

A “Scope, Simulation and Story” approach to identify future spatial development scenarios

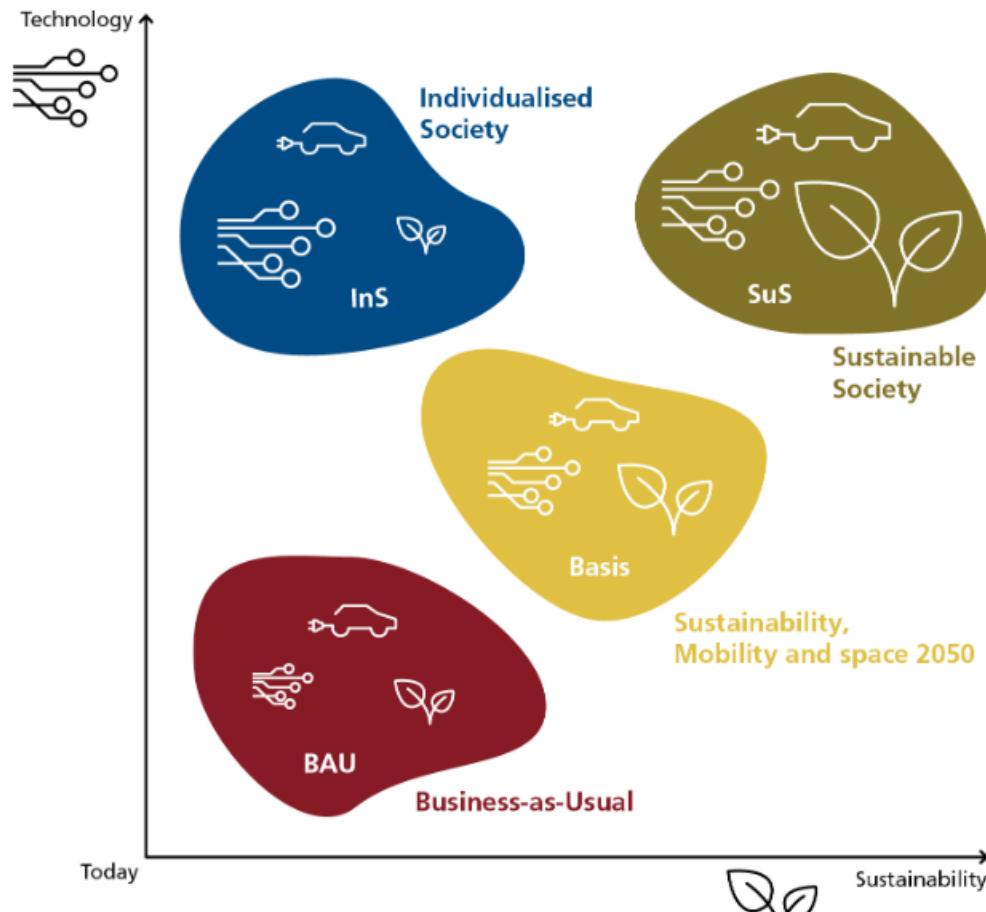
Orlando Roman ¹, Jan Kwakkel ², Bryan T. Adey ¹ and Benjamin Black ¹

¹Department of Civil, Environmental and Geomatic Engineering. ETH Zürich

²Faculty of Technology Policy and Management, TU Delft

Problem Statement

Scenarios

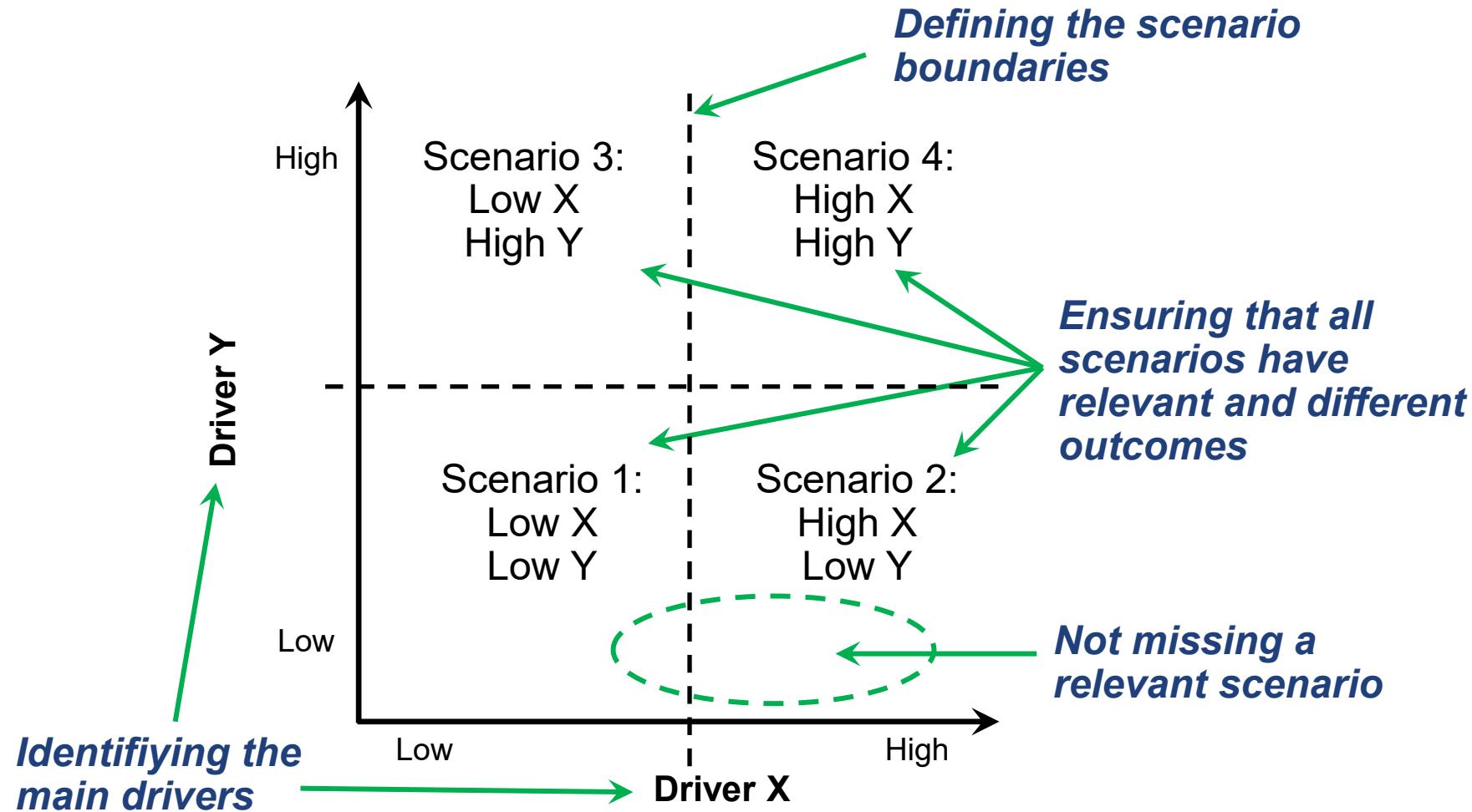


ARE's Transport Outlook 2050 Scenarios:
<https://www.are.admin.ch/are/en/home/mobility/data/transport-outlook/scenarios-methodology.html#1497088217>



The Global Shared Socioeconomic Pathways (SSPs)

Problem Statement – What is challenging



Research Gap

Most scenario-planning exercises remain predominantly qualitative.

***Limited set of
narrative storylines***



***Simulation
models***

“Story and Simulation” approach

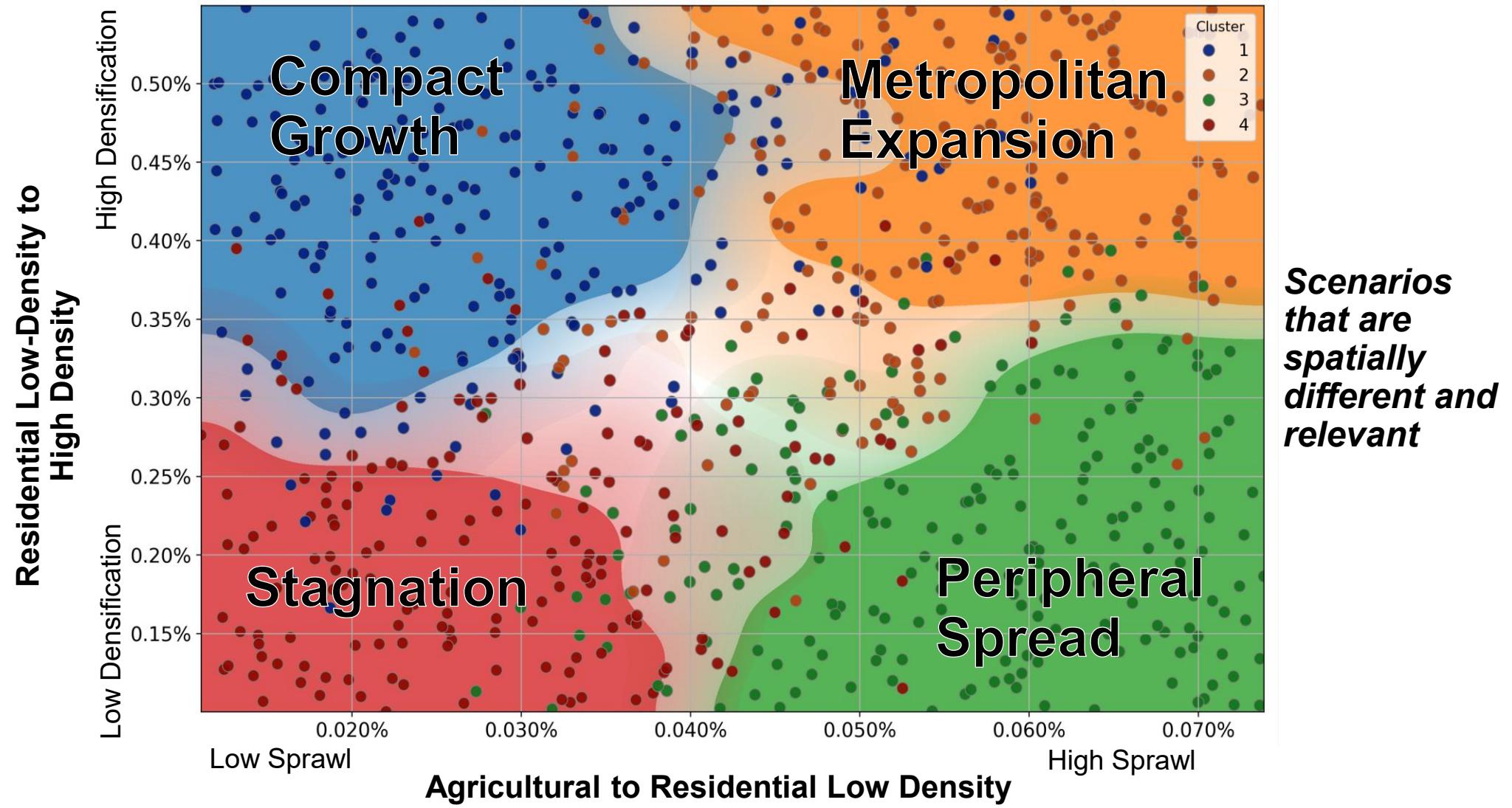
Approaches that invert this workflow, leveraging large ensembles of simulations upfront to quantitatively inform and refine the creation of scenario storylines, are still relatively uncommon.

Proposal

Probabilistic boundaries between scenarios

**Y-axis explain
41% of change**

**X-axis explain
14% of change**



A Scope, Simulation and Story approach

1: Scope

- Spatial and temporal constraints
- Potential drivers of change: Uncertainties and Planning Decisions
- Desirable Futures

2: Simulation

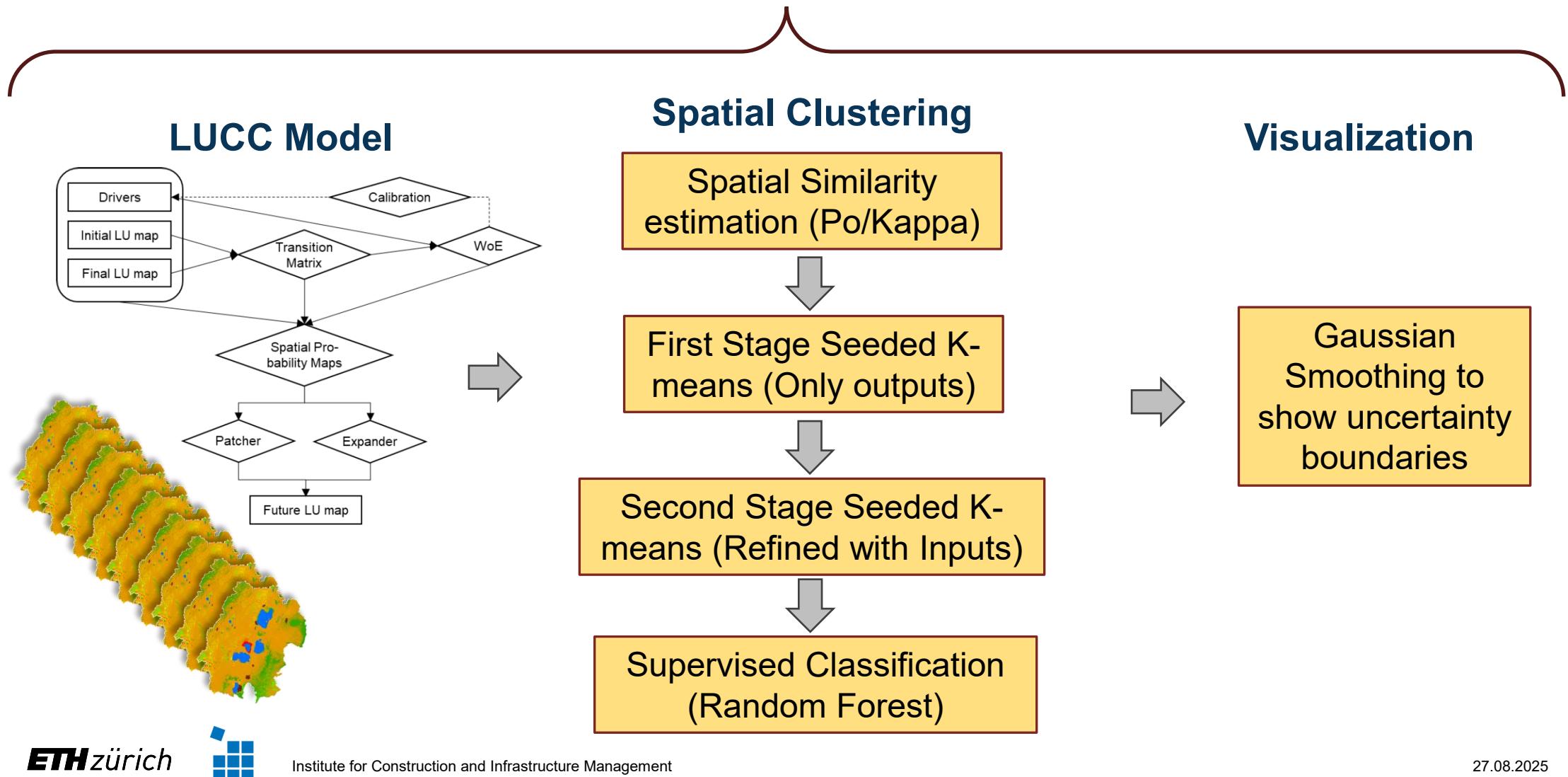
- Data Gathering
- Urban Development Model Selection
- Model calibration and validation
- Simulation of Ensemble of Scenarios
- Spatial clustering
- Visualisation

3: Story

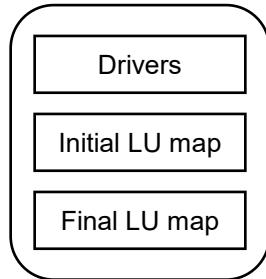
- Iteration with stakeholders
- Scenario refinement
- Storyline development
- Proposal for action plans

A Scope, Simulation and Story approach

2: Simulation

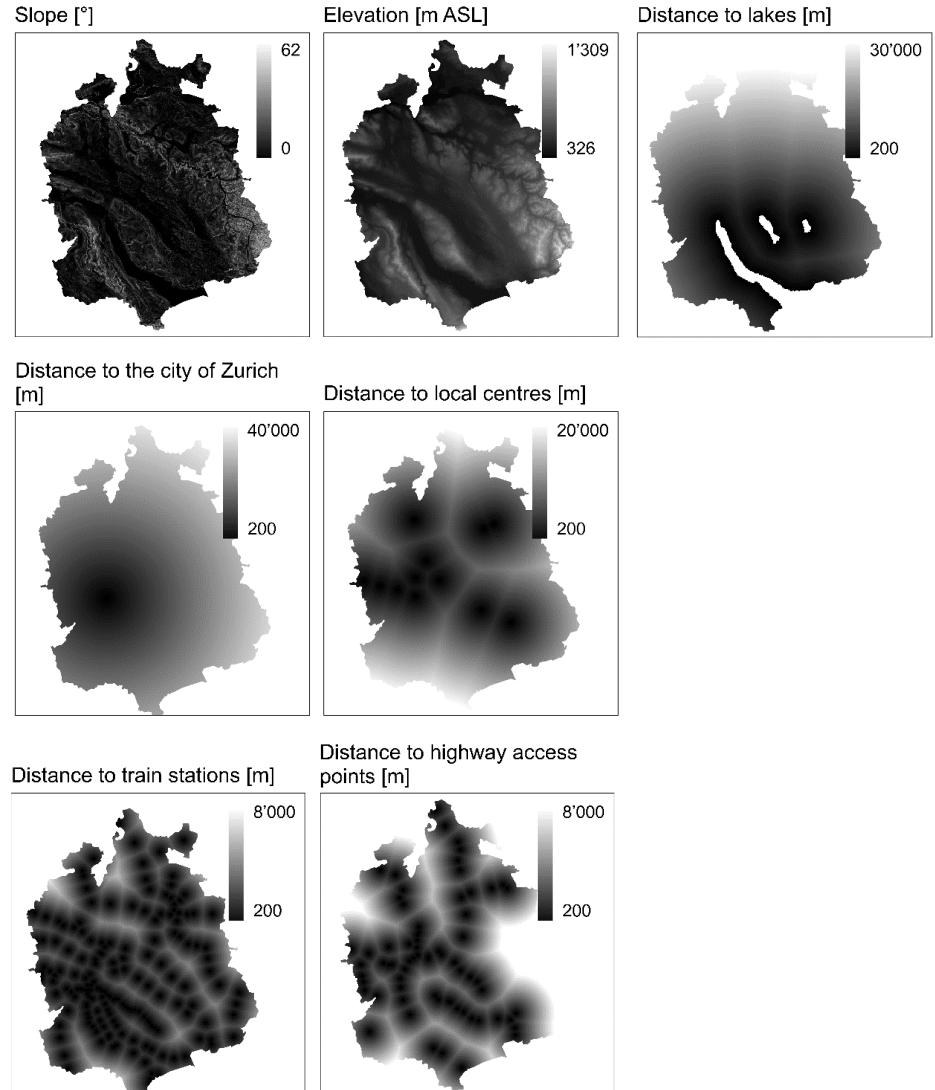
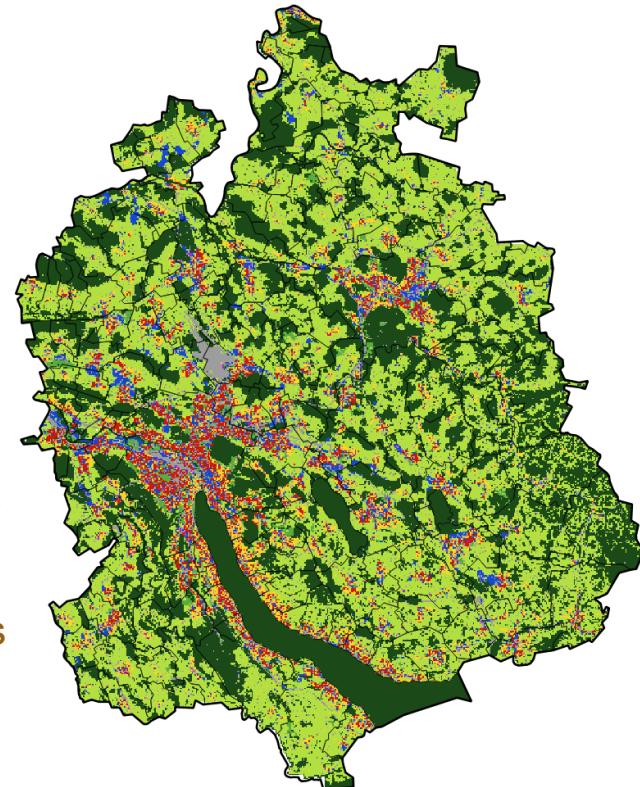


LUCC Model – Dinamica EGO

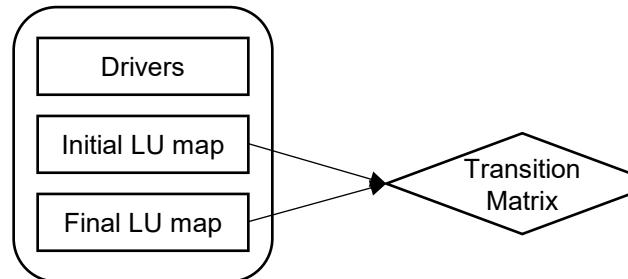


1. Low-density settlement area
2. High-density settlement area
3. Industrial, commercial, and public area
4. Construction sites and fallows
5. Traffic area
6. Recreational and green zones
7. Agricultural area
8. Forest and unproductive area

1985, 1997, 2009, 2018



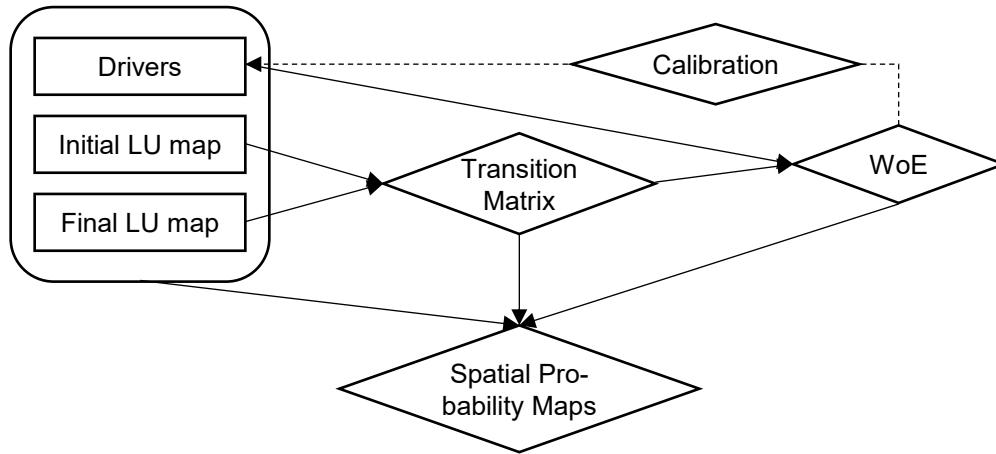
LUCC model – Transition Matrix



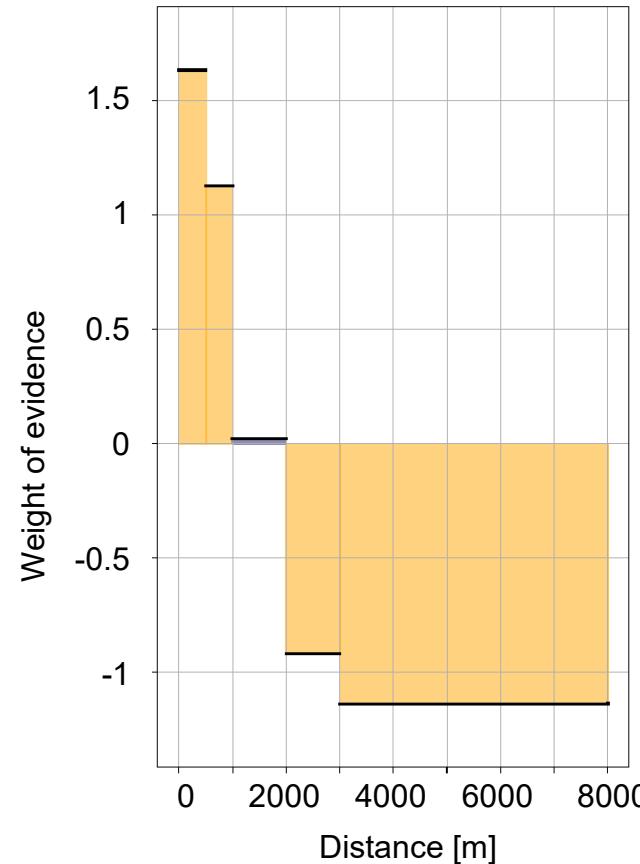
«A pixel of low-density settlement in 2009 will change to high-density settlement by 2018 with a probability of 0.35%»

From:	To:	1 LDS	2 HDS	3 ICP	4 CF	5 T	6 RG	7 A	8 FU
1: Low-density settlement area (LDS)		-	0.35%		0.26%				
2: High-density settlement area (HDS)		0.14%	-		0.20%				
3: Industrial, commercial, and public area (ICP)		0.05%	0.12%	-	0.62%	0.03%	0.02%	0.24%	0.14%
4: Construction sites and fallows (CF)		4.52%	7.20%	2.88%	-	4.40%	1.40%	3.82%	1.21%
5: Traffic area (T)		0.003%	0.004%	0.026%	0.114%	-	0.005%	0.003%	0.009%
6: Recreational and green zones (RG)			0.02%	0.06%	0.32%		-	0.10%	0.02%
7: Agricultural area (A)		0.04%	0.04%	0.04%	0.10%	0.02%	0.02%	-	0.09%
8: Forest and unproductive area (FU)		0.003%	0.001%	0.003%	0.007%	0.004%	0.002%	0.047%	-

LUCC model – Drivers Weights of Evidence

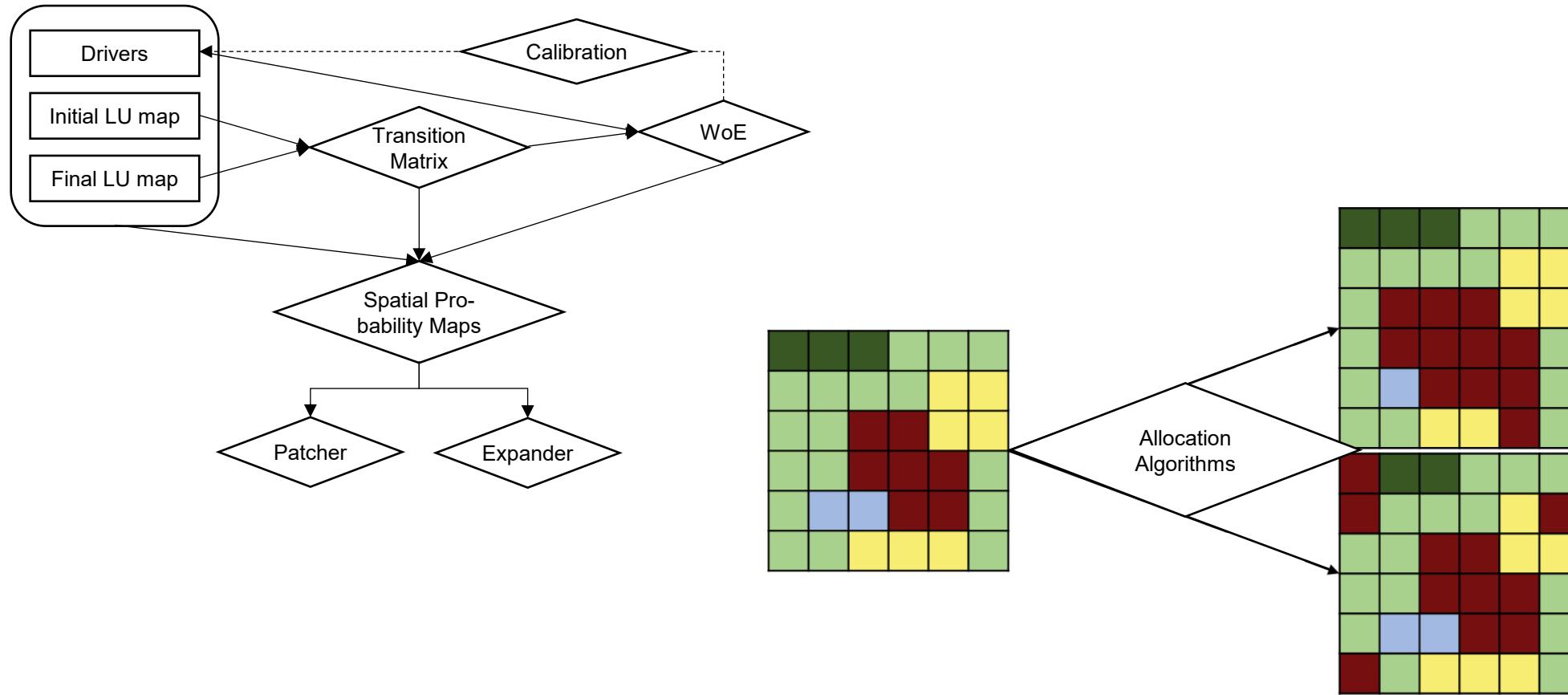


Distance to Train Stations
Agricultural area to high-density
settlement area



«The distance to train stations is significant in explaining the transition from agriculture to high-density settlement. For > 2km from a train station it is more likely that the transition doesn't occur than that it occurs.»

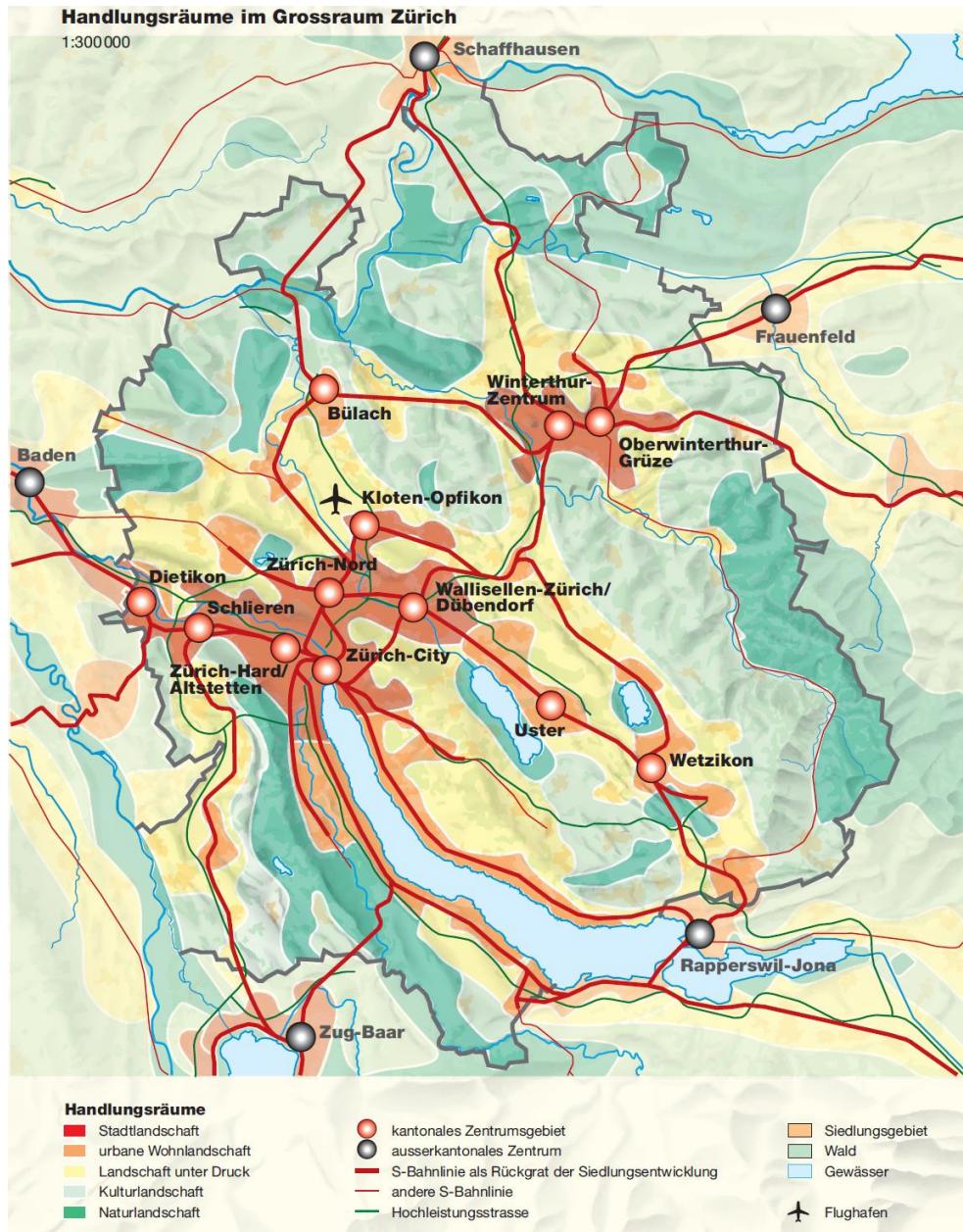
LUCC model – Allocation of predicted change



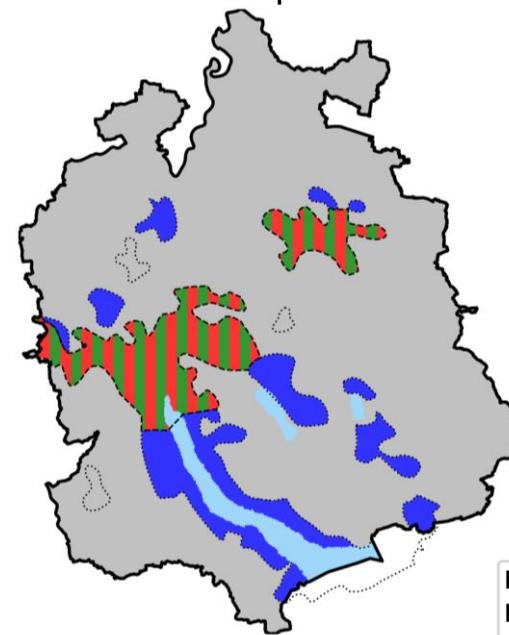
Experimental Design – 800 simulations – Latin Hypercube Sampling

Uncertain Factor	Min. Value	Max. Value	PDF*
Transition rate from low- to high-density settlement	0.25	0.35	Uniform
Transition rate from high- to low-density settlement	0.14	0.24	Uniform
Transition rate from industrial, commercial and public area to low-density settlement	0.04	0.05	Uniform
Transition rate from industrial, commercial and public area to high-density settlement	0.10	0.12	Uniform
Transition rate from agricultural area to low-density settlement	0.04	0.07	Uniform
Transition rate from agricultural area to high-density settlement	0.04	0.05	Uniform
Transition rate from agricultural area to industrial, commercial and public area	0.04	0.06	Uniform
Weights of Train Stations/Highway access points for probability surfaces	-50%/+50%	+50%/-50%	Uniform
Weights of Smaller cities/Big cities for probability surfaces	-50%/+50%	+50%/-50%	Uniform

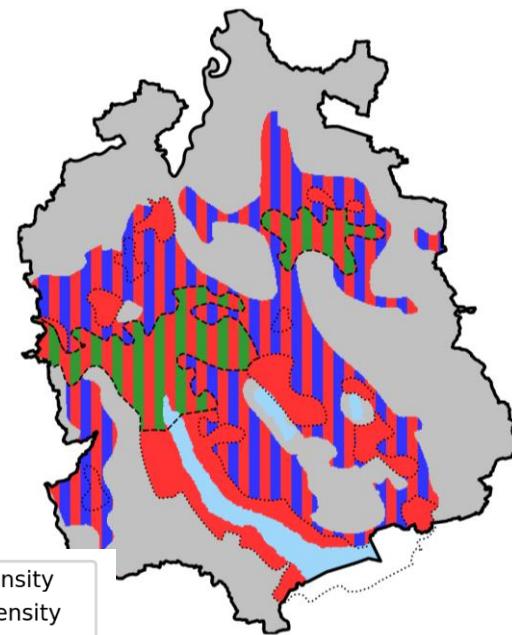
Spatial outputs - Normative



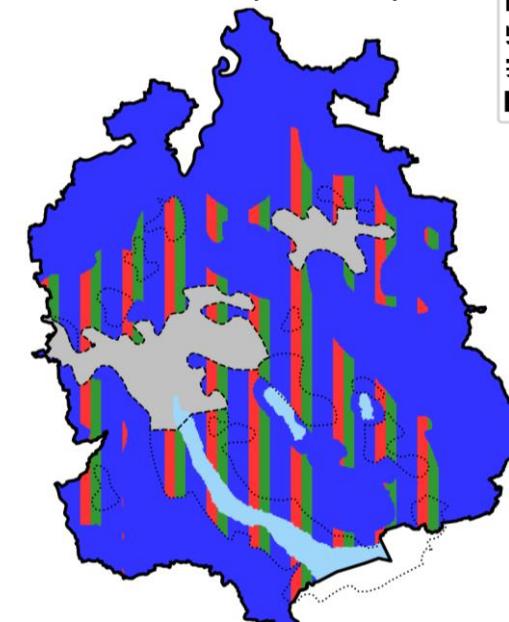
Scenario 1: Compact Growth



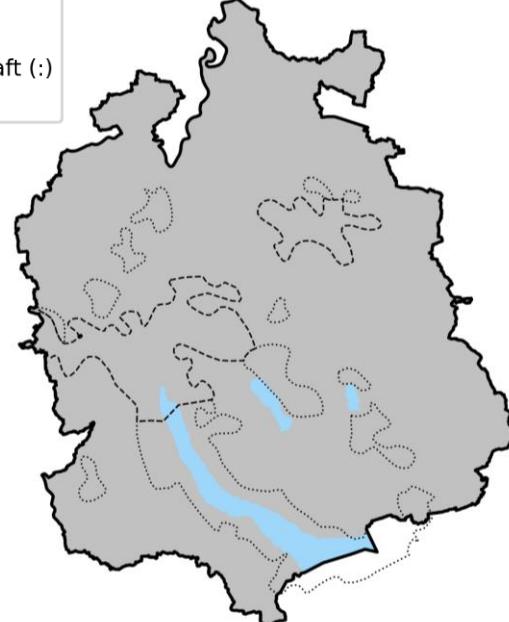
Scenario 2: Metropolitan Expansion



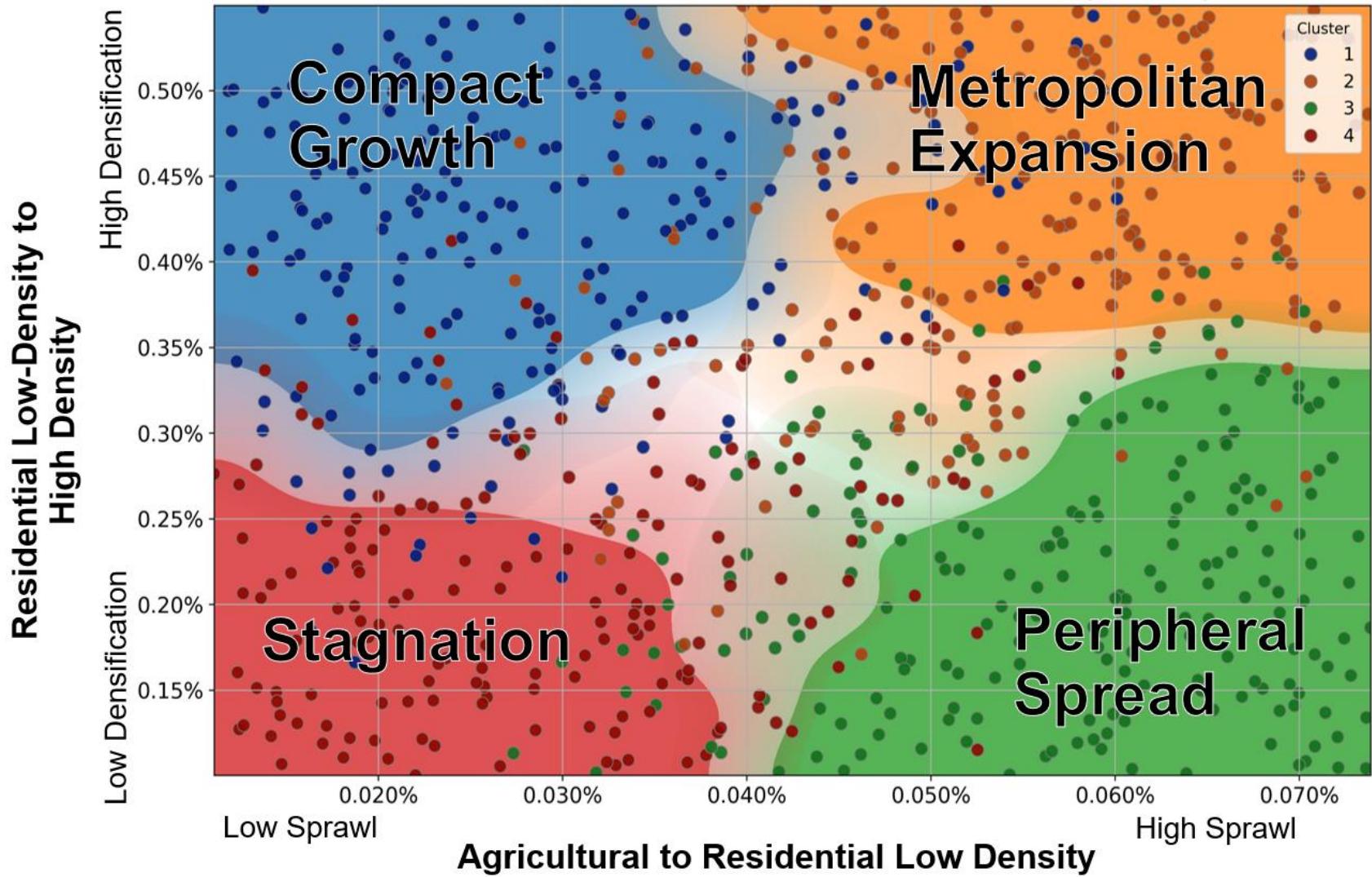
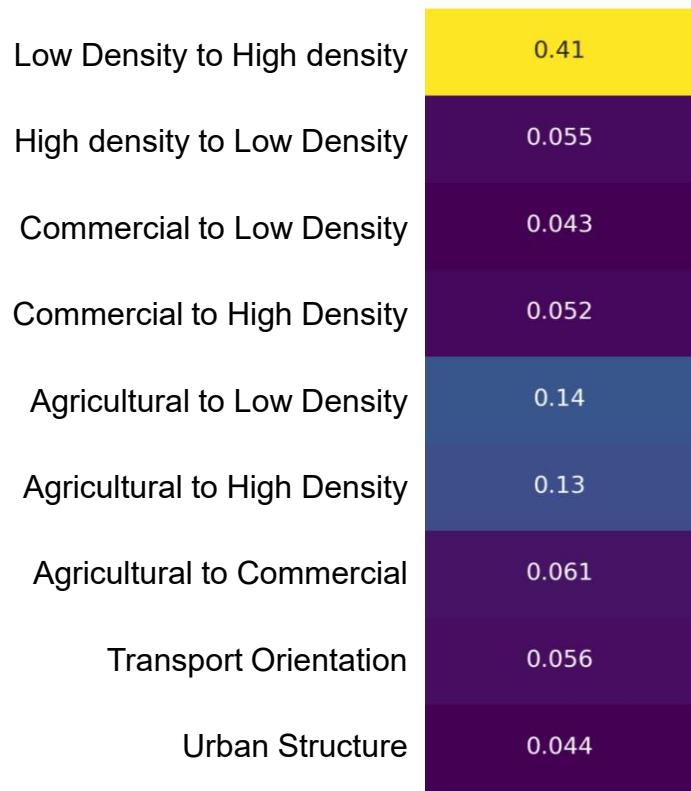
Scenario 3: Peripheral Spread



Scenario 4: Stagnation

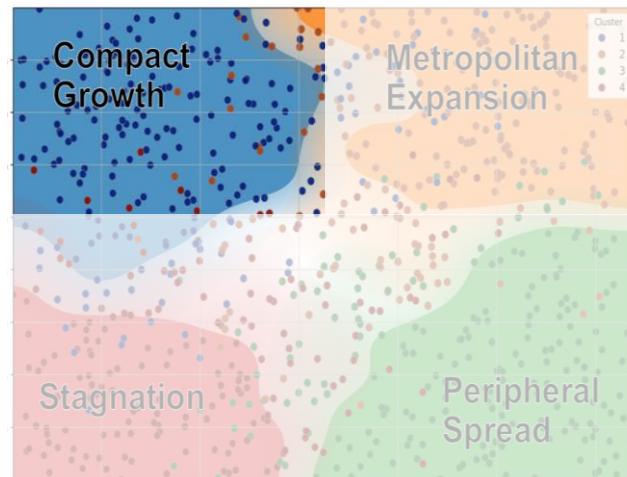


Scenario Map

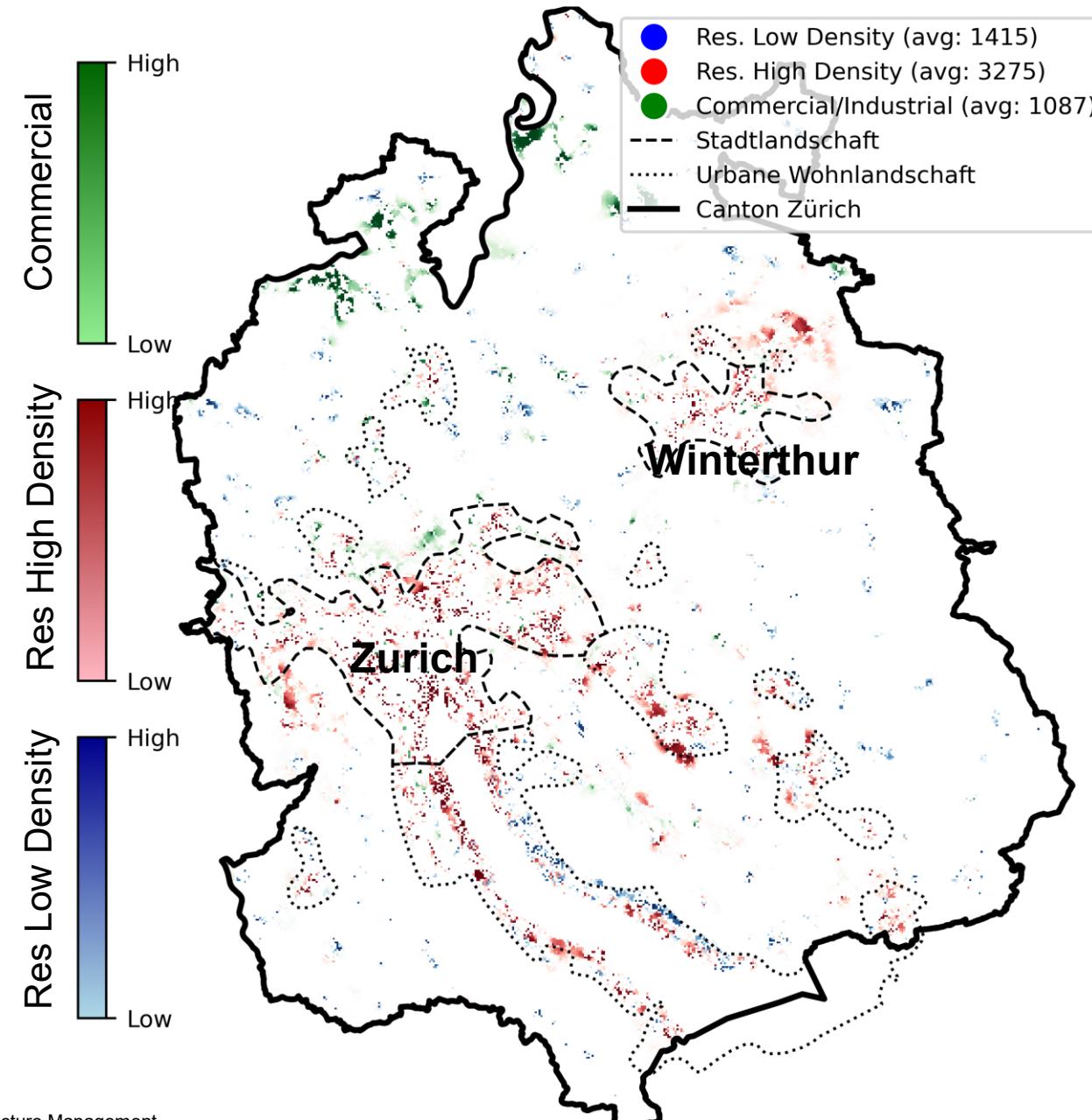


Spatial Scenarios

Compact Growth

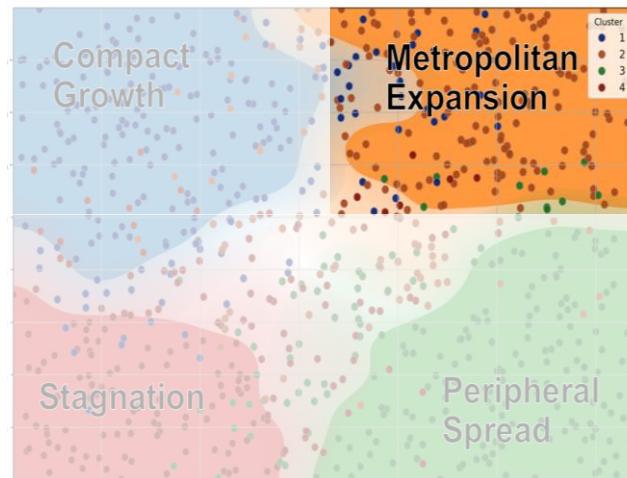


Cluster 1 (146 maps) - Aggregated Transitions

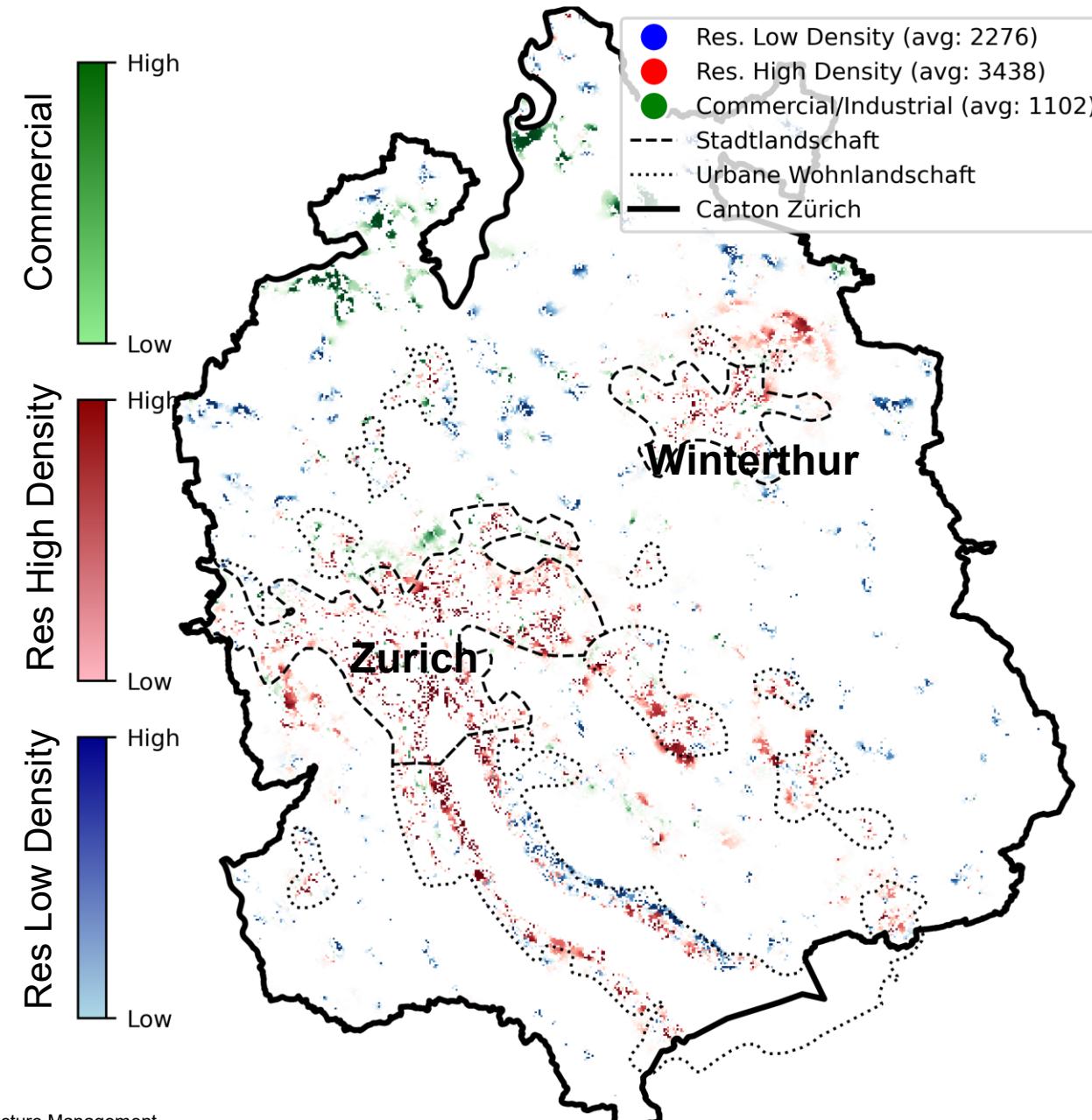


Spatial Scenarios

Metropolitan Expansion

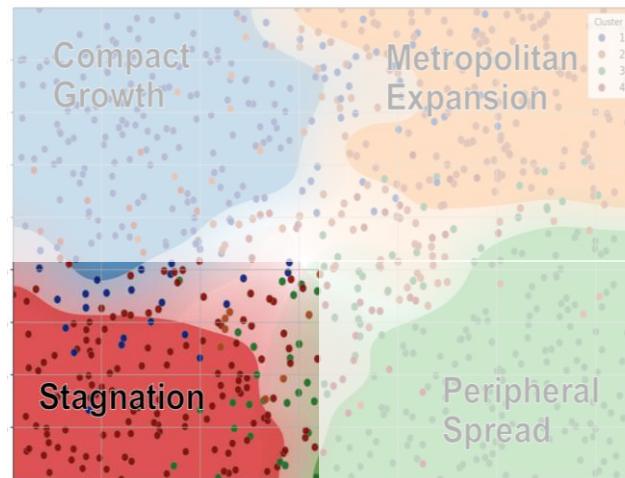


Cluster 2 (125 maps) - Aggregated Transitions

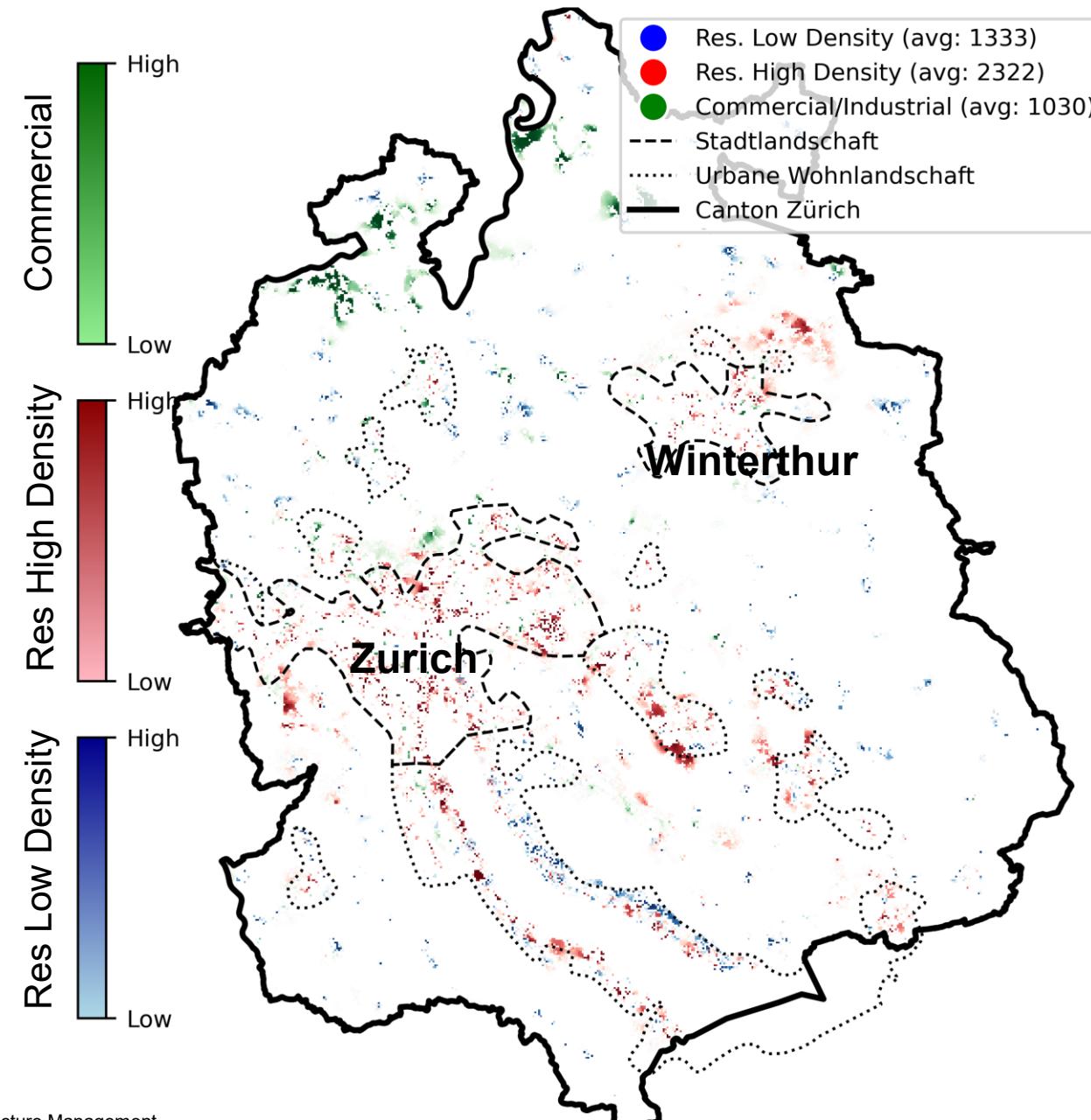


Spatial Scenarios

No Growth

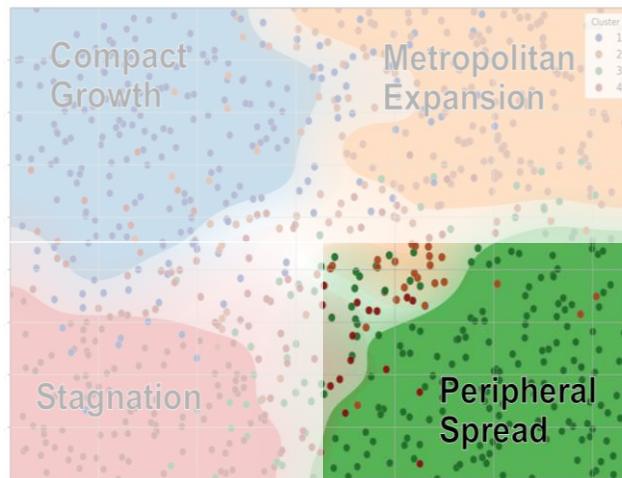


Cluster 4 (95 maps) - Aggregated Transitions

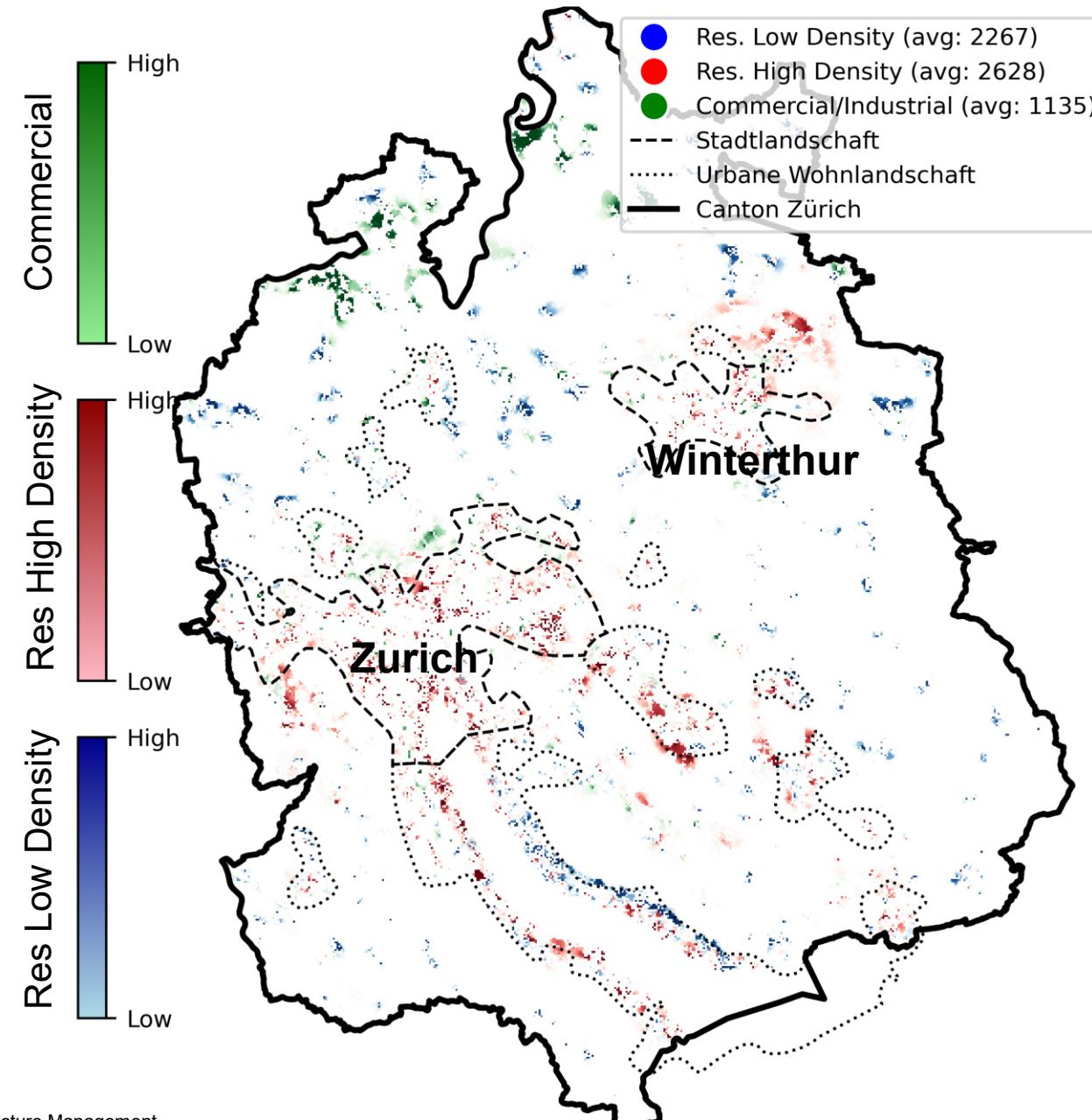


Spatial Scenarios

Peripheral Spread

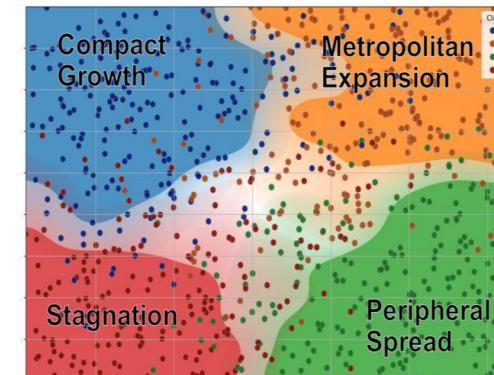
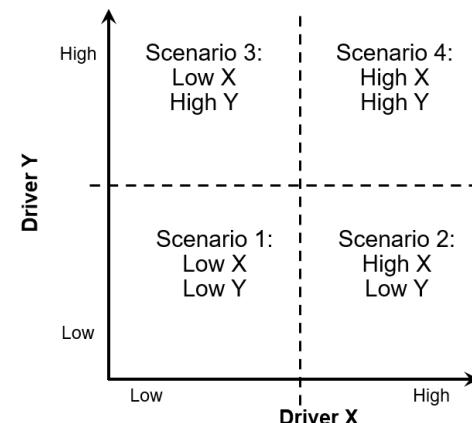


Cluster 3 (146 maps) - Aggregated Transitions



Conclusions

- We are proposing a methodology to develop scenario narratives that are informed by simulations. A “Scope, Simulation and Story” approach.
- Different and relevant spatial patterns emerge from the simulations providing insights for planning (i.e. which are the main factors producing such spatial patterns).
- There are technical challenges to measure spatial similarity, cluster spatial simulations and visualize spatial scenarios. This work contributes with a workflow and an example of how to implement such an approach to identify scenario narratives. The methodology could be adapted for other Urban Development model, not necessarily cellular automata models
- To embed this workflow into real planning processes, we propose to include participatory feedback at the beginning (the Scope) and the end (the Story) of the Simulation process.



Orlando Roman
oroman@ethz.ch

ETH Zurich
Institut für Bau- und Infrastrukturmanagement
Stefano-Francini-Platz 5
8093 Zürich

www.ibi.ethz.ch