

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

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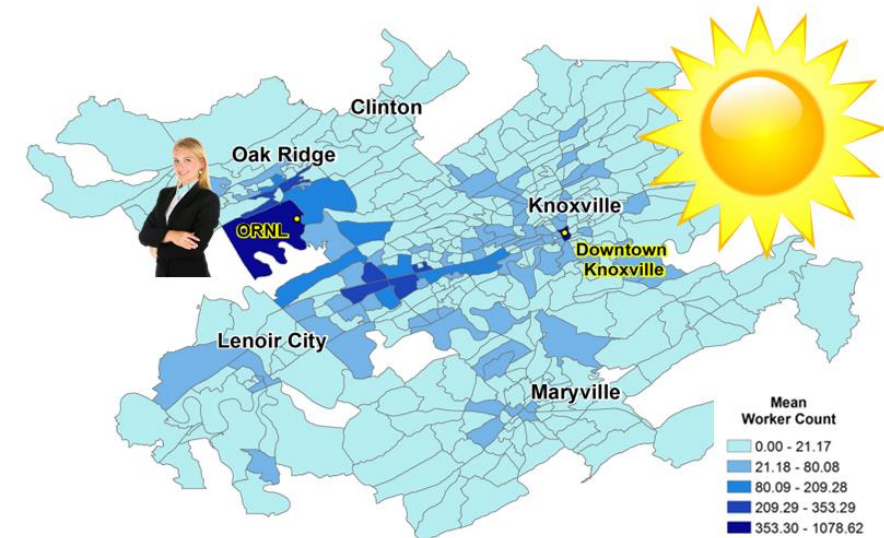
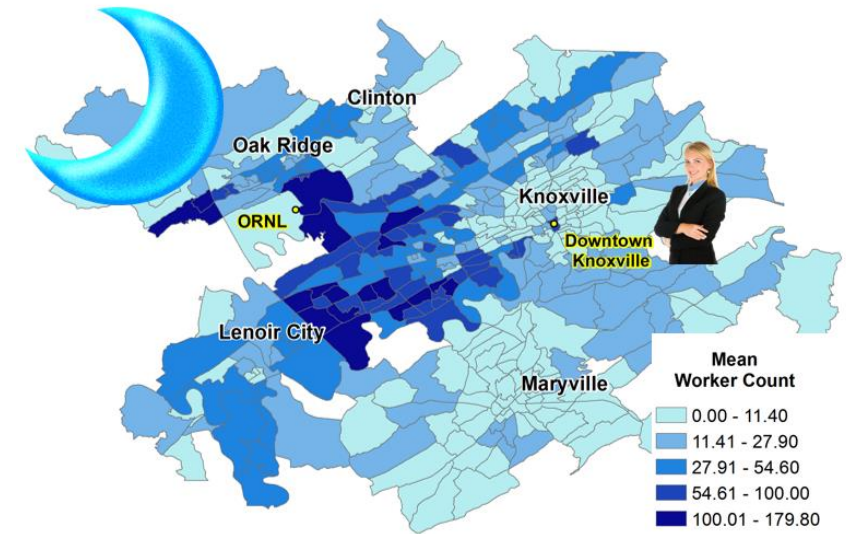
The UrbanPop Spatial Microsimulation Framework

- **Capabilities**

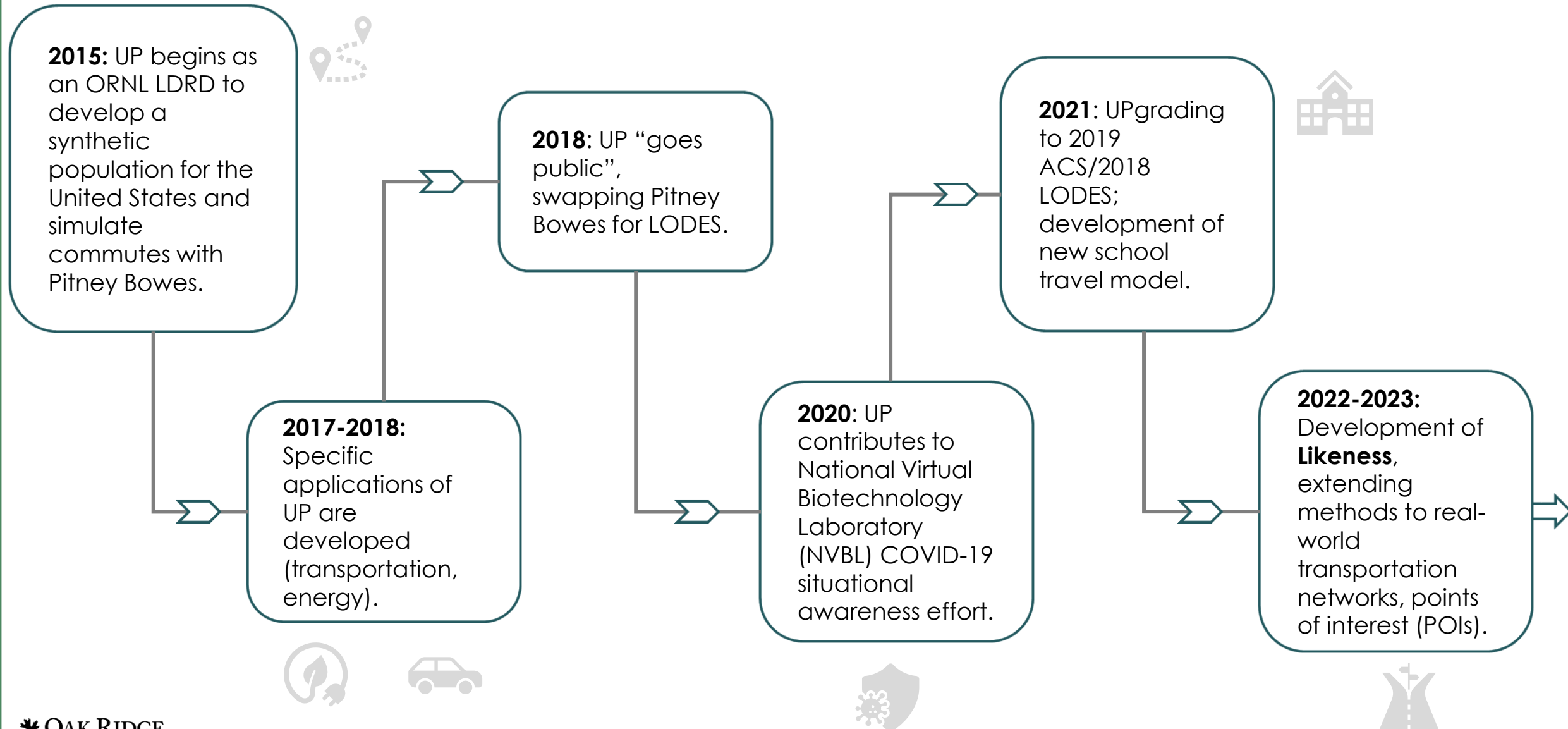
- Coupled individual-neighborhood profiles (*how people live*)
- Daytime/Nighttime dynamics (*how people move*)
- Agent Generation -> patterns of Life/Social contacts (*how people [inter]act*)

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

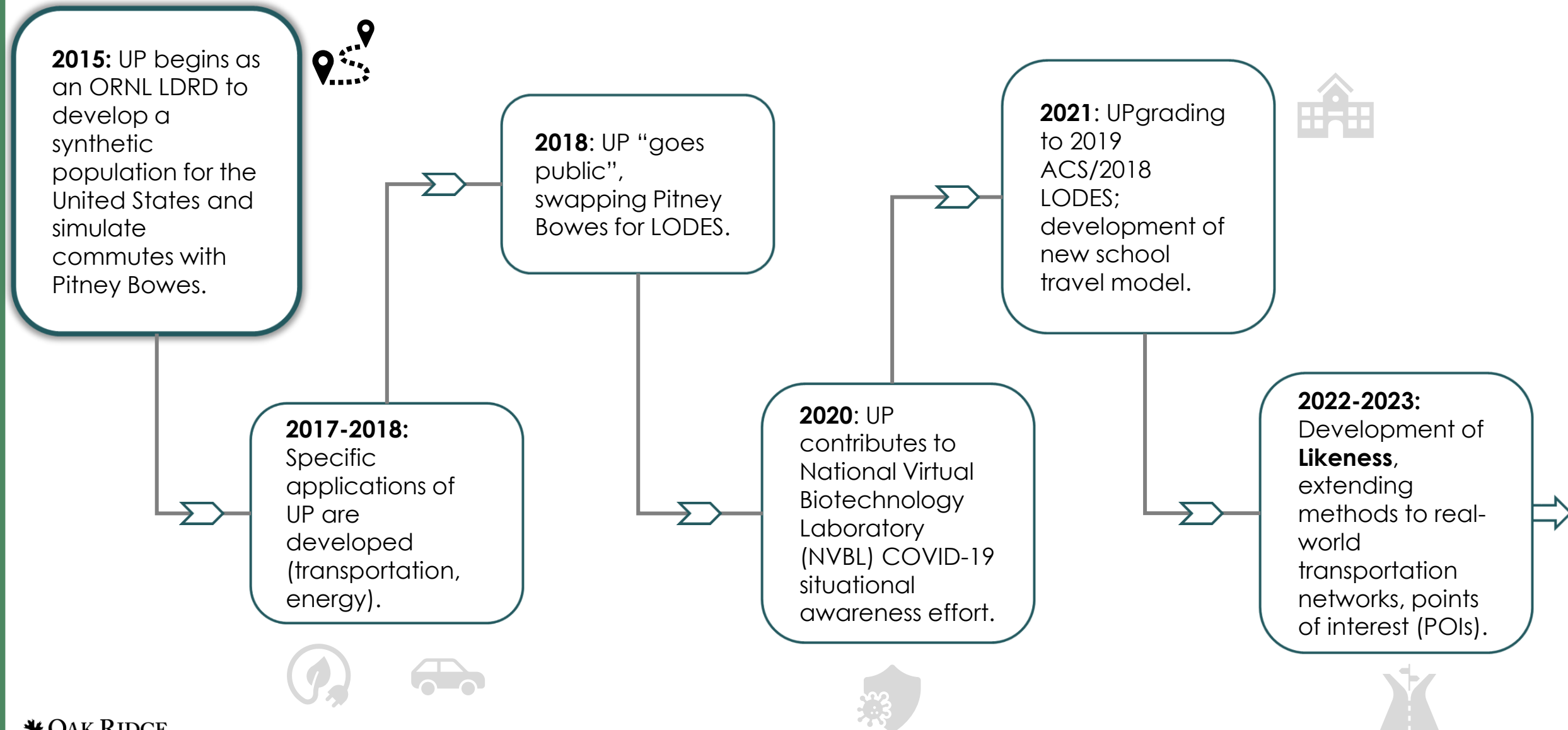
- **Available for any location in the United States 2016 – 2019**



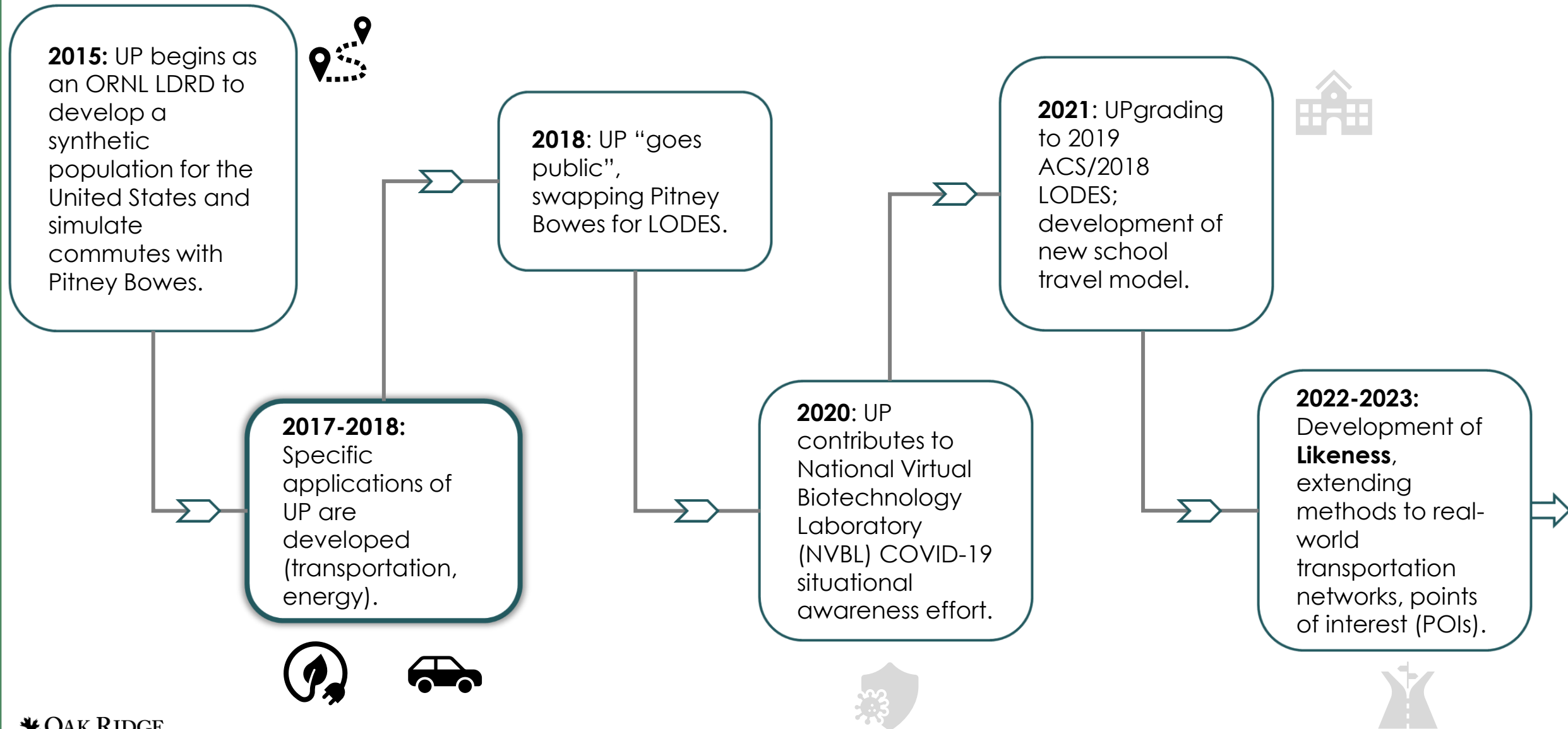
UrbanPop History



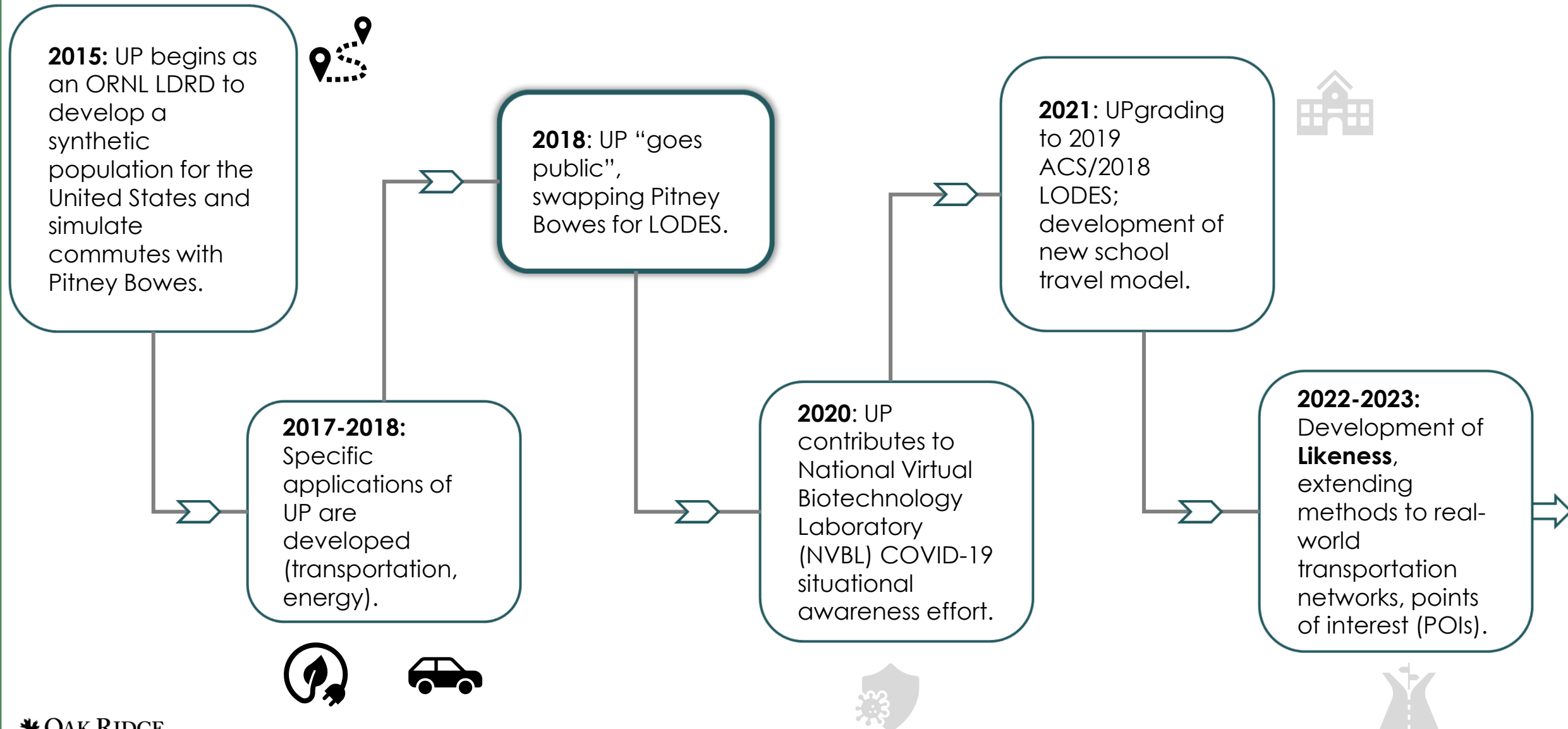
UrbanPop History



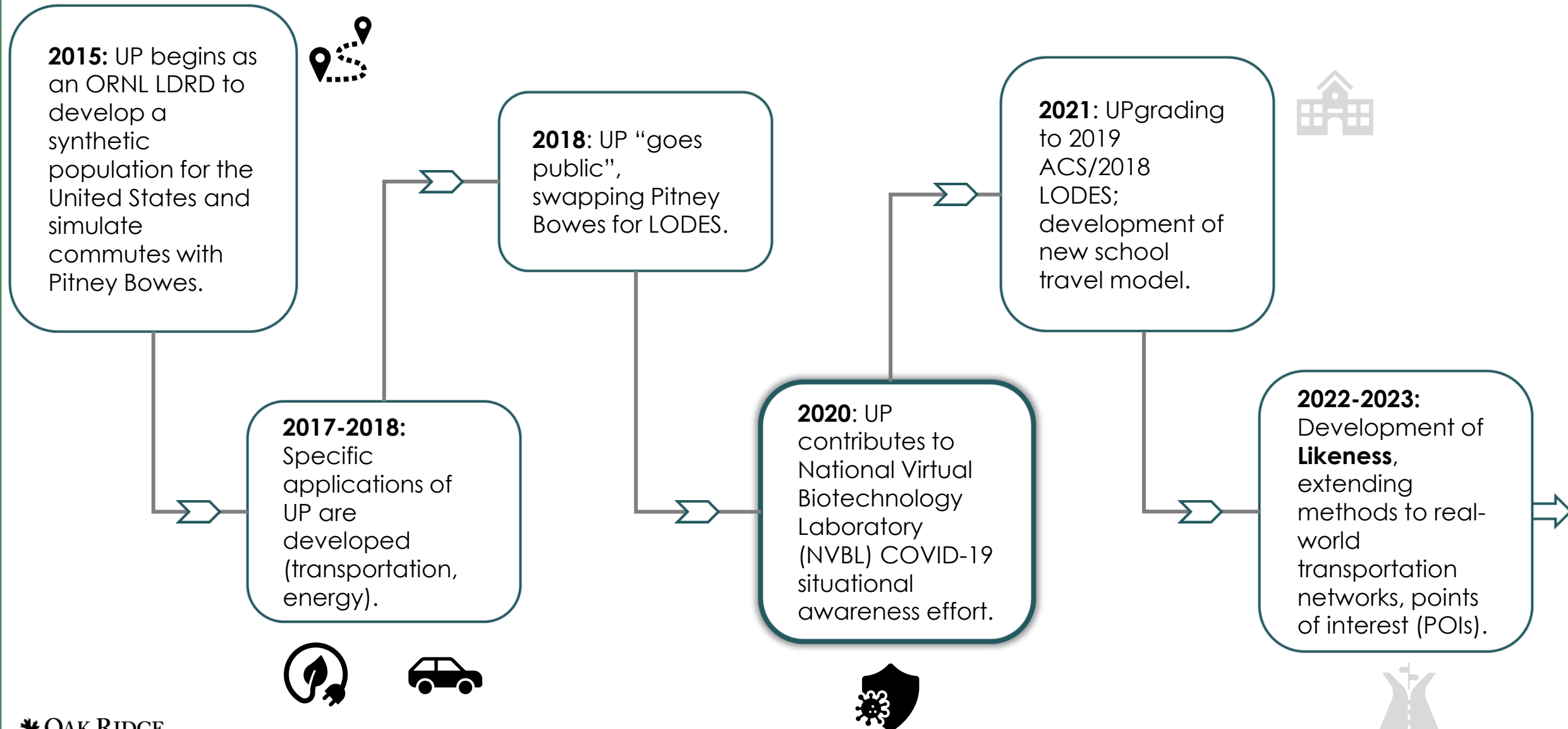
UrbanPop History



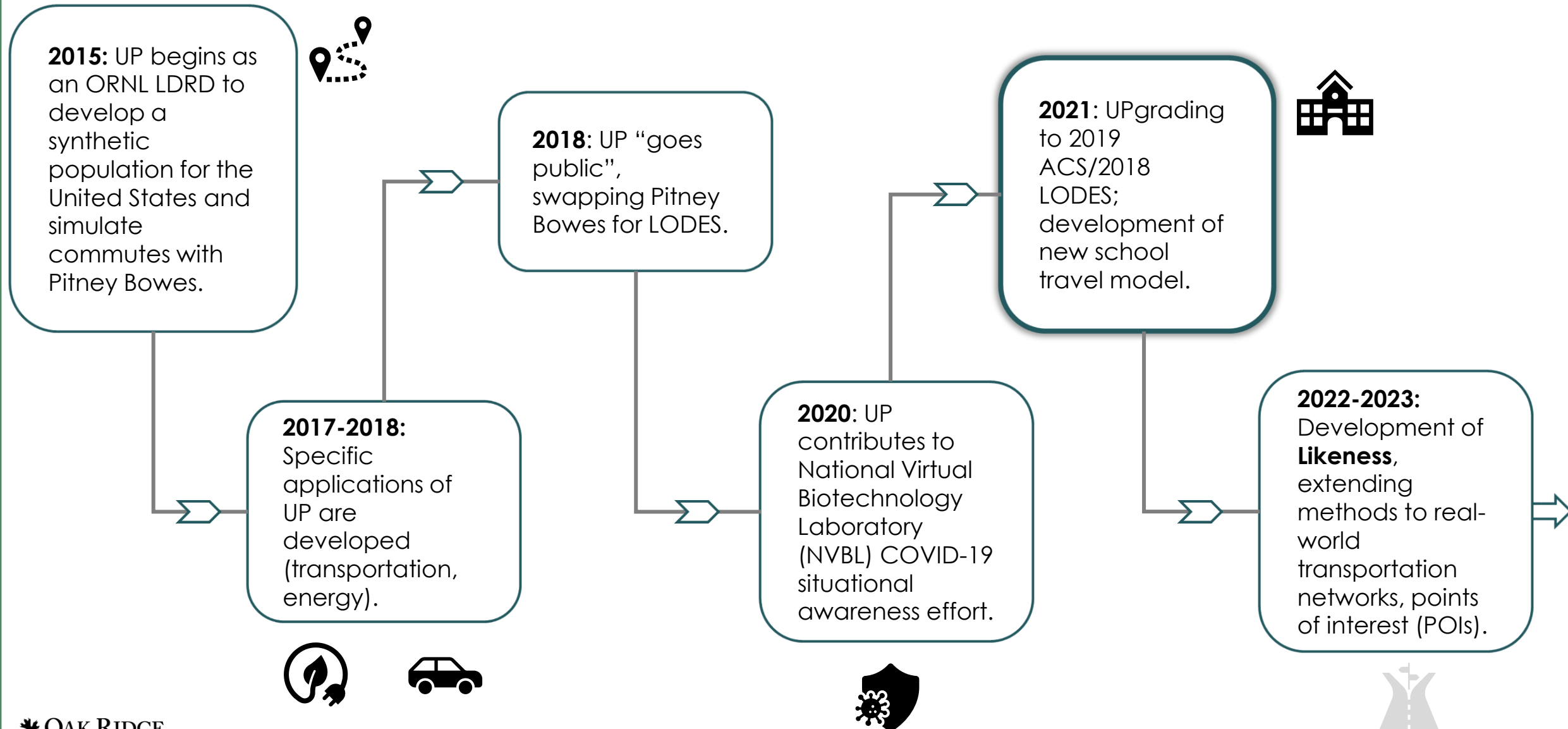
UrbanPop History



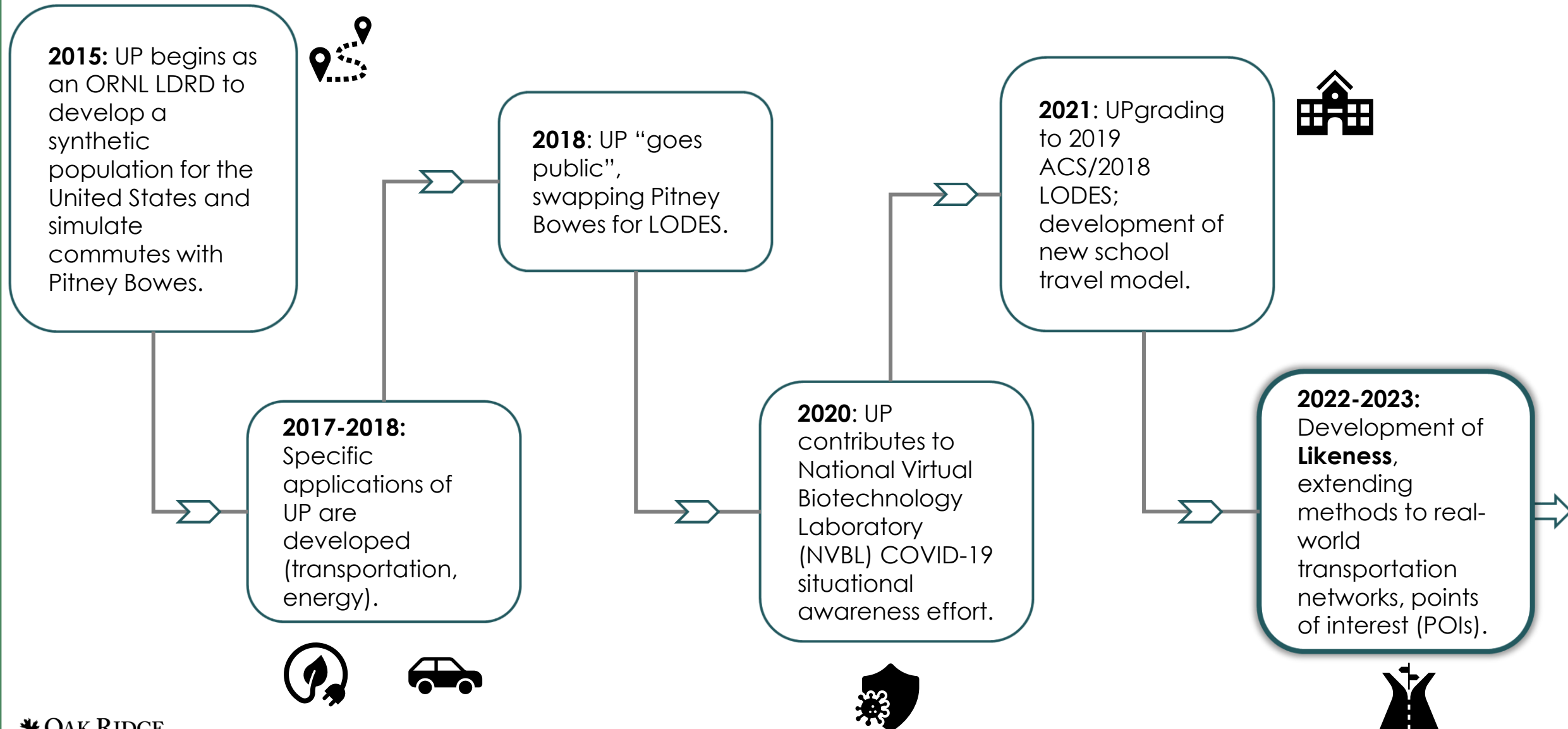
UrbanPop History



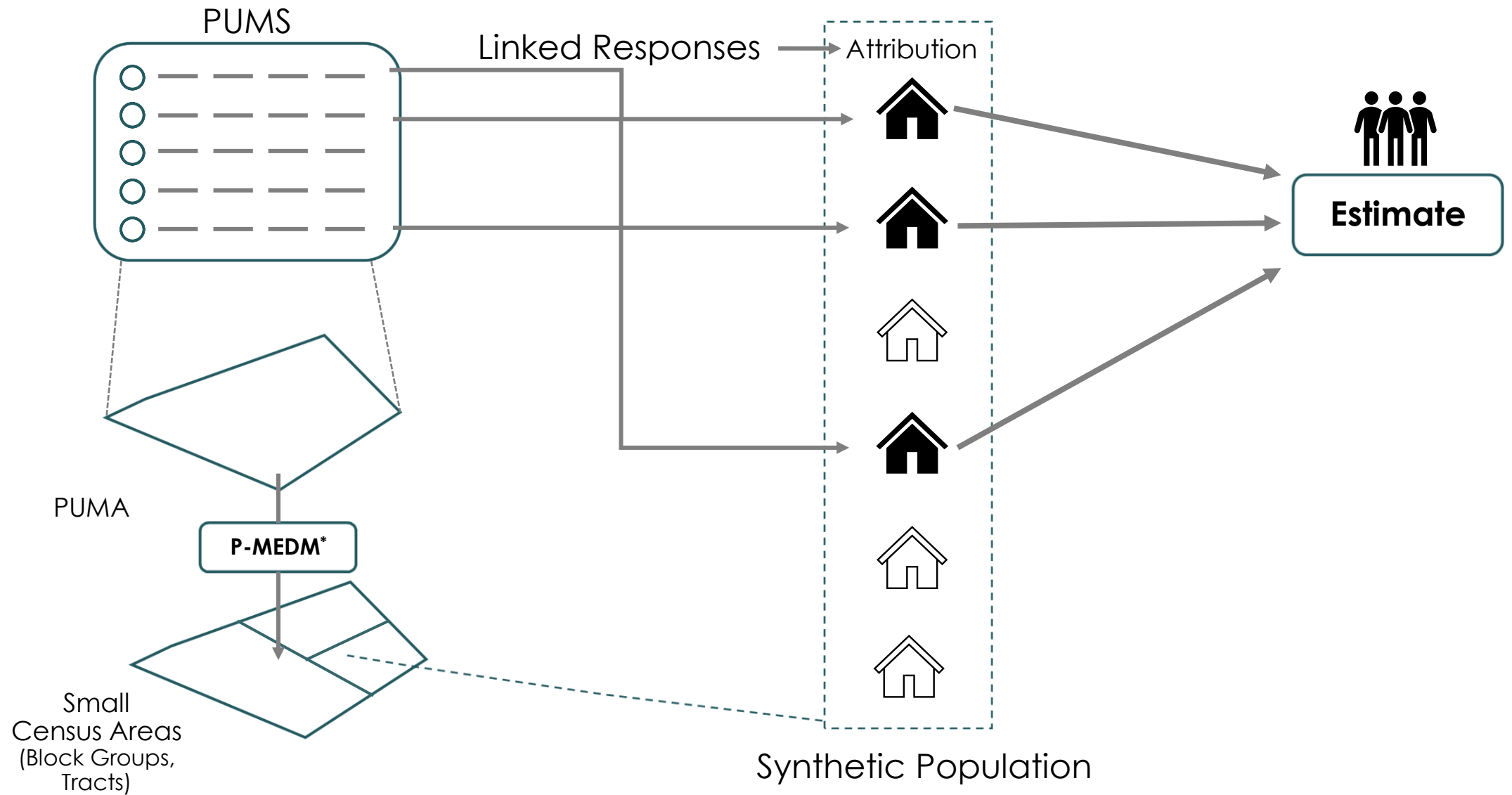
UrbanPop History



UrbanPop History



UrbanPop Core Workflow



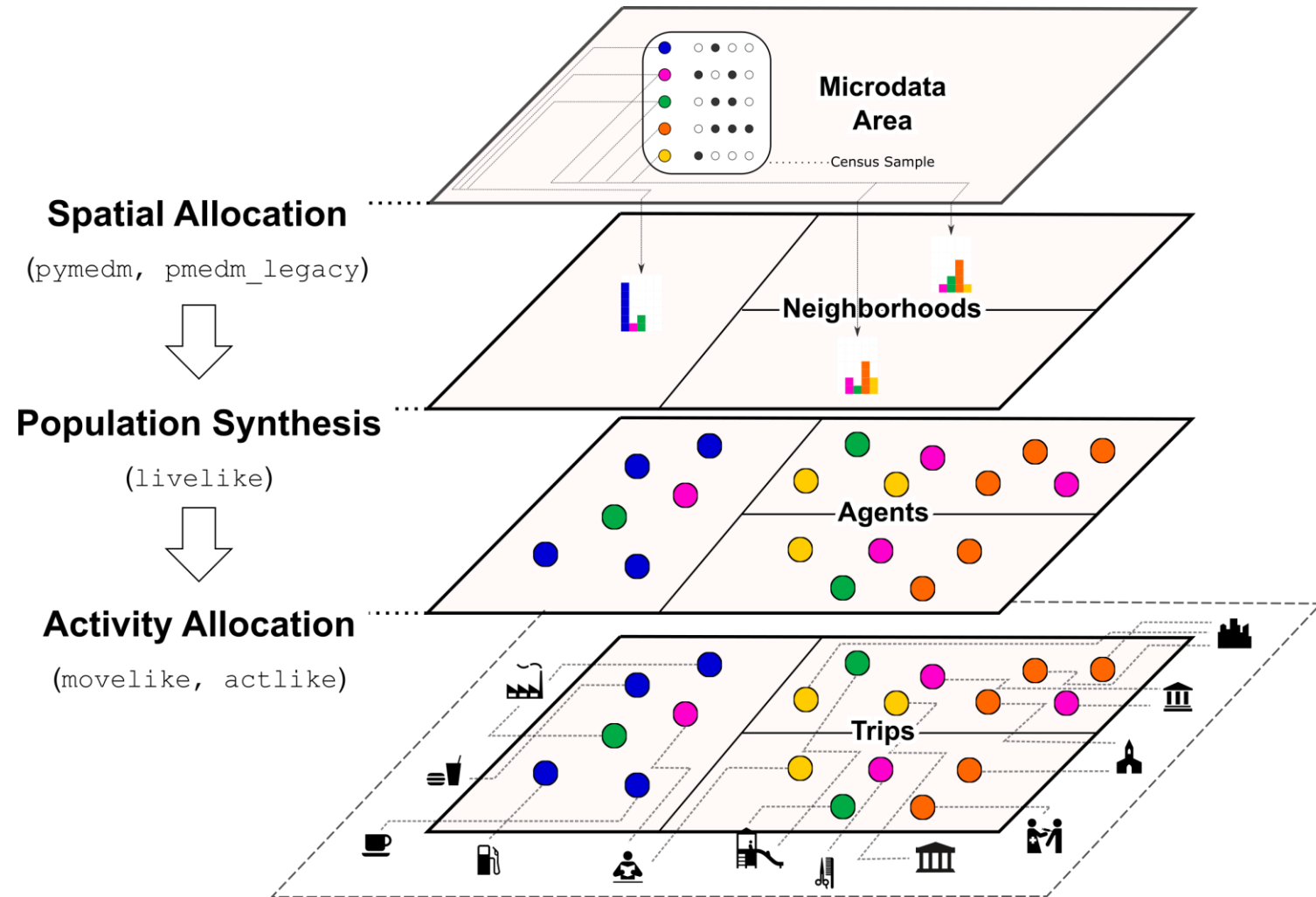
The Likeness Toolkit

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

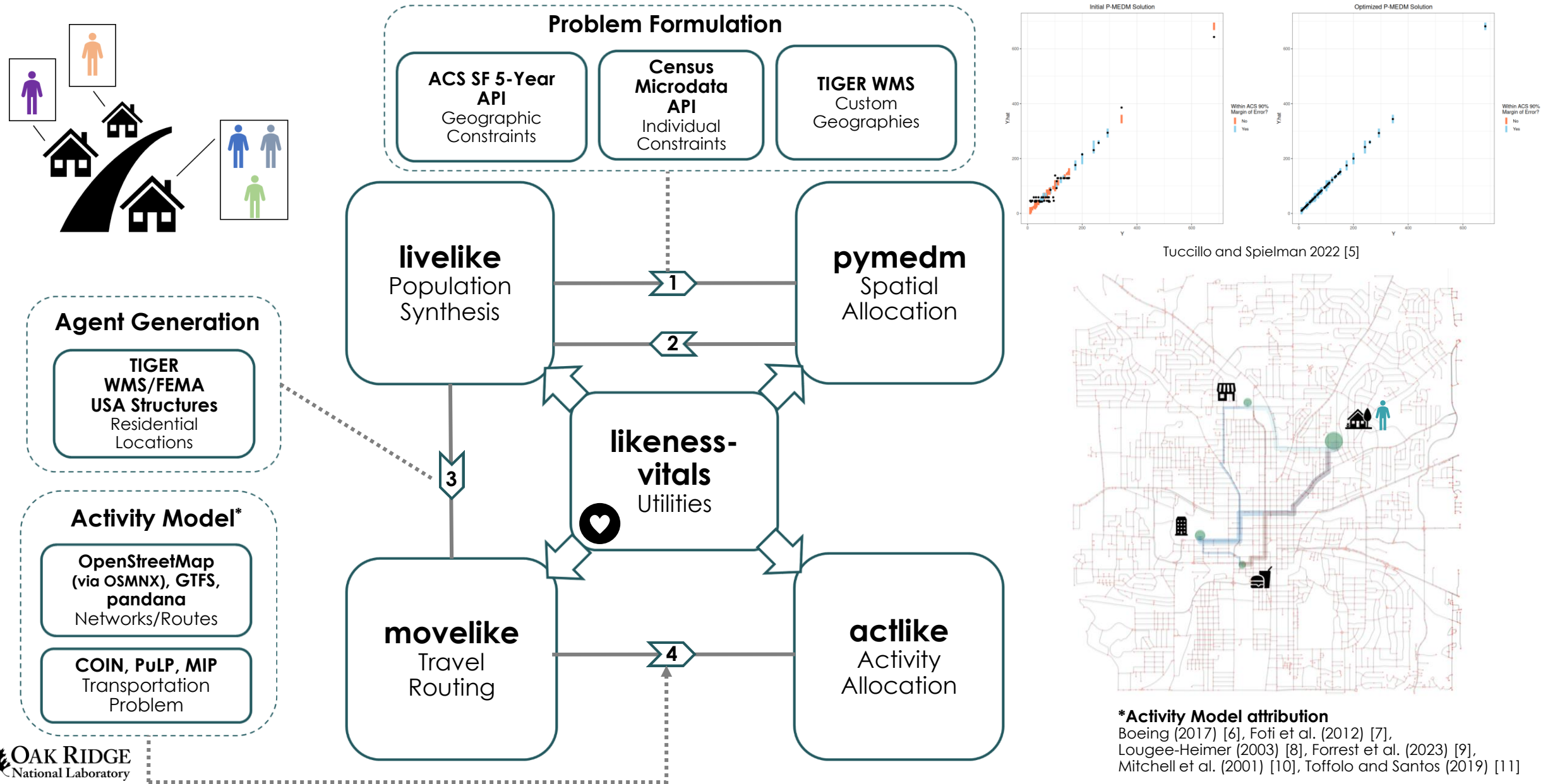
- Synthetic Population Generation
- Small-Area Estimates for neighborhood characterization

- **Enhancements**

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

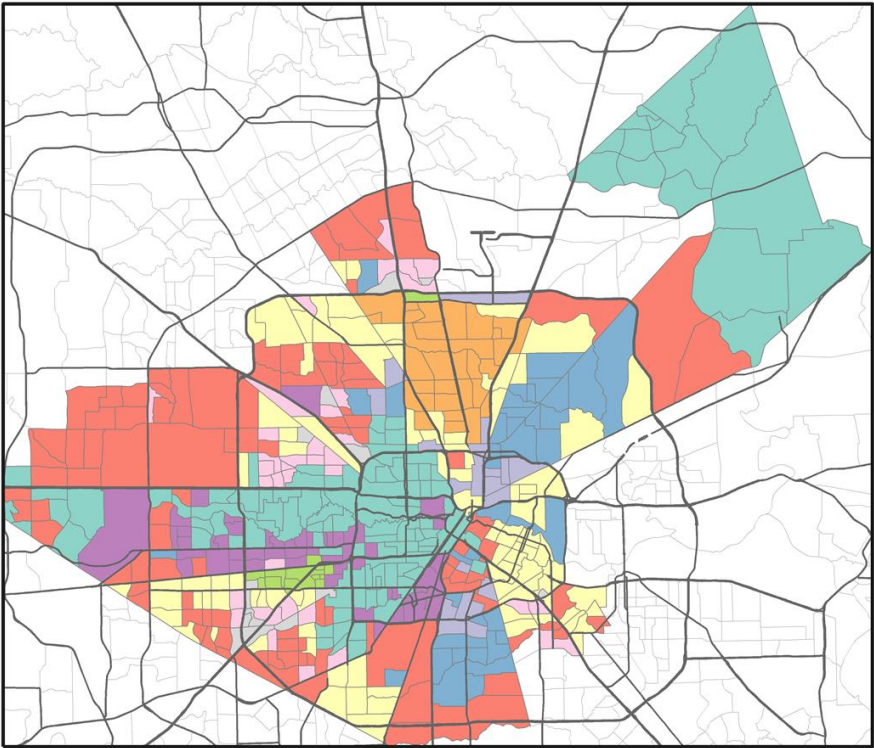


Likeness ecosystem



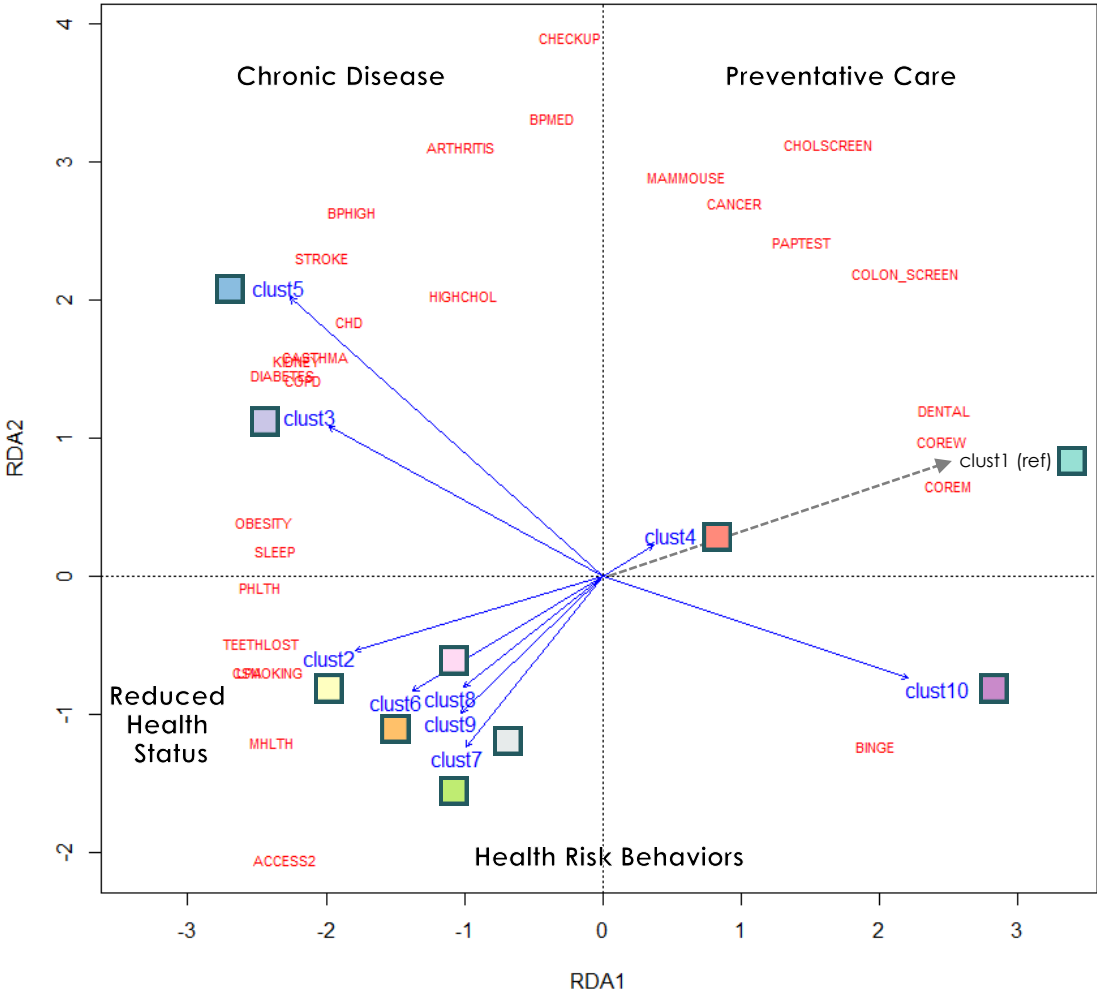
Case Study 1: Neighborhood riskscapes, health, and power outages

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%



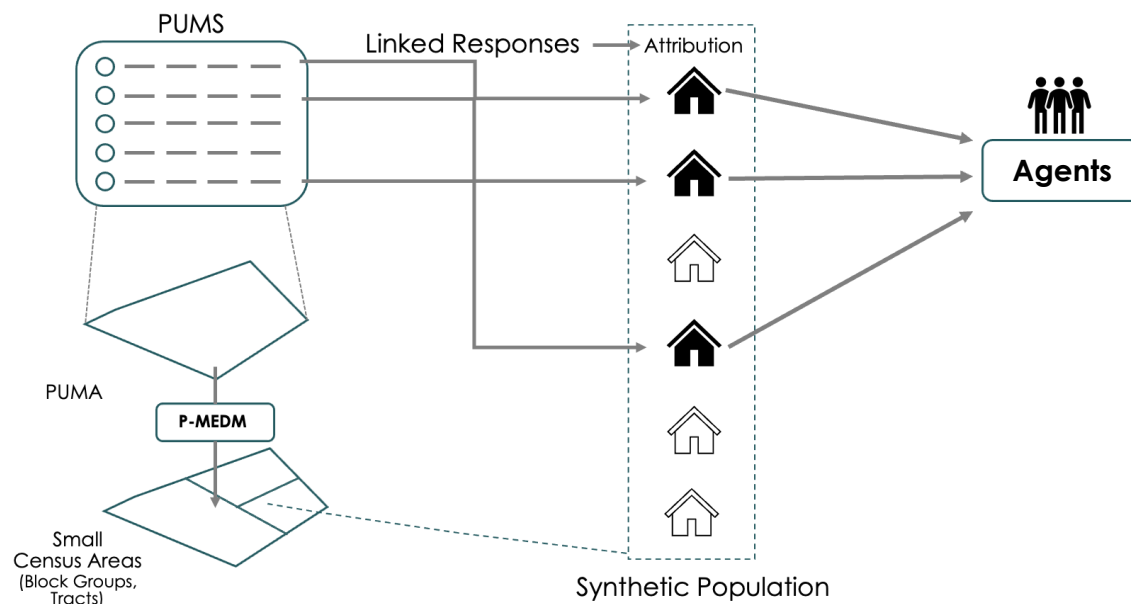
Redundancy Analysis (RDA) on CDC PLACES 2019

$R^2 = 0.65$ ($pr(F) < 0.001$)

Generating Household/Agent Locations

Spatial Allocation

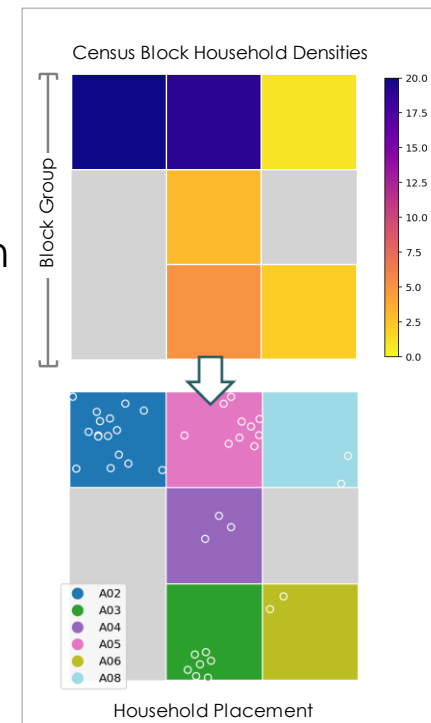
Core Workflow



Downscaling

Working Approach

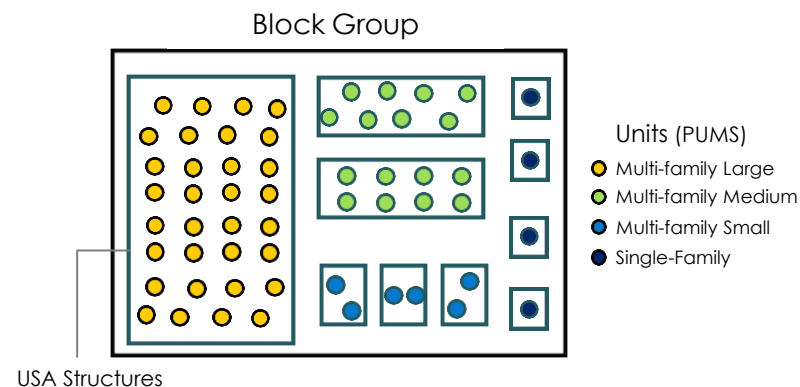
- Weighted random allocation of households to census blocks based on 2010 Decennial Census household density
- Variable-size binpacking problem
- Clustered placement of household points within blocks



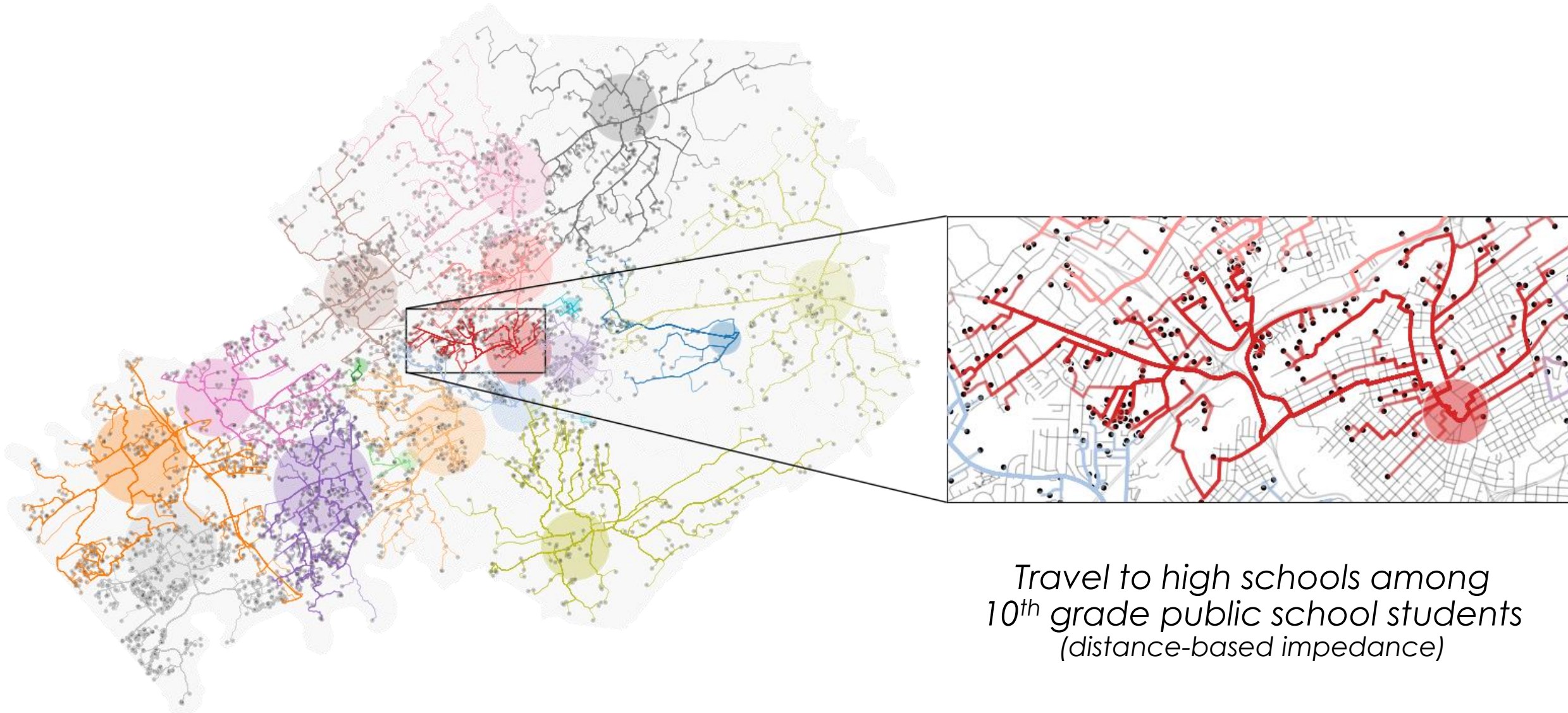
Downscaling

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

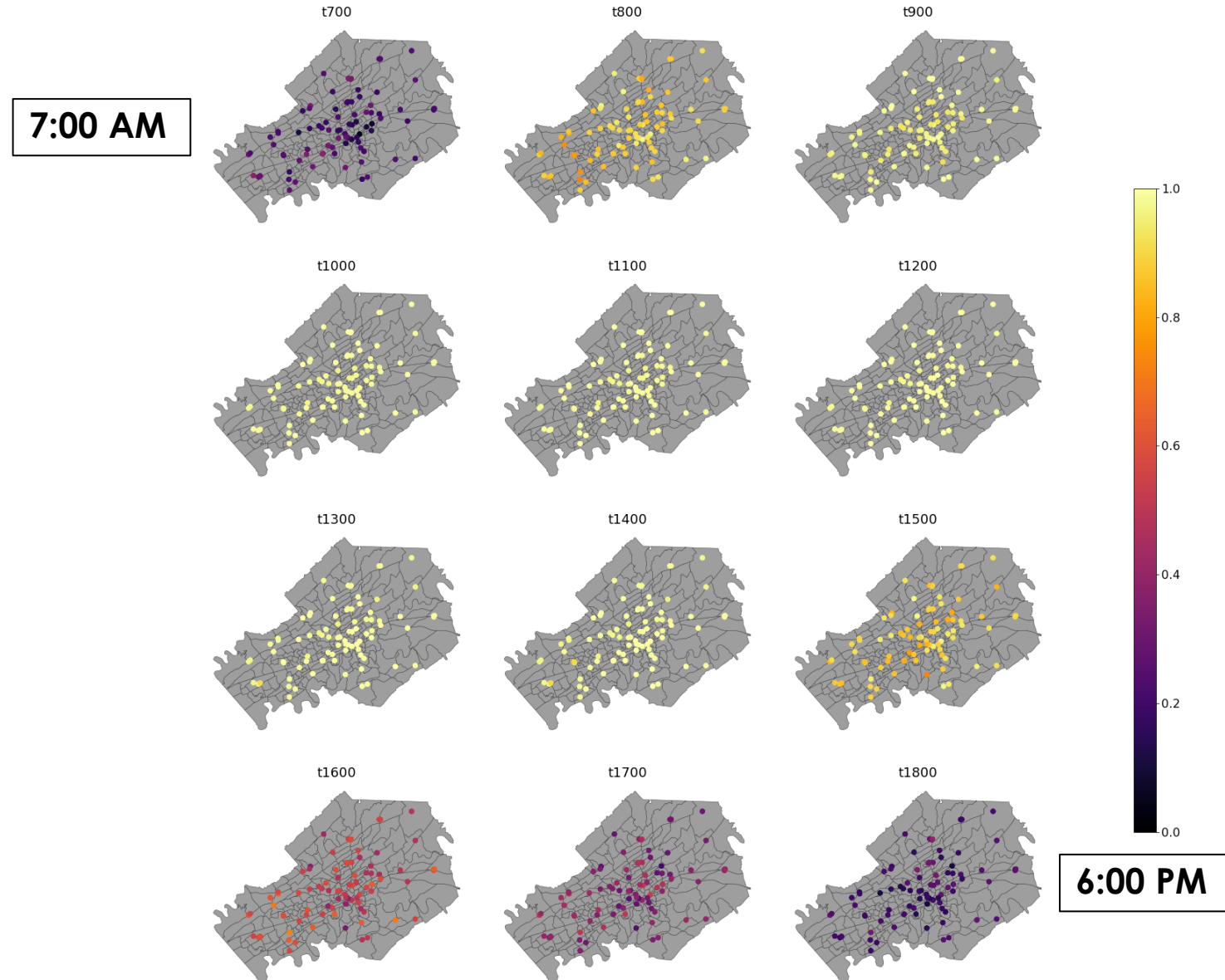
- **Educator arrivals**

- PUMS time of arrival at work (JWAP)

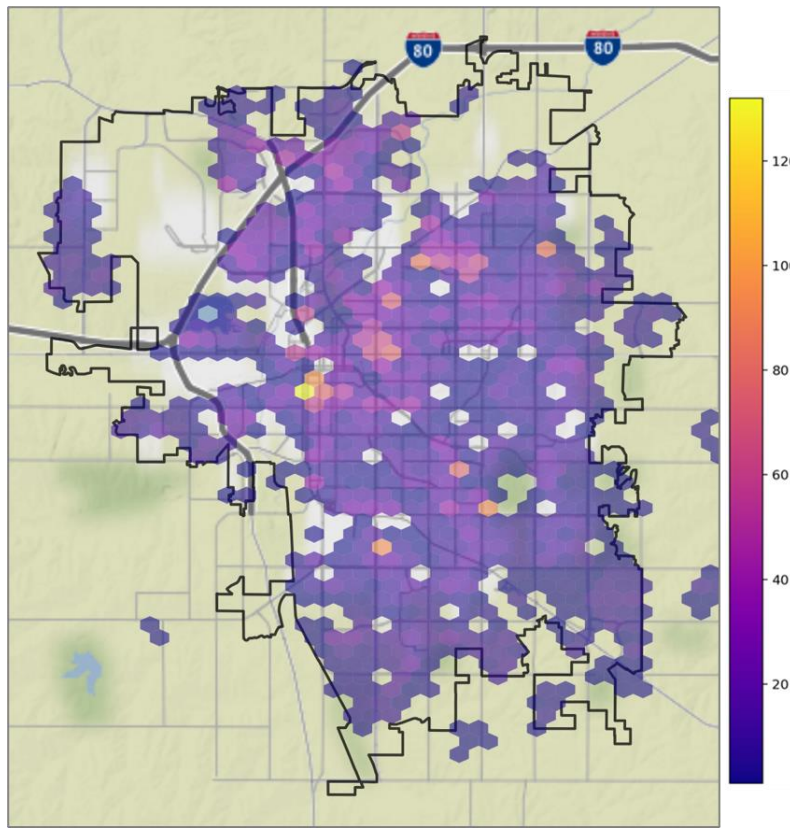
- **Educator departures**

- $\text{PUMS JWAP} + (\text{WKHP} / 5)$

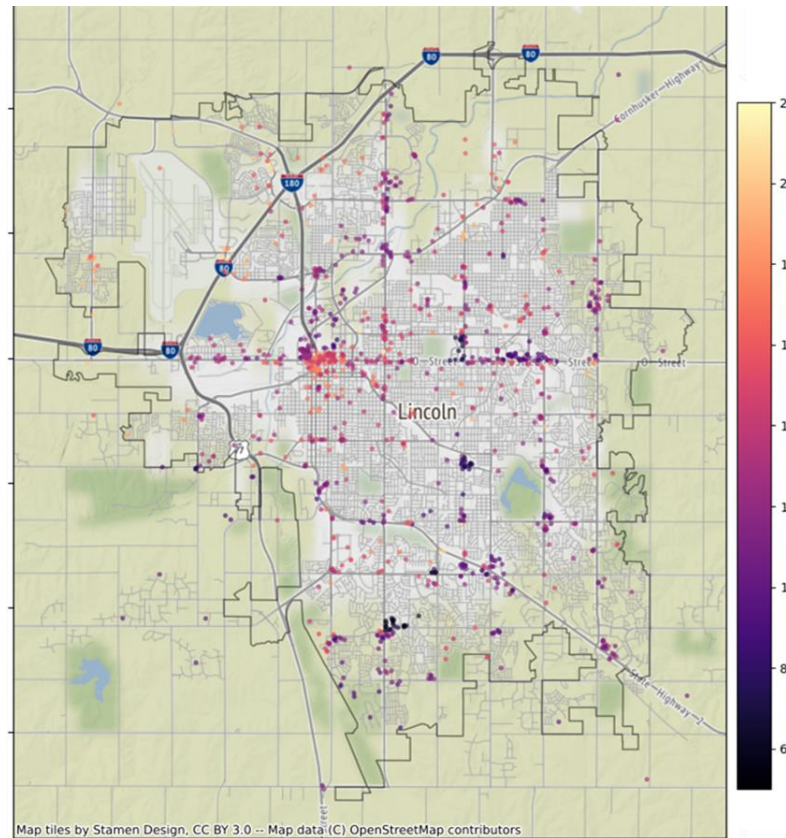
where WKHP represents the number of hours worked per week



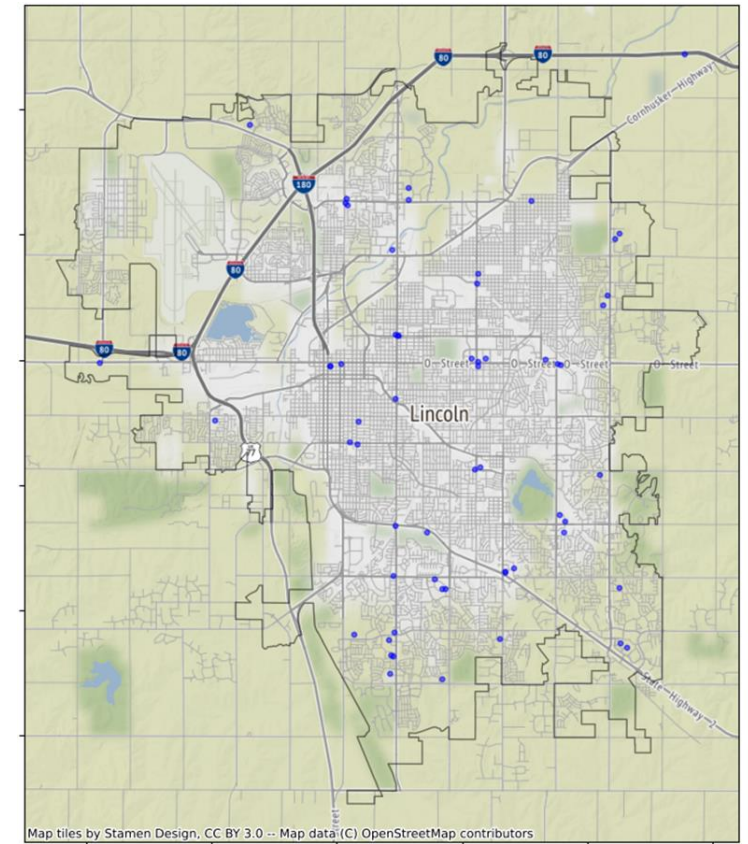
Case Study 3: Essential Activities



Synthetic Population
Food Service Workers



Job Locations
Workers per Establishment



COVID-19 Healthcare Services
Vaccination and Testing Centers

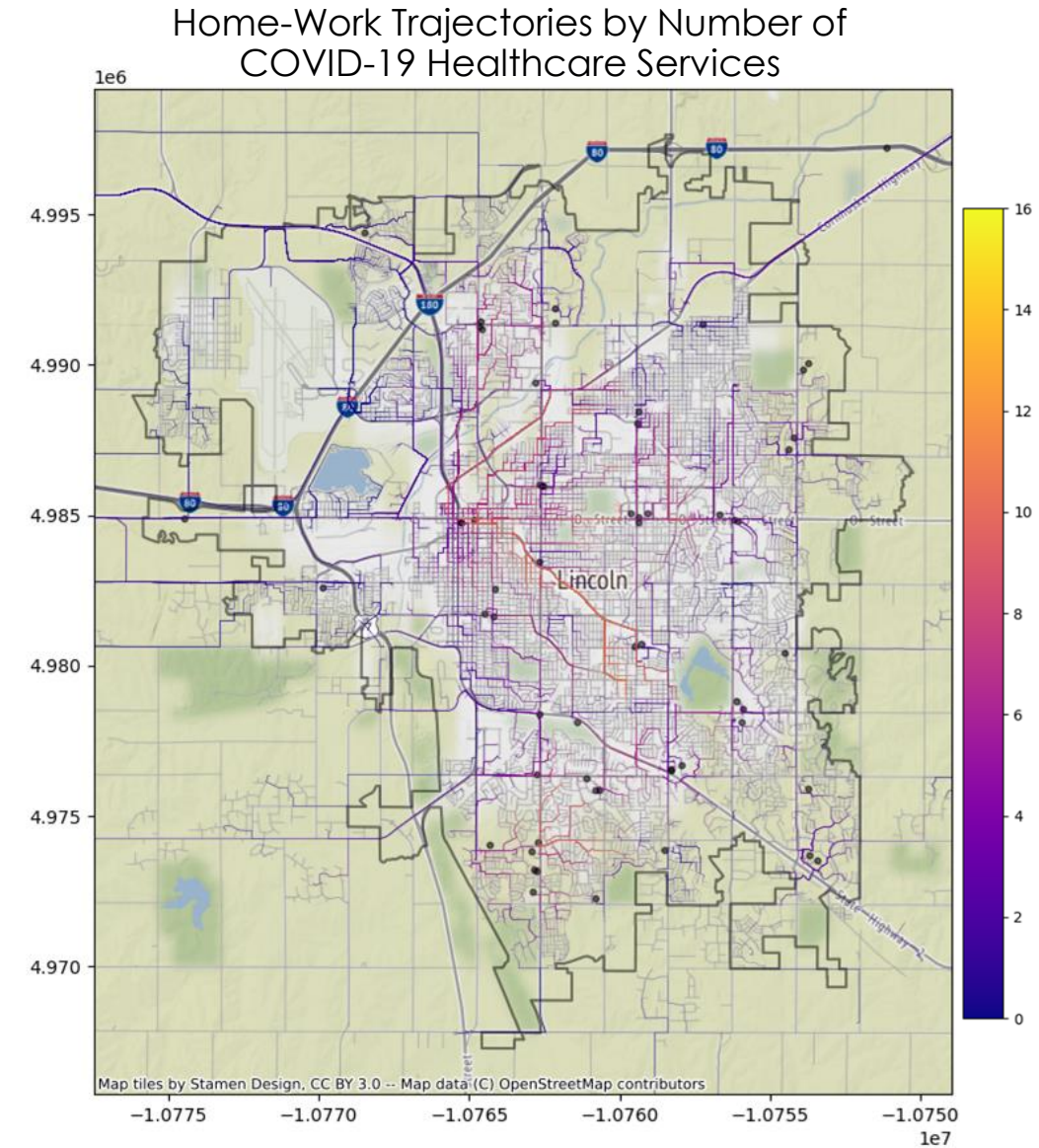
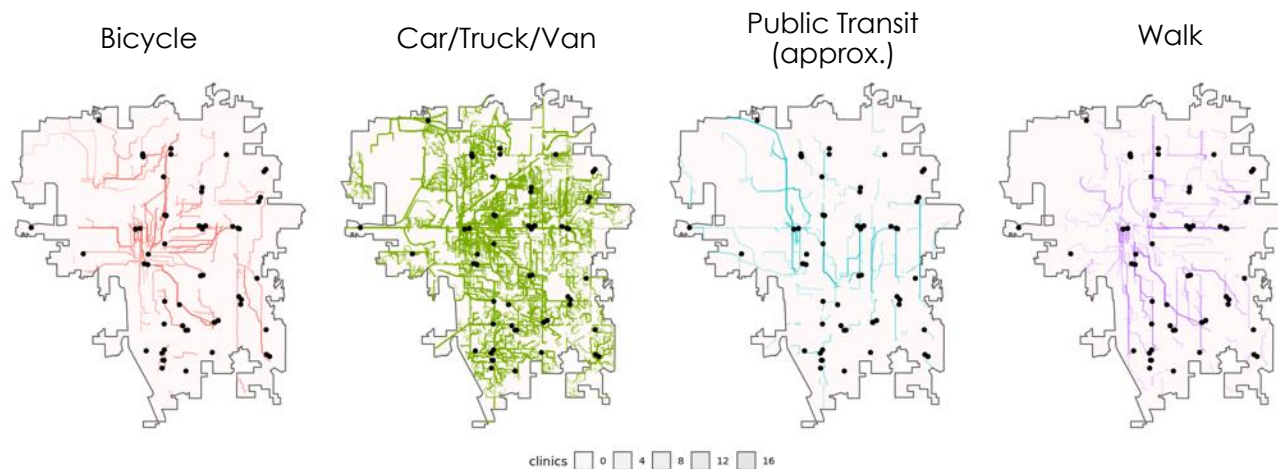


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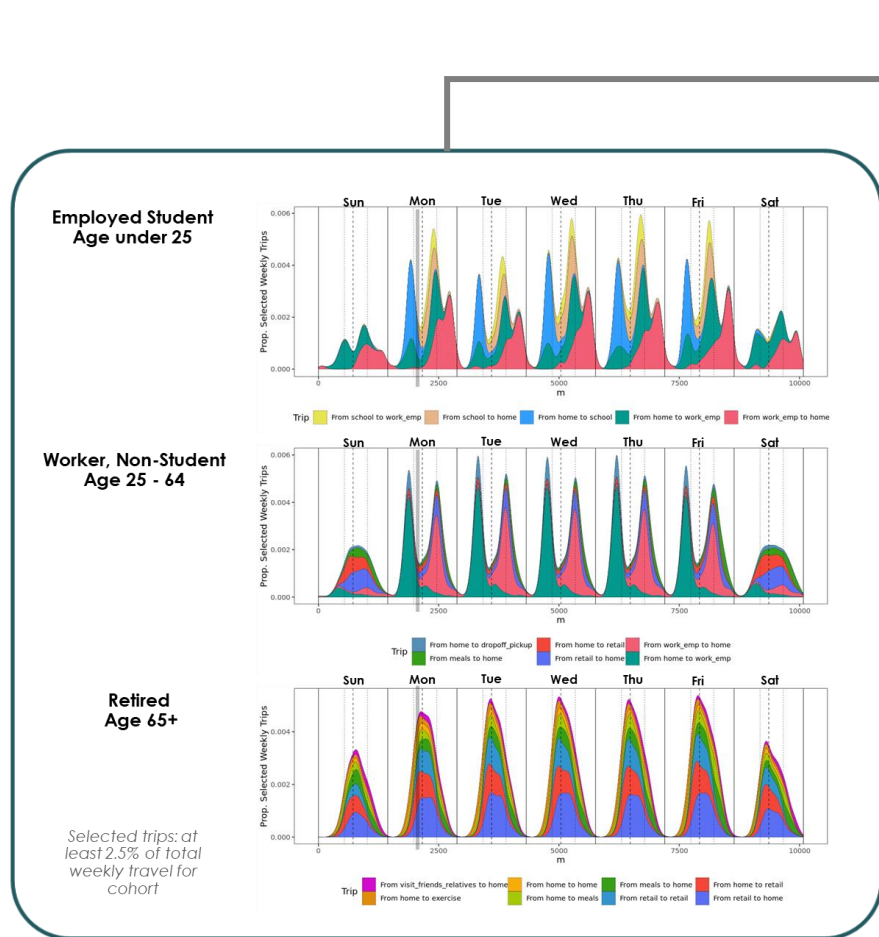
Tuccillo and Gaboardi (2022b) [13]

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

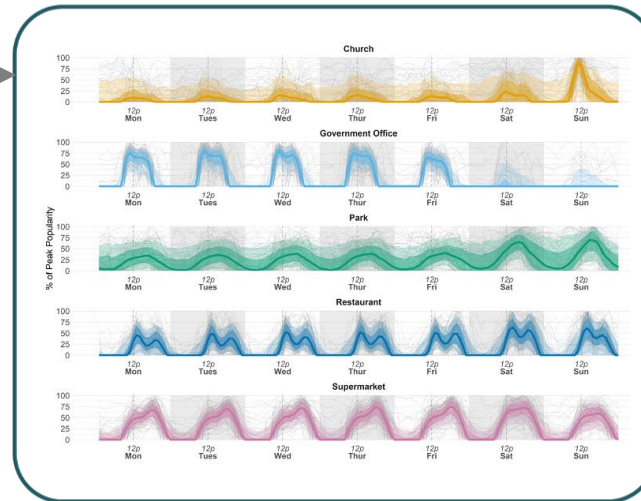


Up Next: Toward 24-Hour Synthetic Populations



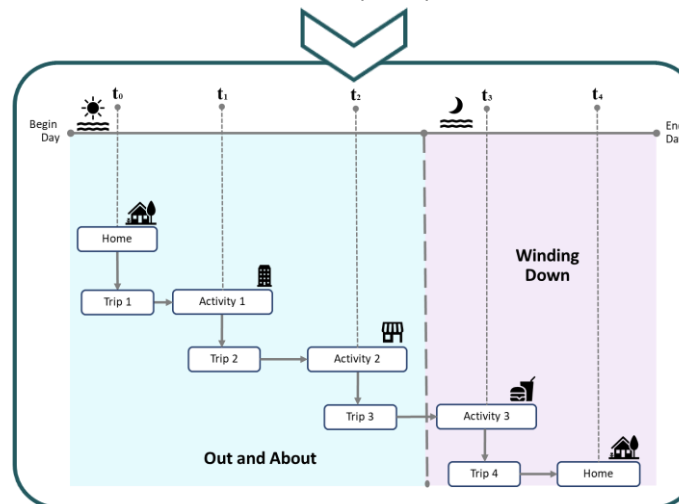
Time-Use and Travel Surveys

Tuccillo and Gaboardi (2023) [15]



POI Popularity Curves

Thakur et. al (2015) [14]

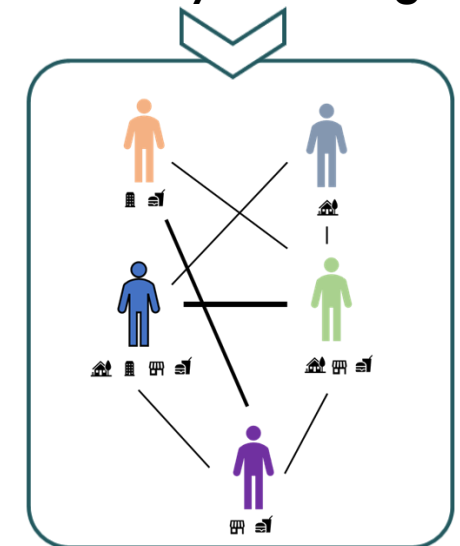


Activity Scheduling

Tuccillo and Gaboardi (2023) [15]



Activity Modeling



Social Contact Networks

Thank you!

Questions?

References

1. Tuccillo, J., Stewart, R., Rose, A., Trombley, N., Moehl, J., Nagle, N., and Bhaduri, B. L. (2023). UrbanPop: A spatial microsimulation framework for exploring demographic influences on human dynamics. *Applied Geography*, 151, 102844.
2. Stewart, R., Aziz, H., Morton, A., Park, H., Duchscherer, S., Piburn, J., Maness, M., and Nagle, N. N. (2015). Population as a Service: Simulating the American population for modeling urban dynamics. Poster presentation, *Oak Ridge National Laboratory (ORNL)*, Oak Ridge, TN (United States).
3. Nagle, N. N., Buttenfield, B. P., Leyk, S., and Spielman, S. E. (2014). Dasymetric modeling and uncertainty. *Annals of the Association of American Geographers* 104 (1), 80–95.
4. Tuccillo, J. V. and Gaboardi, J. D. (2022a). Likeness: a toolkit for connecting the social fabric of place to human dynamics. In Agarwal, M., Calloway, C., Niederhut, D., and Shupe, D. (Eds.), *Proceedings of the 21st Python in Science Conference*, pp. 125 – 135.
5. Tuccillo, J. V., and Spielman, S. E. (2022). A method for measuring coupled individual and social vulnerability to environmental hazards. *Annals of the American Association of Geographers*, 112(6), 1702-1725.
6. Boeing, G. (2017). OSMnx: New methods for acquiring, constructing, analyzing, and visualizing complex street networks. *Computers, Environment and Urban Systems*, 65, 126-139.
7. Foti, F., Waddell, P., and Luxen, D. (2012, April). A generalized computational framework for accessibility: from the pedestrian to the metropolitan scale. In *Proceedings of the 4th TRB conference on innovations in travel modeling*. Transportation research board (pp. 1-14).
8. Lougee-Heimer, R. (2003, January). The Common Optimization INterface for Operations Research. *IBM Journal of Research and Development*, 47(1):57-66.
9. Forrest, J. et al. (2023). coin-or/Cbc. Zenodo. <https://doi.org/10.5281/zenodo.7843975>
10. Mitchell, S., O'Sullivan, M., and Dunning, I. (2001). PuLP: a linear programming toolkit for python. *The University of Auckland*, Auckland, New Zealand, 65.
11. Toffolo, T. A. M. and Santos, H. G. (2019). PYTHON-MIP. <https://www.python-mip.com/>
12. Kahl, A., Tuccillo, J. V., Luke, N., Tran, L., Bhaduri, B. L., and Dumas M. (2022, July). Coupling vulnerability metrics to assess the impacts of large-scale winter storms. *Natural Hazards Workshop Researchers Meeting 2022*.
13. Tuccillo, J. V. and Gaboardi, J. D. (2022b, November). Examining COVID-19 Vaccine Access Profiles through Spatial Microsimulation and Activity Allocation. *69th Annual North American Meetings of the Regional Science Association International*, Montreal, Quebec, Canada.
14. Thakur, G. S., Bhaduri, B. L., Piburn, J. O., Sims, K. M., Stewart, R. N., and Urban, M. L. (2015). PlanetSense: a real-time streaming and spatio-temporal analytics platform for gathering geo-spatial intelligence from open source data. In *Proceedings of the 23rd SIGSPATIAL International Conference on Advances in Geographic Information Systems*, pp. 1–4.
15. Tuccillo, J. V. and Gaboardi, J. D. (2023). Demographic characterization of human dynamics models with time-use and travel surveys. *American Association of Geographers 2023 Annual Meeting (AAG)*, Denver, Colorado, United States. Zenodo. <https://doi.org/10.5281/zenodo.7853705>

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