

# Modeling relocation dynamics under shifting spatial conditions for adaptive post-disaster recovery planning



Hiroaki GOTO, Kaori ISAWA, Saki YOTSUI, Kensuke OTSUYAMA, U HIROI (University of Tokyo)

Case Study of Kumamoto EQ, 2016

## Introduction | Need for Dynamic Models

### Time-Limited Decision Making

Disaster recovery requires rapid choices under strict time constraints[1]

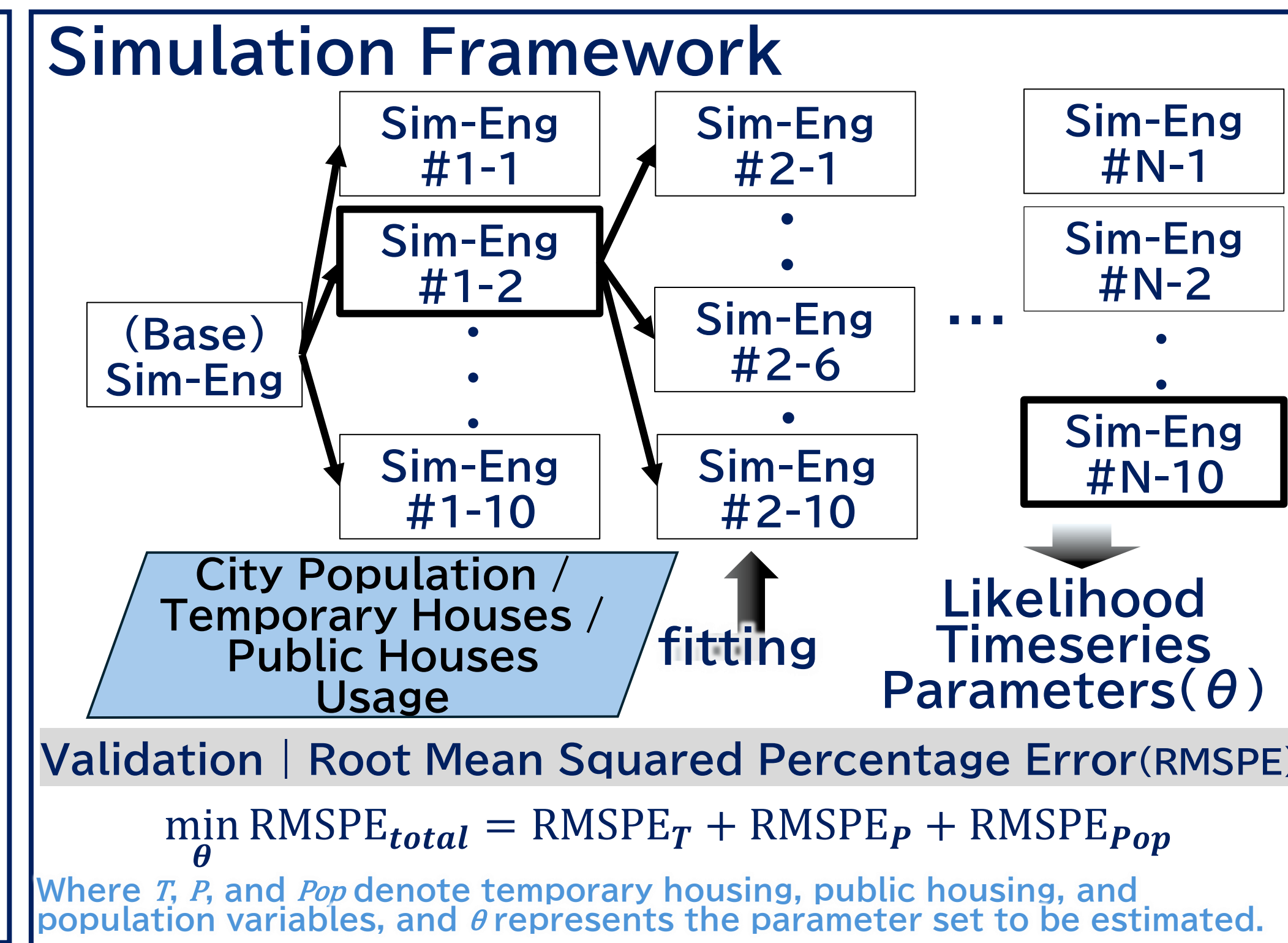
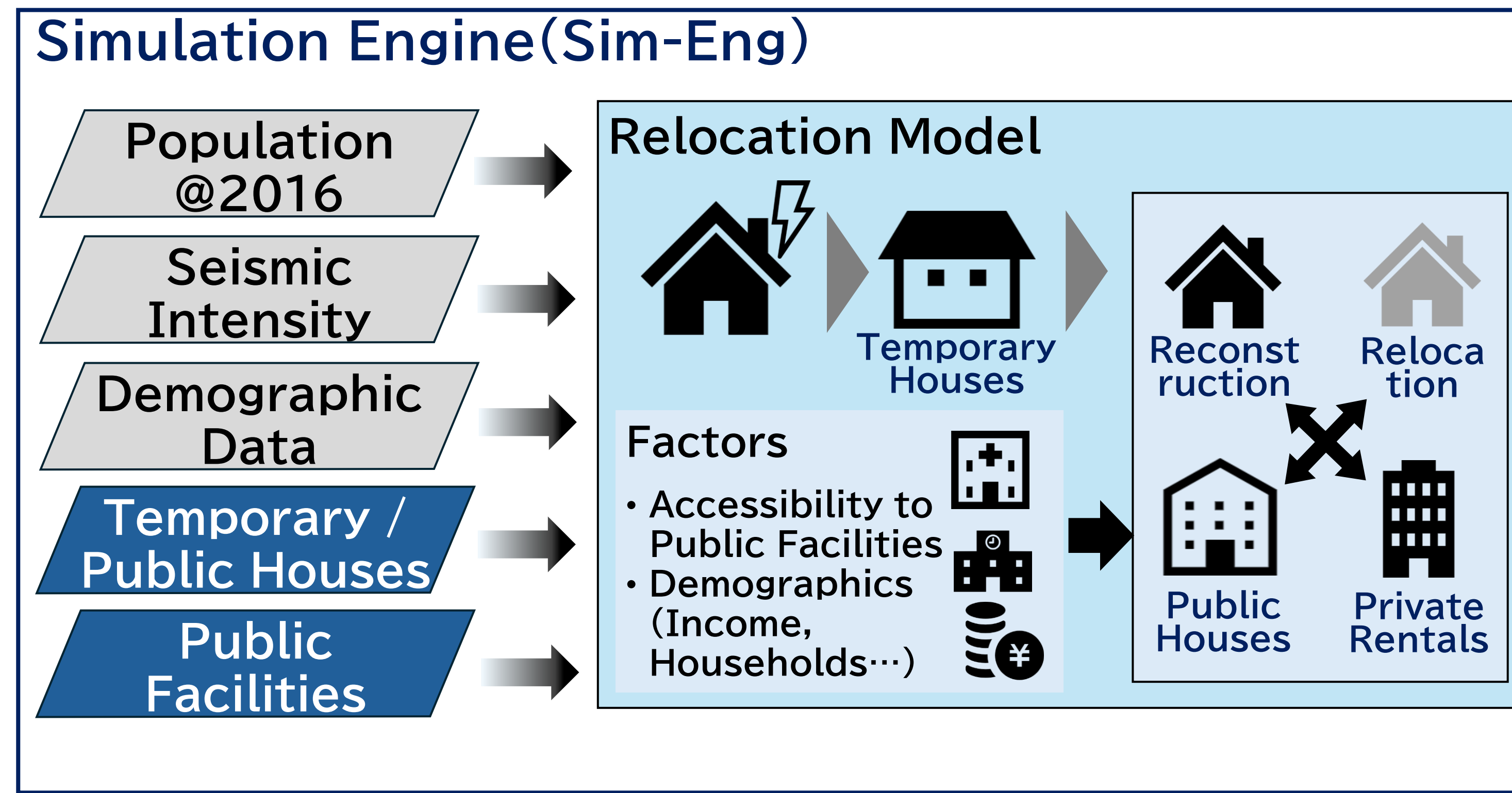
### Dynamic Spatial Change

Living environments shift throughout the recovery phases:



- While various recovery models have been developed (e.g. [2][3][4]), **spatio-temporal dynamics** were overlooked
- lack of models that integrate evolving spatial conditions over time into long-term recovery simulations
- To clarify how **evolving spatial conditions impact residential choice behavior** through a Dynamic Agent-Based Model (ABM).

## Research Method | Agent-Based Modeling with Spatial-Shifts

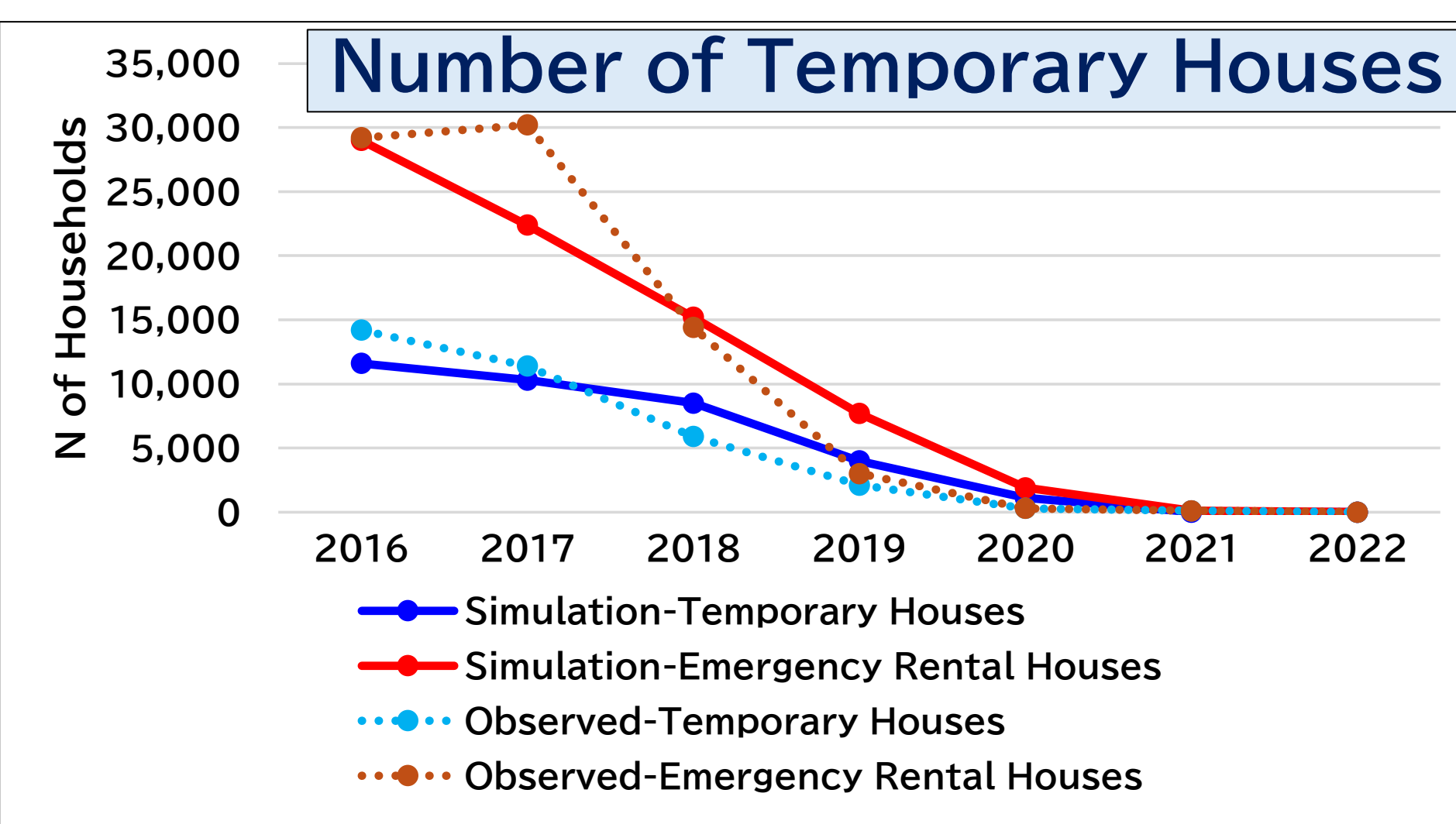
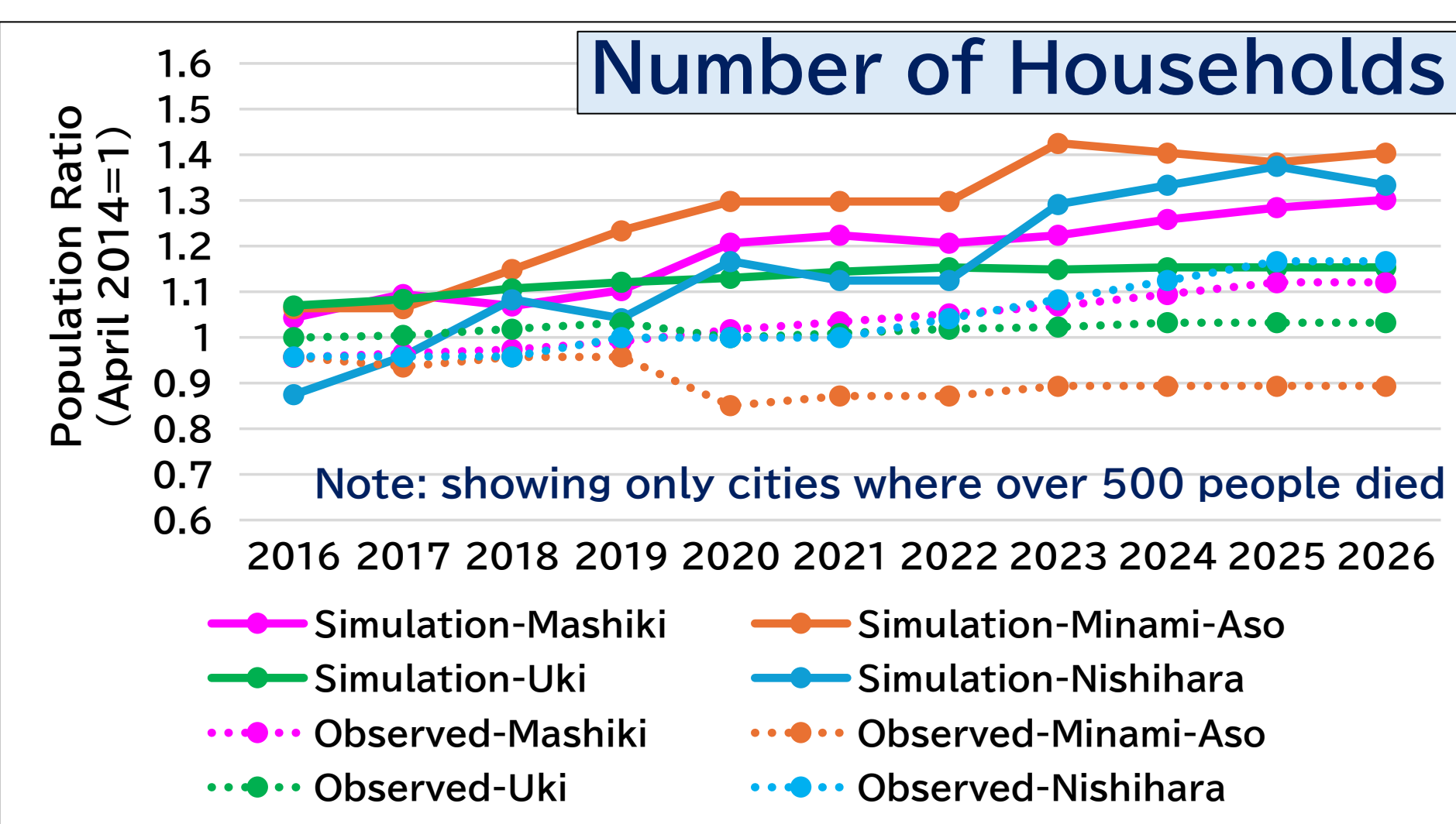


### Utility Calculation

$$U_{t,a,c} = \sum_k \beta_{t,k} x_k + \sum_k \beta_{a,k} x_k + \sum_f \beta_{a,acc} z_{f,c}^{acc}$$

Symbols	Explanation
$t, a, c$	Target household type, area, city (area: in-city, in-prefecture, out)
$k$	Attributes (e.g., age, income, damage level)
$f$	Public facility types
$acc$	Accessibility Type (Proximity, N of facilities)
$x_k$	Attribute dummy variables
$z_{f,c}^{acc}$	Accessibility to facilities $f$ in City $c$
$\beta$	Estimated coefficients

## Results | Key Role of Accessibility



### Accuracy

RMSE (Root Mean Squared Error): Smaller, More Accurate

Index	With Facilities	Without Facilities
RMSE (Number of Households)	14.62	14.67
RMSE (Composition Rate of Households : each city in the Prefecture)	0.28	0.33
RMSE (Number of Temporary Houses)	17.79	15.99

- **General Trends**  
Effectively captured for both population and temporary housing
- **Facility Impact**  
Improved population simulation accuracy, but had little effect on temporary housing due to strict siting constraints
- **Scale Issue**  
Smaller municipalities showed slight deviation from observed data due to high sensitivity

## Conclusion

- **Valuable Approach:**  
Facility accessibility enhances dynamic simulation, though methods should be improved
- **Framework:**  
Establishes a foundation for sequential, dynamic recovery modeling
- **Next Step:**  
Introduce what-if analyses to balance facility scaling with municipal financial capacity

## Applications | Flexible Recovery

- **Policy & Governance**
  - **Location Matters:**  
Careful placement of temporary and public housing is essential to ensure accessibility
  - **Infrastructure:**  
Adjusting public facility layouts is a core element of resilient recovery
  - **Finance:**  
Maintaining public facilities requires municipal financial resources
- **Modeling Strength**
  - **Dynamic Framework:**  
Transitioning from static plans to adaptive, "Living System"
  - **Social Continuity:**  
Tools to evaluate the maintenance of social networks through spatial intervention
- **Beyond Recovery**
  - **Cross-Domain Use:**  
Applicable to climate adaptation and flood risk management
  - **Dual-Use Planning:**  
Integrating recovery strategies into daily urban development

## References

- [1] Olshansky et al. (2012), Disaster and Recovery: Processes Compressed in Time, Natural Hazards Review, Vol.13 No.3
- [2] Moradi and Nejat (2020), RecovUS: An Agent-Based Model of Post-Disaster Household Recovery, Journal of Artificial Societies and Social Simulation, 23(4) 13
- [3] Miles and Chang (2006), Modeling Community Recovery from Earthquakes, Earthquake Spectra, Vol.22 No.2
- [4] Park et al. (2024), Post-disaster recovery policy assessment of urban socio-physical systems, Computers, Environment and Urban Systems, 114

## Acknowledgements

• This research was the result of the joint research with CSIS, the University of Tokyo (No. 1518)  
• The authors would like to thank all those who provided support and guidance during this research.

## Contact Me:

Hiroaki GOTO  
(hiroaki-goto@g.ecc.u-tokyo.ac.jp)

