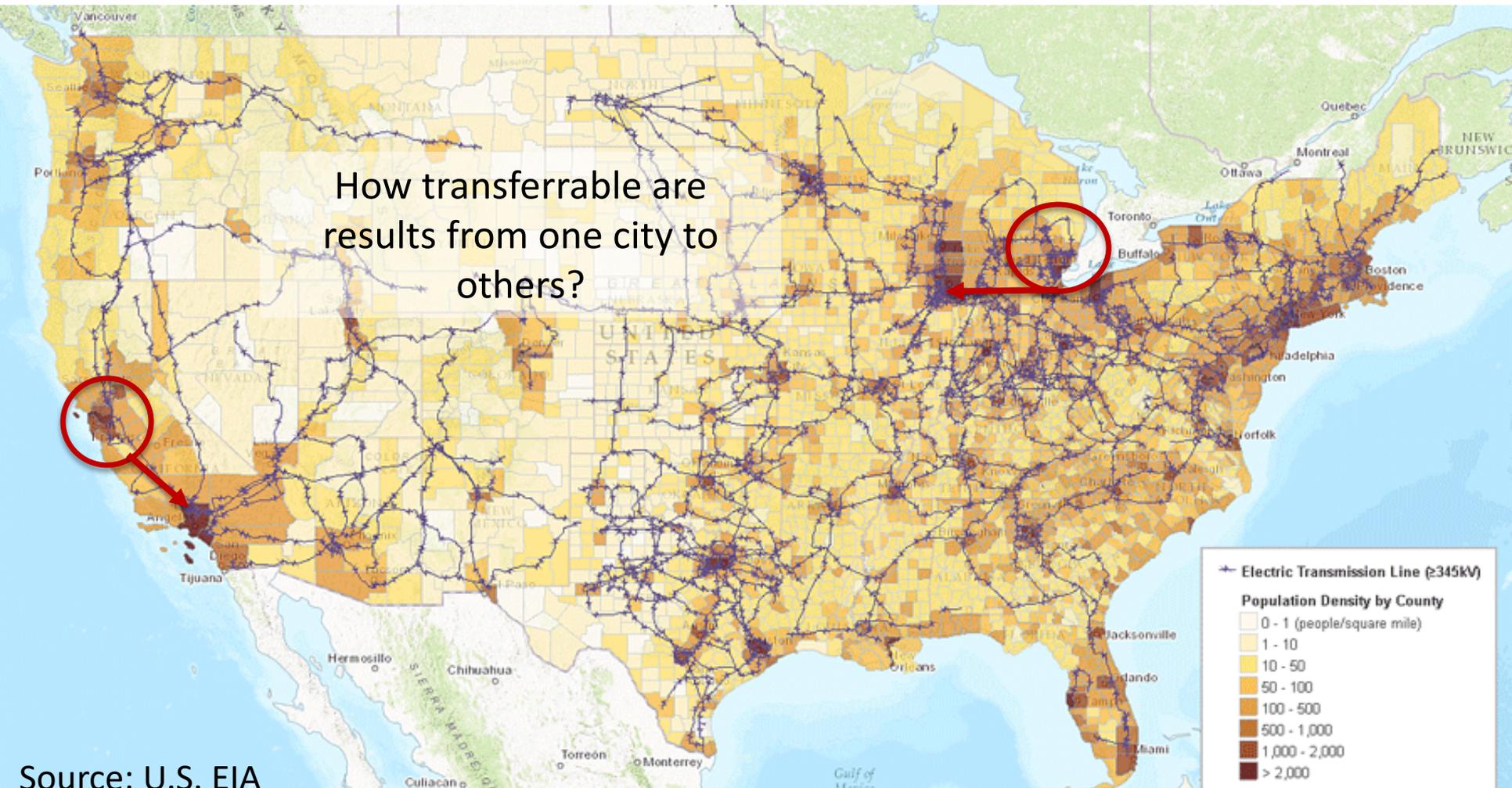
- Scope to incorporate AGM
- Design and implement

### M4: Pilot Analysis

- Execute
- SAS workshop (results)
- Report (high impact journals)



### III. Typology to identify clusters of features in emergent mobility behavior & energy use across settlement types & socioeconomic status groups





# Sustainable

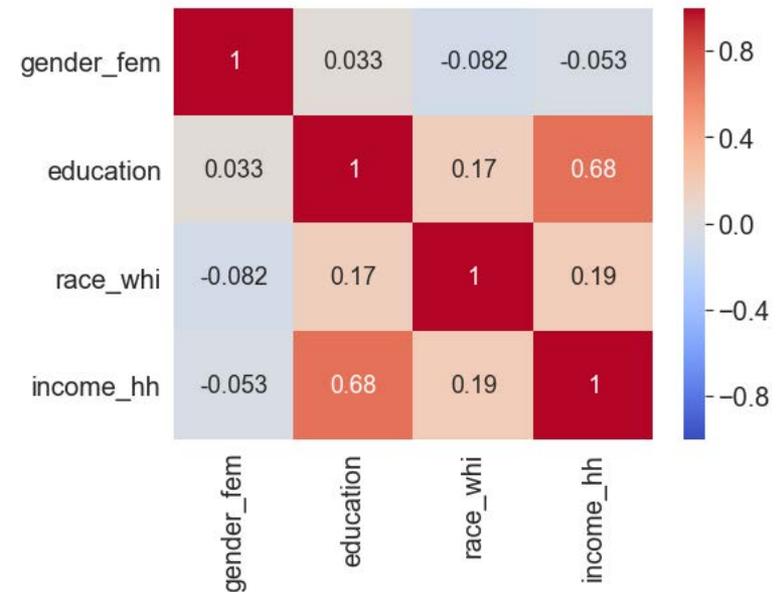
## TRANSPORTATION

U.S. DEPARTMENT OF  
**ENERGY** | Energy Efficiency &  
Renewable Energy

- 1. Create a multi-dimensional typology of adoption and outcomes**
  - E.g. on mobility & energy use
- 2. Engage social science with engineering science**

# II. Methods and data

1. Social, economic, technological environmental, governance (SETEG) indicators (for 10 US states)
2. Standard statistics, factor analysis and cluster analysis
3. Analysis of patterns and variations across population & settlement clusters

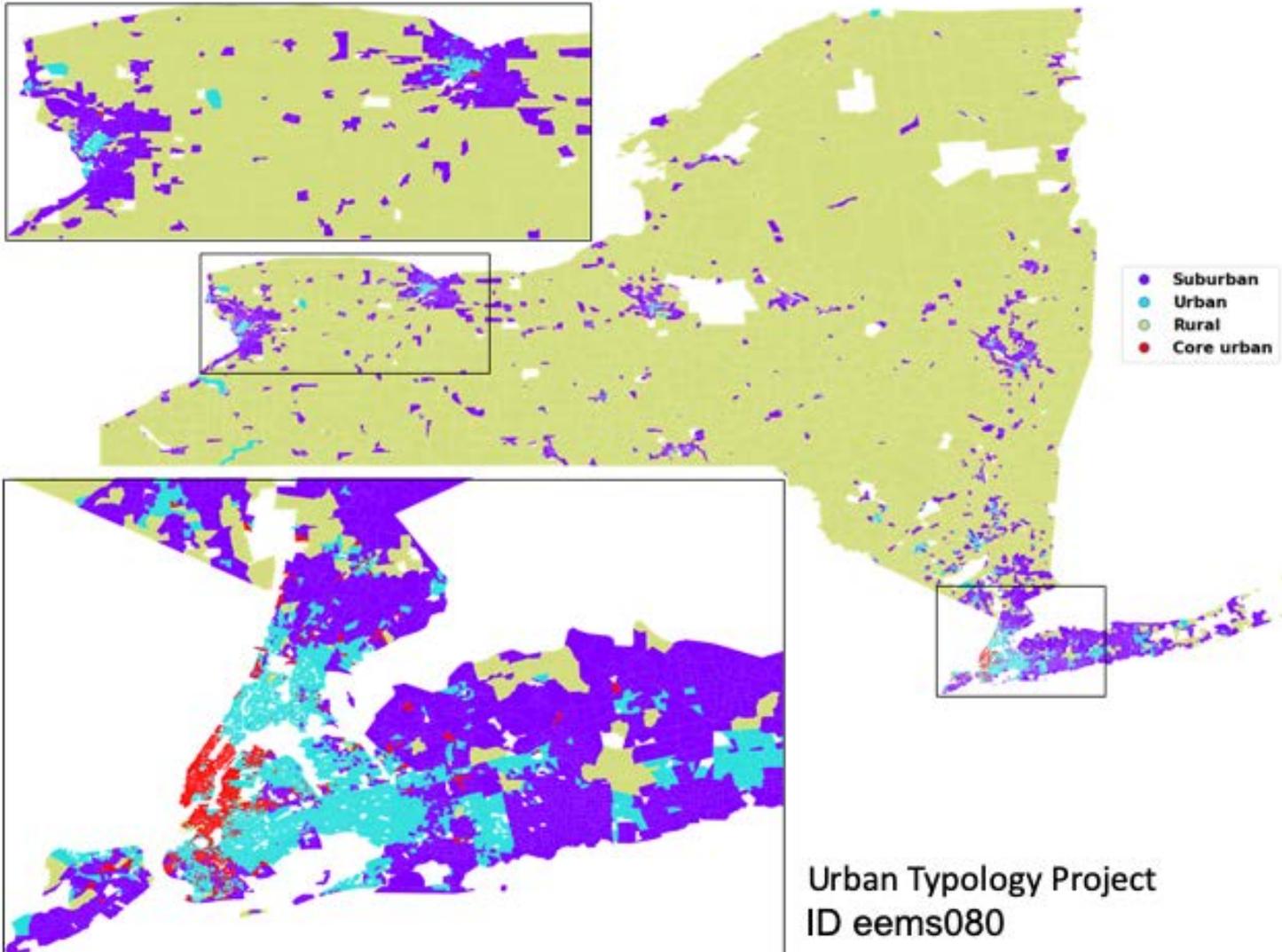


# Dataset Typology

INDICATOR	SETEG	DEP/IND	RESOLUTION	AVAILABILITY
MEP metric	Mobility	DEP	Grid cells	Six cities
Gallons gasoline sold	Energy	DEP	County	Two cities
EV adoption %	Adoption	DEP	Zip code	US
Transit trips/capita	Mobility	DEP	Transit agency	US
Fleet averaged mpg	Energy	DEP	Zip code	US
Gender (female)	S	INDEP	Block group	US
Education (level)	S	INDEP	Block group	US
Race (white)	S	INDEP	Block group	US
Household income	E	INDEP	Block group	US
Employment Access	E	INDEP	Block group	US
House Tenure (own)	E	INDEP	Block group	US
Population Density	T	INDEP	Block group	US
#Vehicles/household	T	INDEP	Block group	US
GHG/MWh electricity	T	INDEP	Block group	US
HTA Index (CNT)	T	INDEP	Block group	US
PM 2.5	<i>E</i>	INDEP	Block group	US
Cancer Hazard	<i>E</i>	INDEP	Block group	US
Respiratory Hazard	<i>E</i>	INDEP	Block group	US
Intersection Density	G	INDEP	Block group	US
Non SOV commuters	G	INDEP	Block group	US

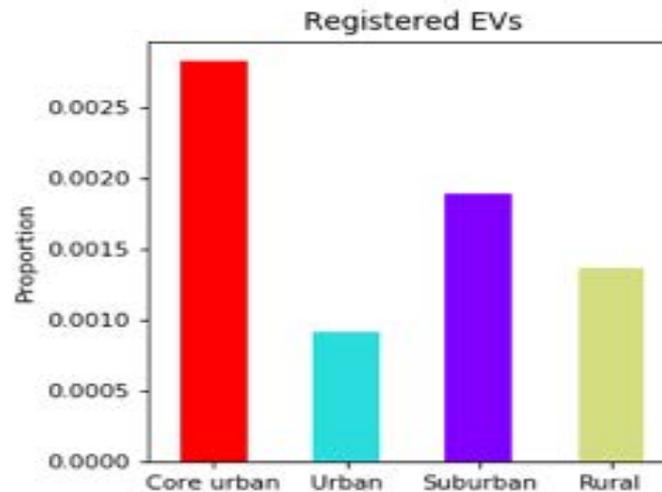
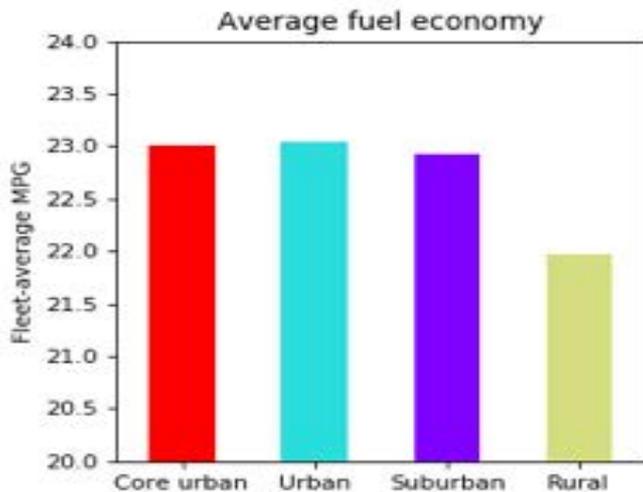
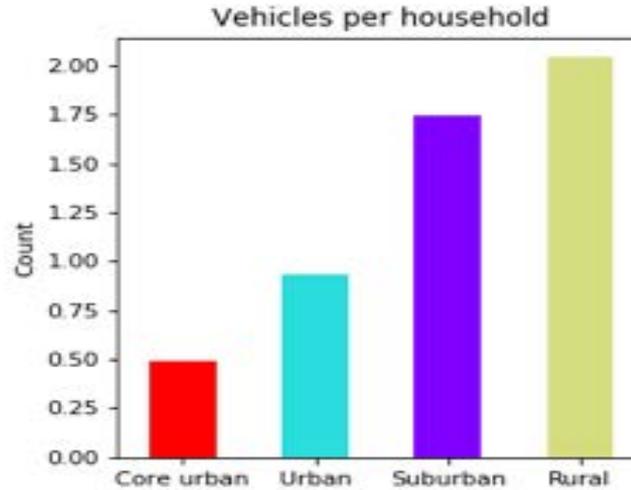
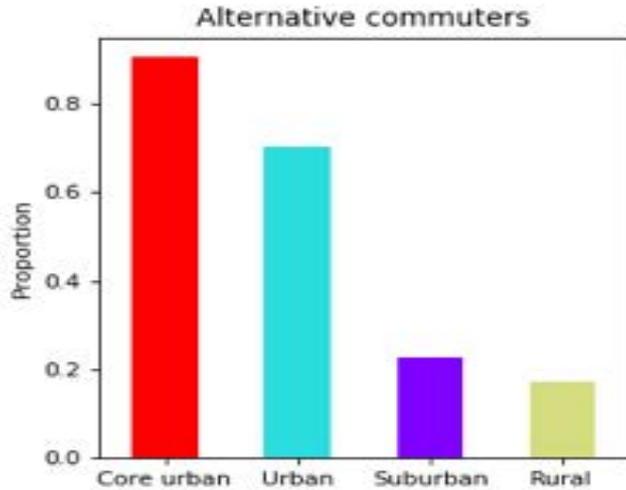
# Preliminary Results

## Areas in New York State Using a Typology Approach

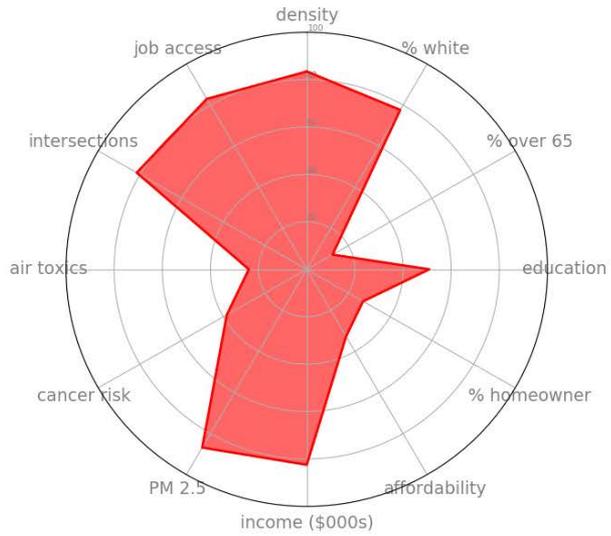


# Preliminary Results

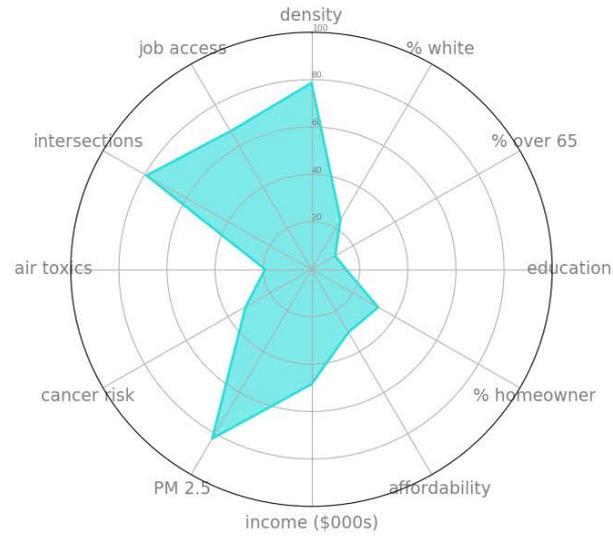
## Mobility and Energy Outcomes across Areas



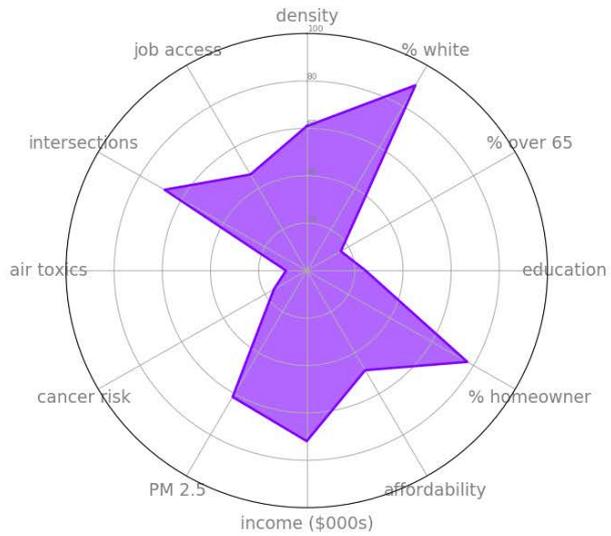
**Core urban**



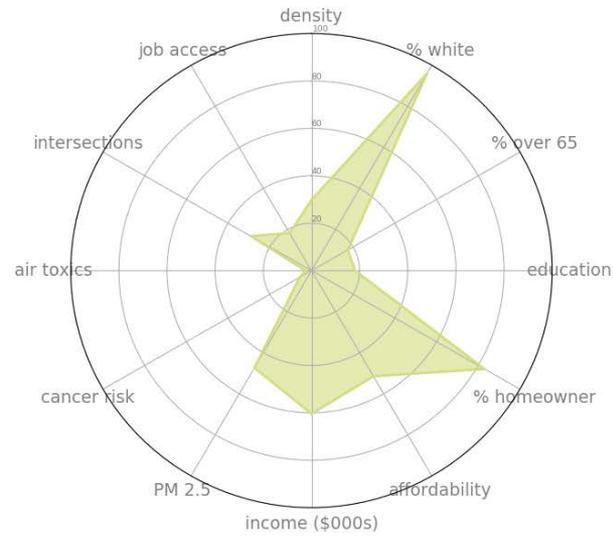
**Urban**



**Suburban**

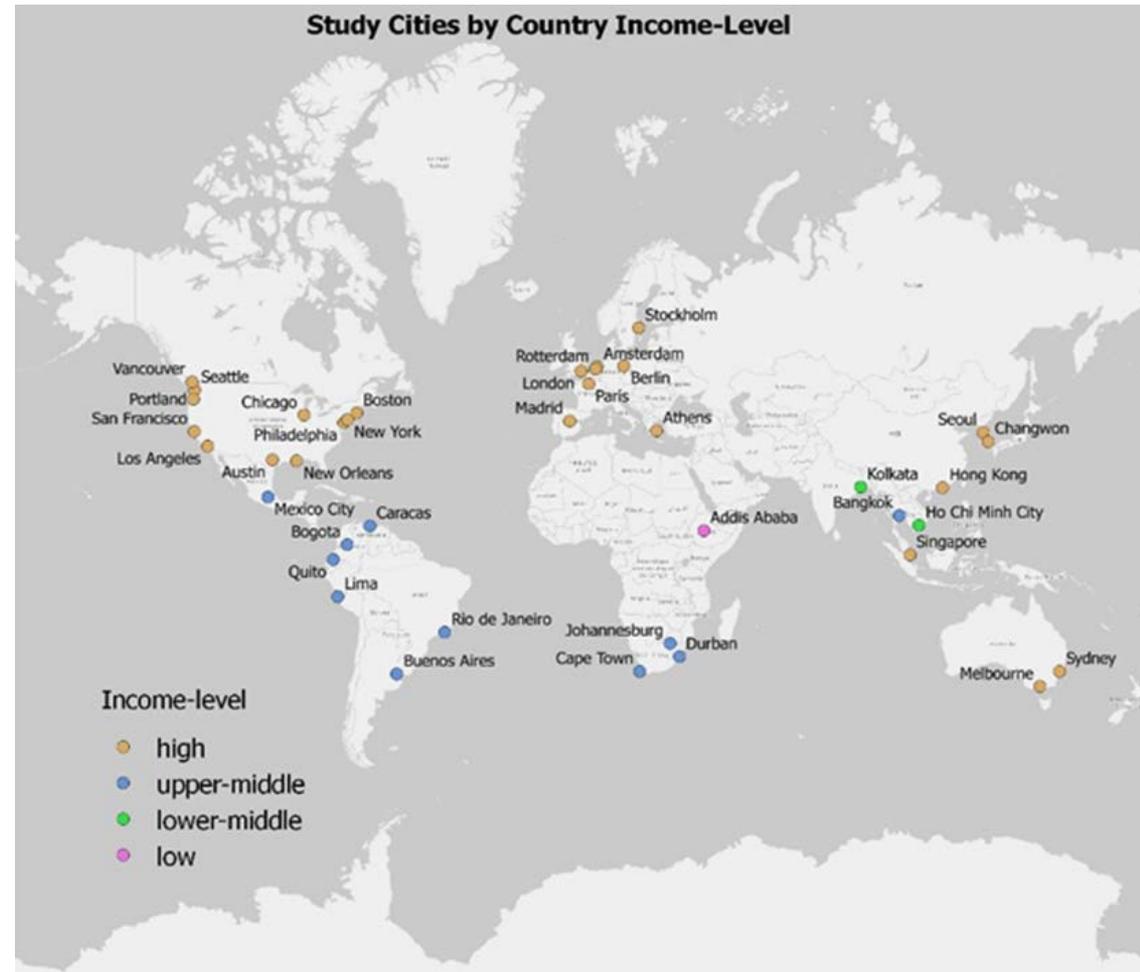


**Rural**



# Other relevant outcomes

- White paper on urbanization, electrification and urban areas (draft)
- Meta-analysis of Risk Inequality and FEW Nexus: A Study of 43 City Adaptation Plans (published in Frontiers)





THE UNIVERSITY OF  
CHICAGO

Thanks!

Paty Romero-Lankao