Epidemiology Modeling for Compliance Graph Analysis

CS-7863: Scientific and Statistical Computing

Final Presentation

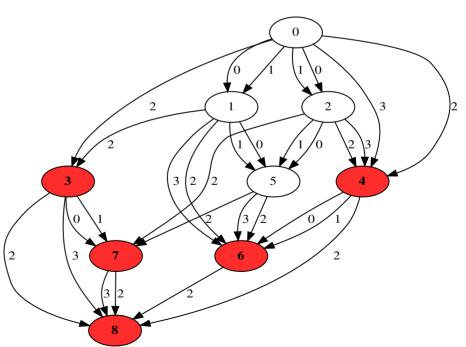
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Introduction to Compliance Graphs

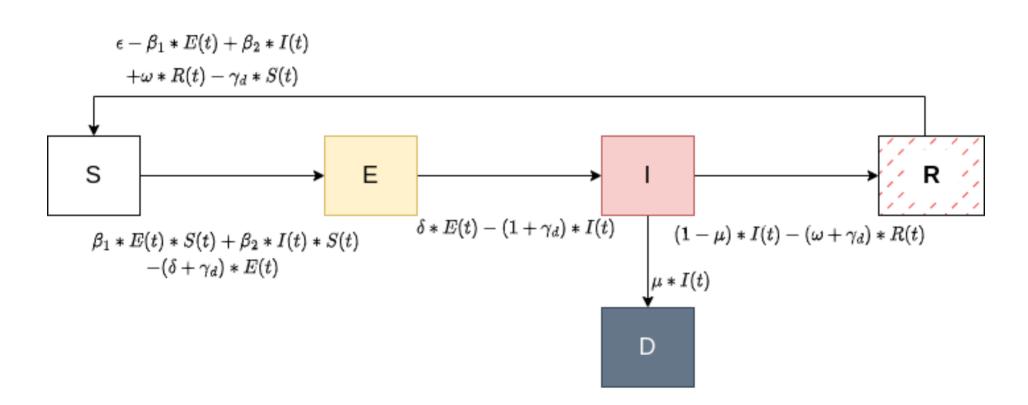
Determine all possible ways systems may fall out of compliance

Directed Acyclic Graph

- (DAG)



SEIRDS Epidemiology Model



Model Parameters

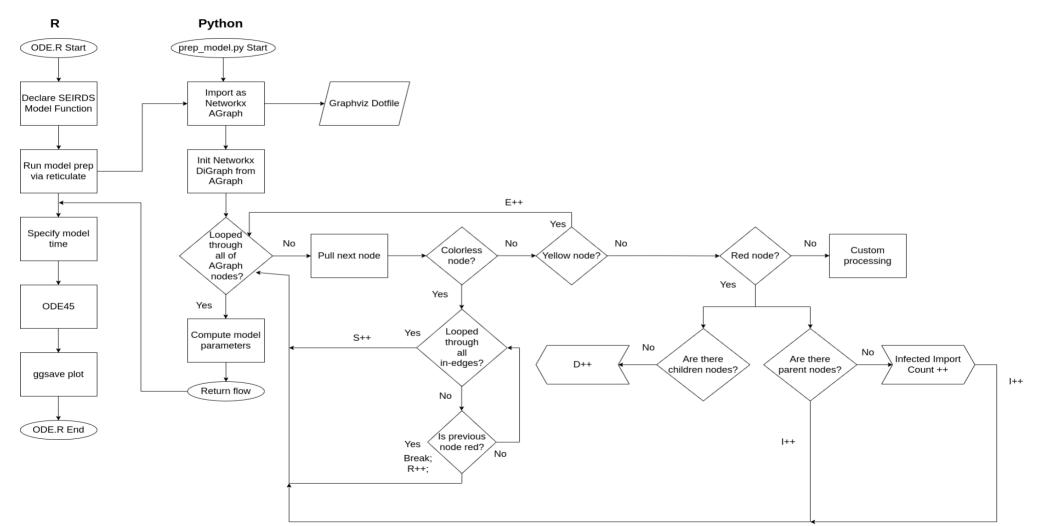
```
\beta = \text{rate of infection}
\delta = \text{symptom appearance rate}
\gamma_R = \text{recovery rate}
\gamma_D = \text{death rate}
\mu = \text{fatality ratio}
\epsilon = \text{infected import rate}
\omega = \text{waning immunity rate}
```

Model Assumptions

 Infection rate from Susceptible contact with Exposed compartment is equal to contact with Infected compartment.

No increase in population.

Program Flow



Results (1)

Initial Model Compartments:

S: 62

E: 8

I_R: 288 I D: 1

R: 36

D: 1

Model Parameters:

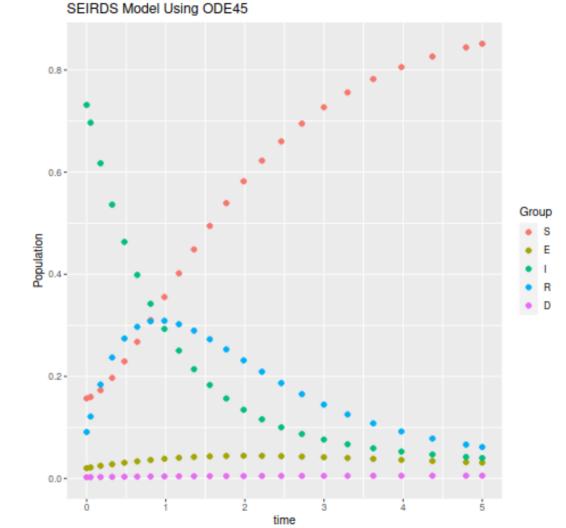
beta: 0.4316

delta: 1

gamma_r: 0.0911 gamma d: 0.0025

mu: 0.0035 epsilon: 0.0

omega: 1



Issues

- Model parameters and compartments are derived from the same network
- Model parameters do not account for the topographical relationships of the network
 - EX: Most 'recovered' nodes are upstream. Almost all (~225) 'infected' nodes are downstream, and don't recover.
 - SEIRDS model does not account for this for future time steps

Trivially Weighted Networks

• For each node: $EdgeWeight = \frac{1}{\text{num edges}}$

• Infection rate and recovery rate now take edge weightings into account .

Results (2) SEIRDS Model Using ODE45

Initial Model Compartments:

S: 62 E: 8

I_R: 288

I_D: 1

R: 36

D: 1

Model Parameters:

beta: 0.2624

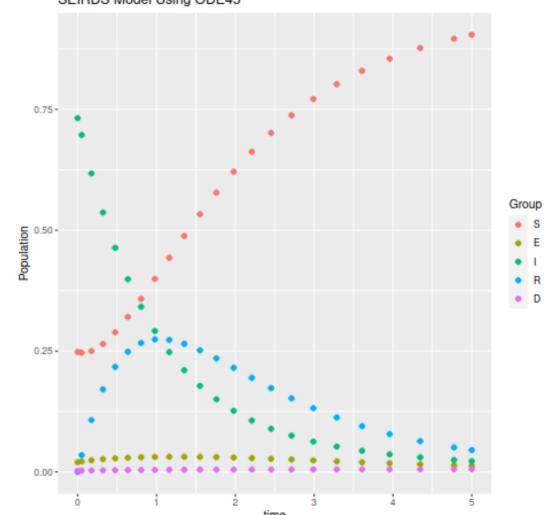
delta: 1

gamma_r: 0.0459 gamma d: 0.0025

mu: 0.0035

epsilon: 0.0

omega: 1



Future Work

- Additional probability parameters for likelihood of recovery
- Multi-strain epidemiology modeling (modeling started)
 - S and R compartments remain the same
 - Split E and I compartments into E₁, E₂, ... I₁, I₂ ...
- Parameter derivation over time (subgraphing started)
 - Using networks that have time-steps, create subgraphs
 - For each subgraph, compute parameters
 - Compartments based on entire network