

# Simulating Urban Flow with Geographically Explicit Synthetic Population

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 **University at Buffalo** The State University of New York



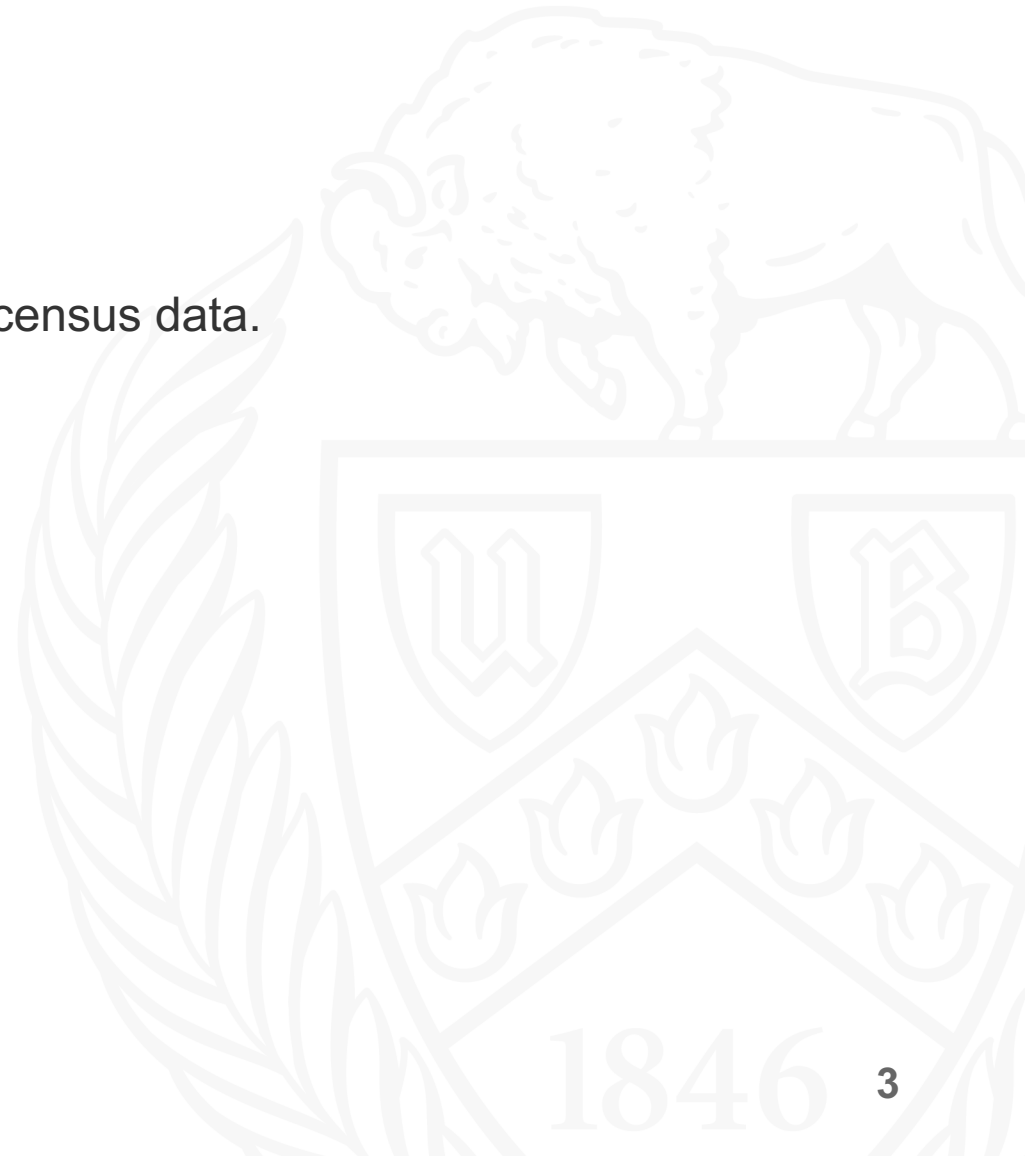
# Introduction

- Urban human mobility is an active research field that studies human mobility patterns in urban areas.
- Through individual's movement, higher level phenomena such as traffic congestion and disease outbreaks emerge.
- Understanding how and why people move around a city plays an important role in urban planning, traffic control, and public health.
- An agent-based model (ABM) is developed to simulate people's daily commute patterns.

 How can we utilize census data in ABM to uncover emerging urban commute patterns?

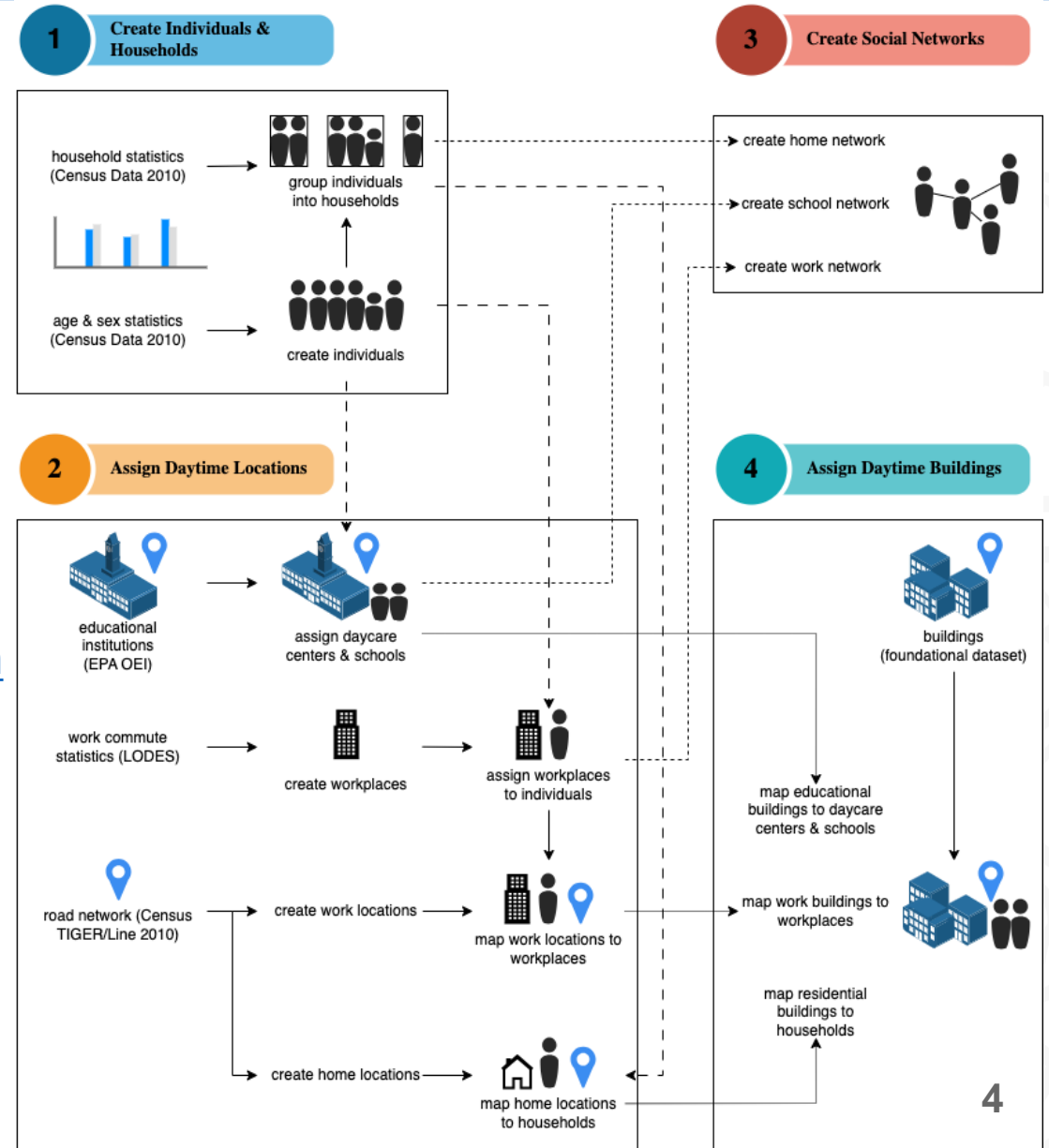
# Approach

- Study area: Manhattan, New York City.
- Agents are initialized with a synthetic population created out of census data.
- Agents' home and work locations are derived from census data.



# Approach: Synthetic Population

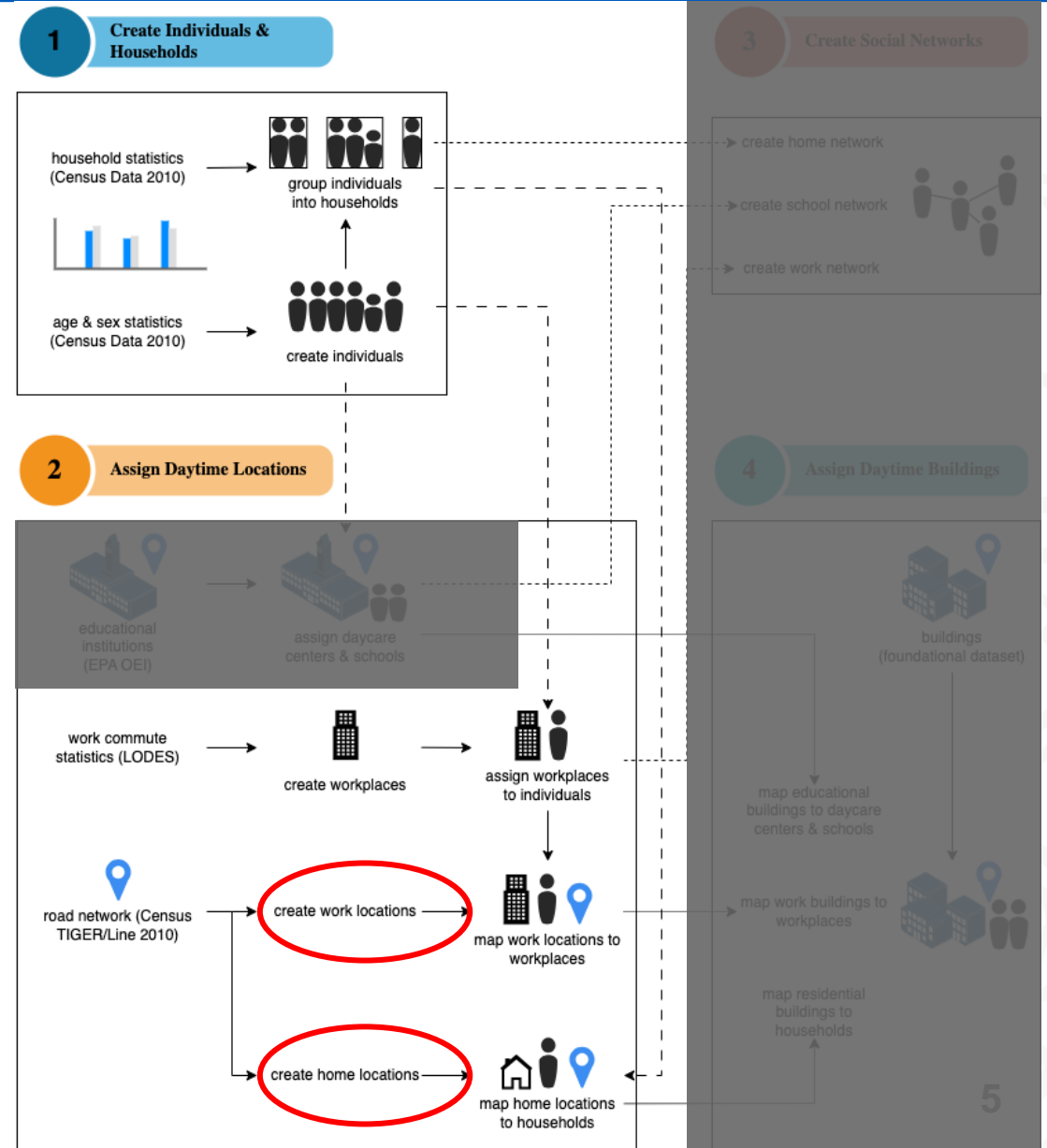
- Jiang, N., Crooks, A.T., Kavak, H. *et al.* A method to create a synthetic population with social networks for geographically-explicit agent-based models. *Comput.Urban Sci.* 2, 7 (2022).
- Source code available at: [https://github.com/njiang8/Create\\_Synthetic\\_Population](https://github.com/njiang8/Create_Synthetic_Population)



# Approach: Synthetic Population

The synthetic home and work locations are sampled from road network based on road types:

- **home locations** are sampled at 50m intervals on local neighborhood road, rural road, city street, and private road for service vehicles
- **work locations** are sampled at 20m intervals on secondary roads



# Approach: Agent-Based Model

Model is developed using **Mesa** & **Mesa-Geo** in Python, both of which are open-source free software, with source code publicly available at <https://github.com/projectmesa>

- The only Python ABM framework with GIS support
- Easy integration with the Python ecosystem
- Expanding community

**Come and join us!**

- NumFOCUS Small Development Grant
- Google Summer of Code



**SCAN ME**

# Approach: Agent-Based Model

## GeoSpace

- Contains a road network layer, constructed from TIGER/Line shapefiles of roads in the study area.
- Implemented by two underlying data structures:
  - a topological network, for shortest path calculation
  - a k-d tree, for nearest node queries

## GeoAgents

- The commuters are the GeoAgents in the model, initialized with the synthetic population created earlier.
- Each agent has a unique id, home and work locations, randomly sampled on road networks.
- Every agent has a random start time to go to work (between 6am to 9am). After 8 hours, agents commute back home from workplaces.
- All agents travel at the same speed of 12km/h (subject to further calibration and validation).

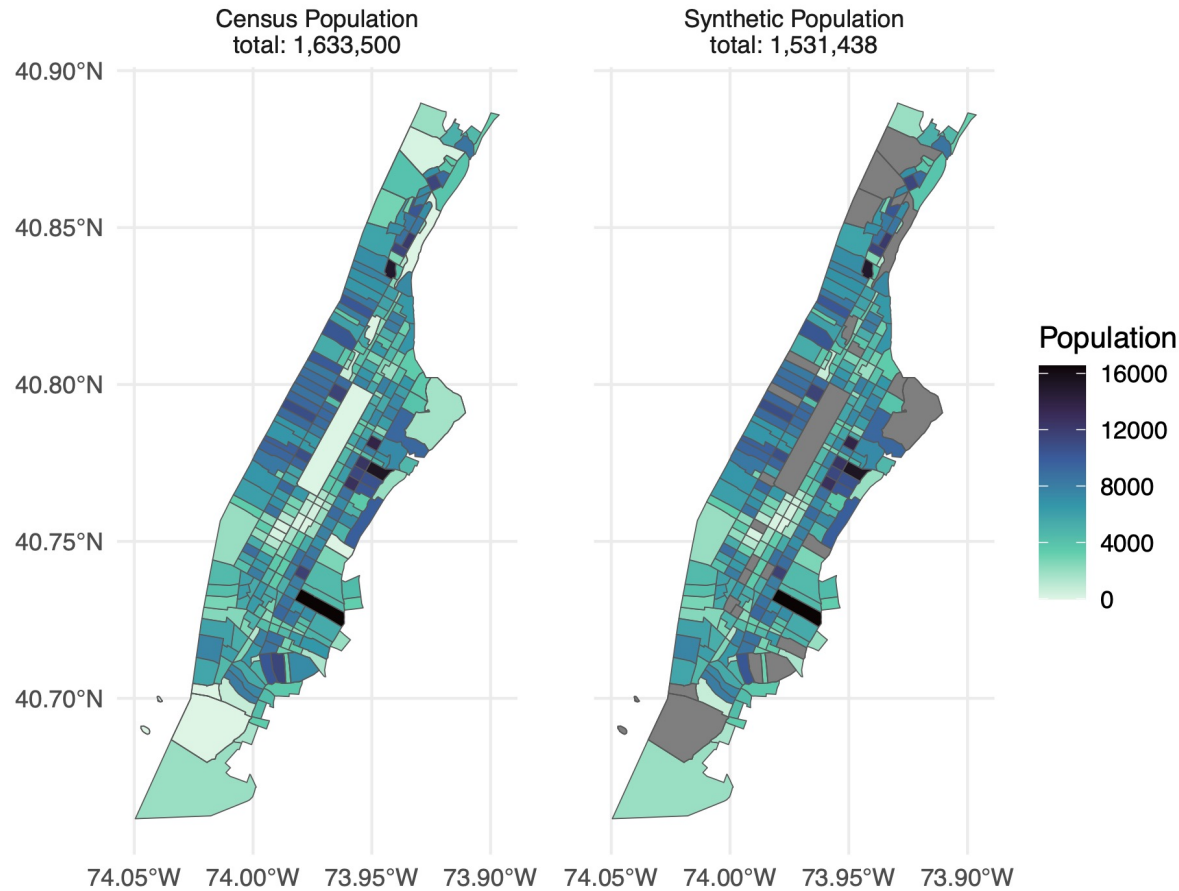
## Results: Synthetic Population

- The total population from census data is 1,633,500. The synthetic population contains a total number of 1,531,438 individuals, which is 6.25% less than true population size.
- The amount of people who both live and work inside the study area is **443,893**. This is the number of agents that are included in the commuting model (ideally).



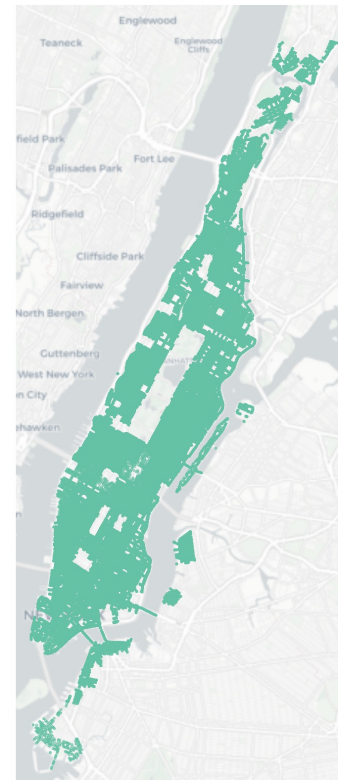


# Results: Synthetic Population



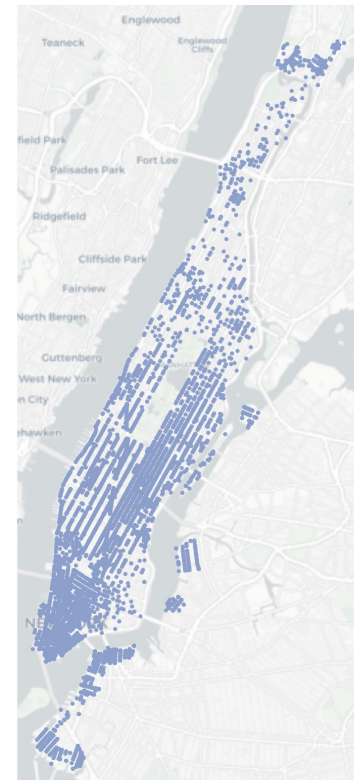
Synthetic population by census tracts in the study area.

**Synthetic Home Locations**  
 total: 289,648



(C) OpenStreetMap contributors (C) CARTO

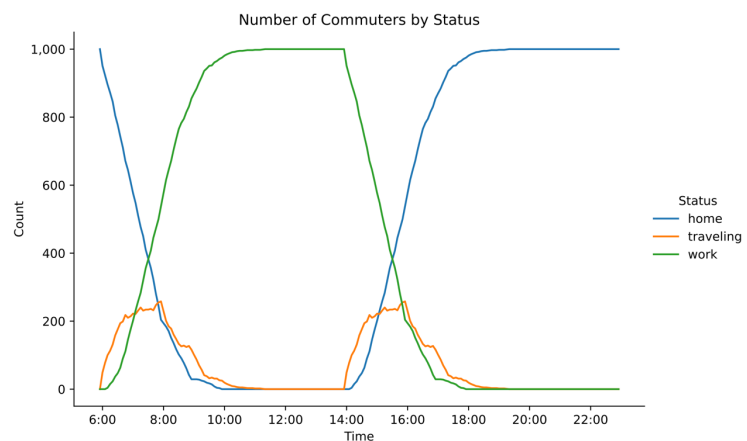
**Synthetic Work Locations**  
 total: 17,800



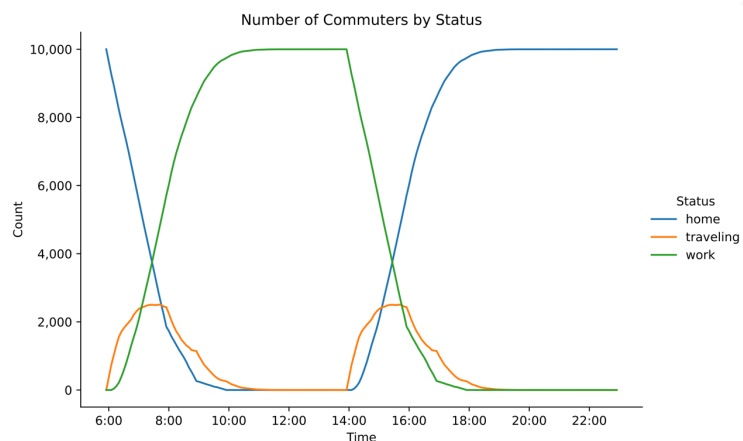
(C) OpenStreetMap contributors (C) CARTO

Synthetic home and work locations, sampled on road network based on road types.

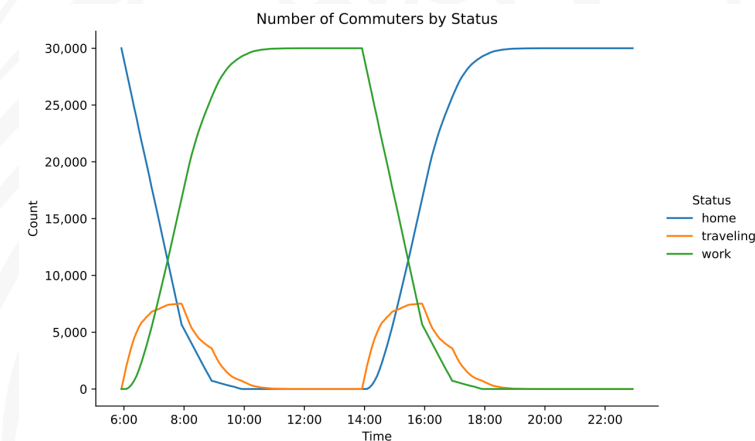
# Results: Agent-Based Model



(a)



(b)

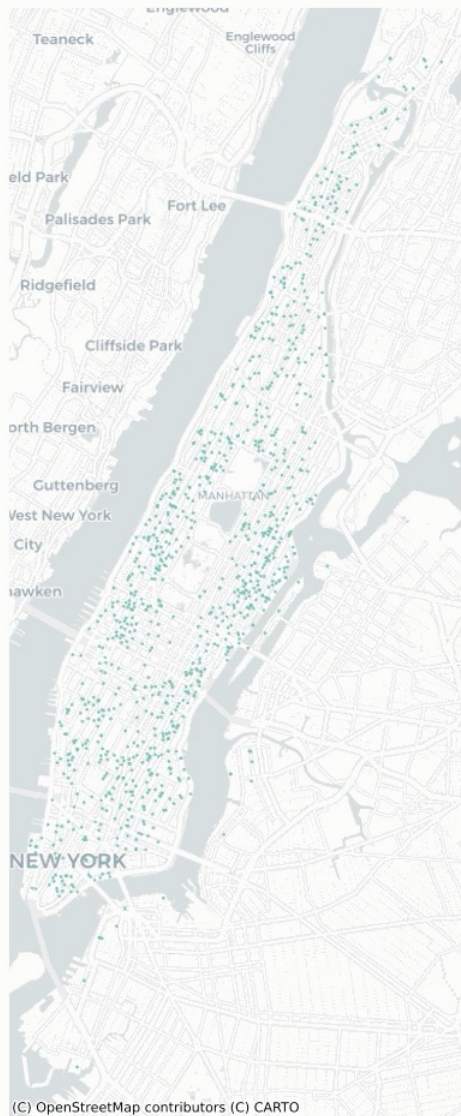


(c)

Number of commuters by status when total number of commuters is (a) 1,000, (b) 10,000, and (c) 30,000.

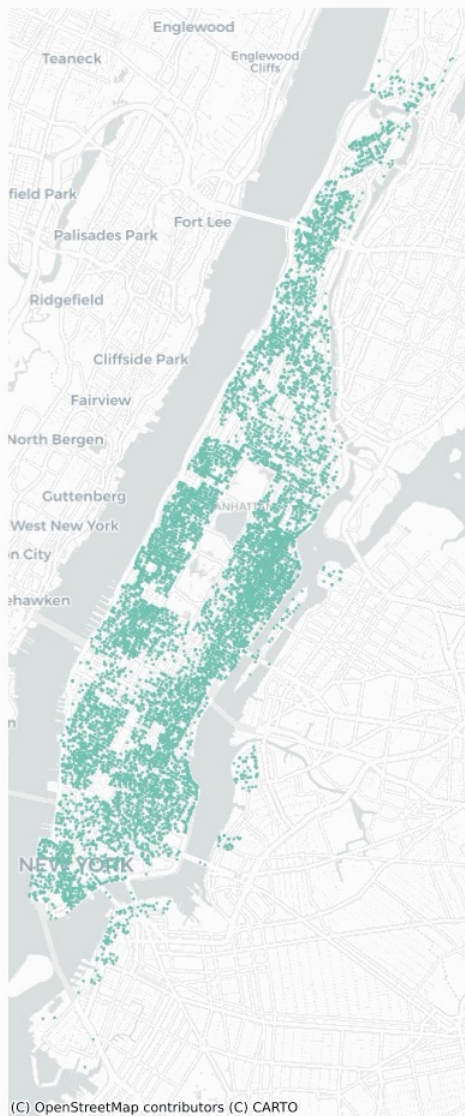
Number of Commuters: 1,000

Time: 05:55



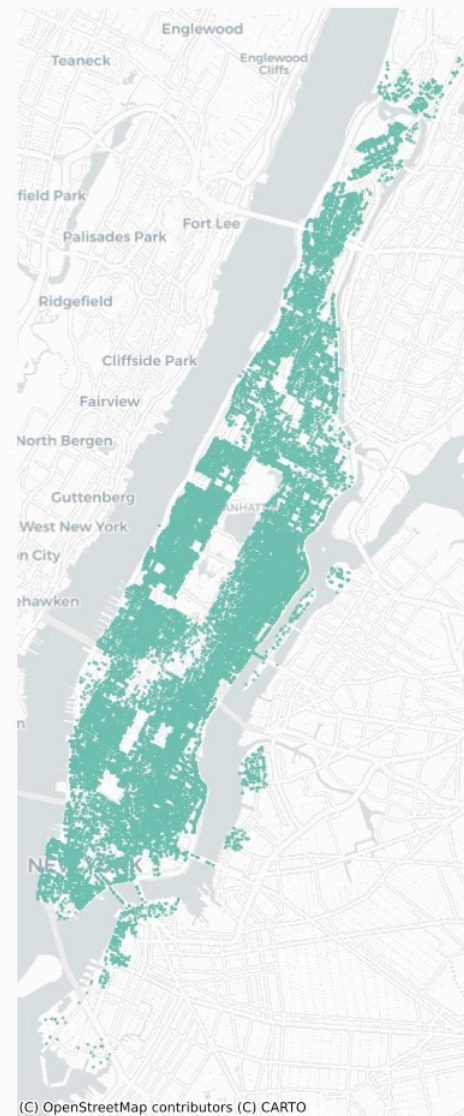
Number of Commuters: 10,000

Time: 05:55



Number of Commuters: 30,000

Time: 05:55

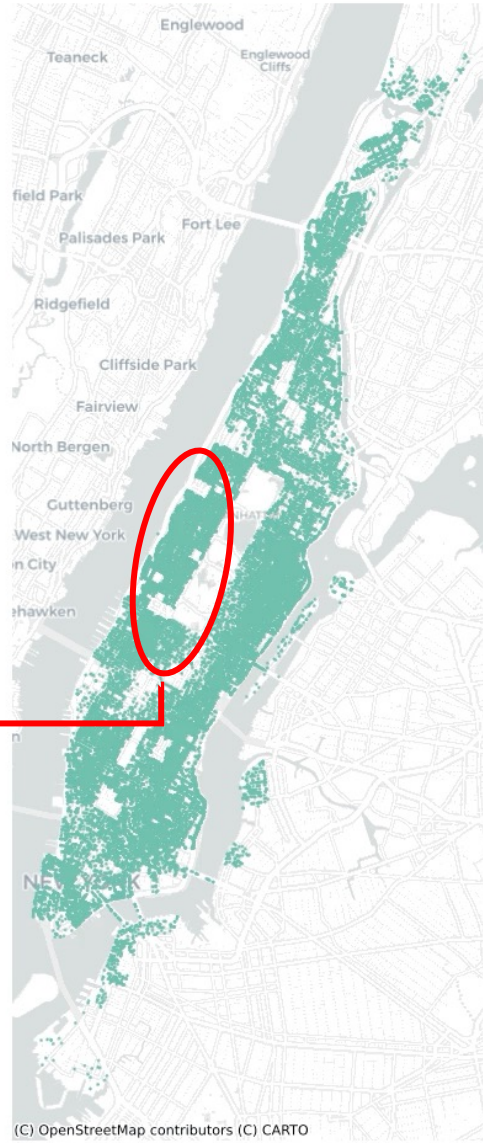


# Emergence

Number of Commuters: 30,000

Time: 05:55

Broadway



Number of Commuters: 62,079

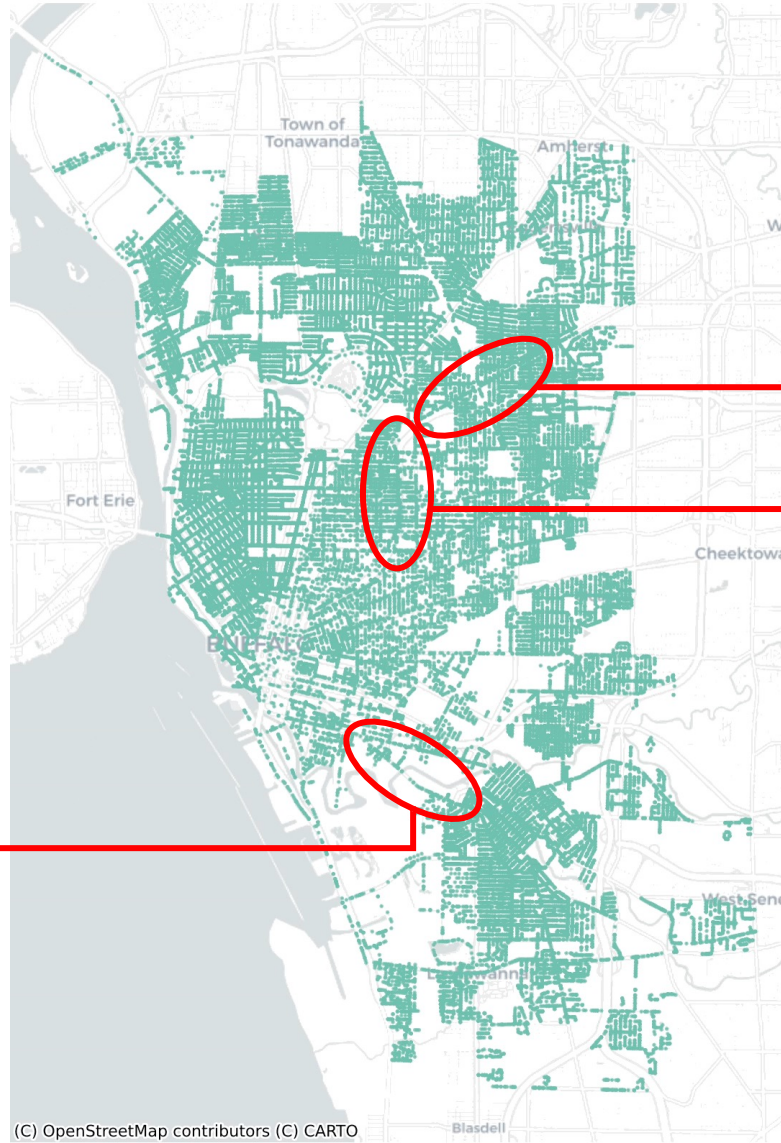
Time: 05:55

# Emergence

“**South Park Avenue** is the most important street in [South Buffalo](#) and is the primary alternative to the Buffalo Skyway ([New York State Route 5](#)) when traveling between the city and the southern suburbs.”

[https://en.wikipedia.org/wiki/South\\_Park\\_Avenue](https://en.wikipedia.org/wiki/South_Park_Avenue)

South Park Avenue



Kensington Avenue

Kensington Expressway

# Improving the Agent-Based Model

Calibrate model parameters (e.g., commuter speed) and validate results with travel surveys

- Average travel time, number of trips, etc.

Expand the geographic boundary

- To include more agents whose home and workplaces are outside the current study area

Expand the types of activities

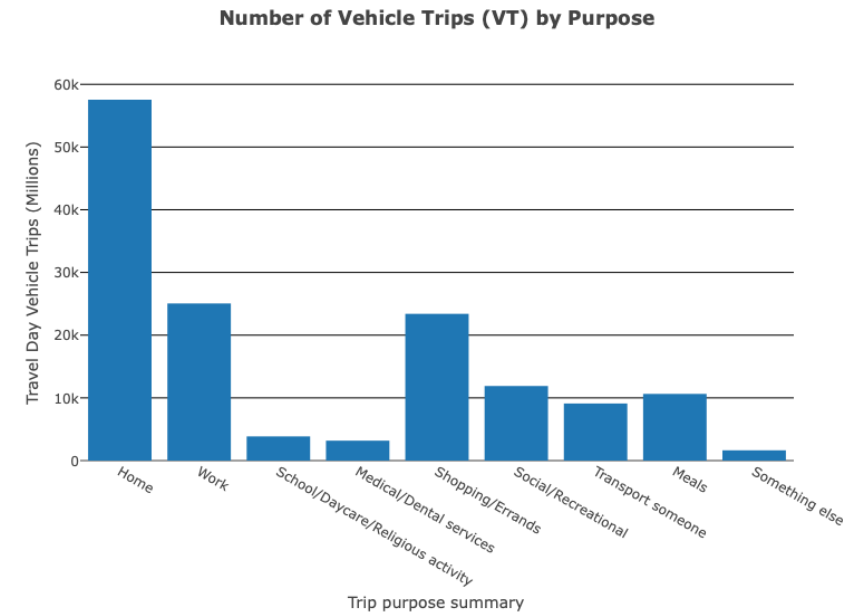
- To include travels between home, workplaces, shopping/errands places, schools, and recreational centers, etc.

Have heterogeneous agents

- People of different characteristics (e.g., age, gender) may behave differently

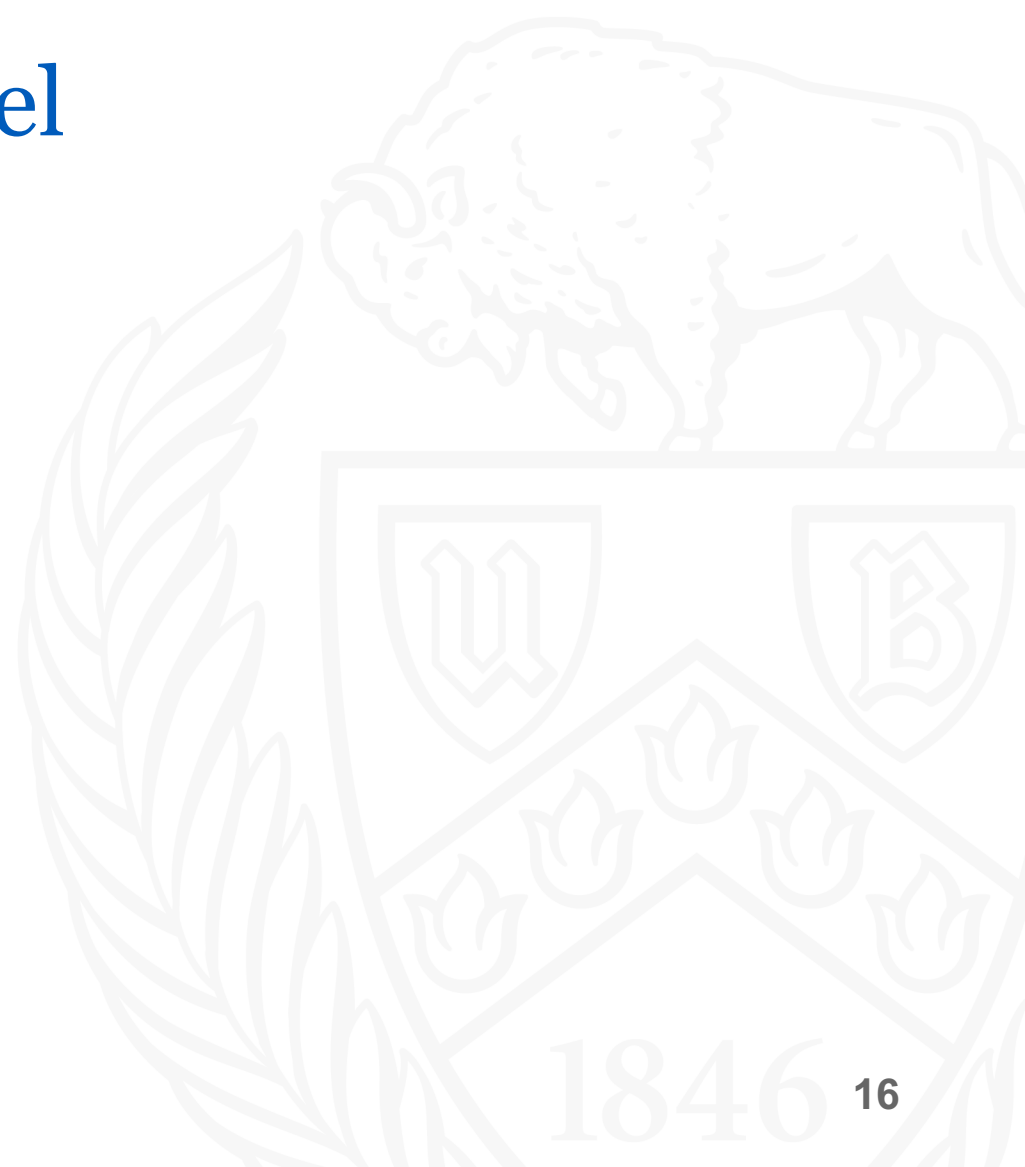
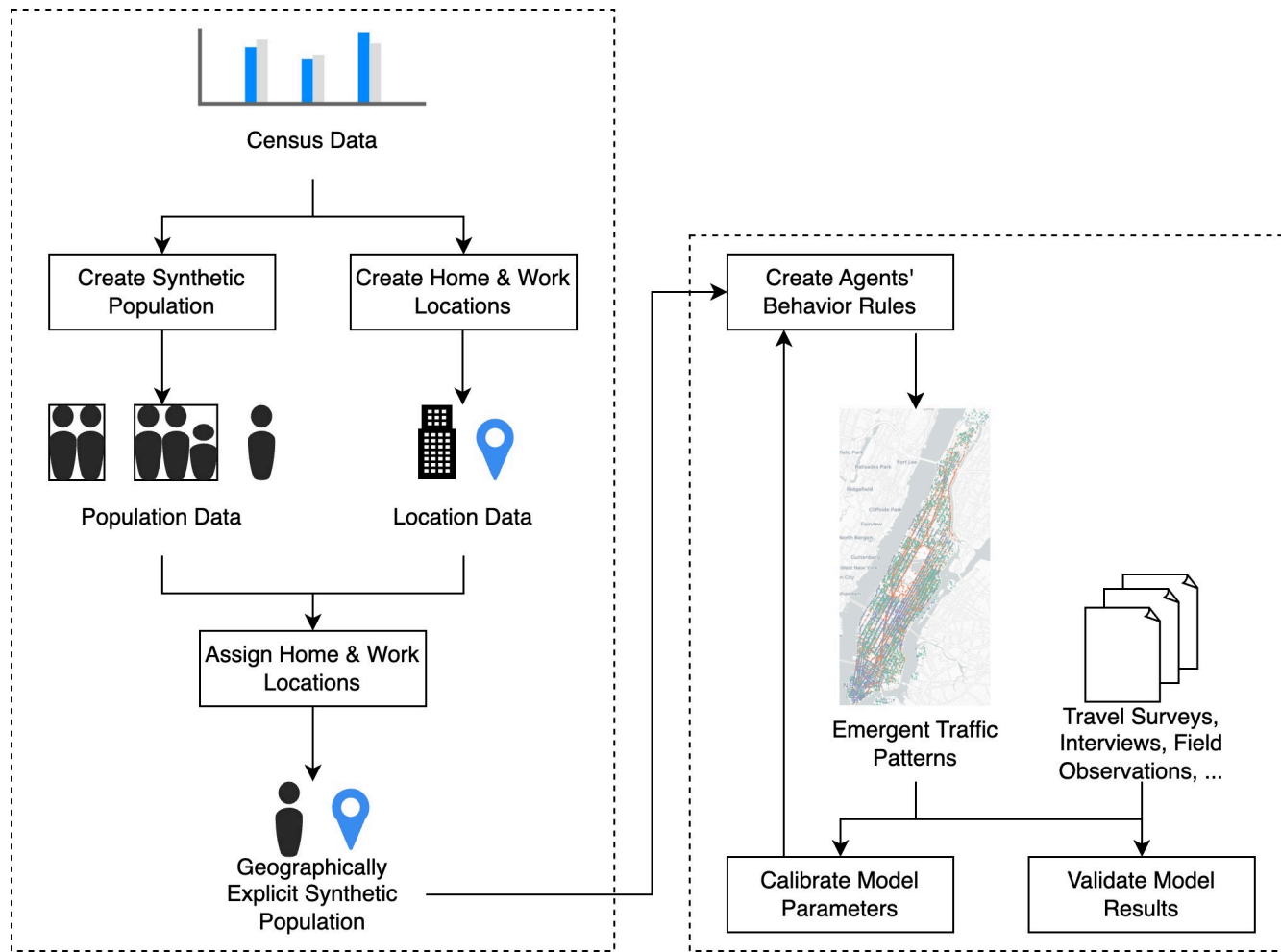
Include social behaviors (i.e., coordinated movements)

- Parents send kids to schools
- Friends go to the same restaurant from different home locations

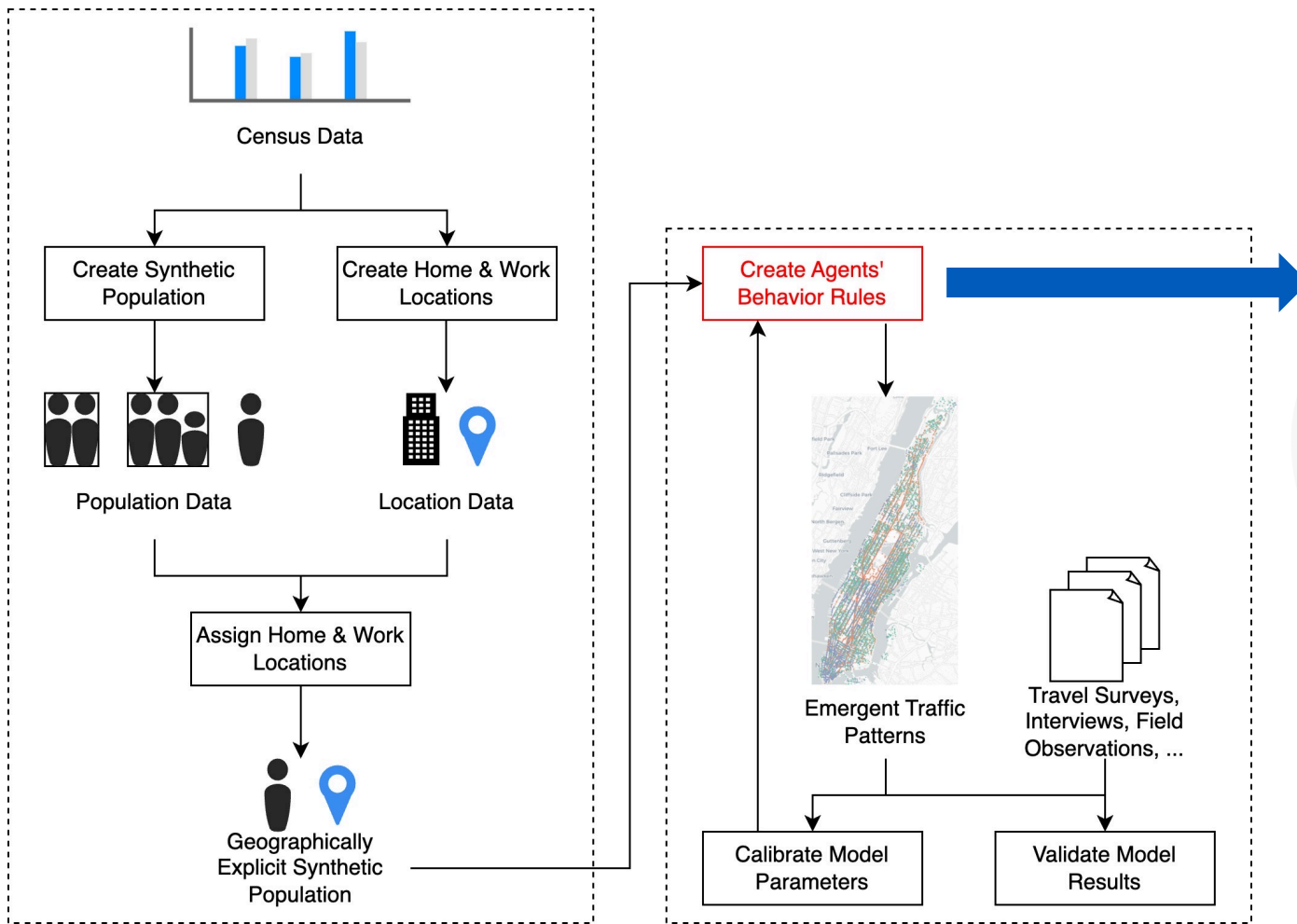


Source: Popular Vehicle Trips Statistics  
<https://nhts.ornl.gov/vehicle-trips>

# Improving the Agent-Based Model



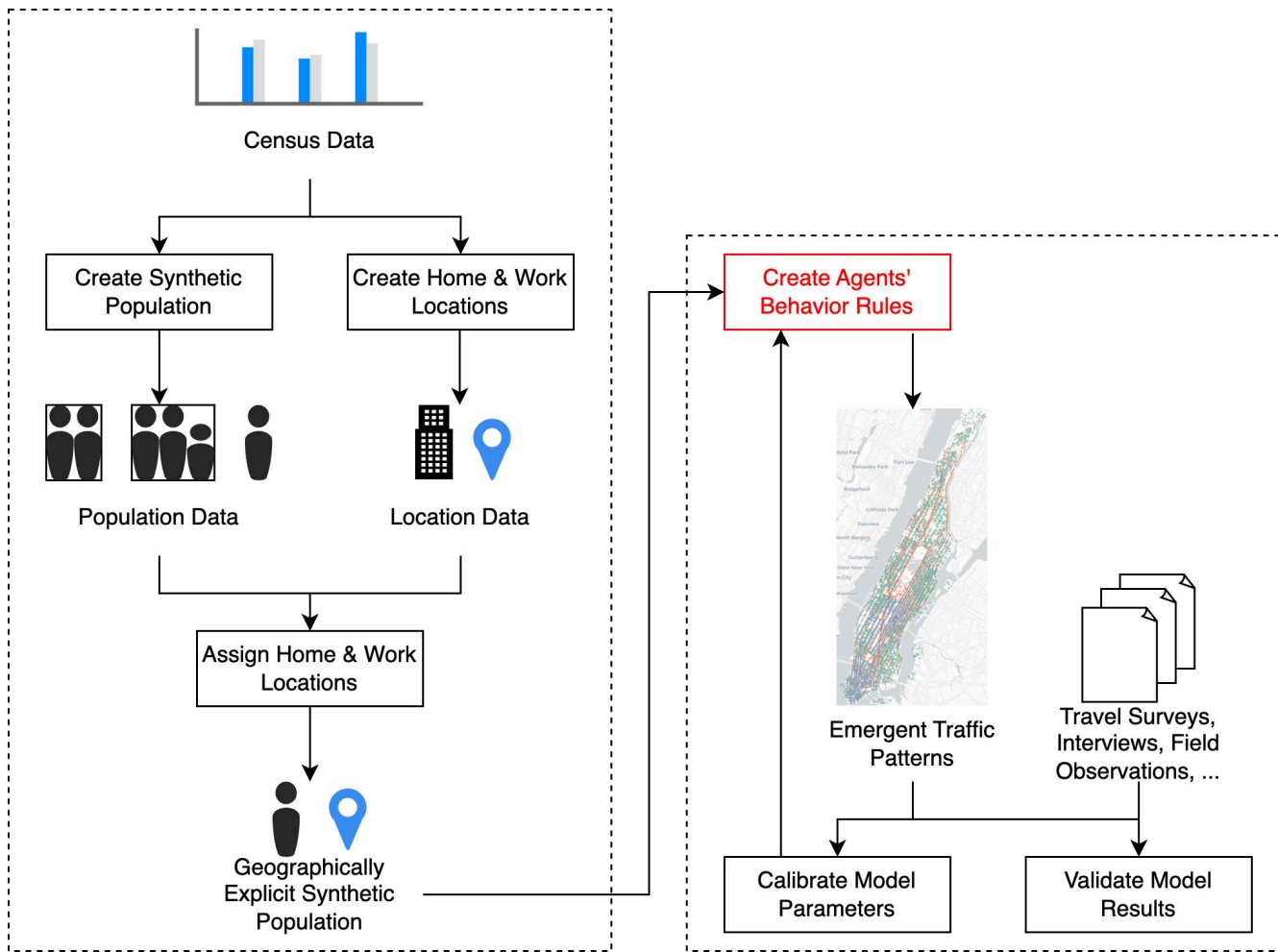
# Improving the Agent-Based Model



- Go to work in the morning by shortest path between home and work locations.
- Travel at the same speed of 12km/h (adjustable).
- Stay at work for 8 hours.
- Travel back home by the same shortest path in the morning.



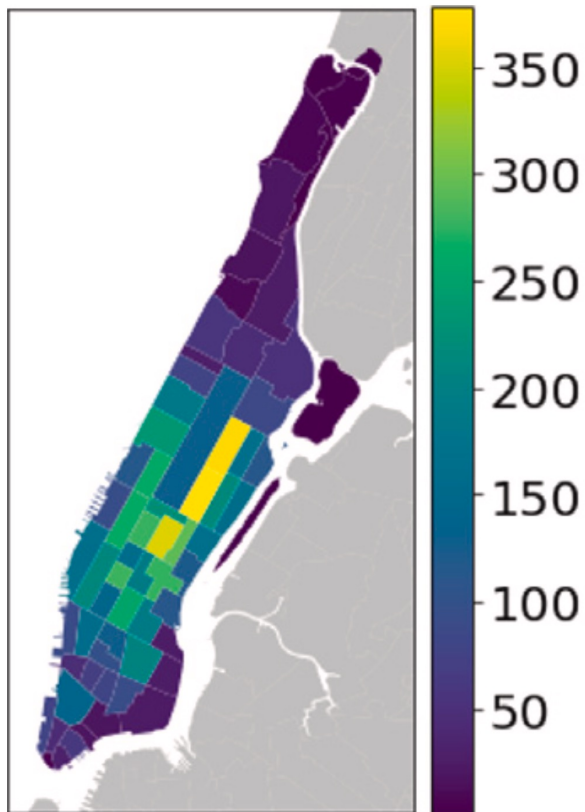
# Improving the Agent-Based Model



A more data-driven approach?

# Improving the Agent-Based Model

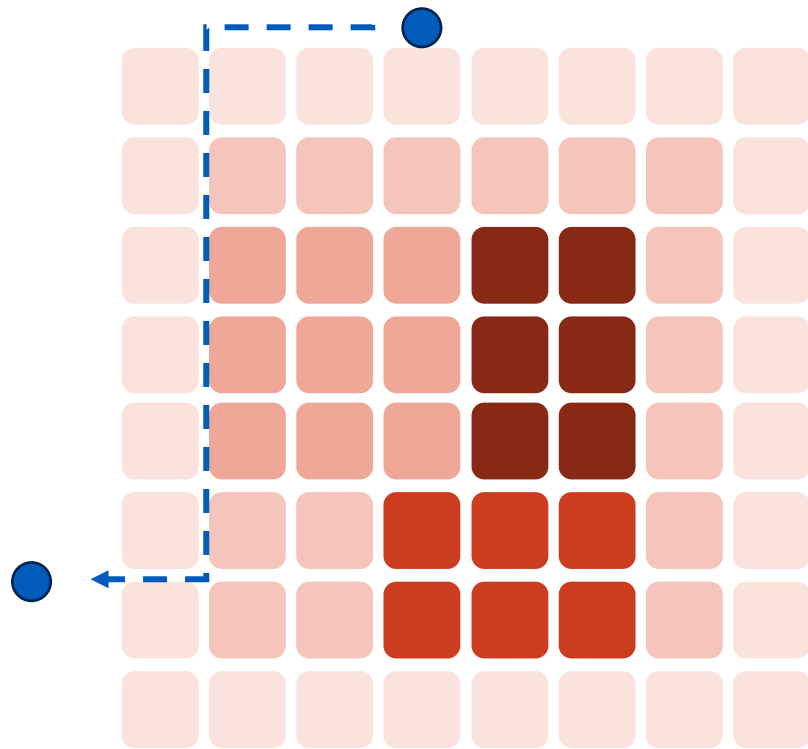
Taxi traffic as a proxy for urban human mobility



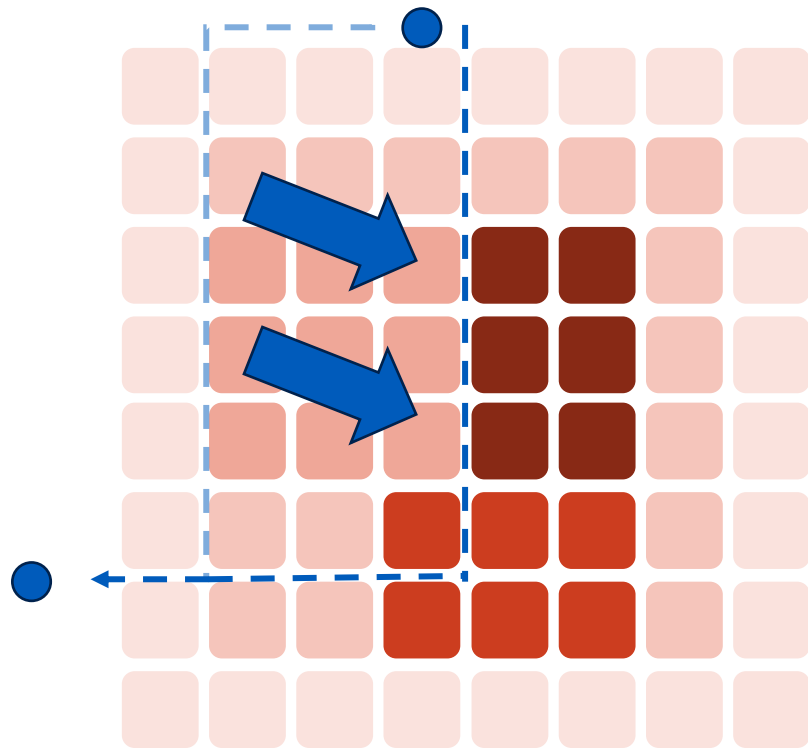
Taxi zones of Manhattan color-coded based on the average hourly activity from NYC Taxi data

Source: Tygesen, M. N., Pereira, F. C., & Rodrigues, F. (2023). Unboxing the graph: Towards interpretable graph neural networks for transport prediction through neural relational inference. *Transportation research part C: emerging technologies*, 146, 103946.

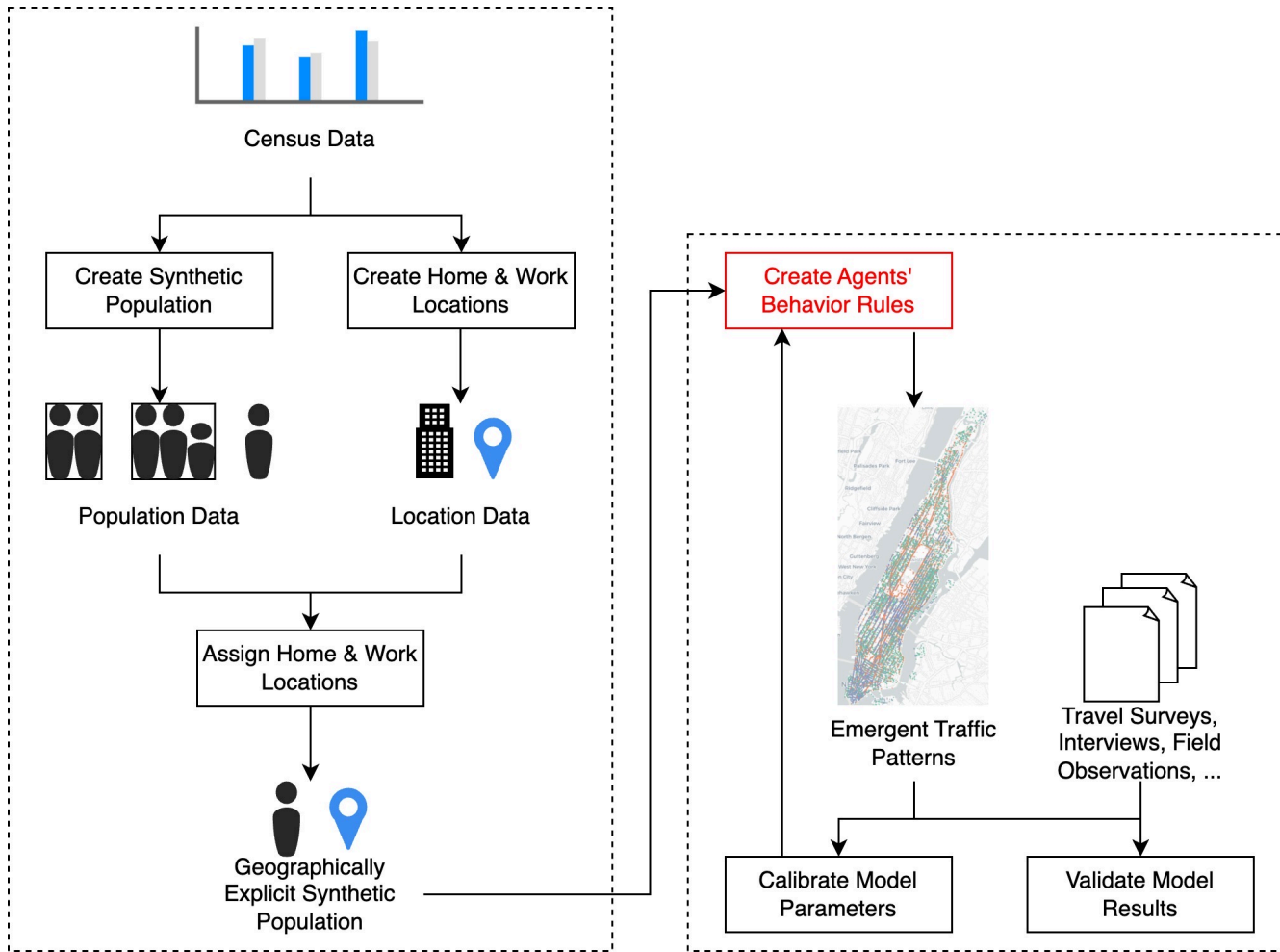
# Most Probable Commute Path



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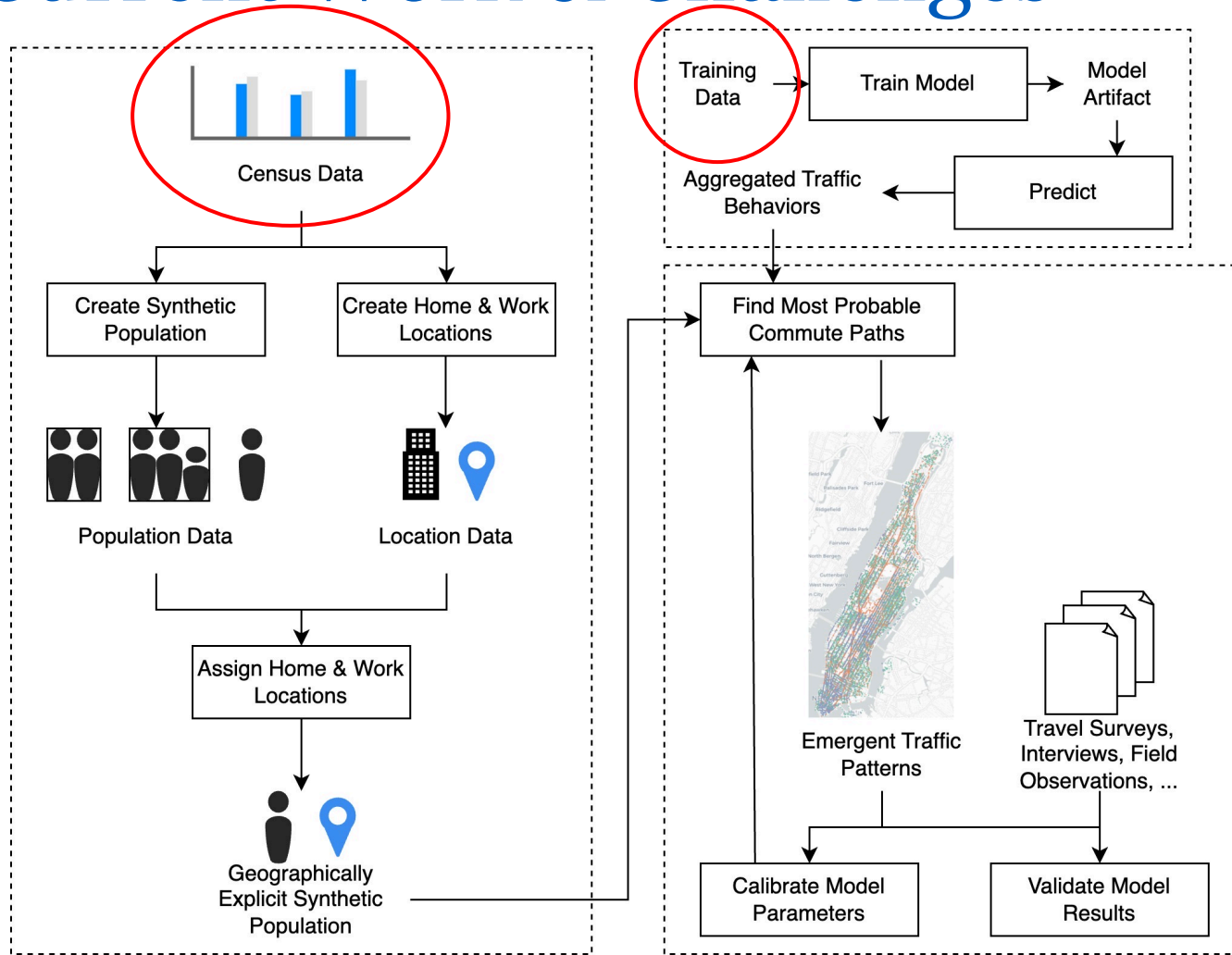


# Improving the Agent-Based Model

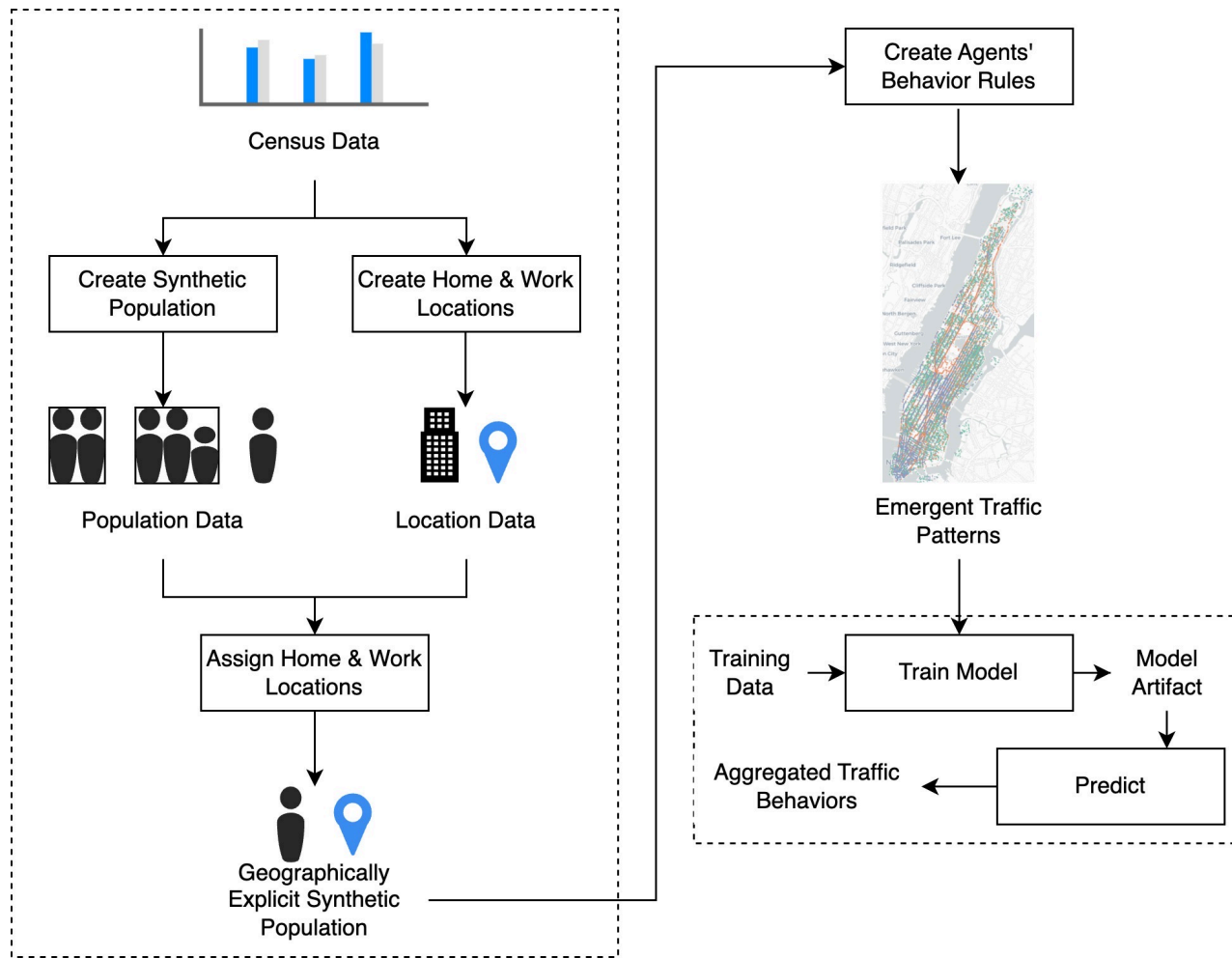


A more data-driven approach?

# Current Work & Challenges



# On A Side Note...



# Conclusion

- An agent-based model of urban commuting in Manhattan is developed to simulate people's daily commute patterns, based on census survey data.
- Emerging traffic patterns can be observed through agents' simple behaviors.
- Such modeling technique can be complemented by recent advancements in GeoAI techniques.
- The model can be used as a basis to answer interesting “what-if” questions.
  - How many more minutes do people have to commute if certain roads are closed for maintenance?
  - Imagine a  $x\%$  adoption rate of autonomous vehicles. Does it worth it to have one or more dedicated lanes for them?



**Thank you for listening!**  
**Welcome comments,**  
**questions and suggestions.**



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