



Numerical Analysis of Discrete-time Networked Competitive Bivirus SIS Models

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Motivation



- **Goal:** understanding virus spread using mathematical modeling and computational simulations.

Problem Statement

- **Goal:** understanding virus spread using mathematical modeling and computational simulations.
- Traditional focus on single-virus models
 - We focus on **competitive bivirus** processes in a population.

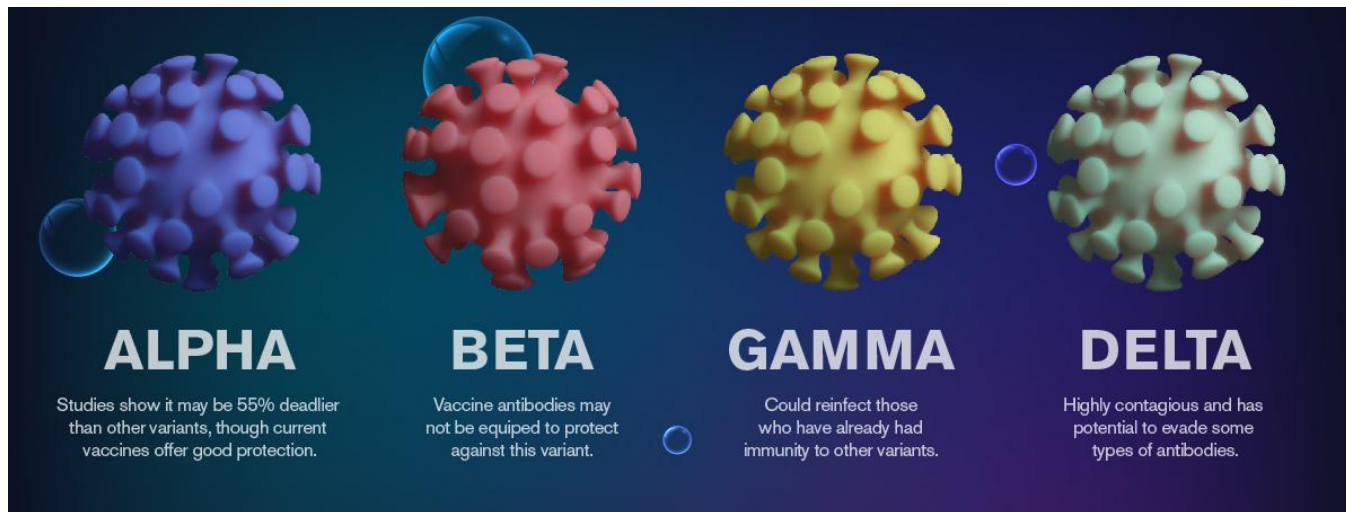


Image from: <https://news.asu.edu/20210812-sars-cov-2-theme-and-variations>

Introduction

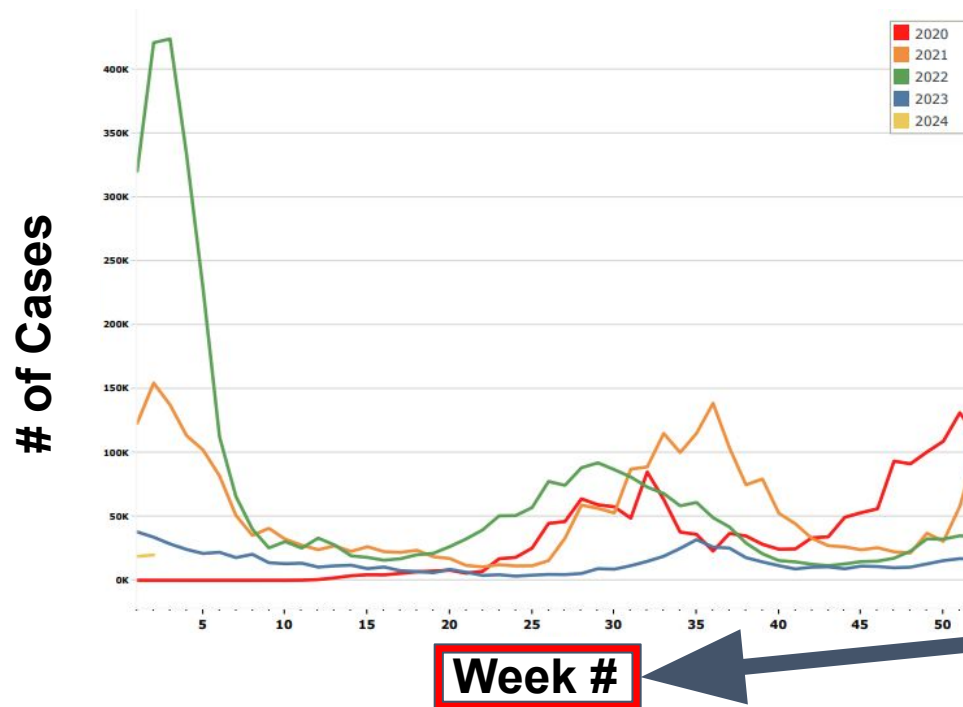
Model Setup

Real-world
Simulations

Conclusion

Problem Statement

- Traditional focus on continuous-time models
 - We highlight the **discrete-time** approach used to match real-world virus data collection.



One data point
per **week**

Image from: <https://www.dshs.texas.gov/sites/default/files/IDCU/disease/COVID-19/2024/2024Wk02Jan18.pdf>

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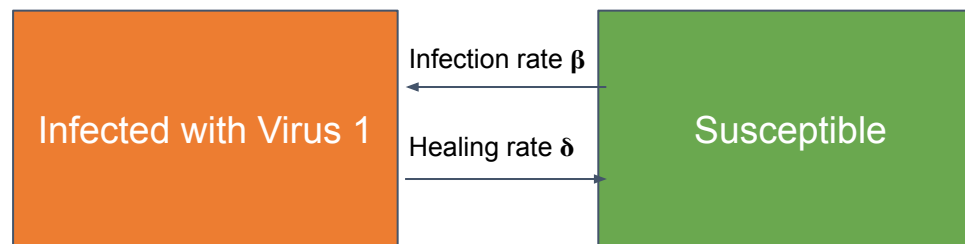
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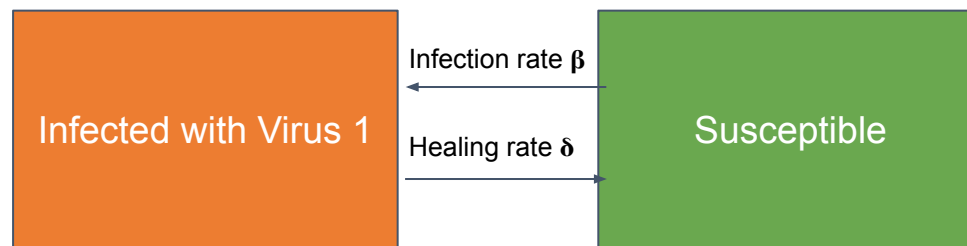
SIS Framework - Individual Setting

- How will people behave when they recover?
- We adopt the susceptible-infected-susceptible model: Individuals become susceptible again after recovery.
- Infection and healing rates are properties measured from the virus itself.



SIS Framework - Individual Setting

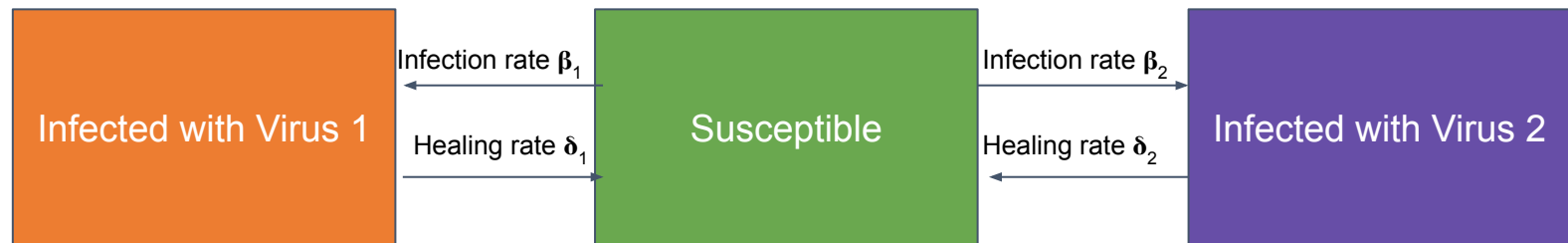
- How will people behave when they recover?
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BUT...

SIS Framework - Individual Setting

- What if there are two (competitive) viruses involved?
- An individual can only be infected with **up to one** virus.
 - e.g., dengue virus VS zika virus
- Infection and healing rates **specific to virus**



SIS Framework - Network Setting

- **Node** represent the **group of people in cities**
- **Edges** represent **Strength of connectivity** between nodes

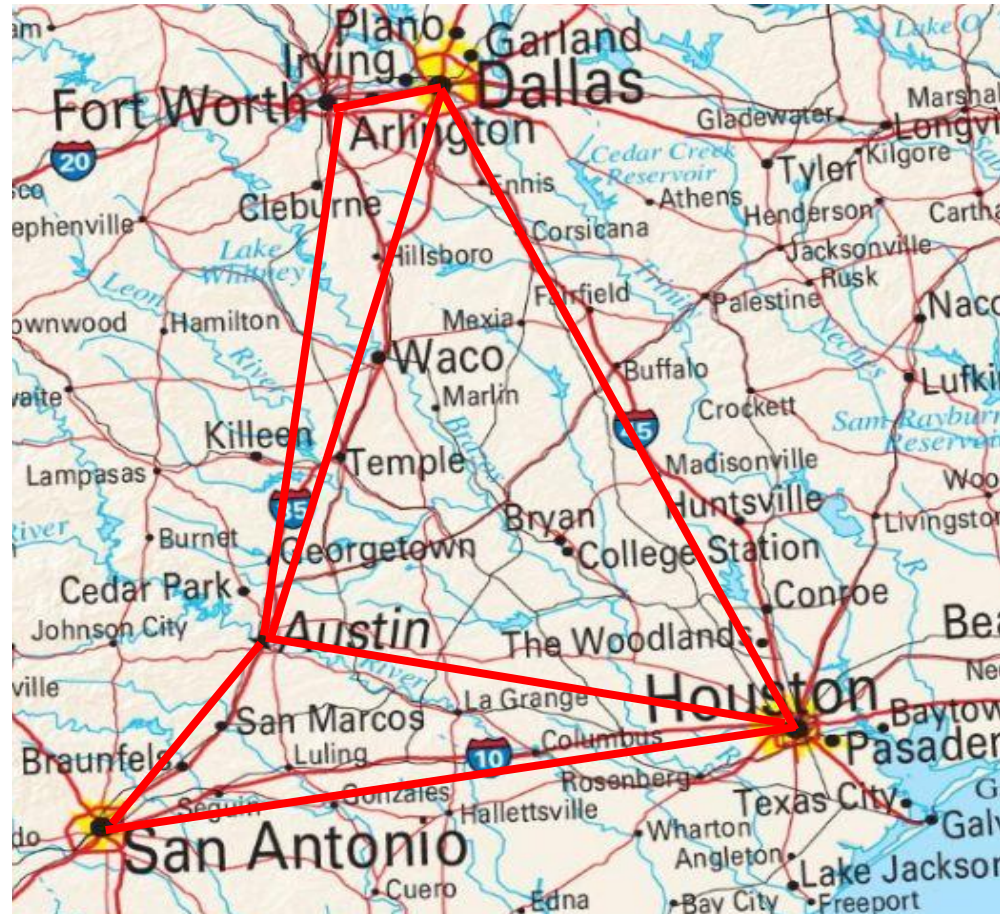
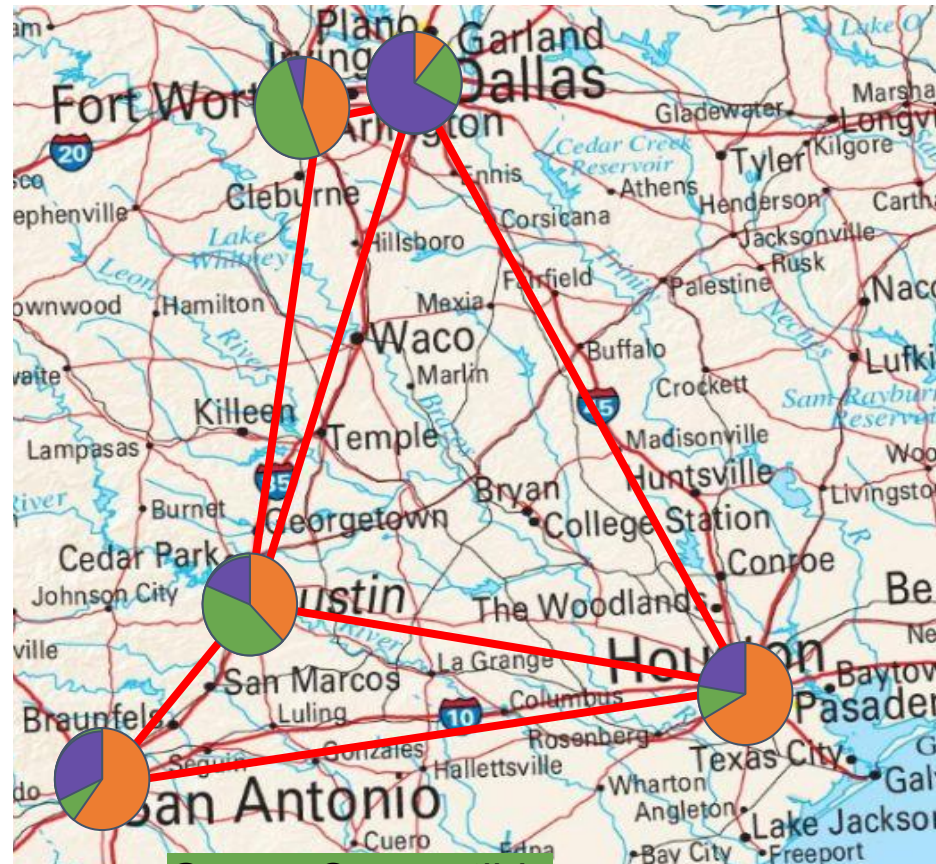


Image from: <https://ontheworldmap.com/usa/state/texas/texas-highway-map.html>

SIS Framework - Network Setting

- For each node (population), a **portion** is susceptible, the rest are infected
- Each portion of infected population will try to **'infect'** its neighbors in with **their respective virus** the next time-step!



Green: Susceptible

Orange: Infected with virus 1

Purple: Infected with virus 2

Image from: <https://ontheworldmap.com/usa/state/texas/texas-highway-map.html>

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SIS Framework - Network Setting



Model Setup

The Single Virus case

- β_i, δ_i denote the infection and healing rates of node i , respectively
- The model captures $x_i[t] \in [0, 1]$, the portion of the population that's infected at node i , at time t .

$$x_i[t+1] = x_i[t] + h \left((1 - x_i[t]) \left(\sum_{j=1}^n \beta_i a_{ij} \right) - \delta_i \right) x_i[t]$$

Diagram illustrating the Single Virus case model equation:

- $x_i[t+1]$: Infection level at time $t+1$
- $x_i[t]$: Infection level at time t
- h : Sampling parameter
- $(1 - x_i[t])$: % of Susceptible at time t
- $\sum_{j=1}^n \beta_i a_{ij}$: Infection Rate (where β_i is Infection Rate and a_{ij} is Edge weight between node i and j)
- δ_i : Recovery Rate
- $x_i[t]$: Infection level at time t

The Competitive Bivirus case

- In the Bivirus case, we consider the portion infected with virus 1, 2 denoted $x_i^1[t]$, $x_i^2[t]$, respectively. β_i^1 , δ_i^1 , β_i^2 , δ_i^2 defined similarly.

$$\left\{ \begin{array}{l} x_i^1[t+1] = x_i^1[t] + h \left((1 - x_i^1[t] - x_i^2[t]) \left(\sum_{j=1}^n \beta_i^1 a_{ij} \right) - \delta_i^1 \right) x_i^1[t] \\ x_i^2[t+1] = x_i^2[t] + h \left((1 - x_i^1[t] - x_i^2[t]) \left(\sum_{j=1}^n \beta_i^2 a_{ij} \right) - \delta_i^2 \right) x_i^2[t] \end{array} \right.$$

Diagram illustrating the competitive bivirus model equations and their components:

- Left side (Infection level of virus k at time $t+1$):** $x_i^1[t+1]$ and $x_i^2[t+1]$ (red boxes).
- Right side (Infection level of virus k at time t):** $x_i^1[t]$ and $x_i^2[t]$ (orange boxes).
- Sampling parameter:** h (yellow box).
- % of Susceptible at time t :** $(1 - x_i^1[t] - x_i^2[t])$ (green box).
- Infection Rate of virus 1:** β_i^1 (blue box).
- Edge weight between node i and j :** a_{ij} (purple box).
- Recovery Rate of virus k :** δ_i^1 and δ_i^2 (blue boxes).
- Infection Rate of virus 2:** β_i^2 (blue box).

What happens “in the end”?

$(t \rightarrow \infty)$



Image from: <https://www.pngegg.com/en/search?q=question+mark>

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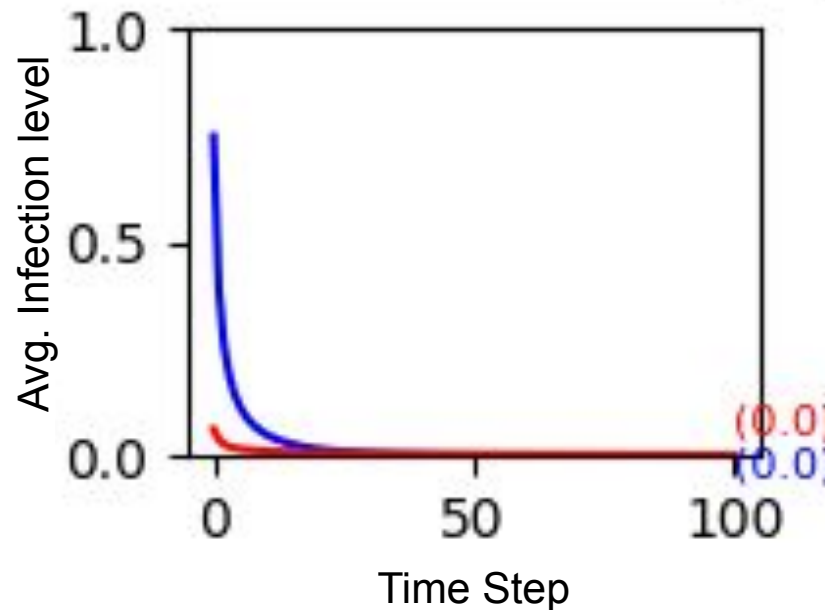
Conclusion

Theoretical Outcomes

- Three possible outcomes (equilibria):
 - Disease Free Equilibrium: $\rho(I - hD^k + hB^k) \leq 1$ for both viruses

spectral radius: largest absolute value among its eigenvalues

“Both viruses die”

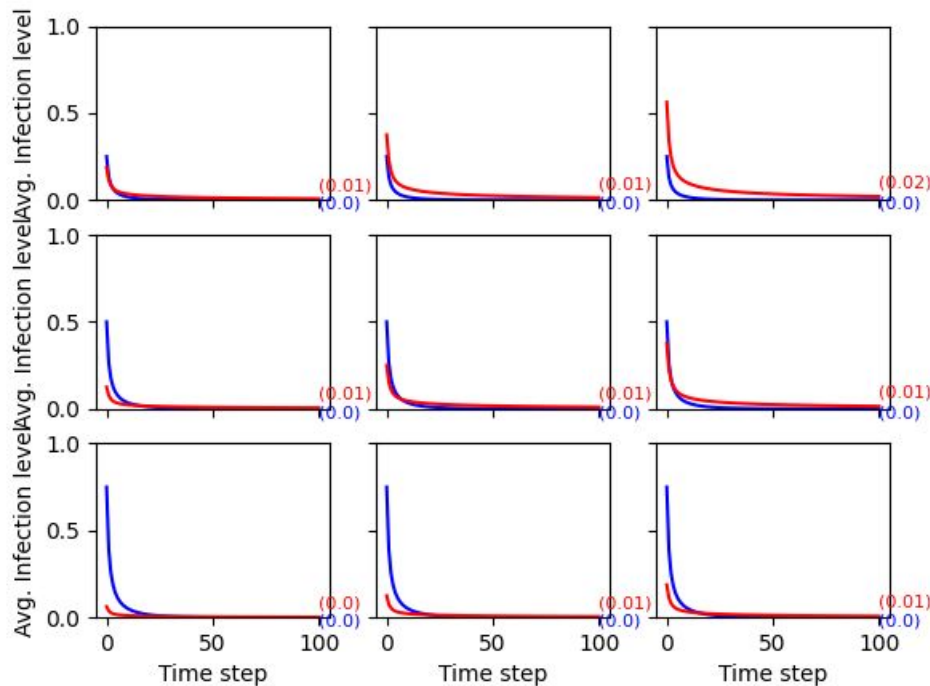


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Average Infection levels across 20 pop. nodes VS Time

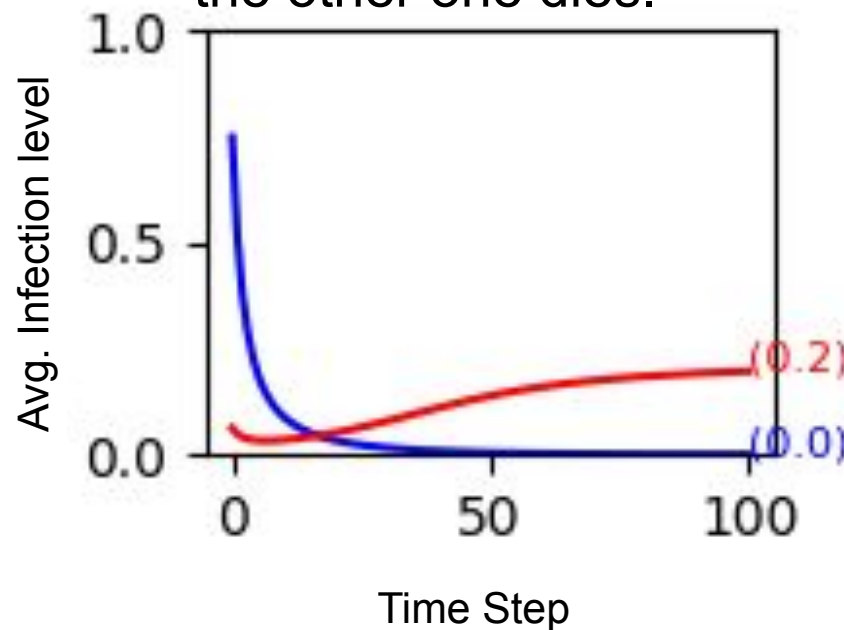


“Both viruses die”

Theoretical Outcomes

- Three possible outcomes (equilibria):
 - Endemic Equilibrium: $\rho(I - hD^k + hB^k) > 1$ for **dominant virus**, $\rho(I - hD^k + hB^k) \leq 1$ for **diminishing virus**

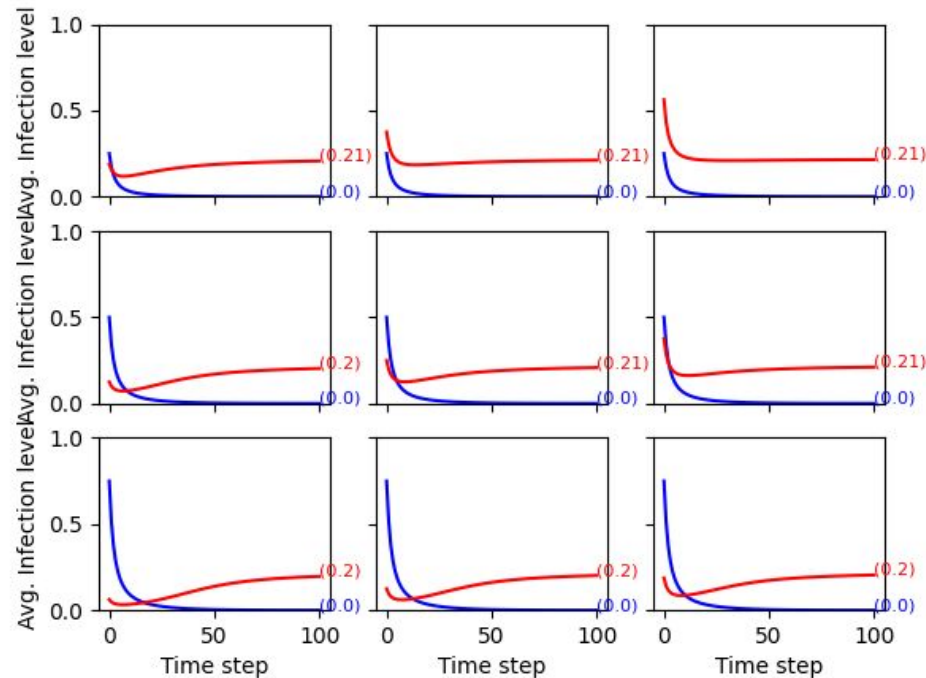
“One virus survives,
the other one dies.”



Theoretical Outcomes

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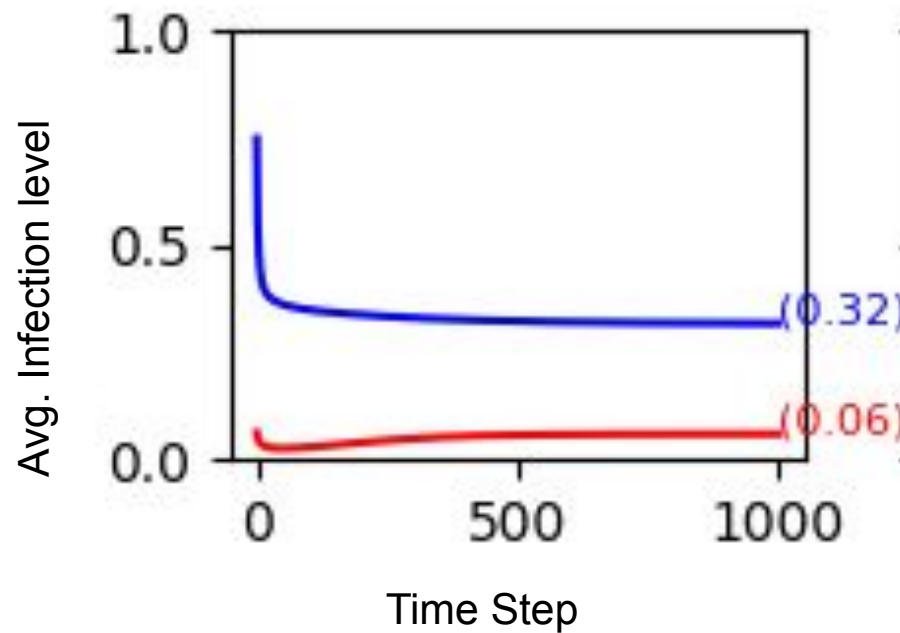


“One virus survives,
the other one dies.”

Theoretical Outcomes

- Three possible outcomes (equilibria):
 - Coexistence Equilibrium: $\rho(I - hD^k + hB^k) > 1$ for both viruses

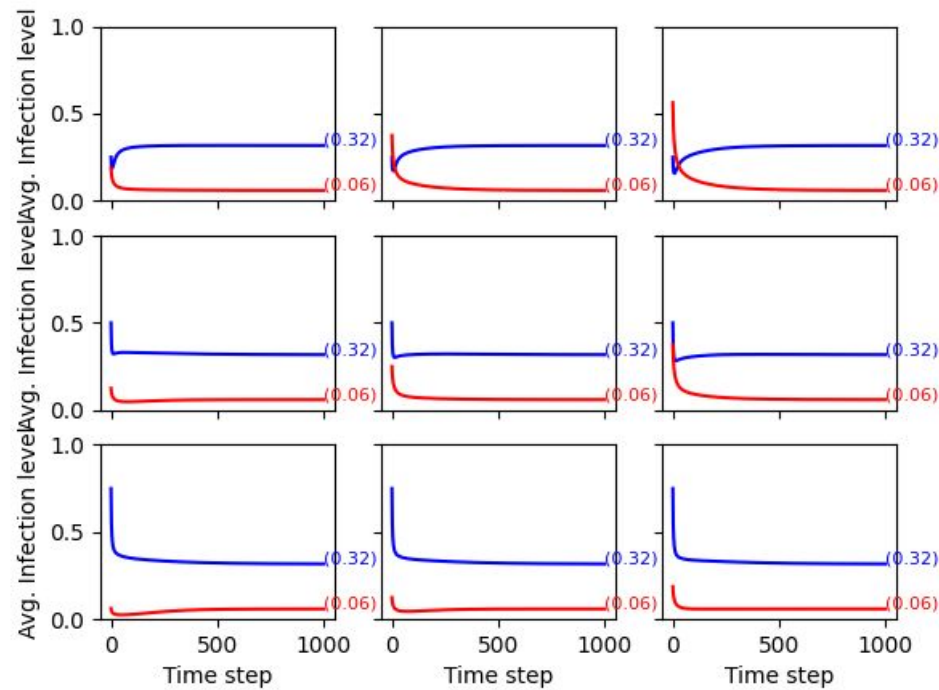
“Both viruses survive”



Theoretical Outcomes

- Three possible outcomes (equilibria):
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Average Infection levels across 20 pop. nodes VS Time



“Both viruses survive”

Real-world Simulations

Real-world Network Simulations

- Two real-world datasets were used
 1. The Massachusetts County Travel Graph
 2. The New York City Taxi Graph

Massachusetts County Travel Graph(Demo)

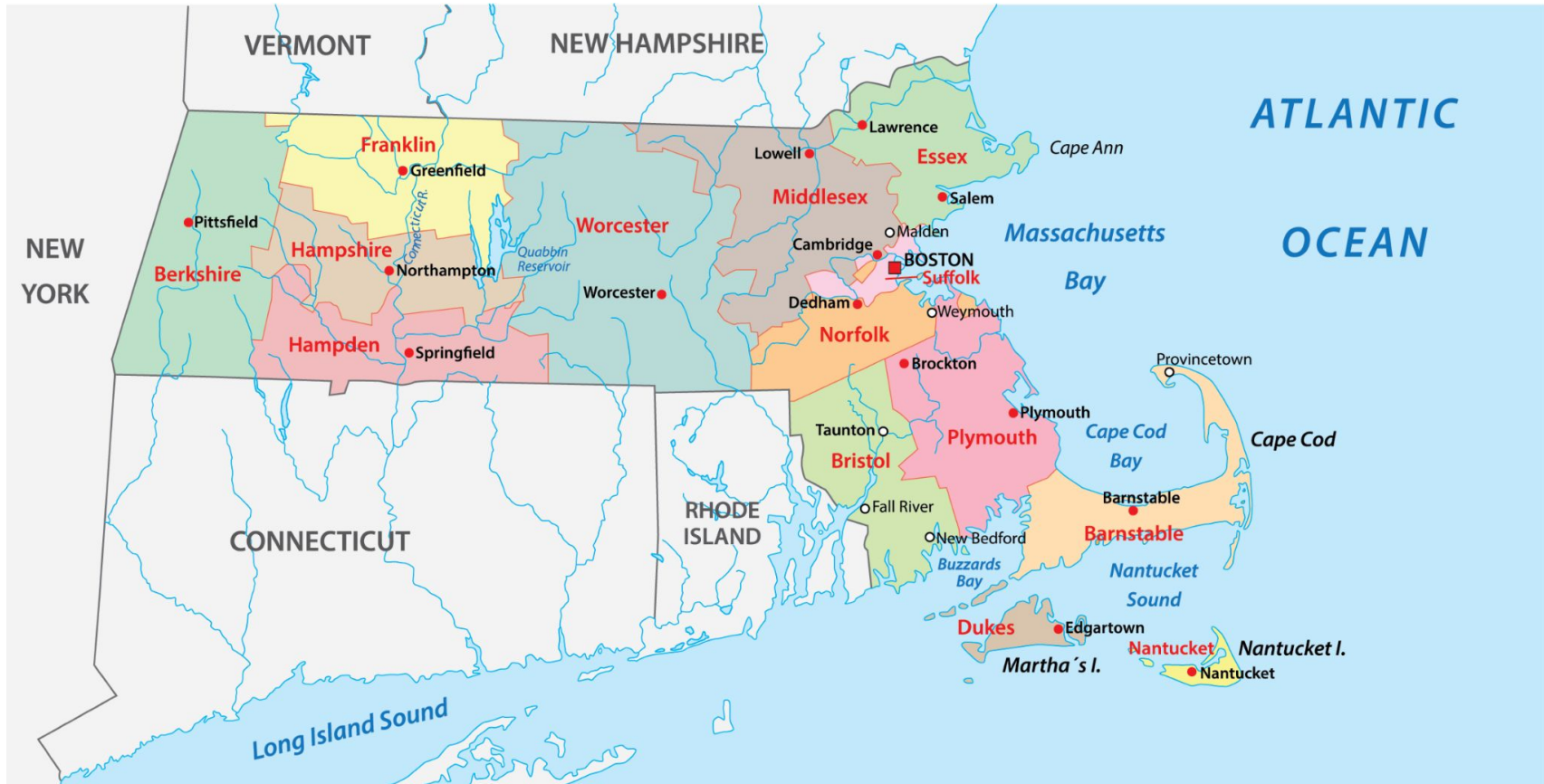


Image from: https://www.reddit.com/r/massachusetts/comments/171vp2h/if_the_massachusetts_counties_were_to_be_changed/

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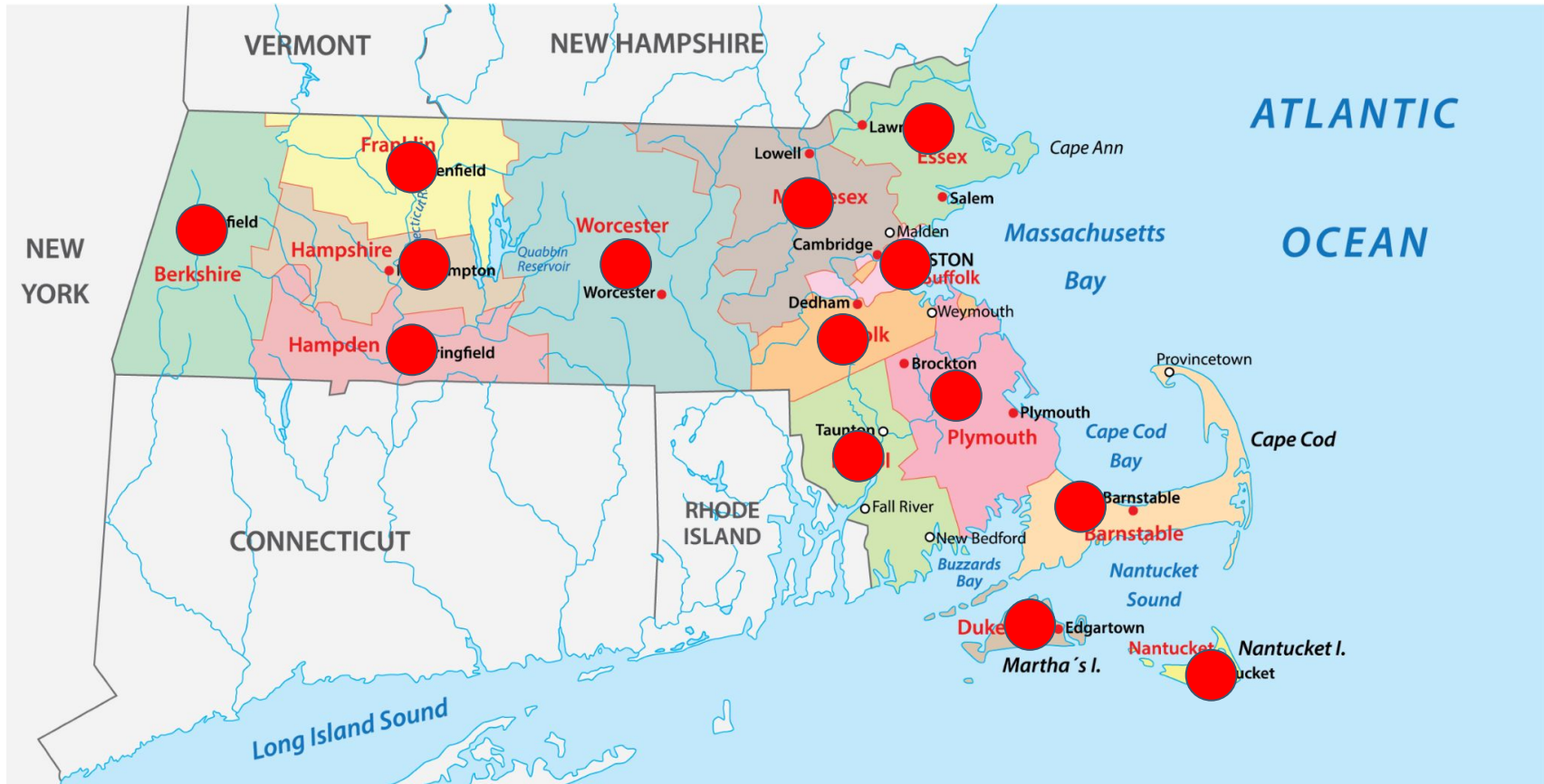


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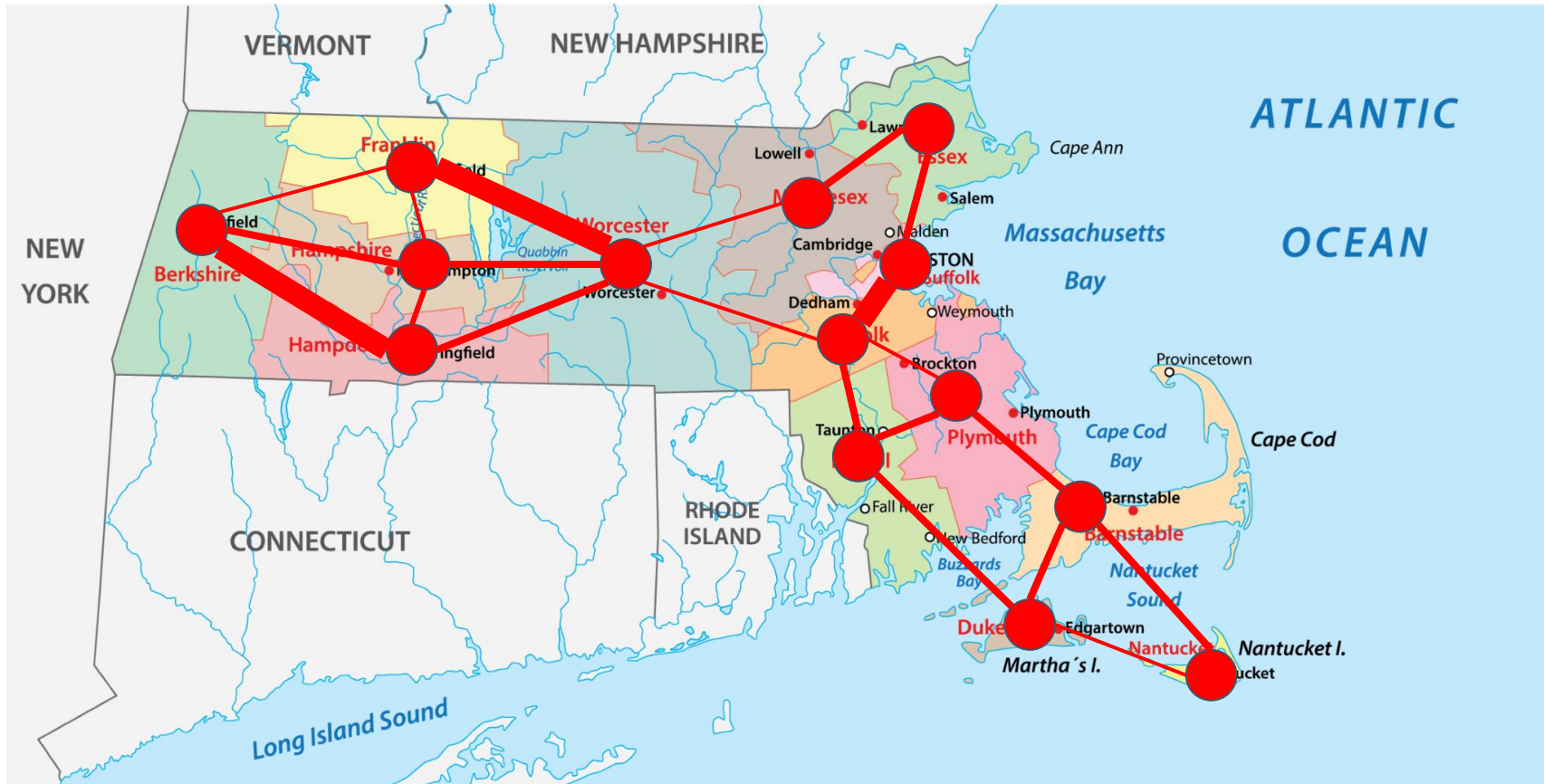
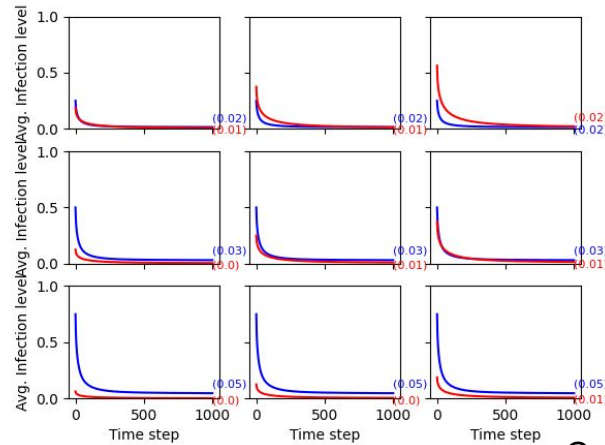


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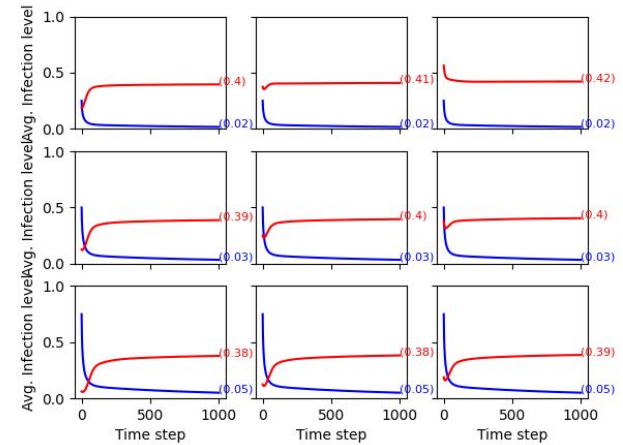
Disease-free Equilibrium

Average Infection levels across 20 pop. nodes VS Time



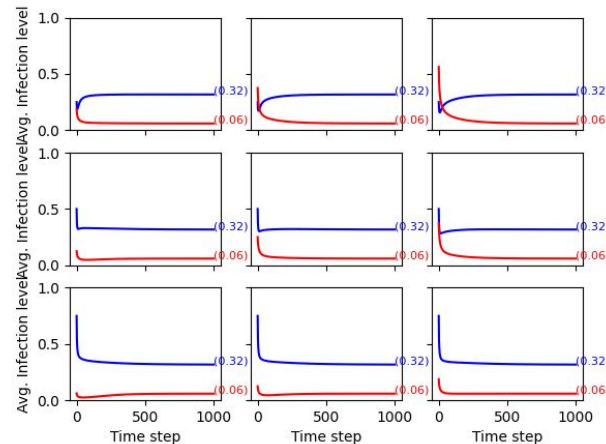
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Coexistence Equilibrium

Average Infection levels across 20 pop. nodes VS Time



The New York City Taxi Graph(Partial)

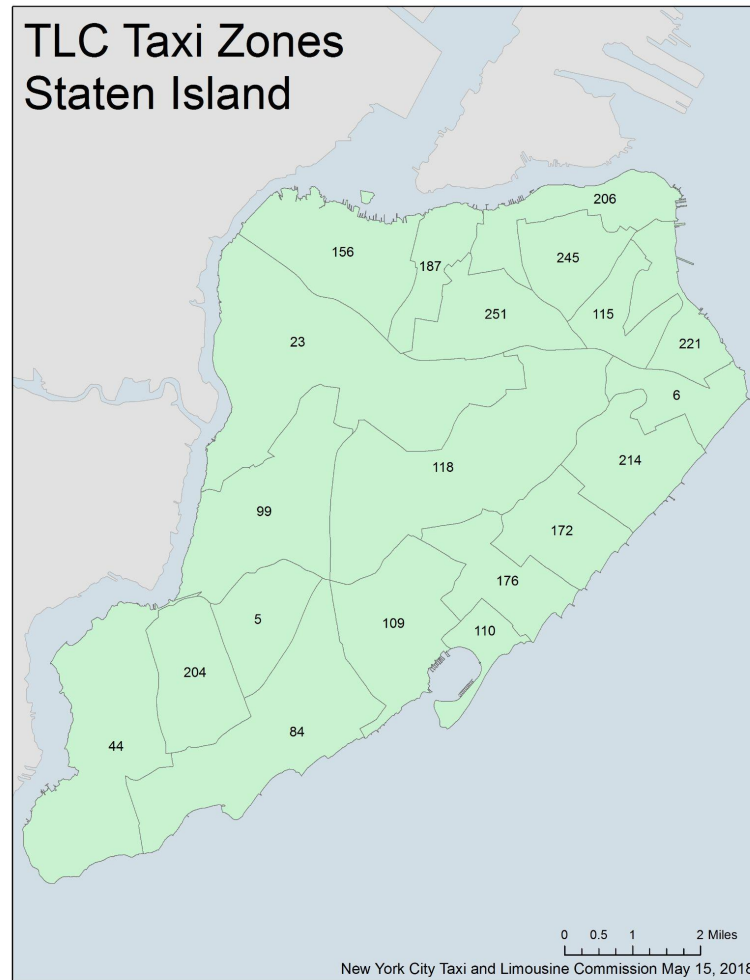


Image from: <https://www.nyc.gov/site/tlc/about/tlc-trip-record-data.page>

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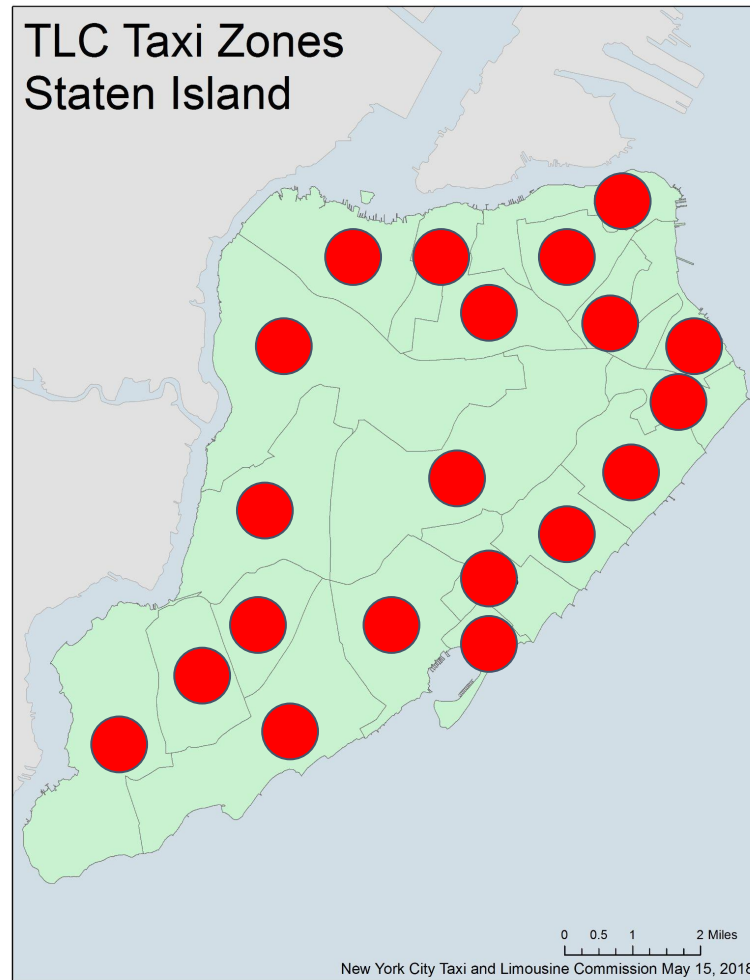


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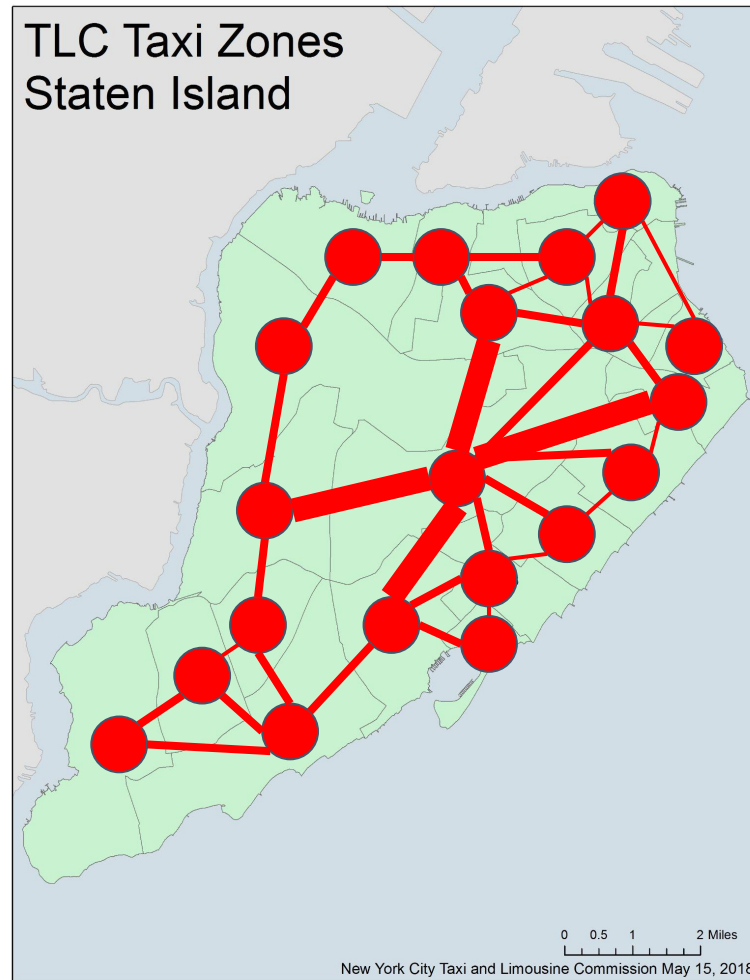
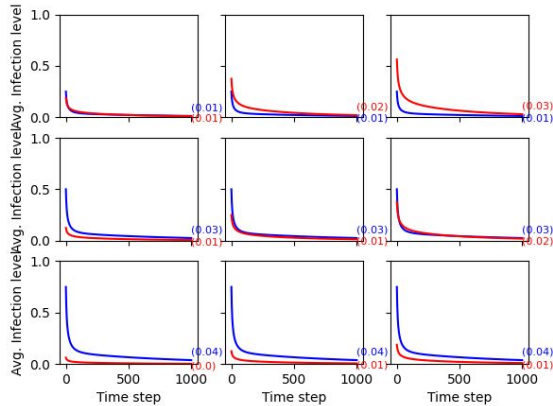


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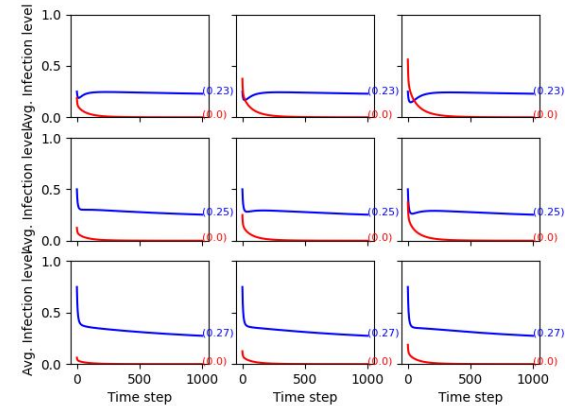
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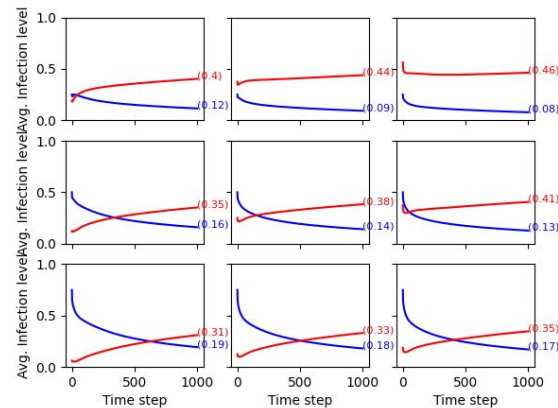
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Average Infection levels across 20 pop. nodes VS Time



Coexistence Equilibrium

Average Infection levels across 20 pop. nodes VS Time



Conclusion & Future Work

- We verified the theoretical predictions regarding the various outcomes, under their respective conditions.

In the future, we can...

- Incorporate time-varying network dynamics for even **more realistic** predictions.
- Performing numerical simulations on larger, more complex datasets can yield **more convincing** results.



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