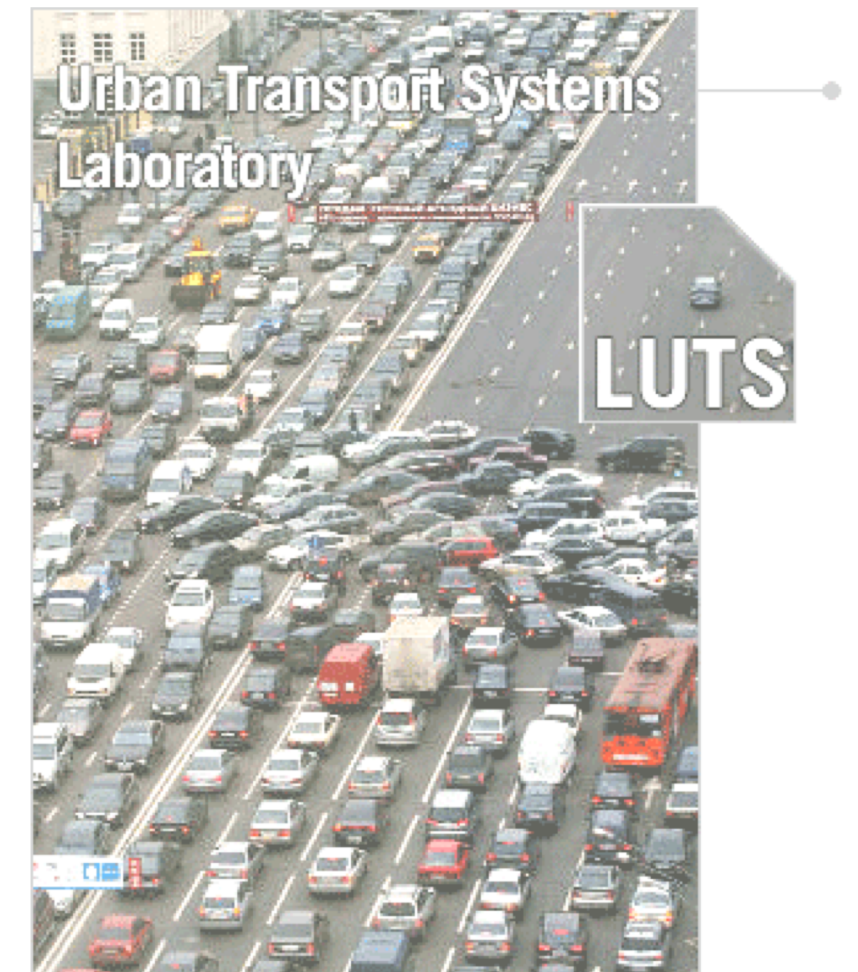


Network-level traffic management and control with MFDs

Intro to traffic flow modeling and ITS

Prof. Nikolas Geroliminis

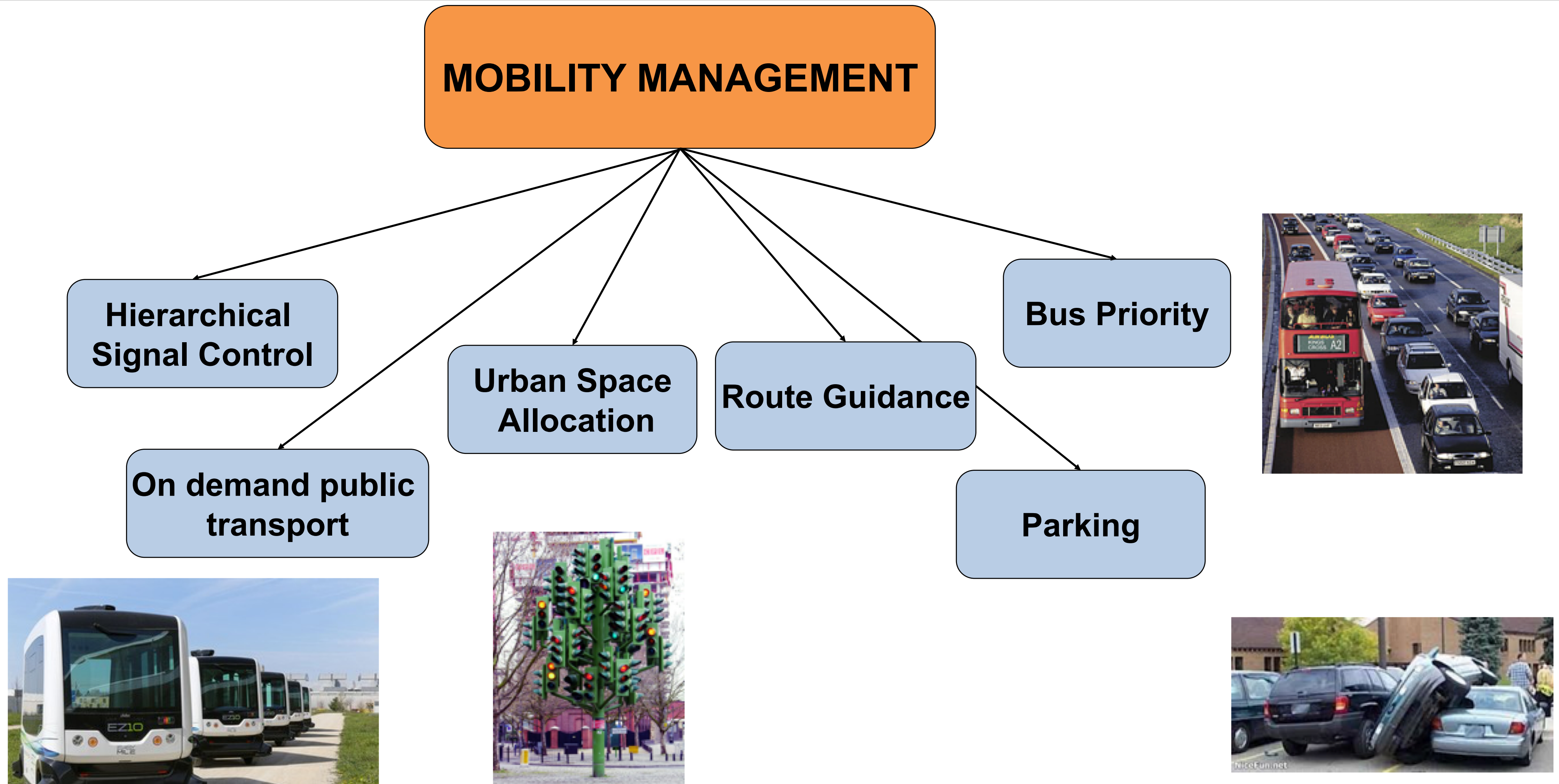


Welcome

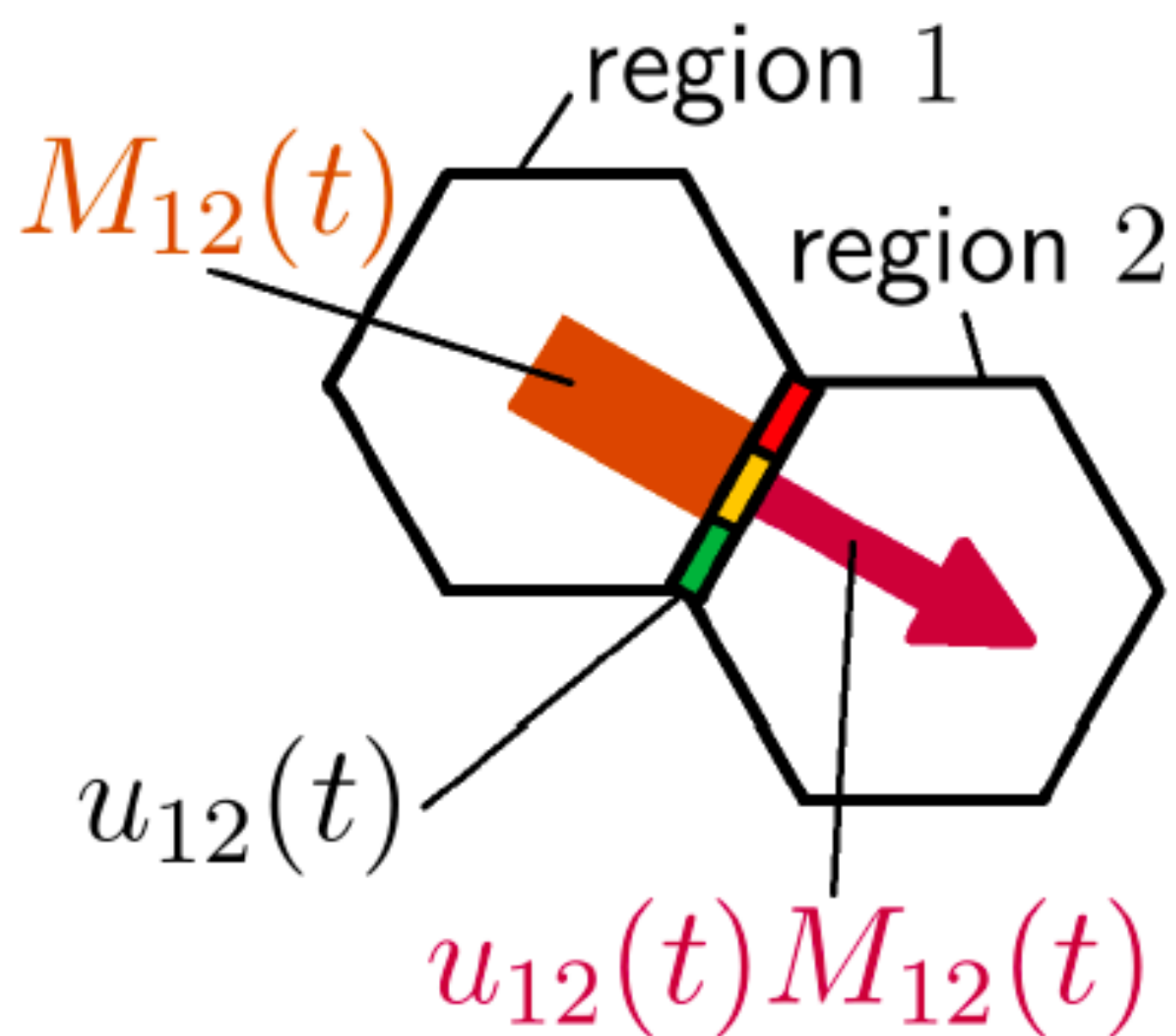
EPFL

- Week 5.1
 - Control Logic
 - Single-region perimeter control
 - Multi-region perimeter linear control
- Week 5.2
 - Model Predictive Control with MFDs

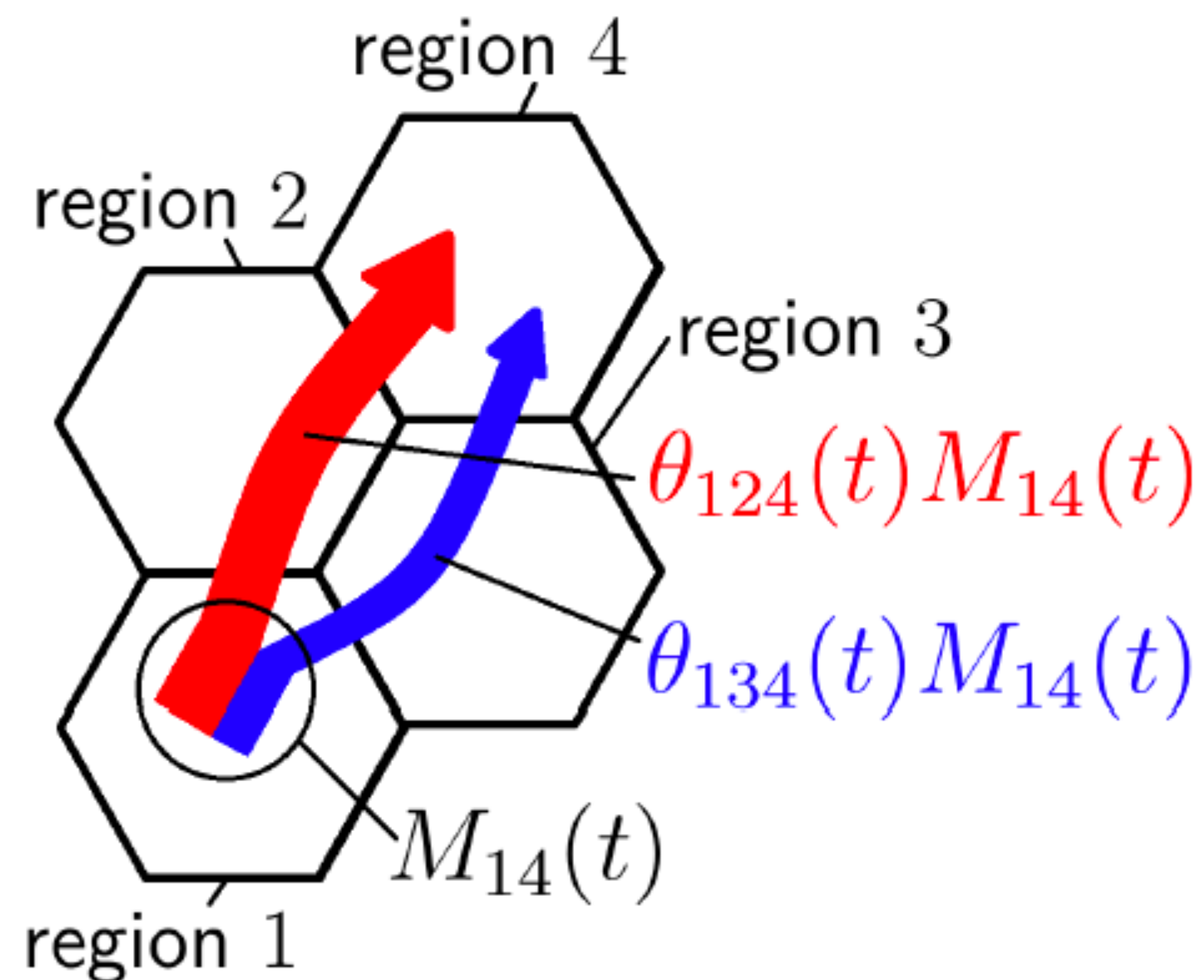
A System-of-Systems approach



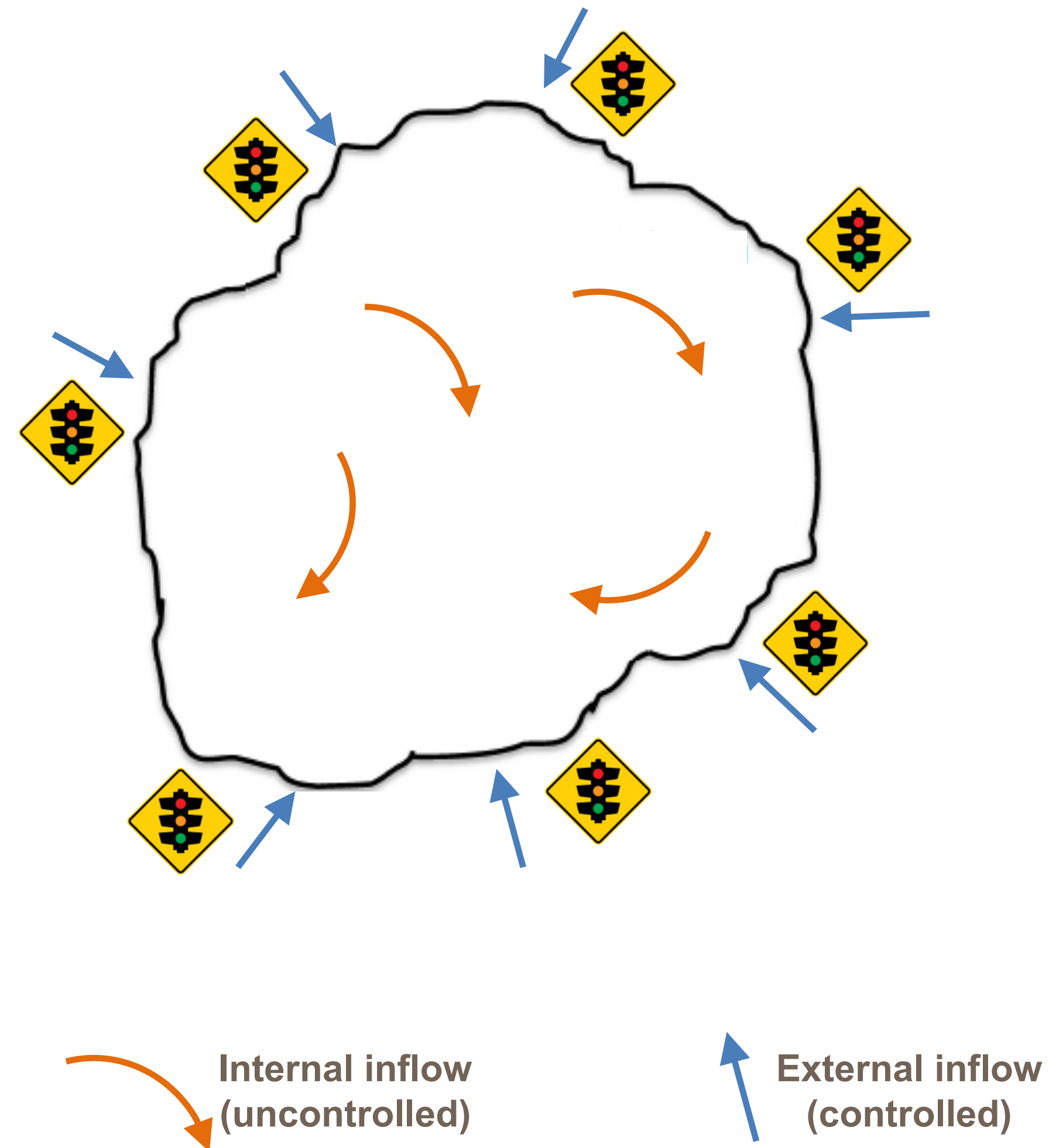
Perimeter control



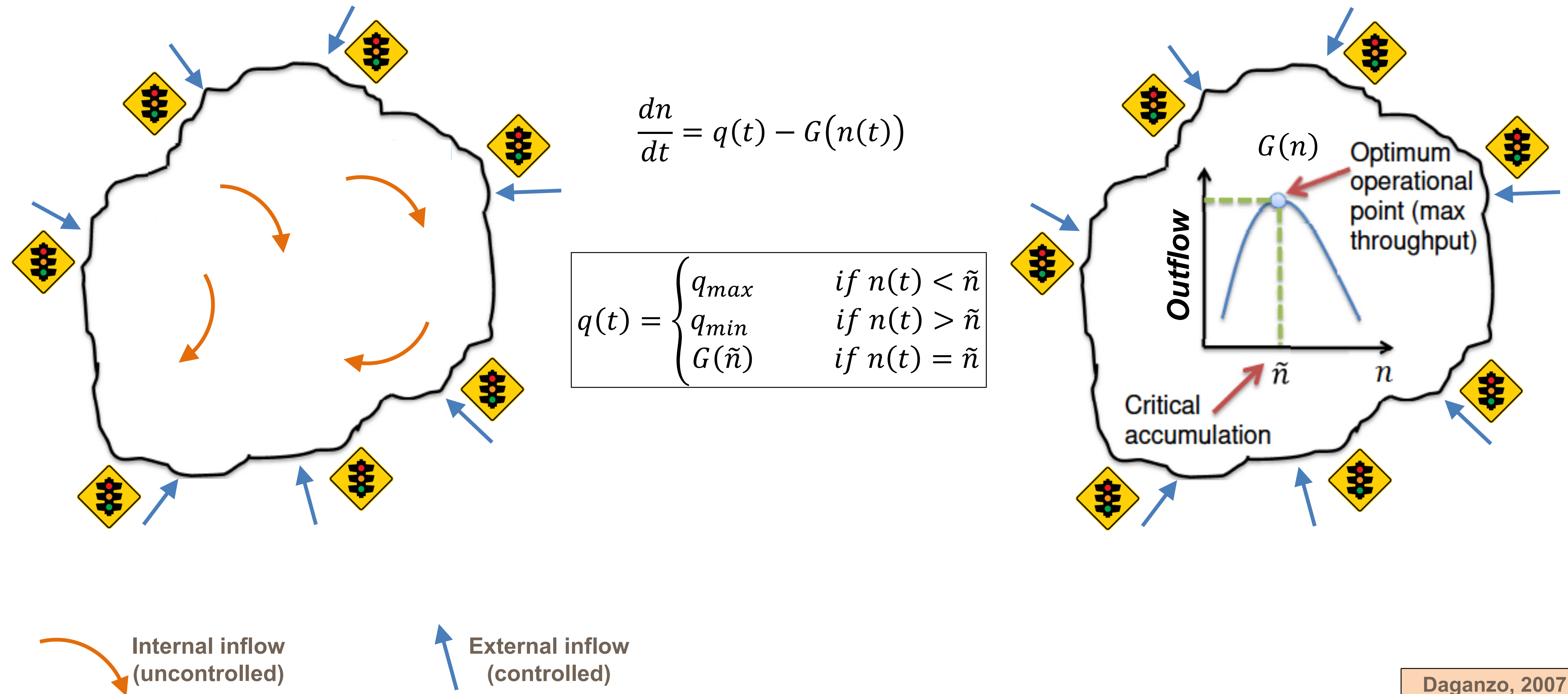
Route guidance



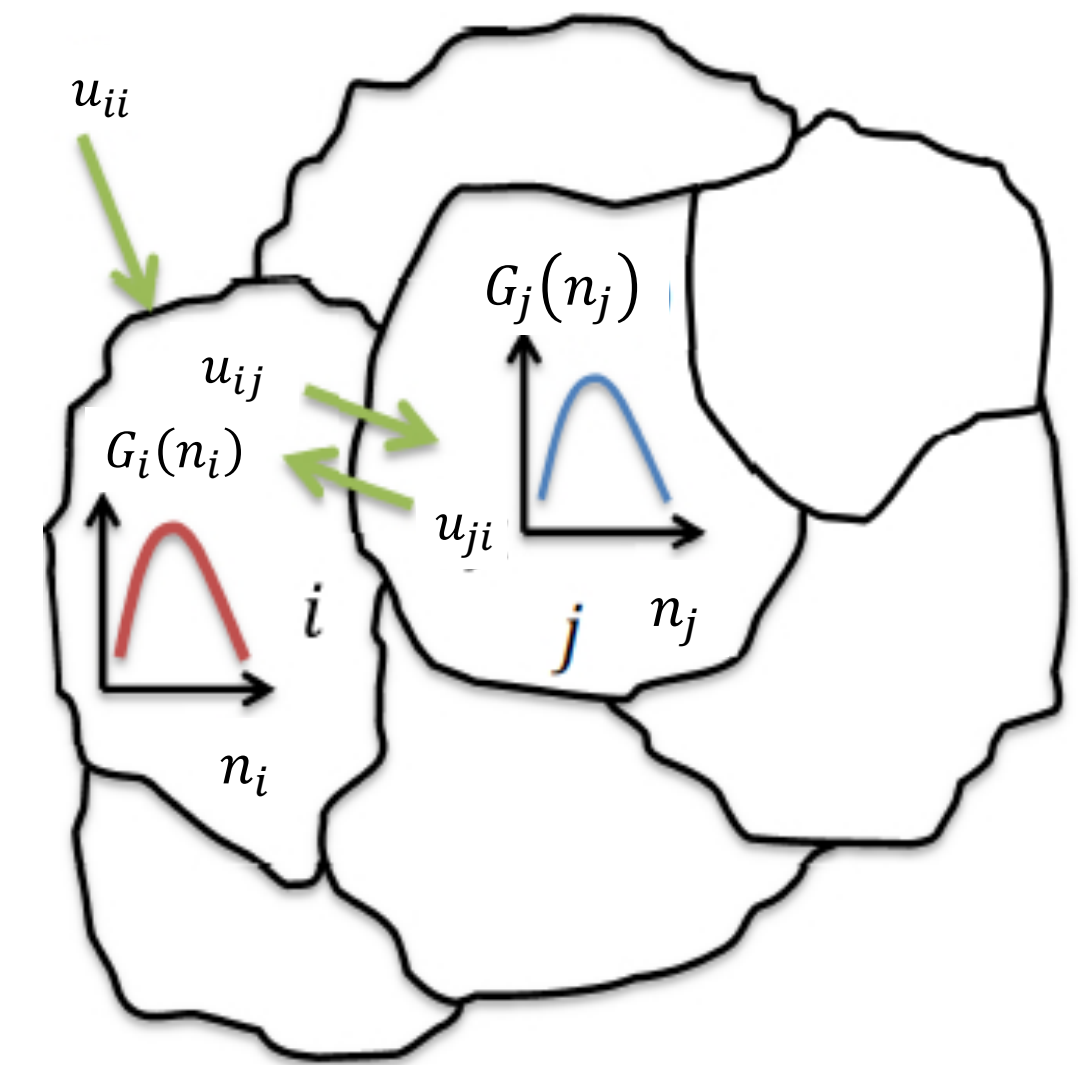
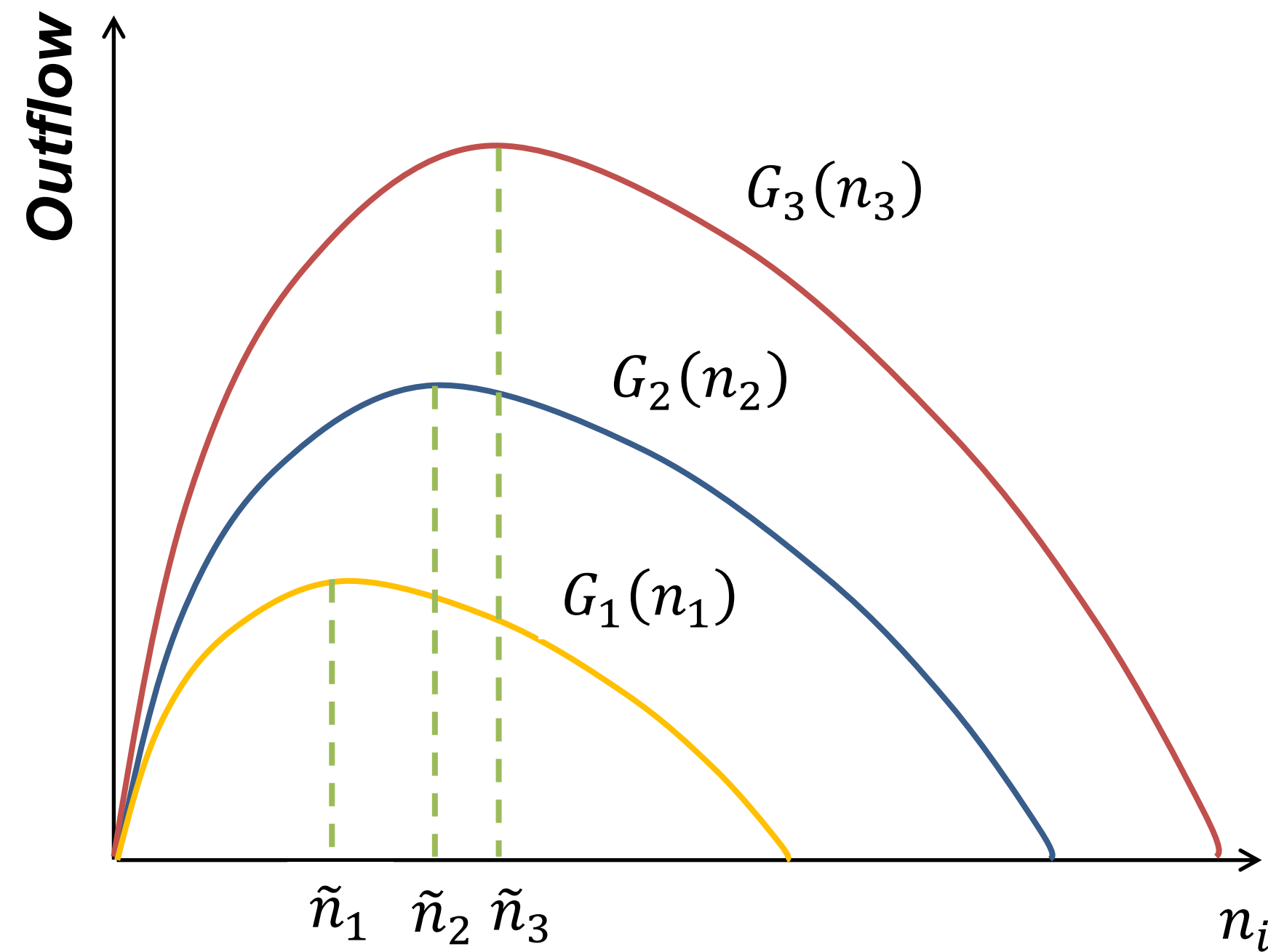
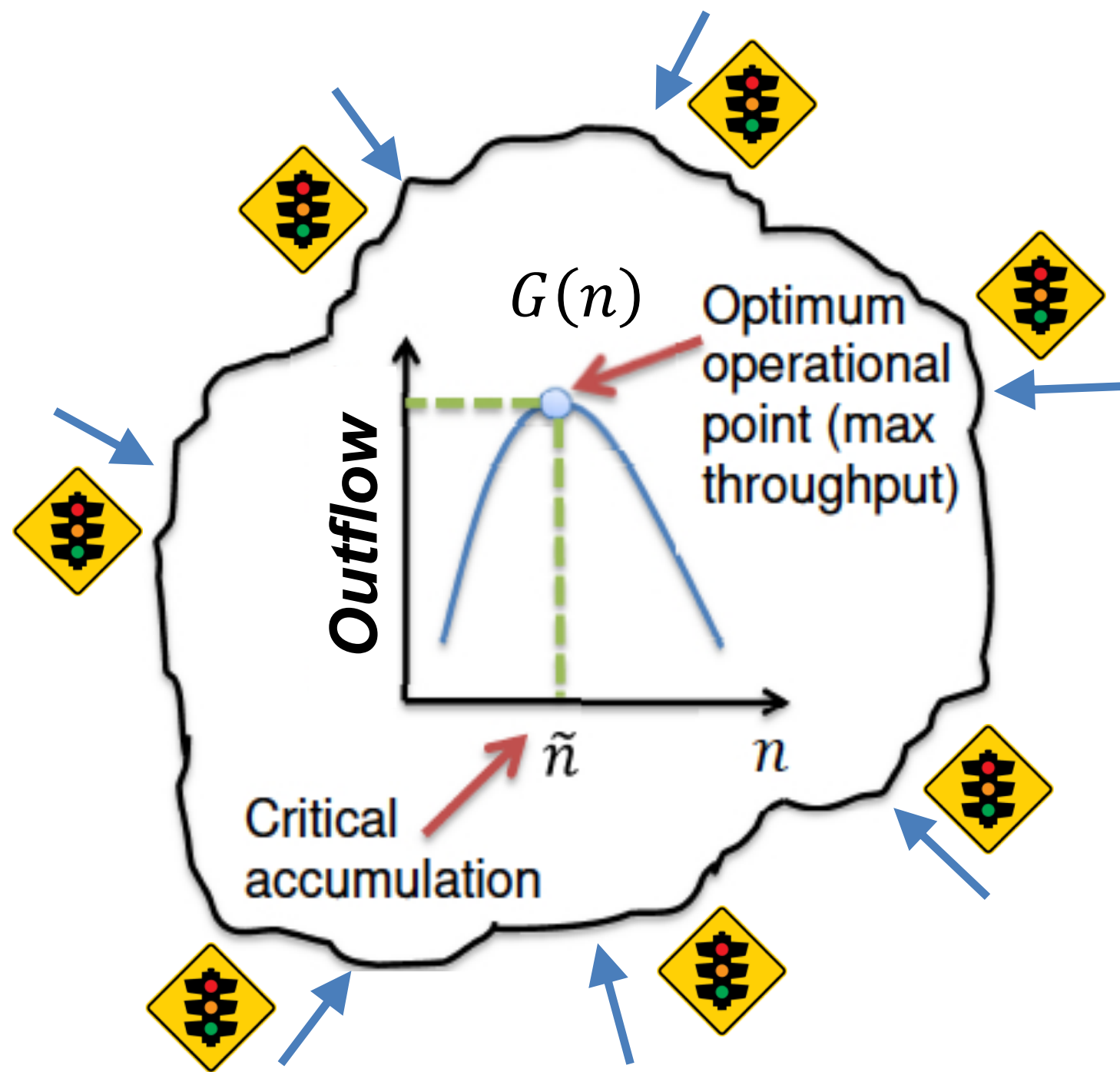
Single-region perimeter control



Single-region perimeter control



Control logic: from single- to multi-region



WHY MULTI-region?

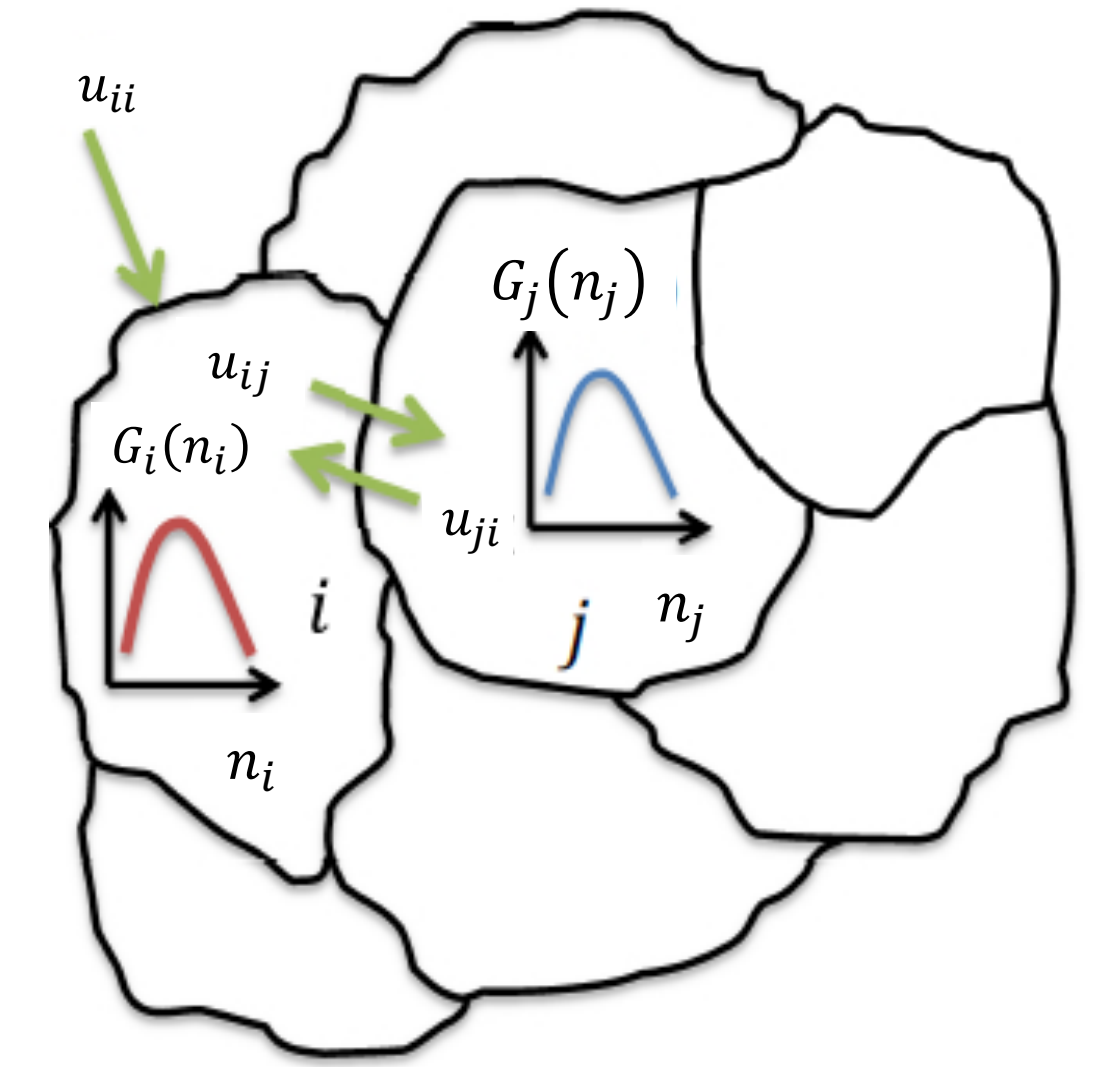
- heterogeneously congested cities with multiple pockets of congestion (partitioning)
- High internal inflows (controllability)
- Avoid long queues at boundaries (equity)

Multi-region perimeter control (LQI control)

- **Given:** MFDs for each reservoir $G(n)$
- **Observe:** Accumulation $n_j(t)$
- **Control:** Boundary flows and external inflows
- **Maintain** Throughput Optimal Operational Point
- **Linear-Quadratic-Integral control** (multivariable PI-regulator)

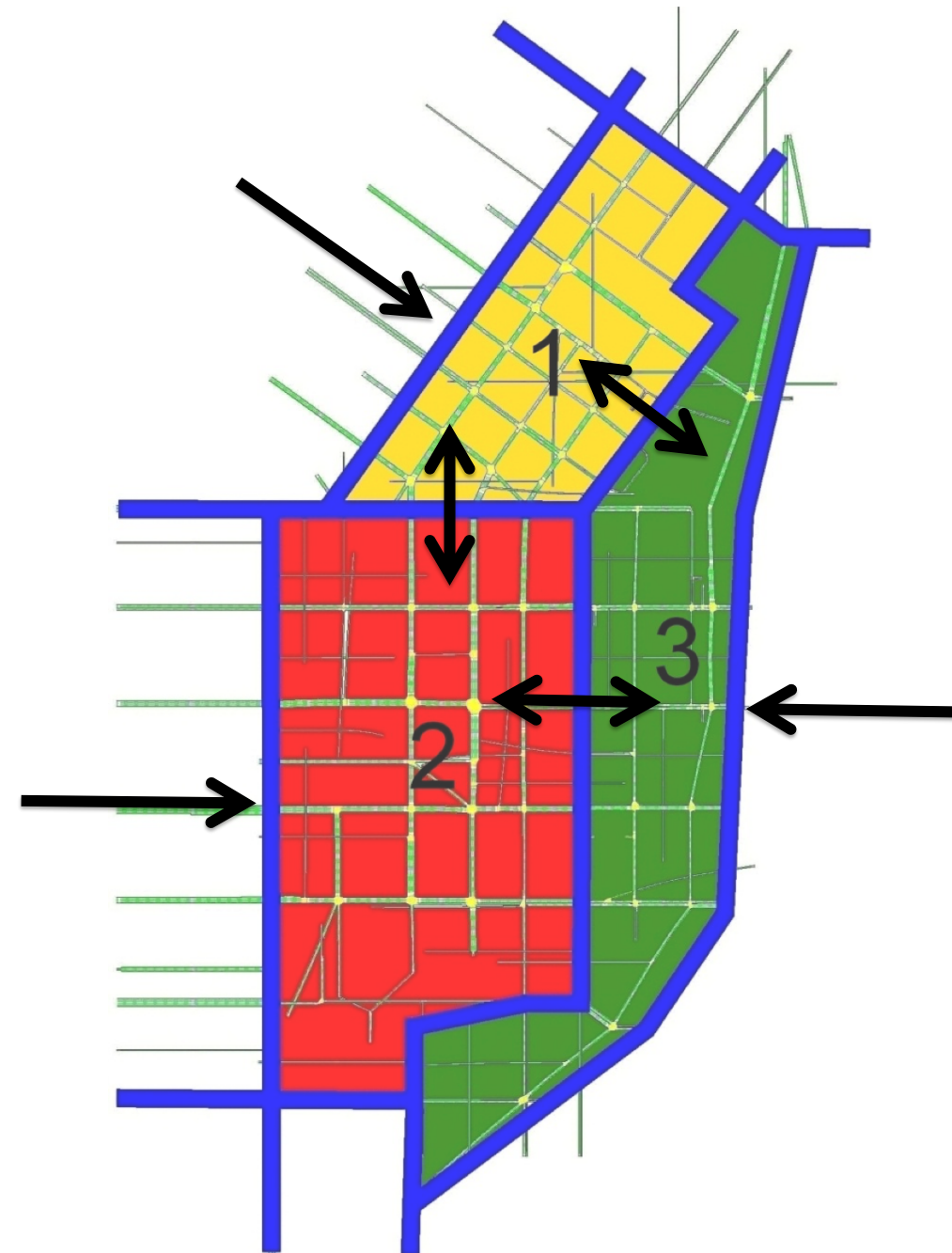
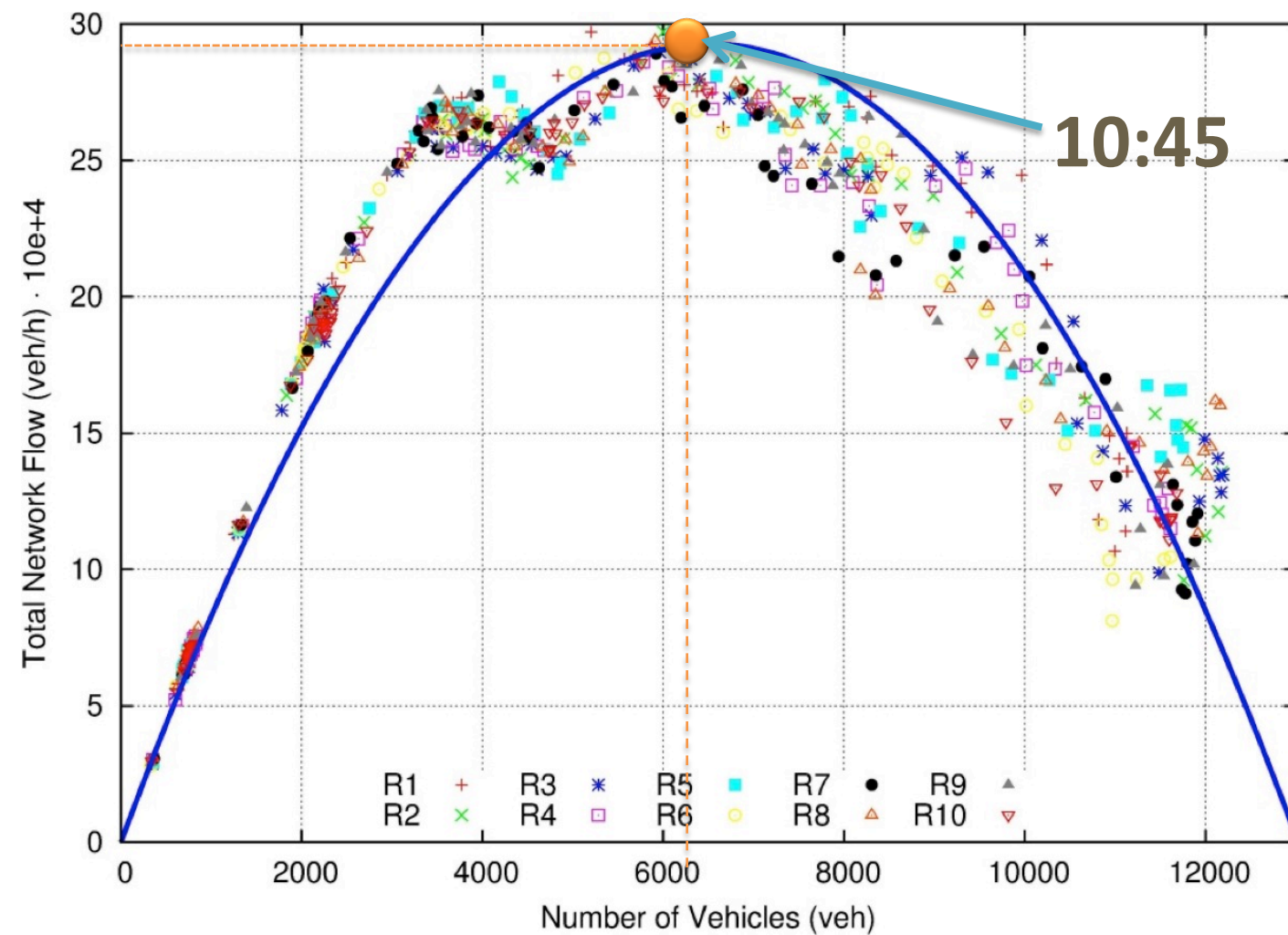
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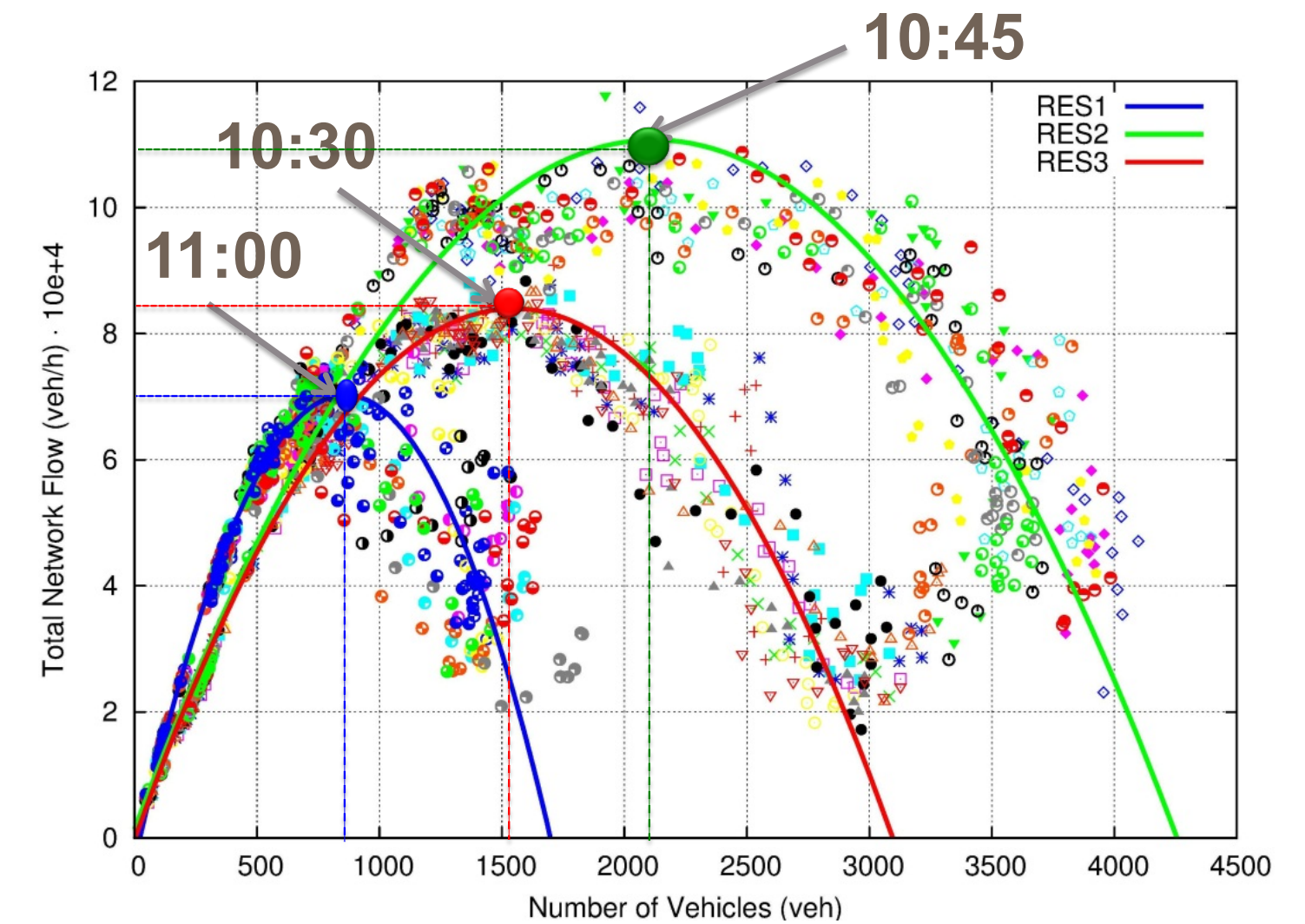


A multi-region case study

MFD for the original network



MFDs for each reservoir

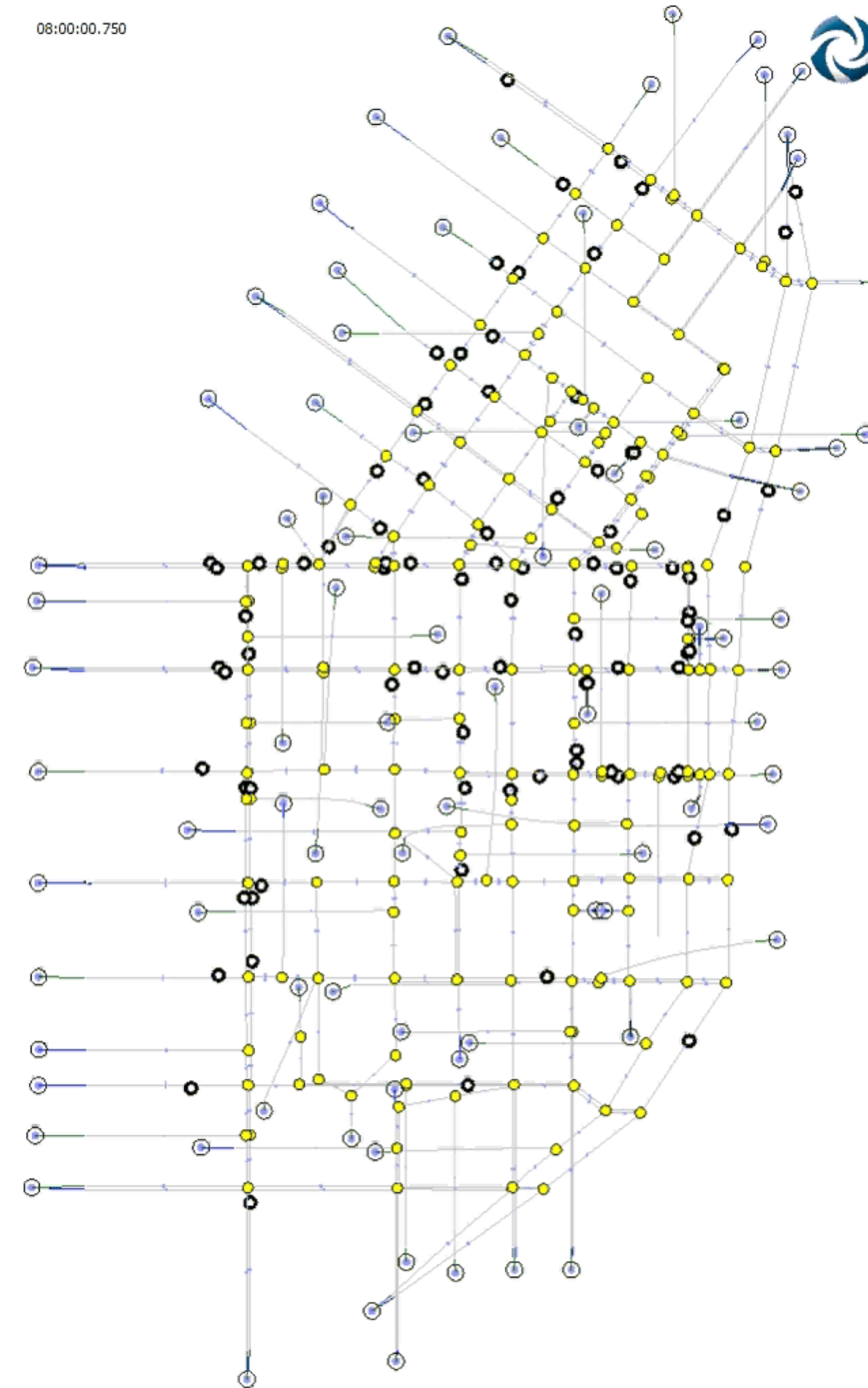
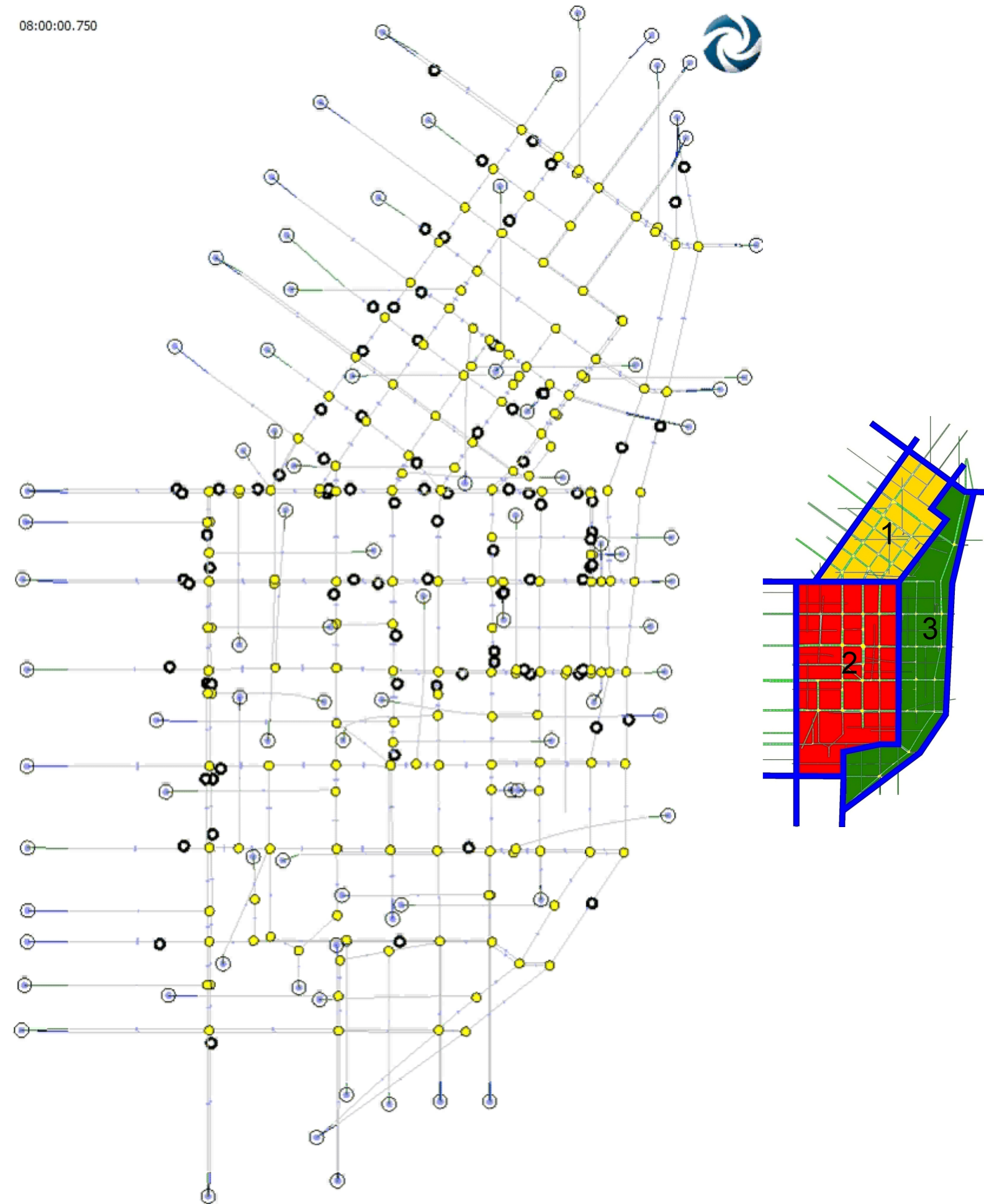


Observations:

- MFD: regions exhibit MFDs with quite moderate scatter
- Heterogeneity: regions reach the congested regime at different times

Micro-simulation of different controllers

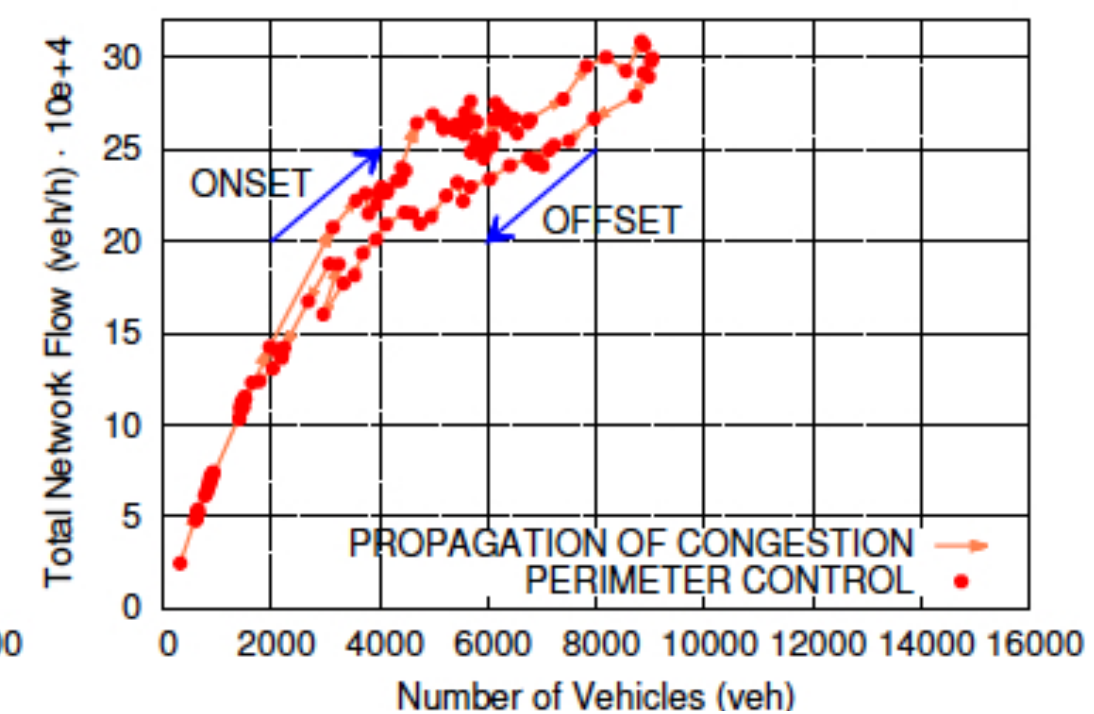
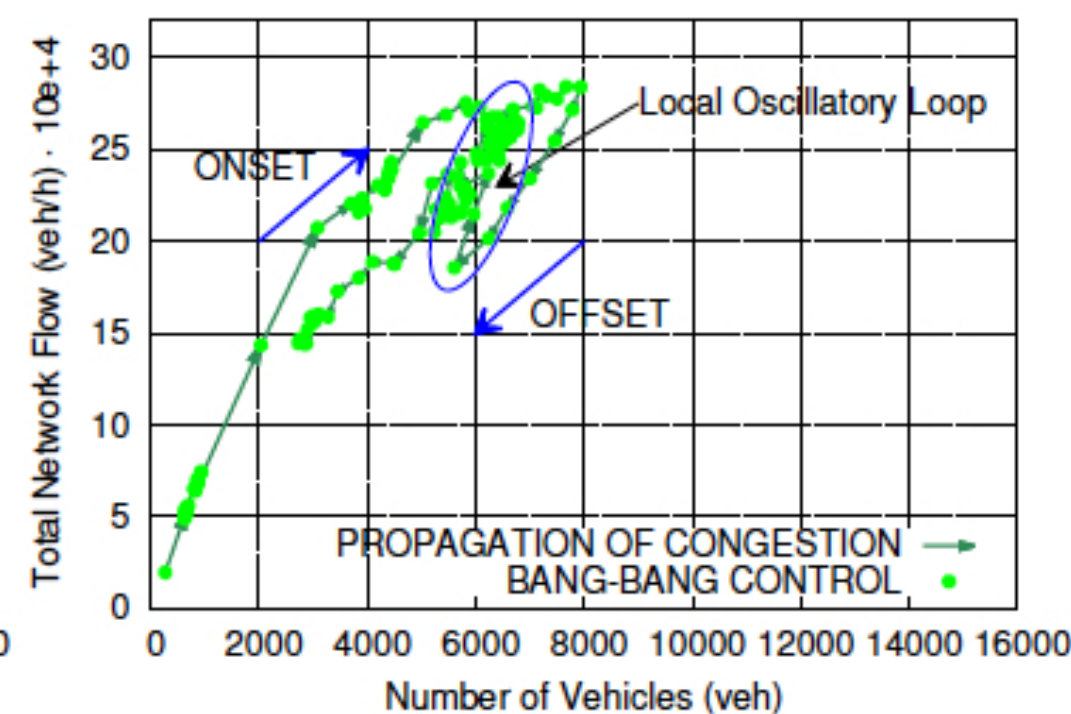
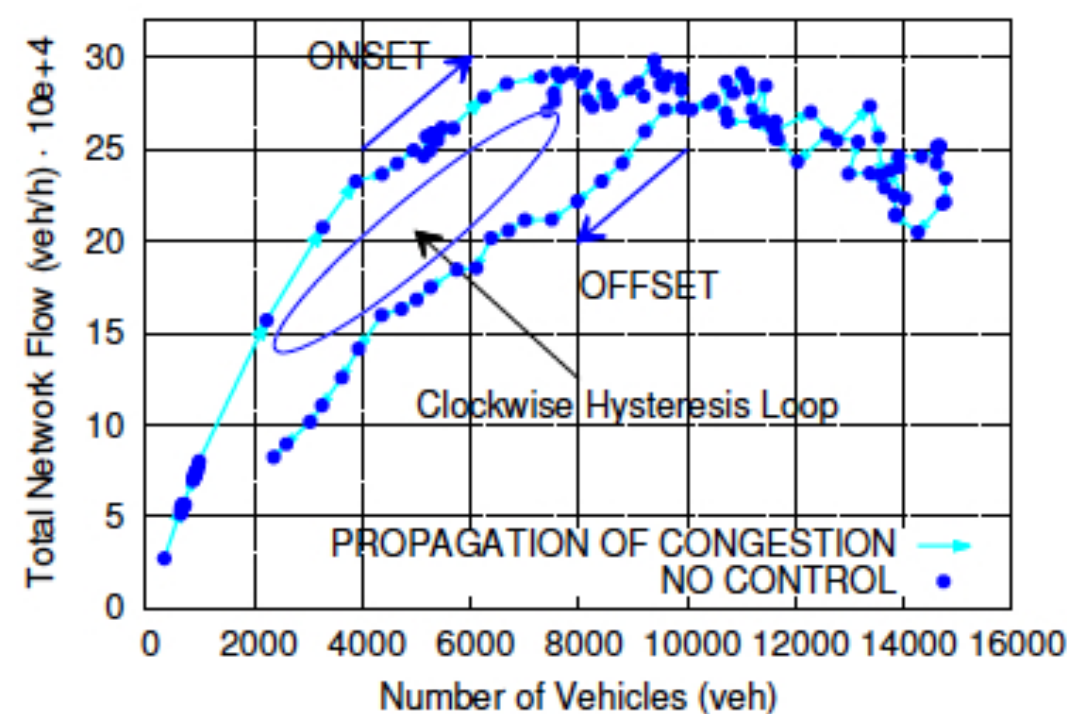
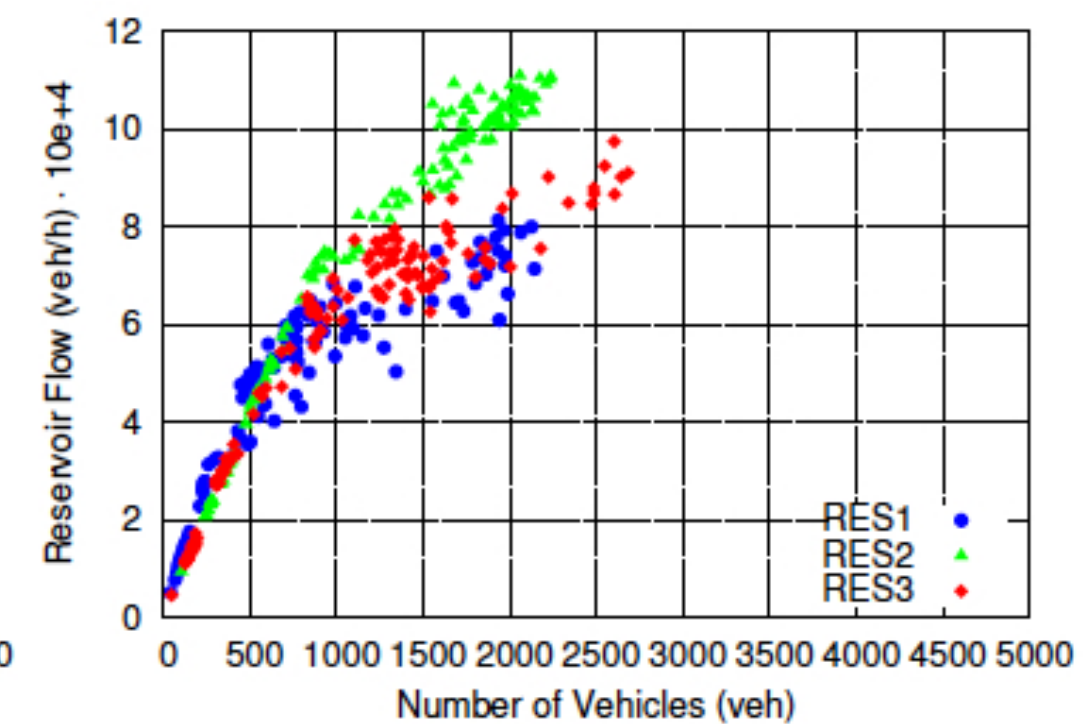
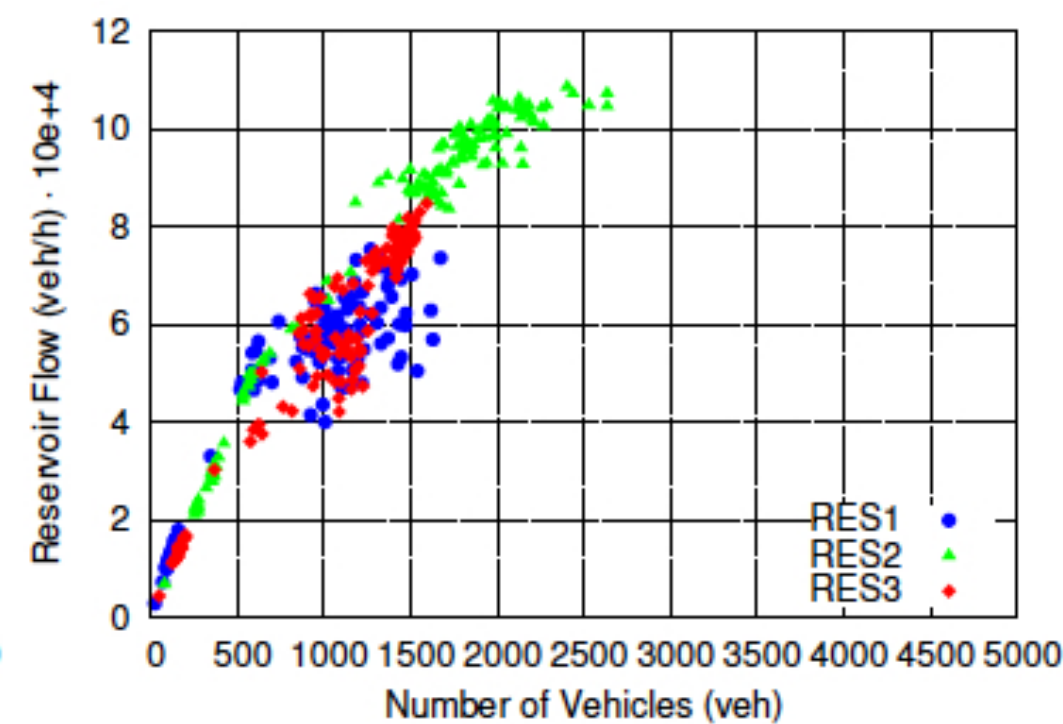
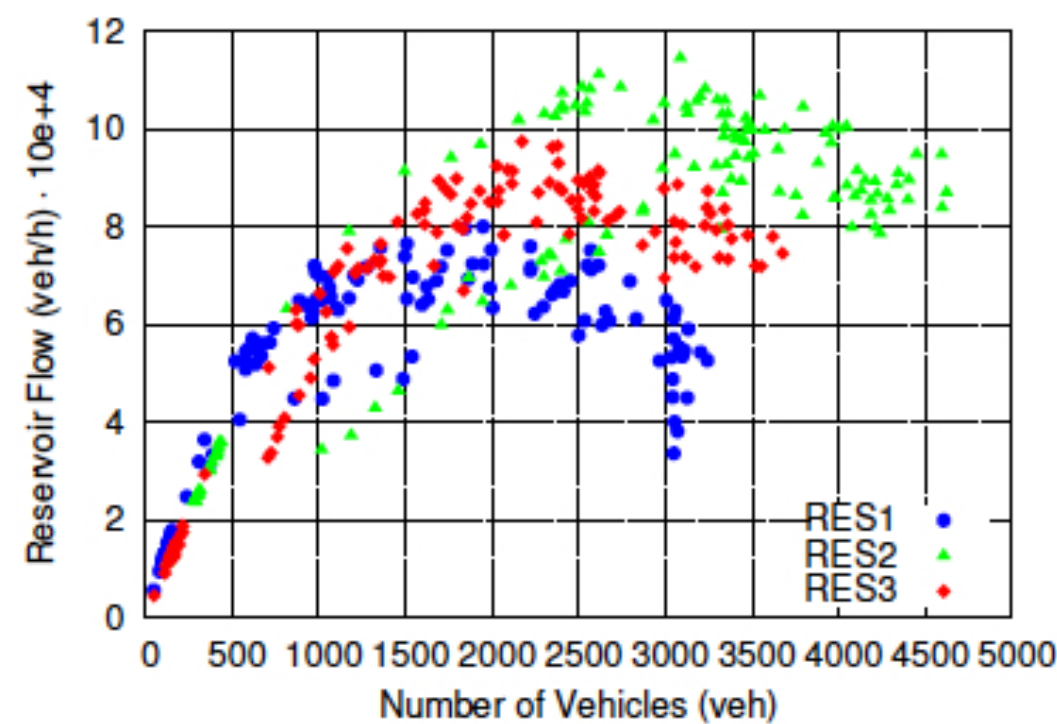
Pre-timed Traffic Control



Smart Traffic Control

Comparison of controllers

- Comparison with pre-timed fixed signals: improvement 33% for total travel time
- Comparison with Bang-bang control: Improvement 12%
- Multi-region PI: maintains throughput; respect reservoirs' homogeneity



Pre-timed fixed signals

Single-region Bang-Bang control

Multi-region PI control

A more advanced controller (PI + parameter fine tuning)

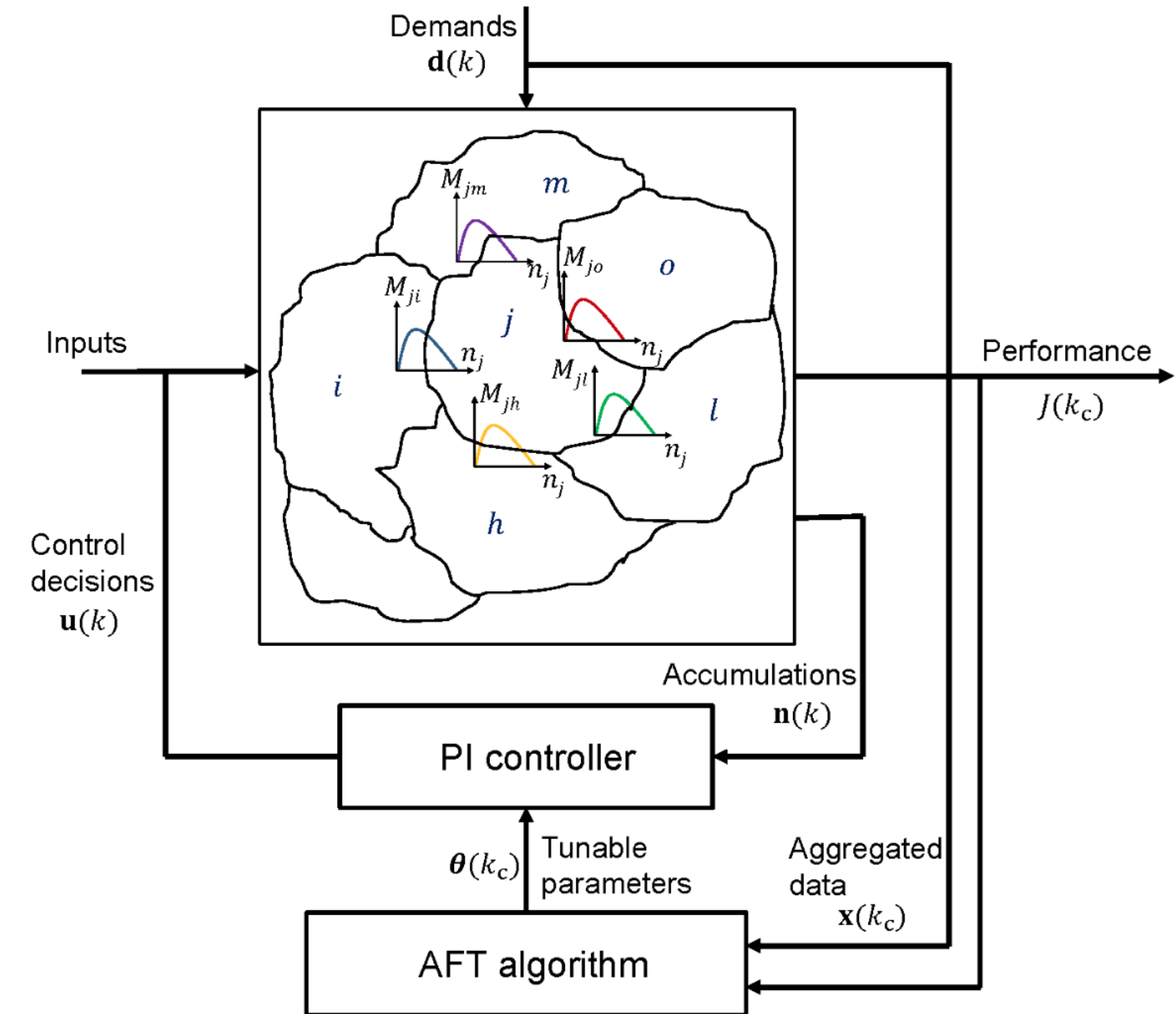
(a) Model-based multivariable PI regulator

$$\mathbf{u}(k) = \mathbf{u}(k-1) - \mathbf{K}_P [\mathbf{n}(k) - \mathbf{n}(k-1)] - \mathbf{K}_I [\mathbf{n}(k) - \hat{\mathbf{n}}]$$

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- $\mathbf{K}_P, \mathbf{K}_I$: proportional and integral gains.

(b) Data-driven online adaptation

- At each iteration, AFT algorithm receives
 - the measured performance index J (total delay of the system)
 - the measurable external disturbances $\mathbf{x}$ (aggregated demand)
- Using the samples of the measured quantities AFT calculates new tunable parameter values to be applied at the next iteration.



A more advanced controller (PI + parameter fine tuning)

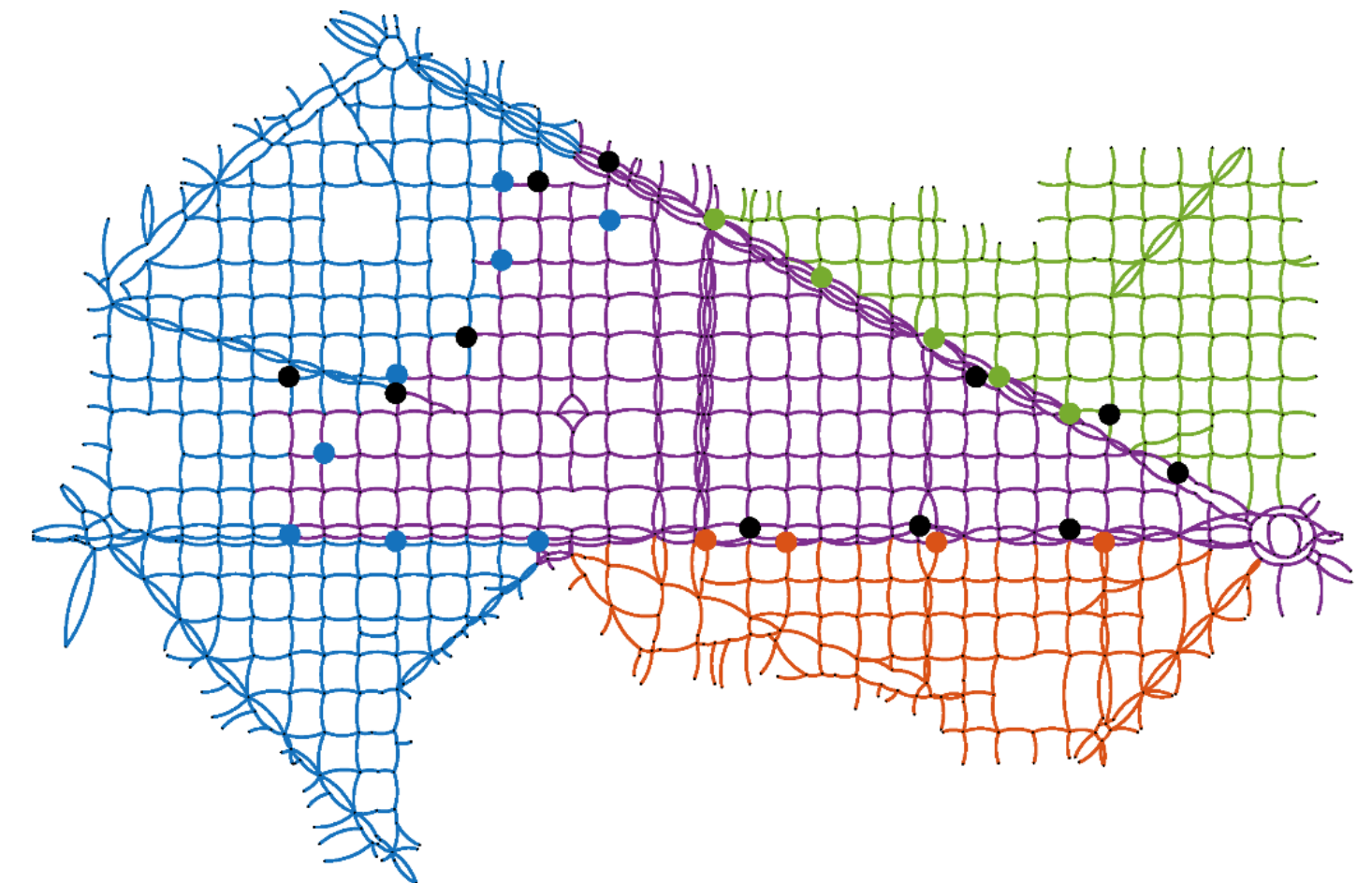
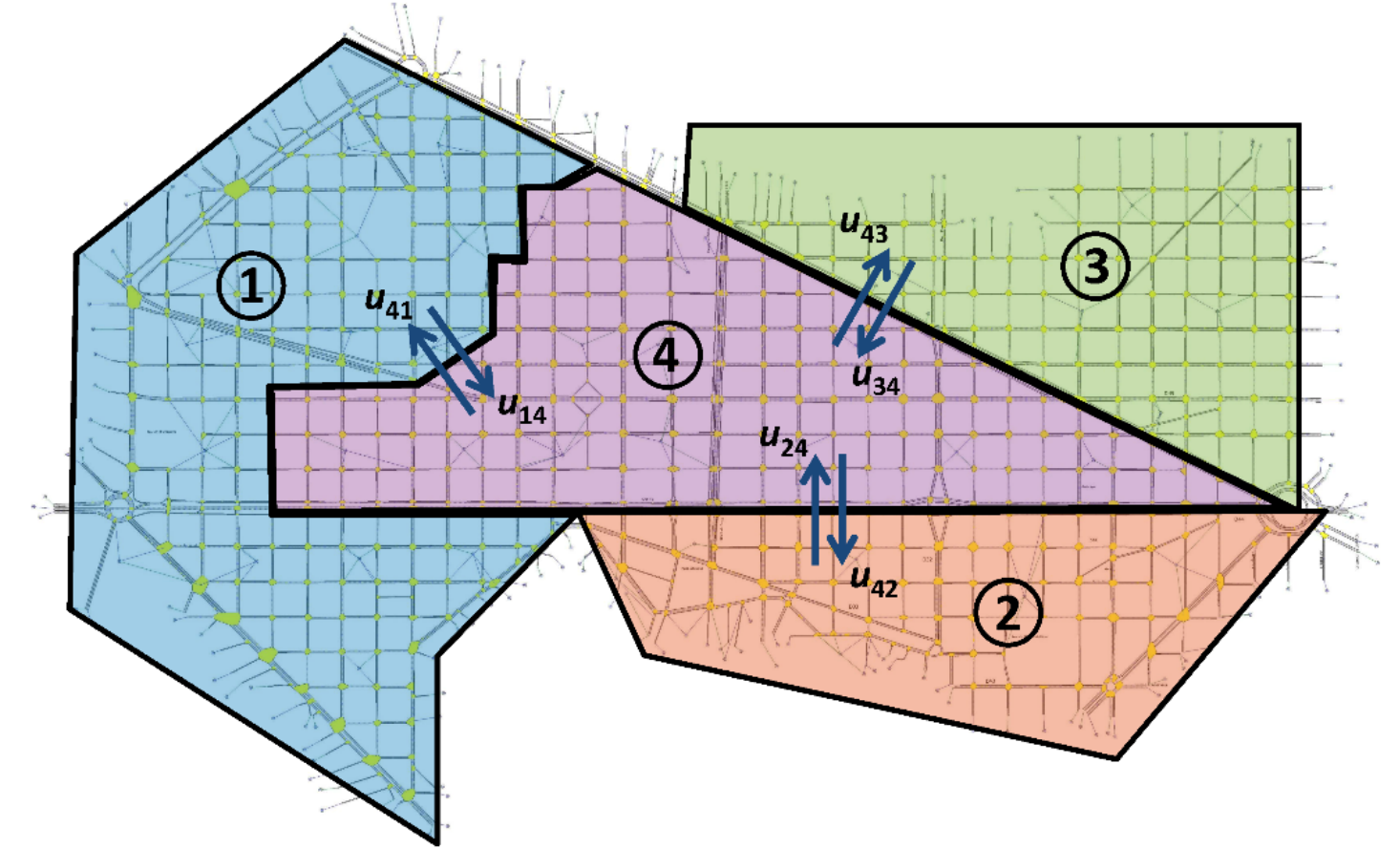
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Micro-simulation results

