

Simulating covid spread at William Jewell and the effects of testing and intervention

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Outline

- Goal
- The Network
- The Simulation
- The method of intervention
- The Results
- Discussion

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- Scaled network density is a ratio of edges to nodes represented by $d = \sqrt{m}/n$
- We use scaled network density because it allows us to compare networks with different size n .

Goal

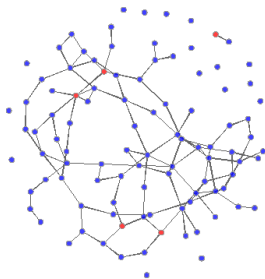
- Simulate the spread of Covid-19 at William Jewell and the effects of testing and quarantines

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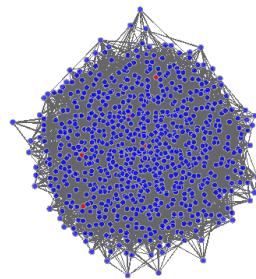
- Simulate the spread of Covid-19 at William Jewell and the effects of testing and quarantines
- We were interested in finding a reasonable testing rate that would prevent the peak of infections greater than 150.

The Network

- The network is an Exponential Random Graph network



(a) nodes = 100

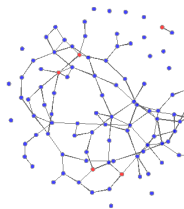


(b) nodes = 750

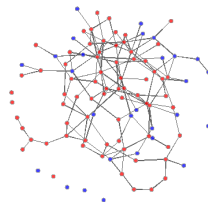
Figure: Blue is susceptible, Red is infected, Green is recovered

The Network

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- The network is a Stochastic network, meaning it changes over time.



(a) time step = 1

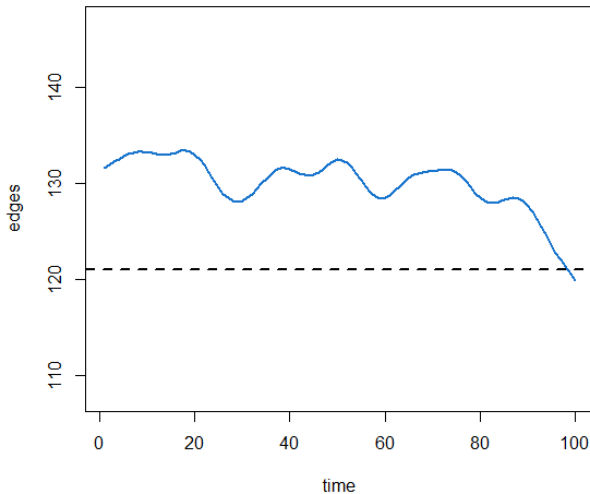


(b) time step = 30

Figure: Blue is susceptible, Red is infected, Green is recovered

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Edges vs Time



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 - ★ passengers in an airport $d = .11$
 - ★ network of co-workers $d = .13$

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Intervention

- ① We first tested a proportion of the population.
- ② If the student tests positive they are sent to quarantine and the probability of transmission was set to zero.
- ③ Once the student recovered, they were let out of quarantine.

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- No accounting for vaccinations.

Results

First we check that the networks are forming as we expect, for $d = .11$ and $n = 750$ we expect $m \approx 6800$.

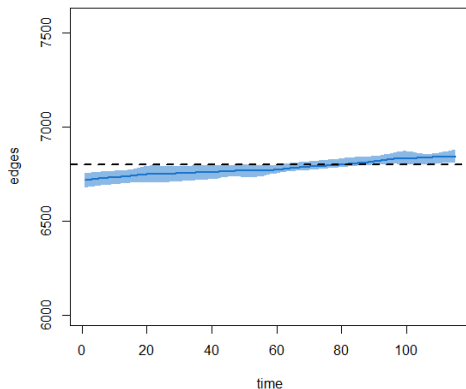


Figure: Total number of edges versus time

Results

We'll take a look at the lowest and highest testing rate graphs

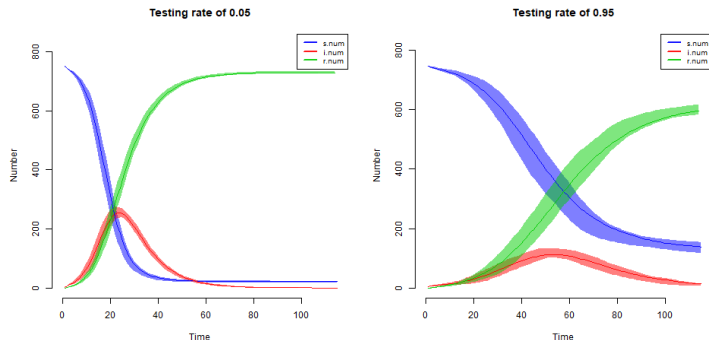
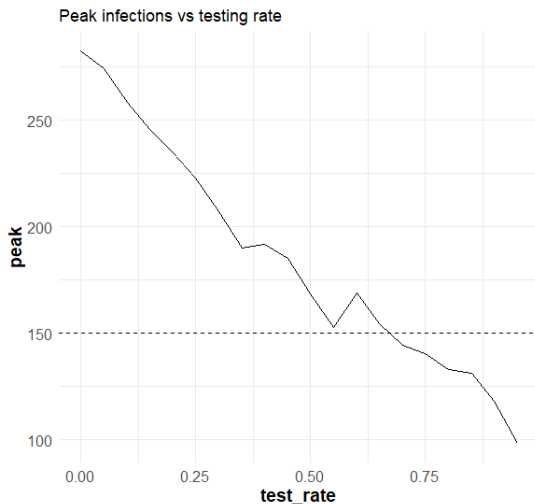


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We can see a distinct difference in the height of peak infections and at what time it occurs.

Summary of results



We can see that, for the target of 150, a testing rate of 50 percent might be viable, but a testing rate closer to 70 percent would be much more reliable.

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- With more information would could also test the necessary level of Covid polices in common areas.

References

- Guy Melançon. Just how dense are dense graphs in the real world? A methodological note. BELIV 2006: BEyond time and errors: novel evaluation methods for Information Visualization (AVI Workshop), May 2006, Venice, Italy, pp.75-81. fflirrm-00091354f
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- Lelieveld J, Helleis F, Borrmann S, Cheng Y, Drewnick F, Haug G, Klimach T, Sciare J, Su H, Pöschl U. Model Calculations of Aerosol Transmission and Infection Risk of COVID-19 in Indoor Environments. International Journal of Environmental Research and Public Health. 2020; 17(21):8114. <https://doi.org/10.3390/ijerph17218114>