

Bridging population, activity spaces, and social interactivity with the Likeness software stack

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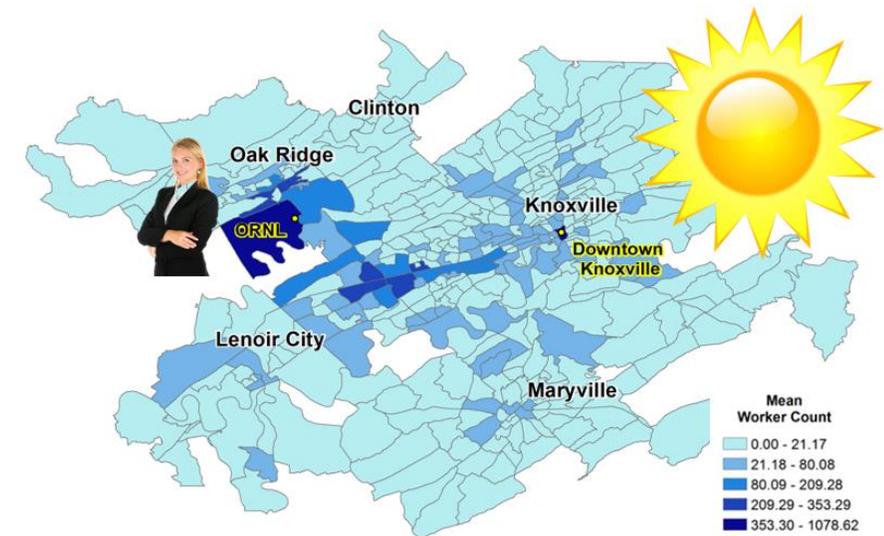
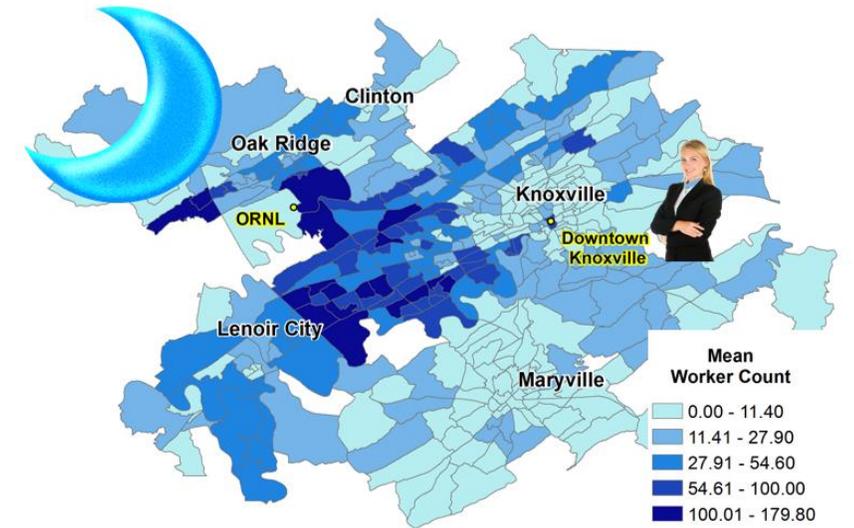
The UrbanPop Spatial Microsimulation Framework

- **Capabilities**

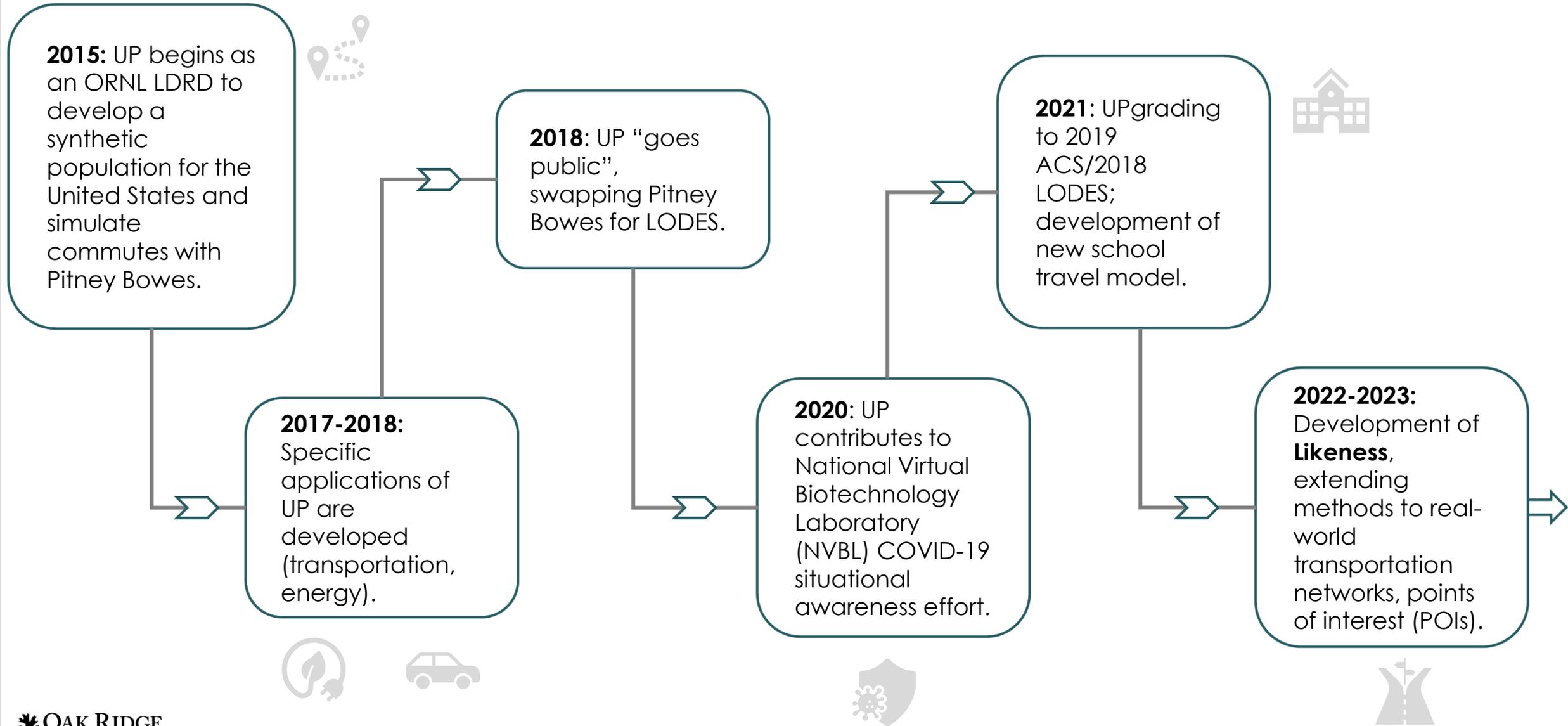
- Coupled individual-neighborhood profiles (*how people live*)
- Daytime/Nighttime dynamics (*how people move*)
- Agent Generation -> patterns of Life/Social contacts (*how people [inter]act*)

- Based on the **American Community Survey (ACS)** and **Public-Use Microdata Sample (PUMS)**

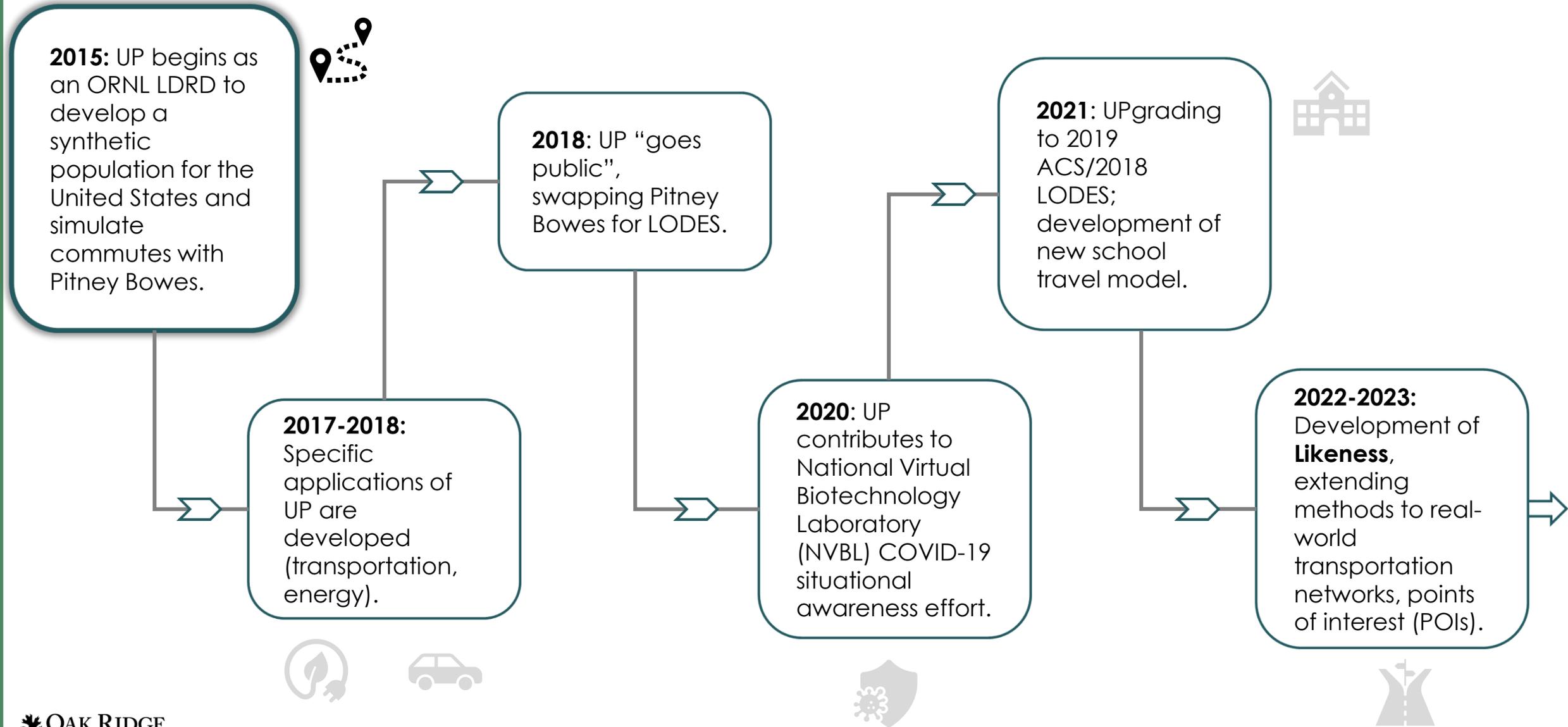
- **Available for any location in the United States 2016 – 2019**



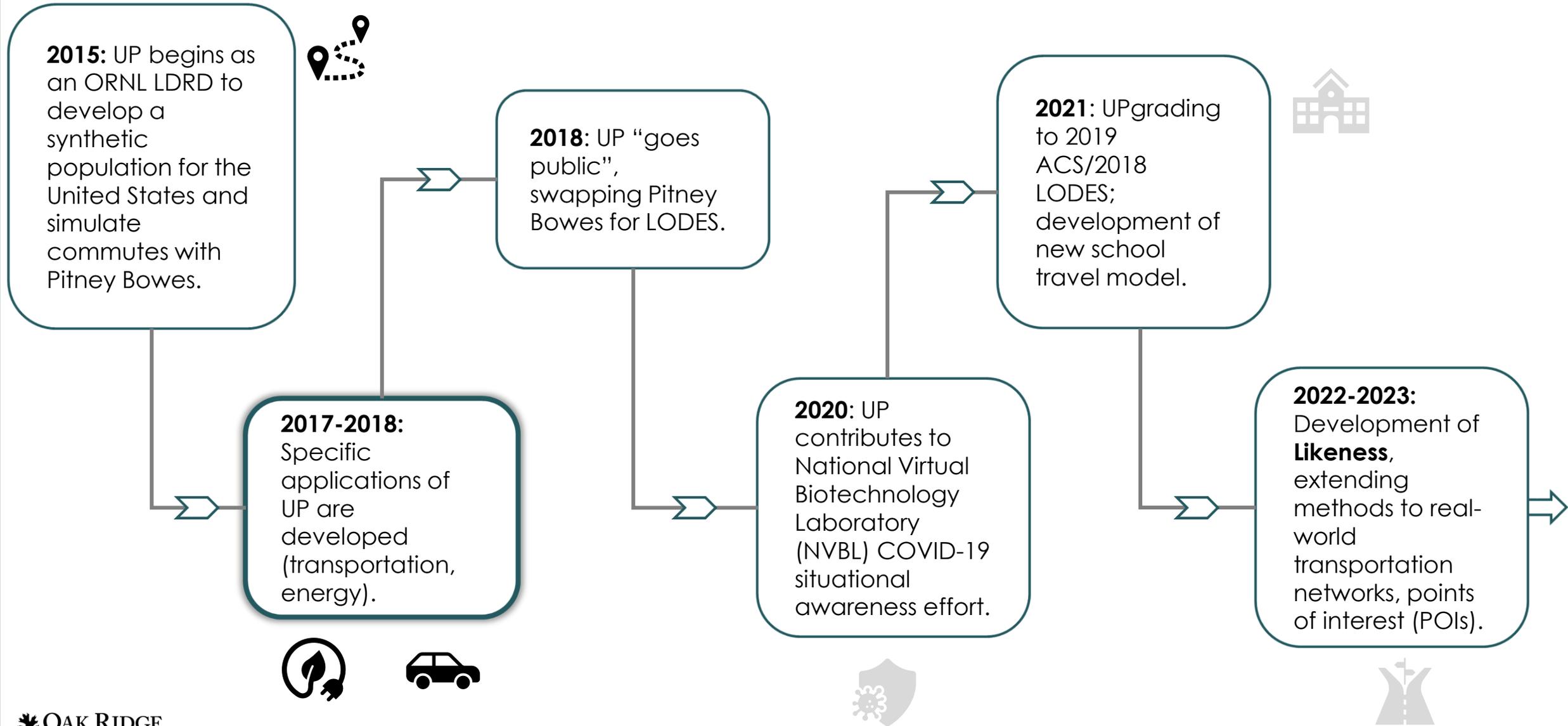
UrbanPop History



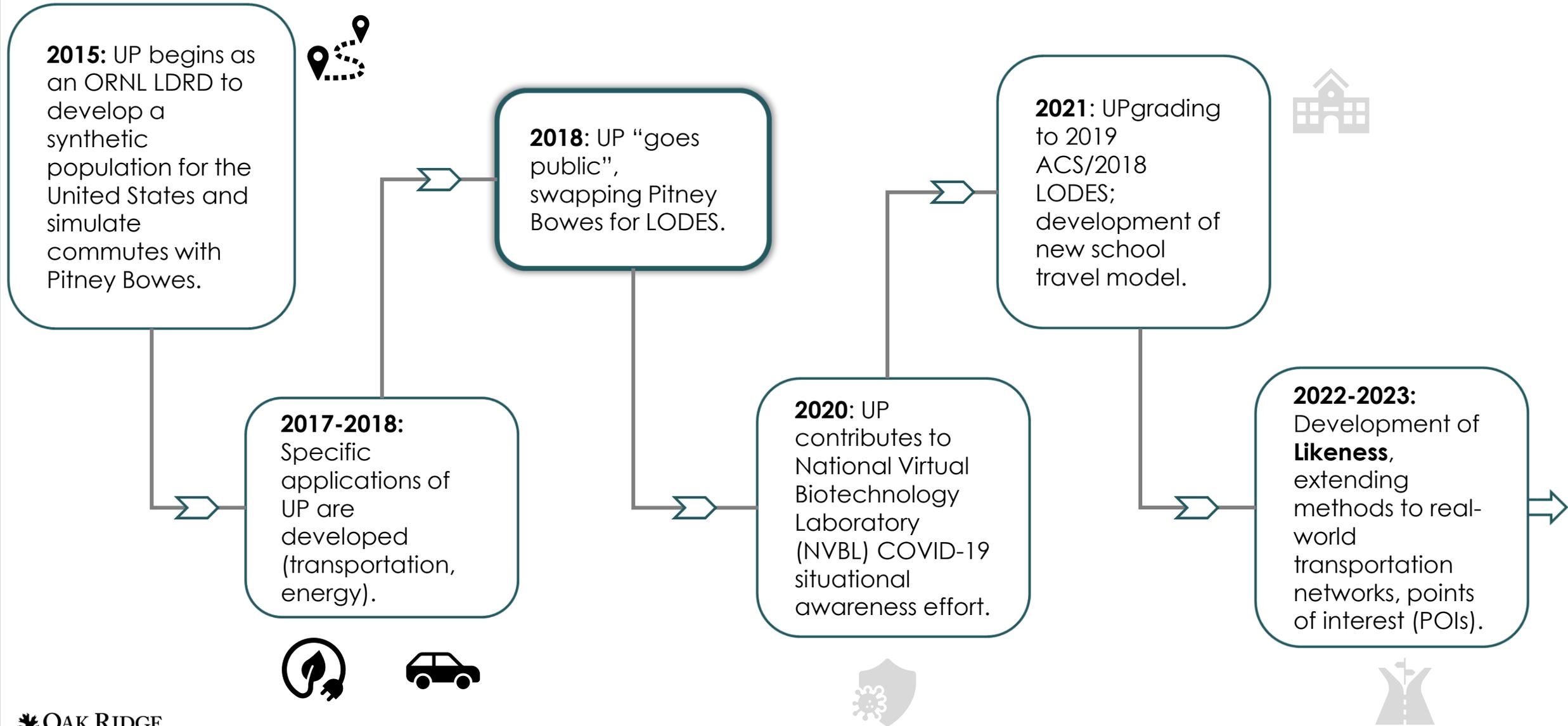
UrbanPop History



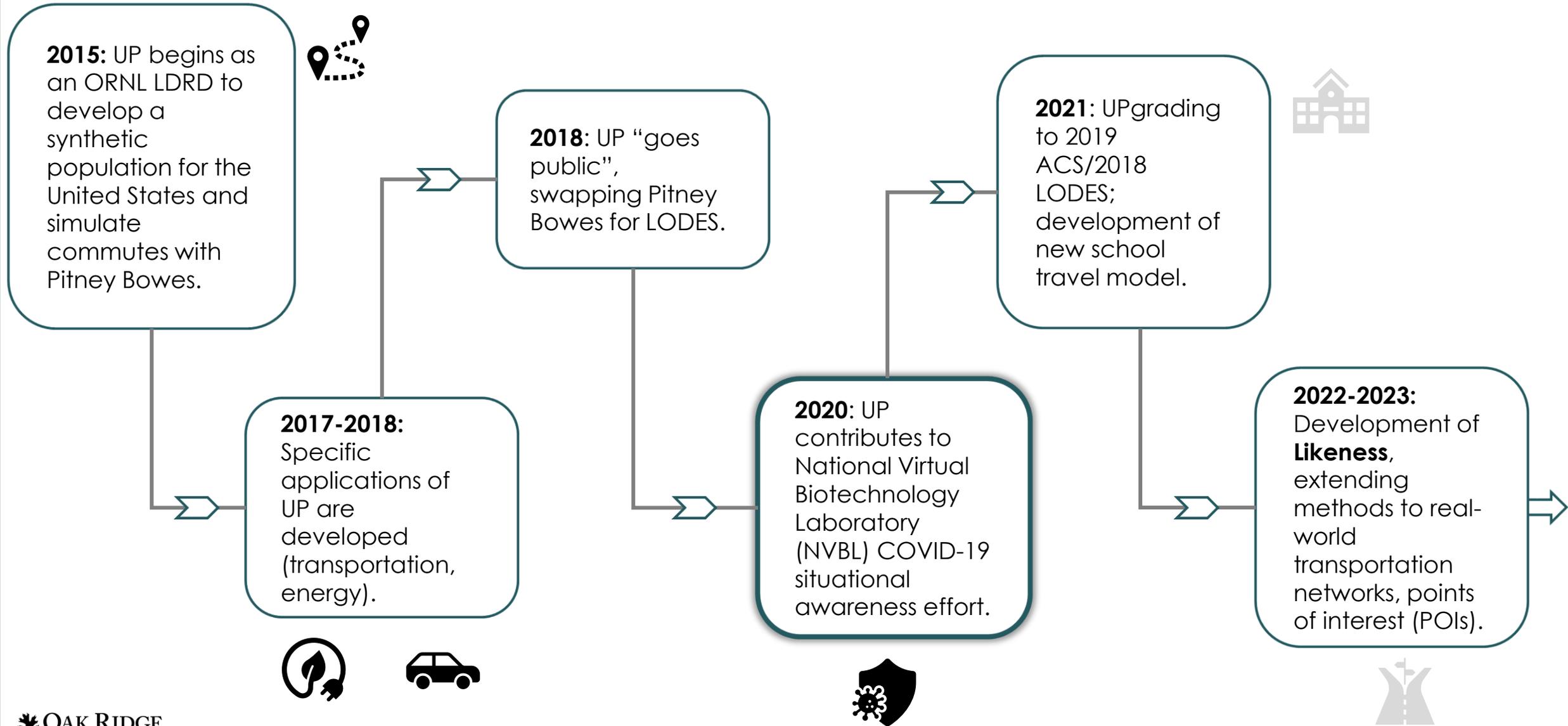
UrbanPop History



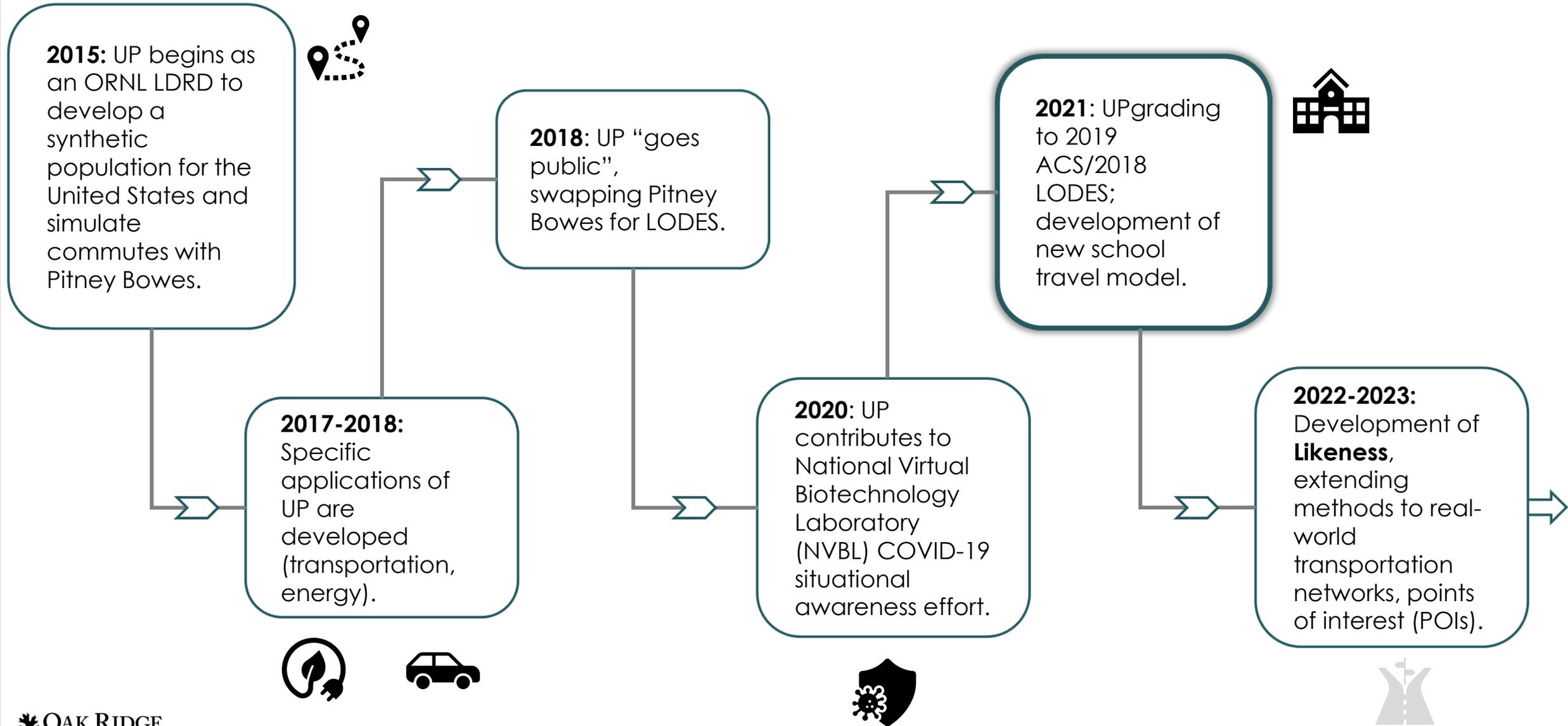
UrbanPop History



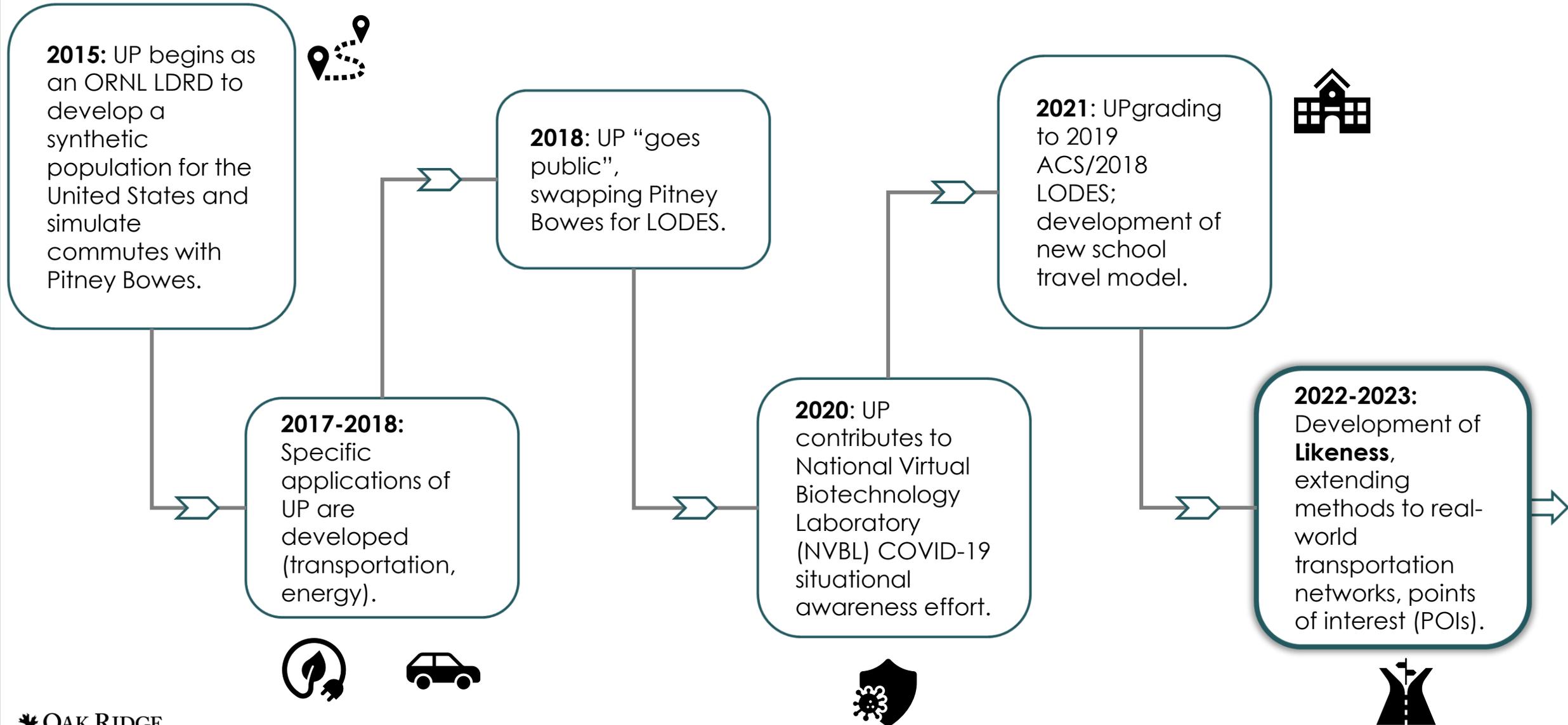
UrbanPop History



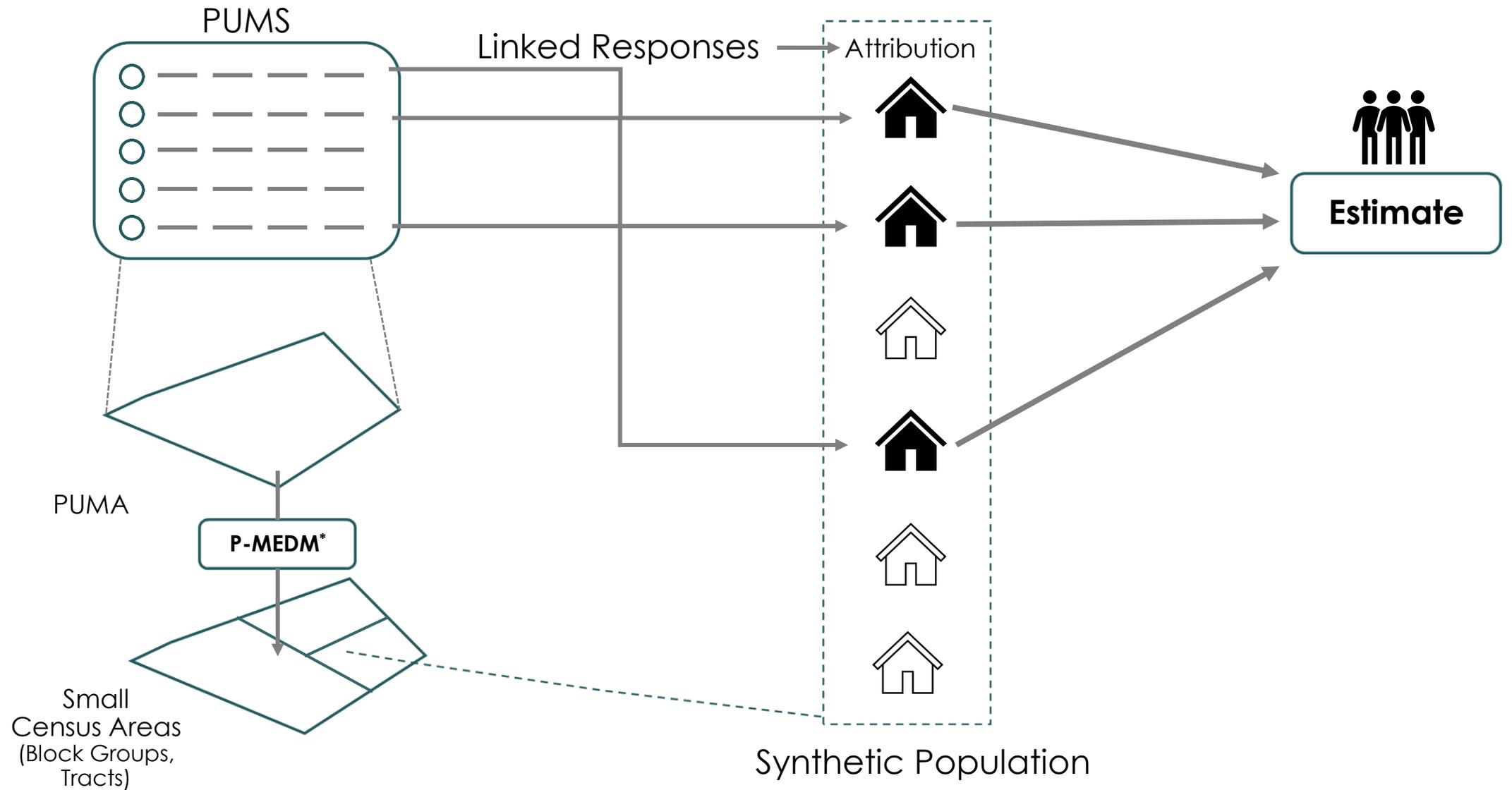
UrbanPop History



UrbanPop History



UrbanPop Core Workflow



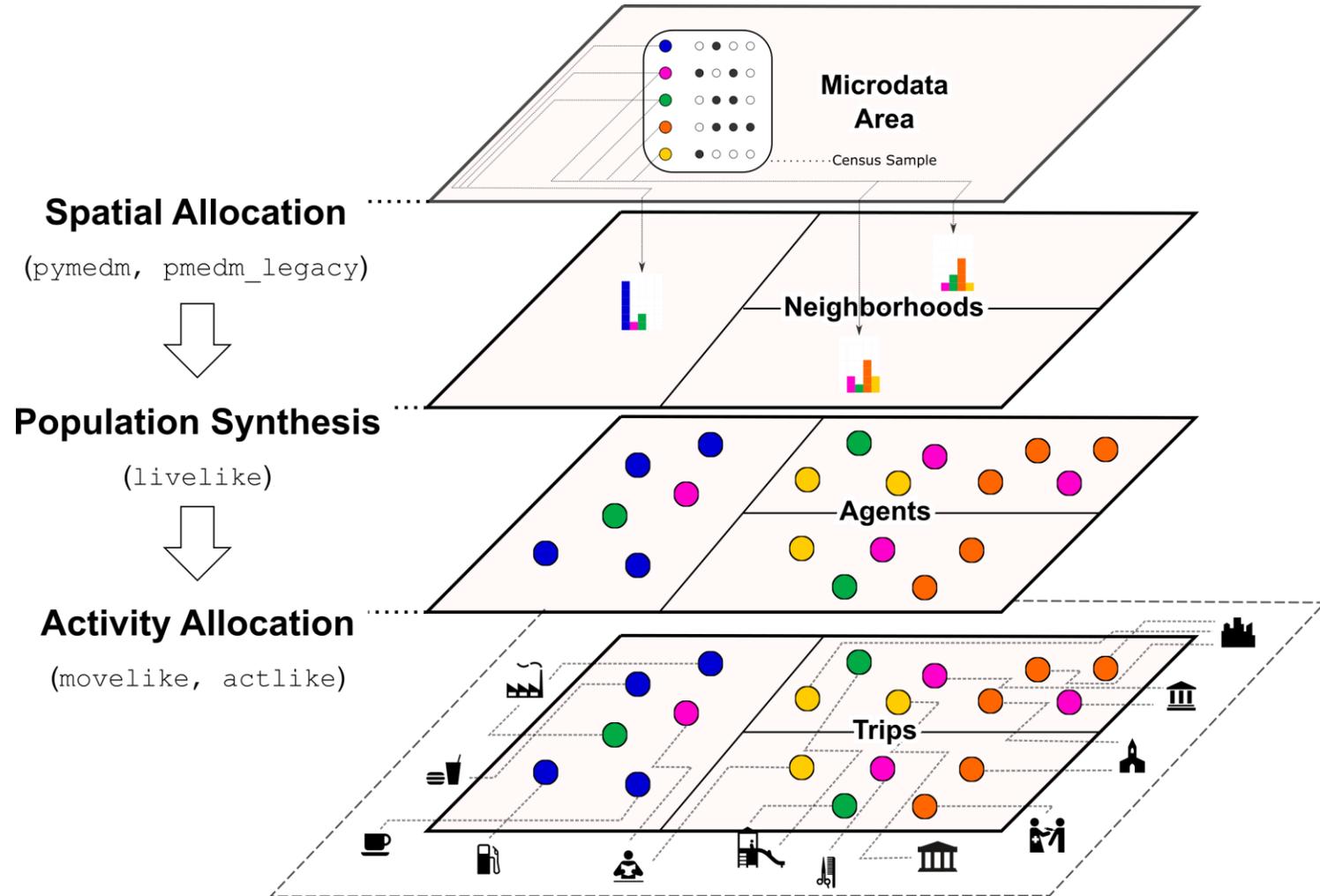
The Likeness Toolkit

- **UrbanPop's core capabilities in a Python software stack**

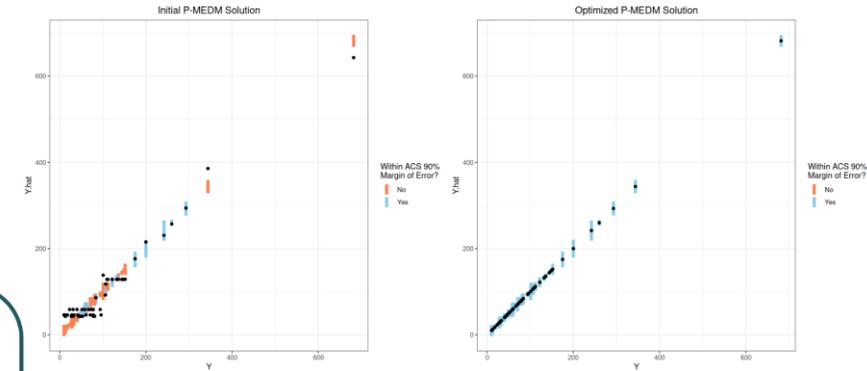
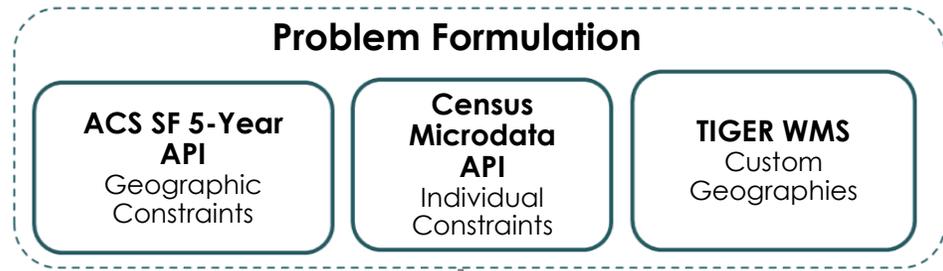
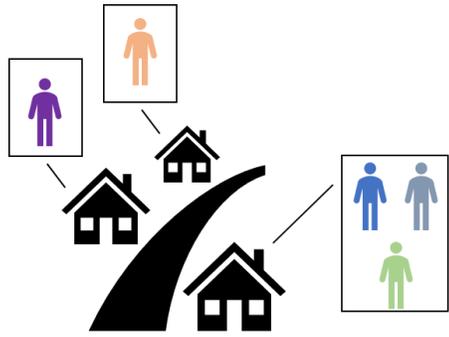
- Synthetic Population Generation
- Small-Area Estimates for neighborhood characterization

- **Enhancements**

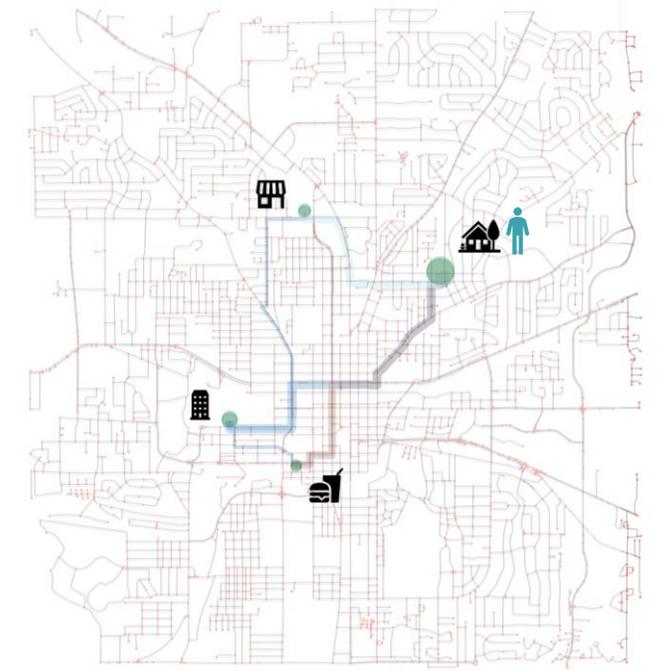
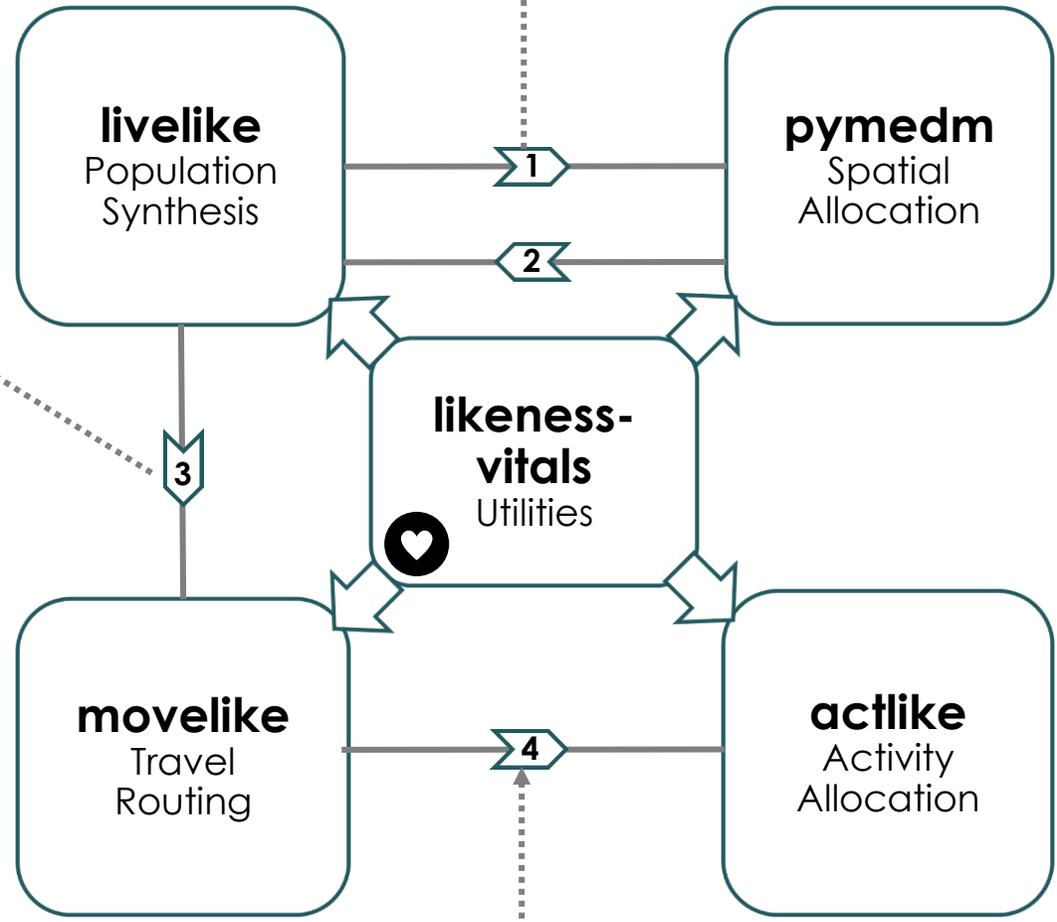
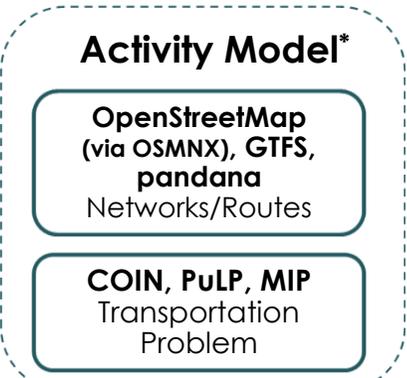
- Travel along real-world transportation networks from OSM
- POI-based destinations for a variety of activity types (work/school, social, recreation, errands, health...)



Likeness ecosystem



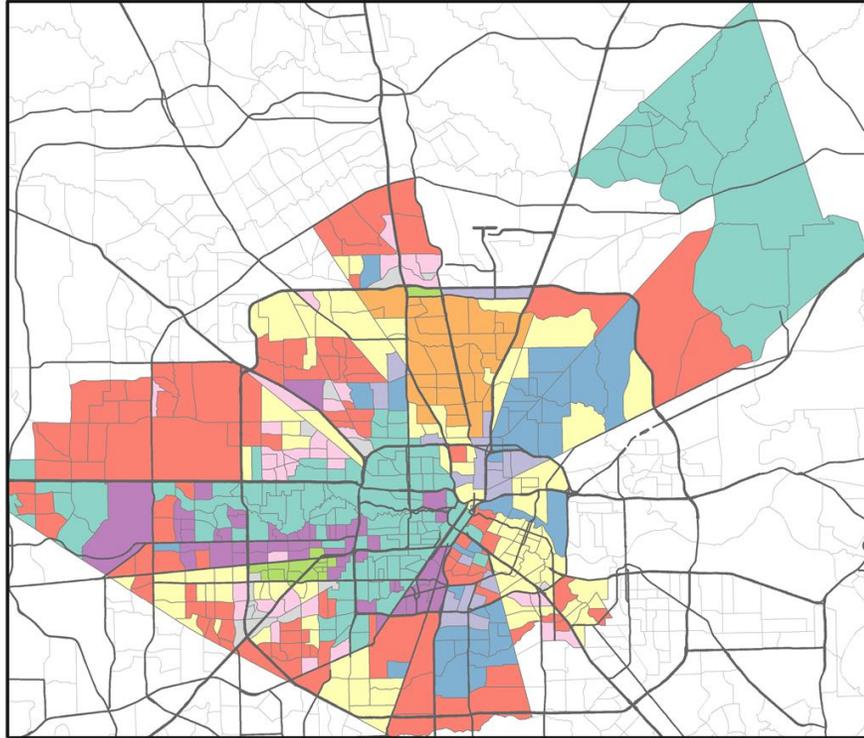
Tuccillo and Spielman 2022 [5]



***Activity Model attribution**
Boeing (2017) [6], Foti et al. (2012) [7], Lougee-Heimer (2003) [8], Forrest et al. (2023) [9], Mitchell et al. (2001) [10], Toffolo and Santos (2019) [11]

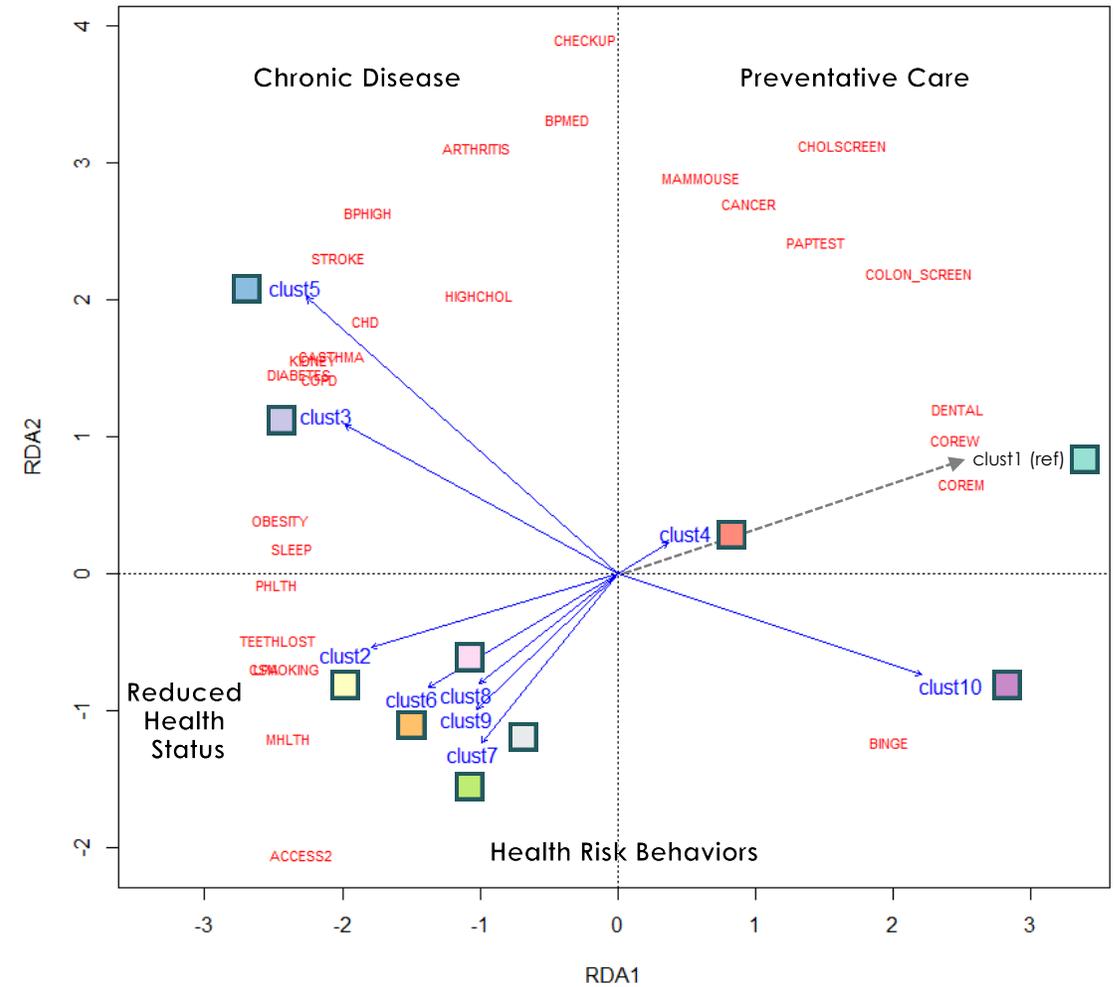
Case Study 1: Neighborhood riskscape, health, and power outages

Harris County, TX 2019



Social Vulnerability Profiles by Individual Vulnerability Theme

	7	6	9	3	8	2	5	4	10	1	Avg. Prop.
Few SV Indicators	13%	15%	17%	25%	26%	29%	38%	55%	60%	87%	49%
Independent Living Sensitivity	2%	7%	3%	14%	5%	10%	16%	9%	4%	4%	7%
Poverty/Density/Limited Mobility	49%	51%	43%	26%	34%	38%	18%	17%	12%	3%	23%
Communication Sensitivity/Unemployment	28%	22%	27%	16%	22%	16%	14%	12%	14%	4%	14%
Single Caregiver Relationships	8%	5%	11%	19%	12%	6%	13%	8%	9%	2%	7%



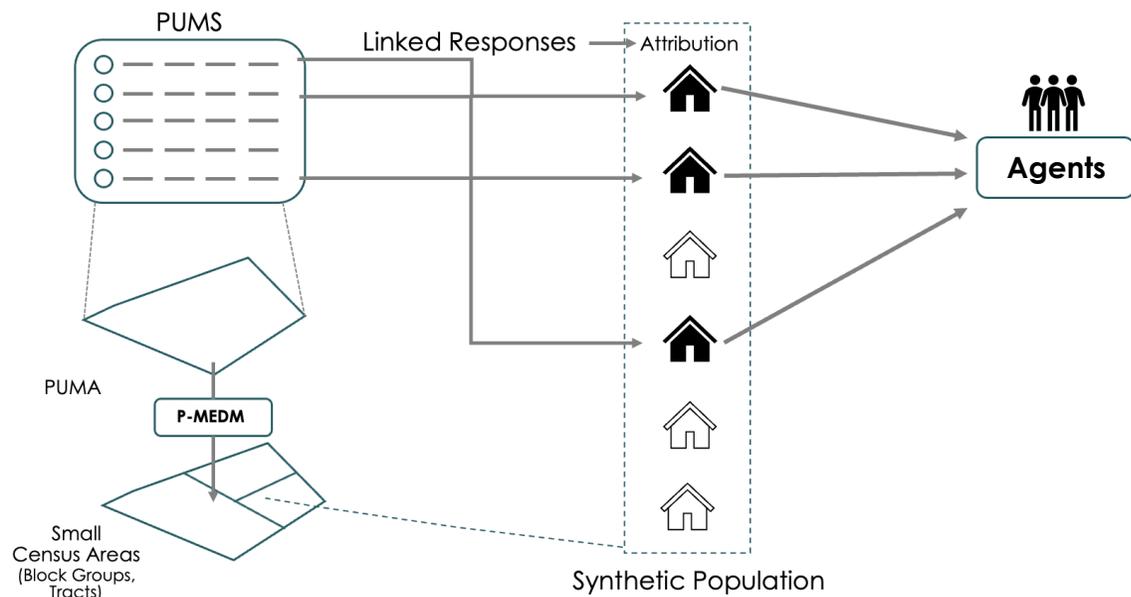
Redundancy Analysis (RDA) on CDC PLACES 2019

$$R^2 = 0.65 \text{ (pr(F) < 0.001)}$$

Generating Household/Agent Locations

Spatial Allocation

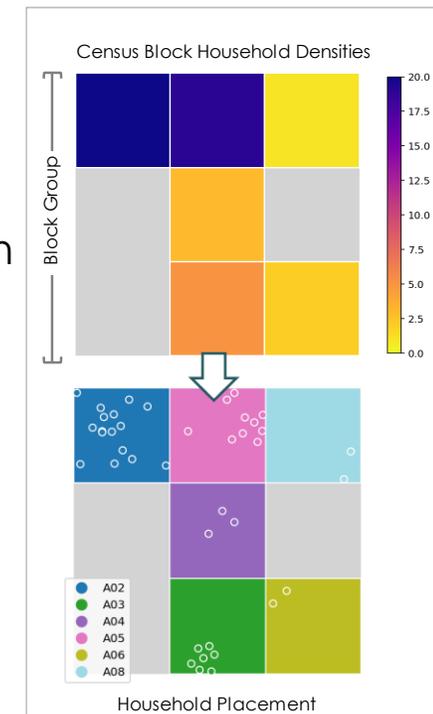
Core Workflow



Downscaling

Working Approach

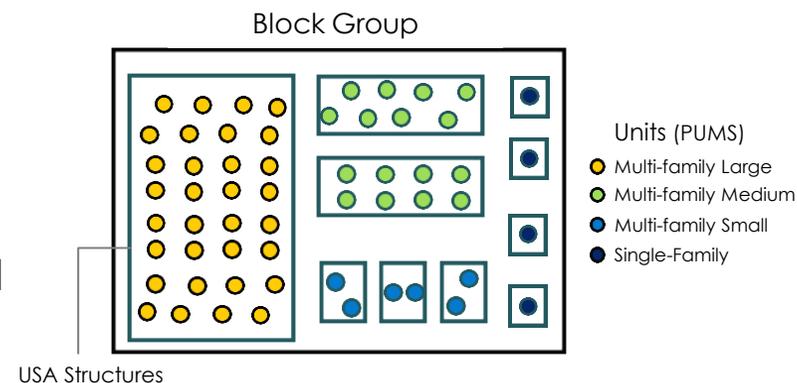
- Weighted random allocation of households to census blocks based on 2010 Decennial Census household density
- Variable-size binpacking problem
- Clustered placement of household points within blocks



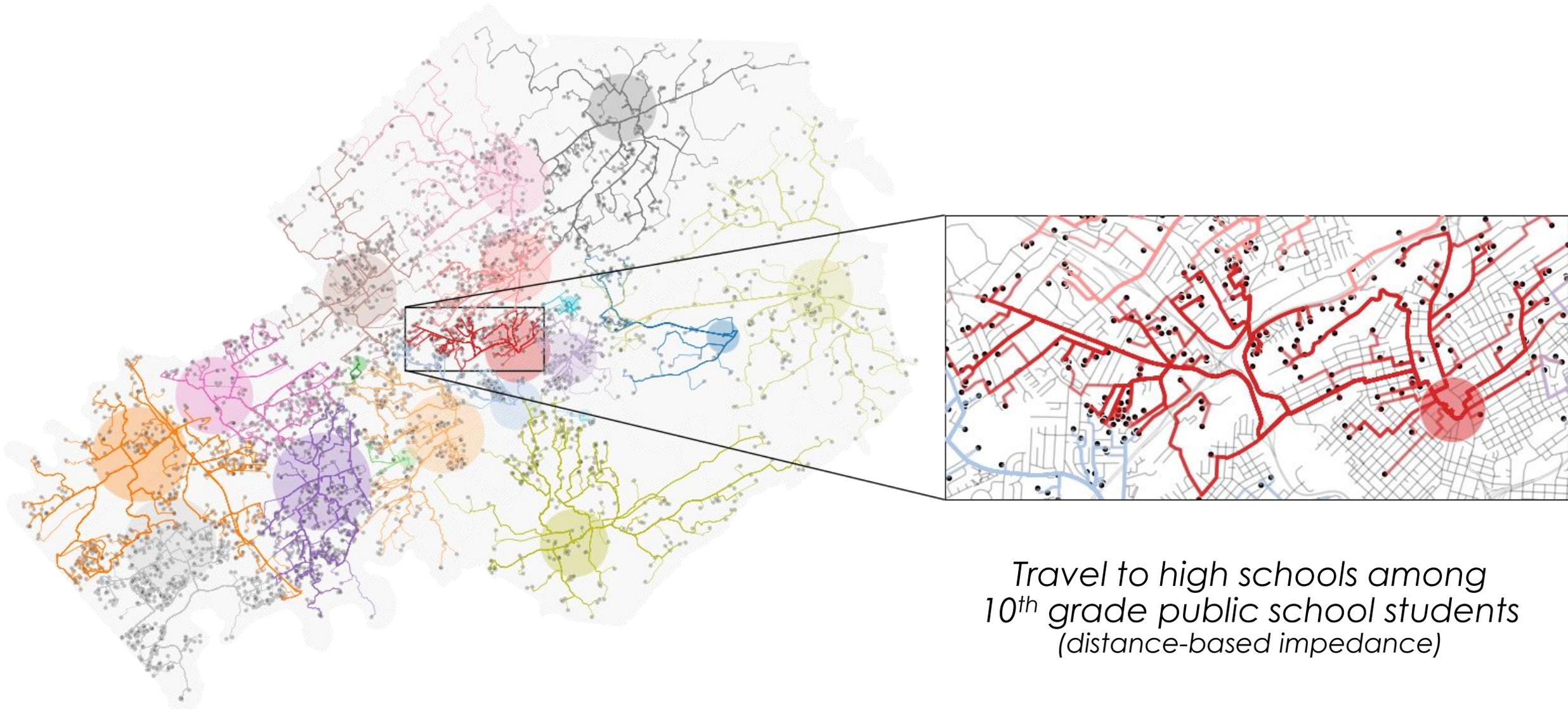
Downscaling

In Development

- Downscaling resolved to USA Structures
- Weighted random assignment of building IDs to synthetic households scored on similarity in unit counts between household (PUMS) and local USA Structures



Case Study 2: K-12 Public Schools in Knox County, TN



*Travel to high schools among
10th grade public school students
(distance-based impedance)*

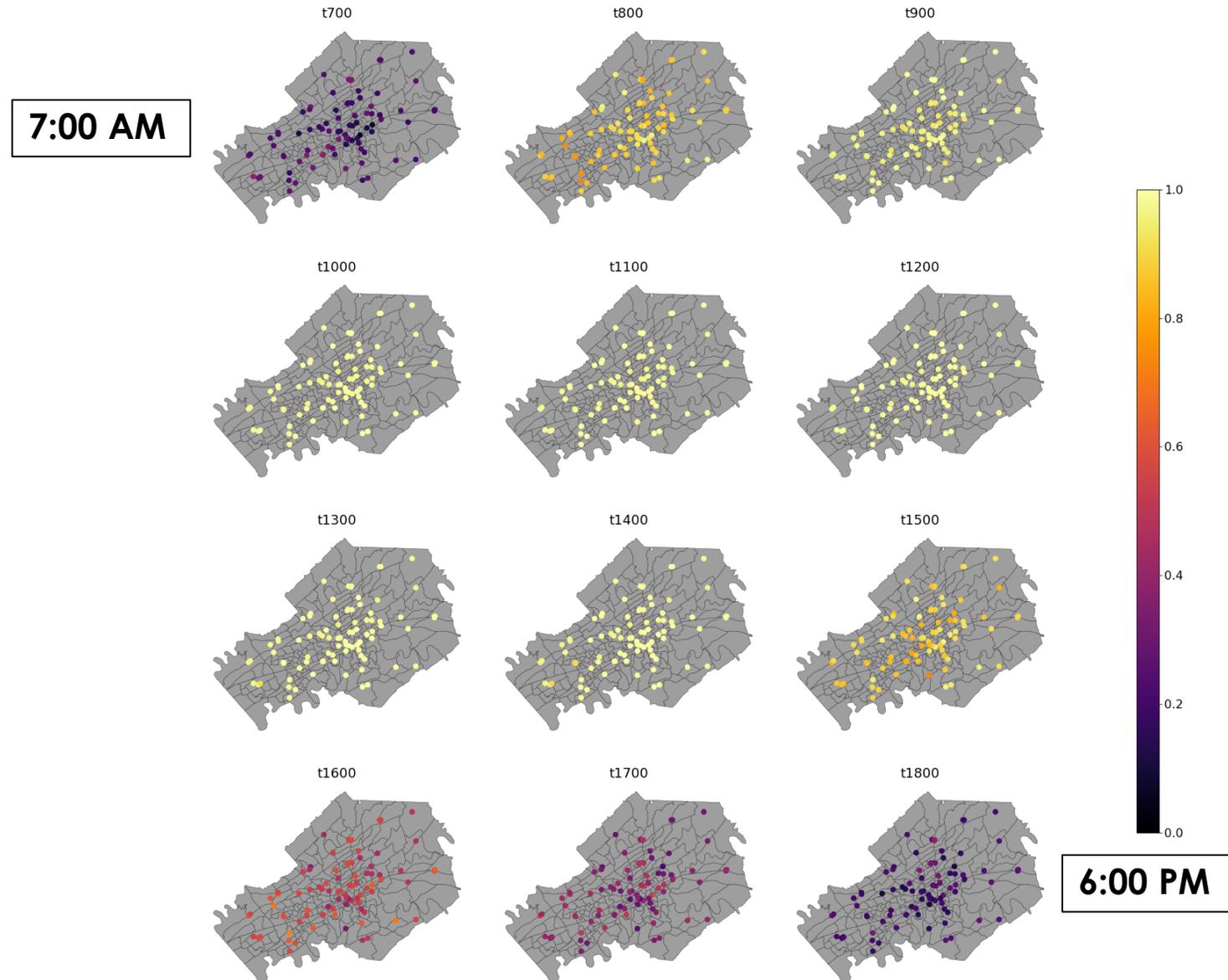
Case Study 2: K-12 Public Schools in Knox County, TN

- **Educator arrivals**

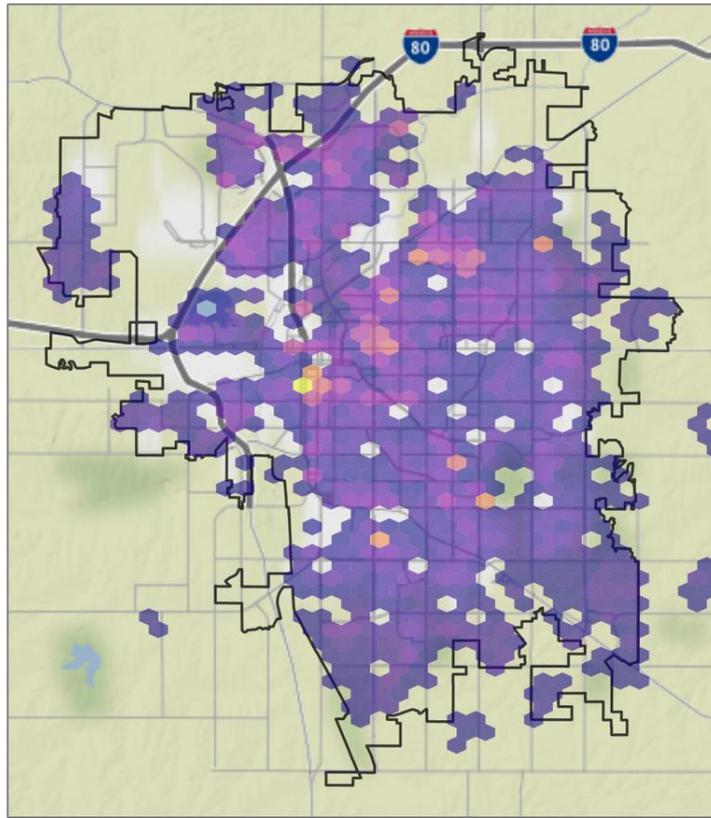
- PUMS time of arrival at work (JWAP)

- **Educator departures**

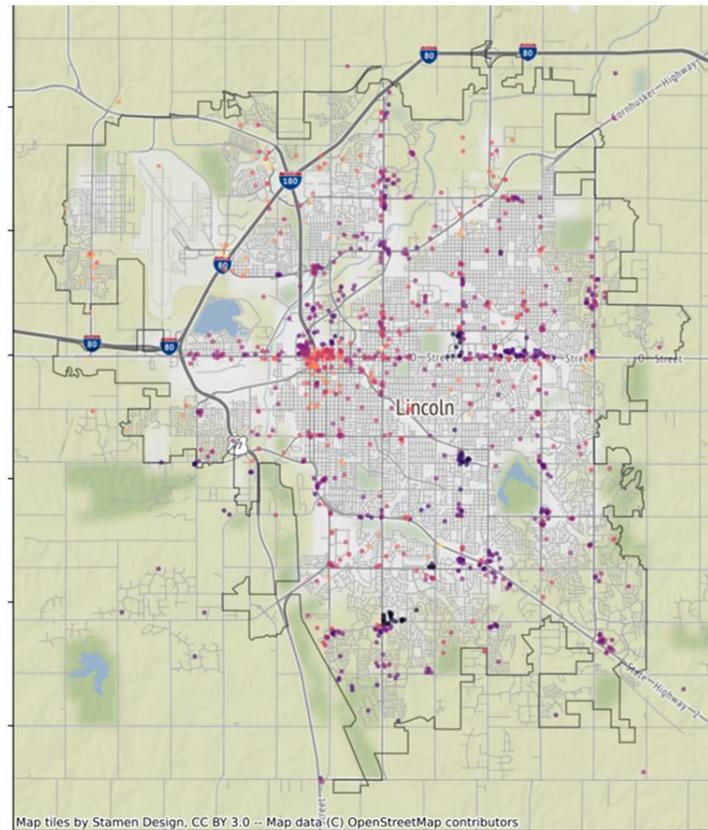
- $\text{PUMS JWAP} + (\text{WKHP} / 5)$
where WKHP represents the number of hours worked per week



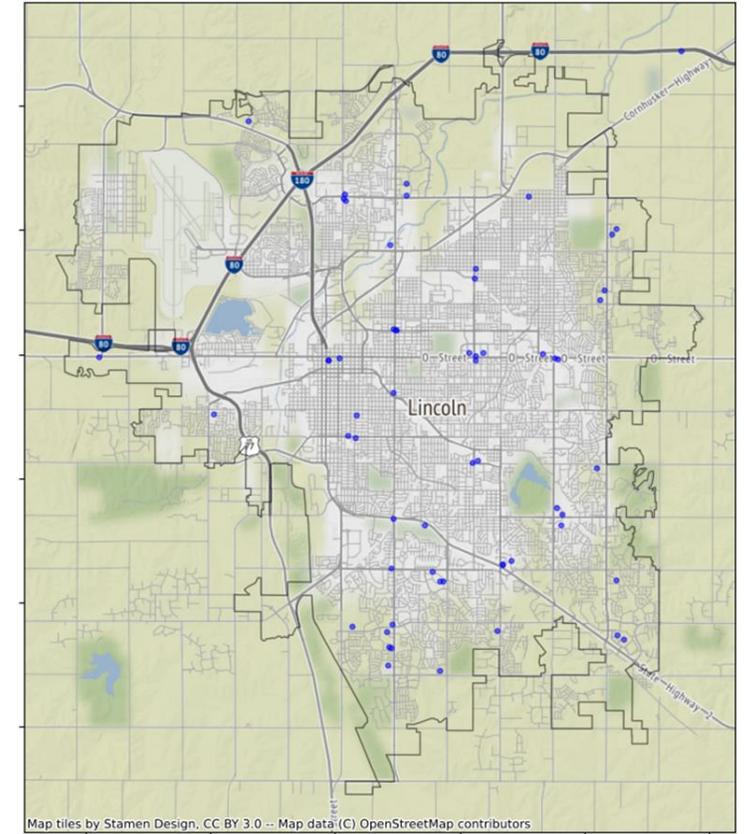
Case Study 3: Essential Activities



Synthetic Population
Food Service Workers



Job Locations
Workers per Establishment



COVID-19 Healthcare Services
Vaccination and Testing Centers

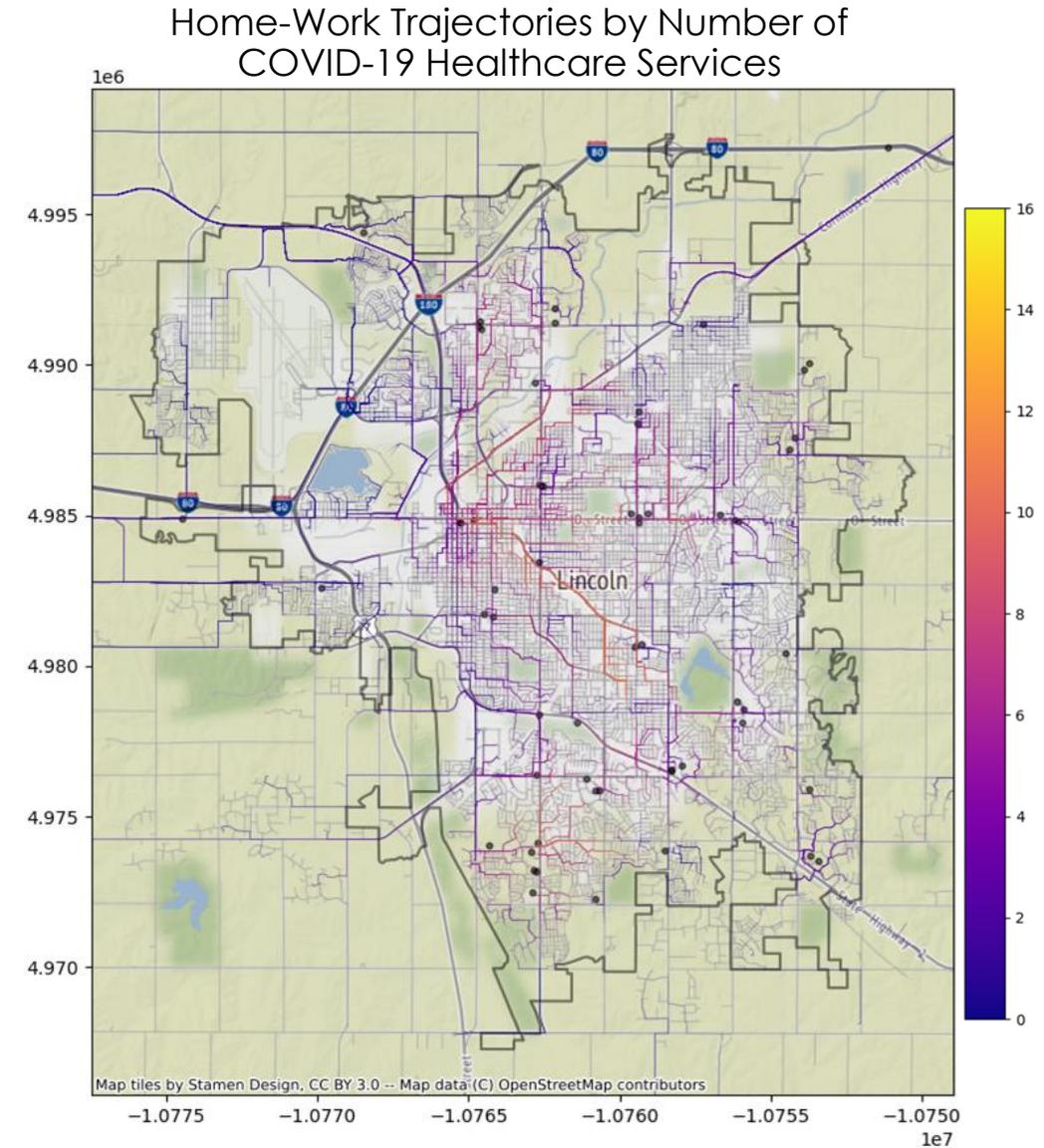
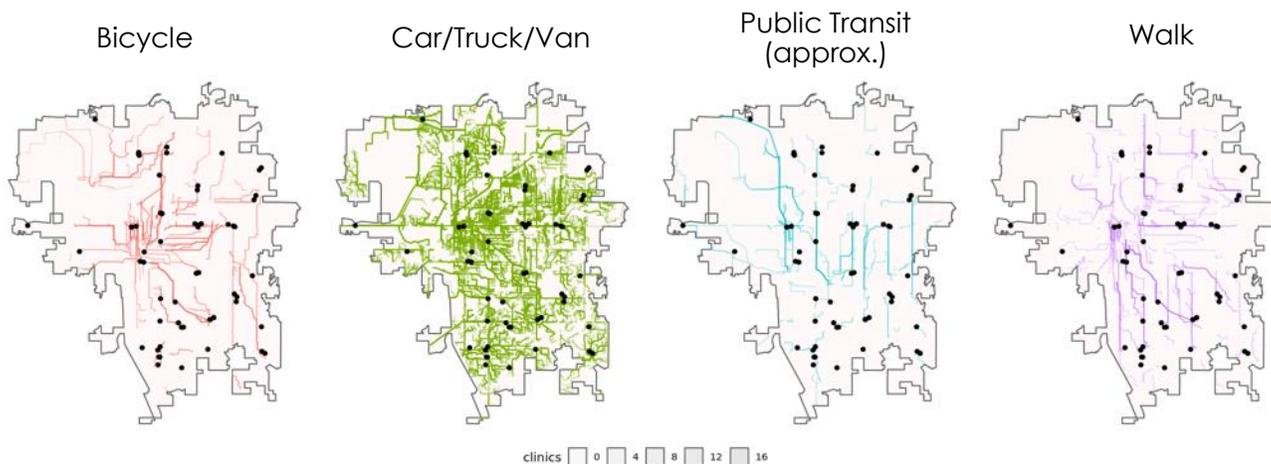


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https://commons.wikimedia.org/wiki/File:Map_of_Nebraska_highlighting_Lancaster_County.svg

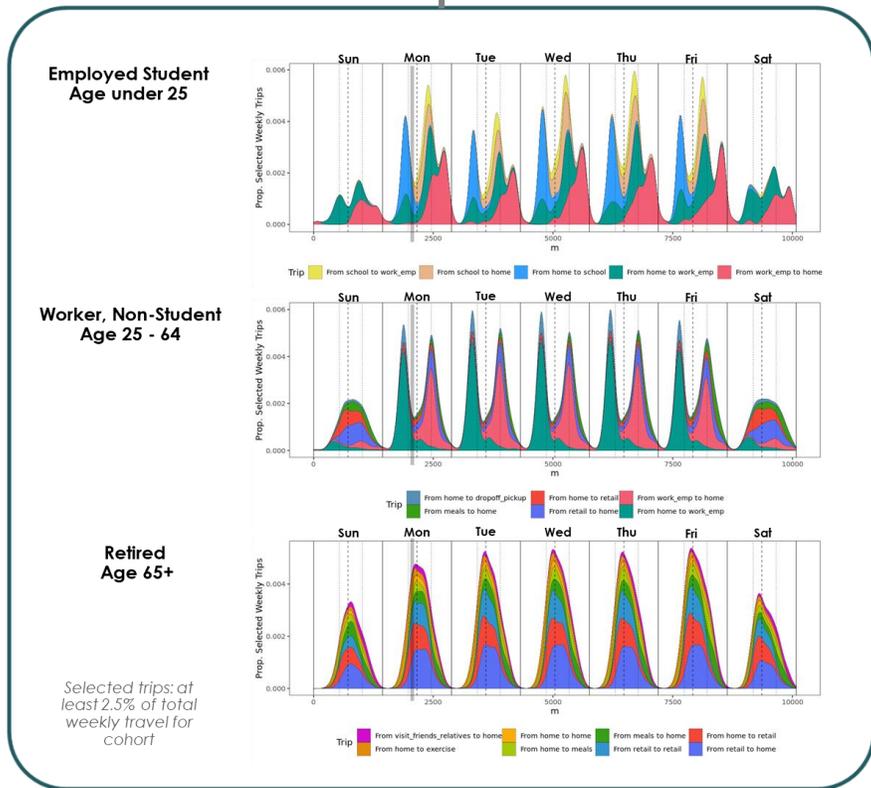
Tuccillo and Gaboardi (2022b) [13]

Case Study 3: Essential Activities

- Greatest access to COVID-19 healthcare services for commutes into downtown core, along major commercial corridors
- Access decreases with urban intensity, particularly in W/NW, SE Lincoln

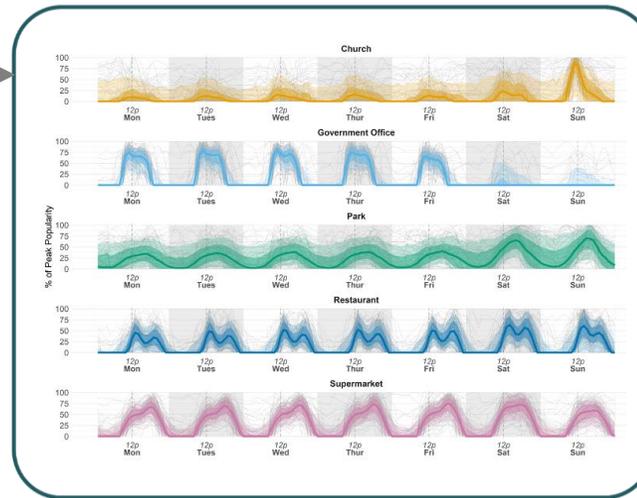


Up Next: Toward 24-Hour Synthetic Populations

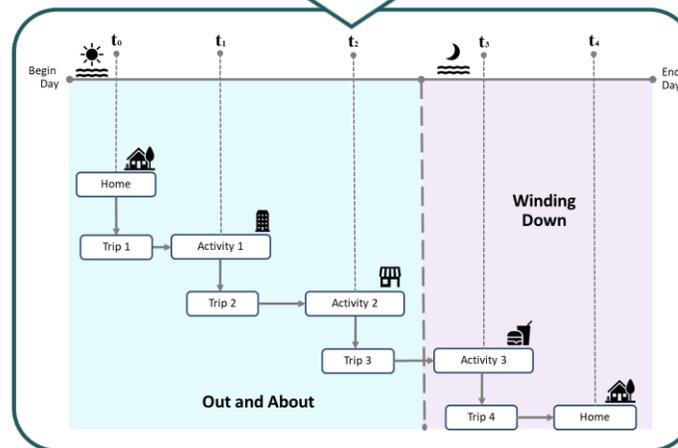


Time-Use and Travel Surveys

Tuccillo and Gaboardi (2023) [15]



Thakur et. al (2015) [14]

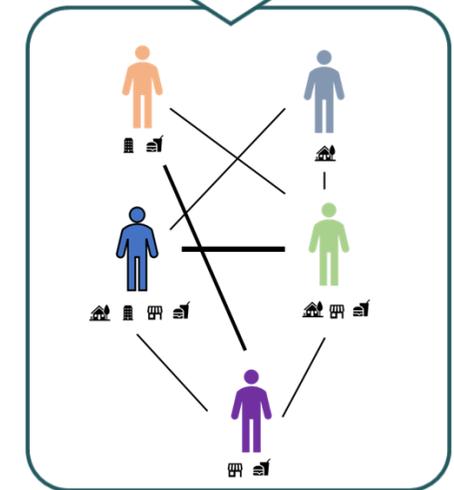


Activity Scheduling

Tuccillo and Gaboardi (2023) [15]



Activity Modeling



Social Contact Networks

Thank you!

Questions?

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