

Agent-based models of non-pharmaceutical interventions for epidemic control

Robert Brian Milligan and Supervised by Julian Garcia Gallego & Buser Say

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Abstract

Currently there is much interest in modelling diseases to understanding ways governments, workplaces and other decisions makers may seek to control the spread of COVID-19. Many various computational models of disease spread exist and this paper builds upon an network agent based model and changes the ways in which agents can comply or not comply with various actions. Both non-strategic and strategic methods of agent compliance were looked into and it was found that this suggests that

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1 Introduction

Mathematical and Computational models are important in preparing policies to deal with pandemics. These models typically do not incorporate behaviour or if they do, do so in simple ways such as compartmental models that incorporate "aggregate states" this was implemented as having 6 different behavioural scenarios which would change based on the time of the simulation or having a certain threshold of positive test, positive cases or deaths per day [1]

Cite a few examples and explain them.

This paper investigates the modification of an existing model disease spread produced by Ryan McGee and used a part of various peer reviewed research articles [4] [5]. I have looked at applying the model to the situation of COVID-19 in childcare and involves creating both a non strategic and strategic behavioural models. The Behaviour relates to compliance to various actions the agent can choose to do such as doing an additional COVID rapid antigen test if they have symptoms of the virus. The non strategic model looks at giving agents a fixed cost of complying with requested actions with a reward based on a mix of local situations such as if they are in close contact with an agent who is in isolation and a global situation such as the percentage of the group that reported a positive test within the last 2 weeks. The strategic model looks at agents where they know how all other agents will act and use this information to decide if it is in their interest to comply.

Add a few more paragraphs: go slowly.

The rest of this paper is organised as follows. Section 2 describes the basic model, Section....

2 Description of Model

The model used is an Agent Based model where agents can belong to one of multiple compartments as well as being in a state of isolation or not in isolation. Agents progress through states with the only sink node being Recovered and Fatality the states of Susceptible, Exposed, Pre-Infectious, Infectious-Asymptomatic, Infectious-Symptomatic, Hospitalised, Fatality and Recovered exist. All these except Hospitalised and Fatality can have agents in a mirror state where they are also isolated, meaning they cannot acquire the disease or spread it to any other agents in the network.

Describe the model at large, what compartments, say it is an agent based model, etc.... cite the paper and explain how they use their model.

- During the day agents can spread the contagion to each other and can progress through the stages of the disease if they have it. They additionally have the choice to participate in contact tracing.
- Agents will also be asked to isolate for one of six reasons. They or a group member develop a symptomatic case, returns a positive test or is told they are a close contact through using contact tracing
- These systems can effectively be disabled by overriding the compliance for them to be a large negative number, for example compliance with contact tracing can be set to -10 to disable the use of that system and this done to limit the model to a smaller amount of variables which compliance can affect.

The model has a variety of limitations including

- Having all agents always test on the same day, so if semiweekly is chosen, all agents will be asked to test on Monday and Thursday
- the network itself is unchanging, but this is not necessarily bad

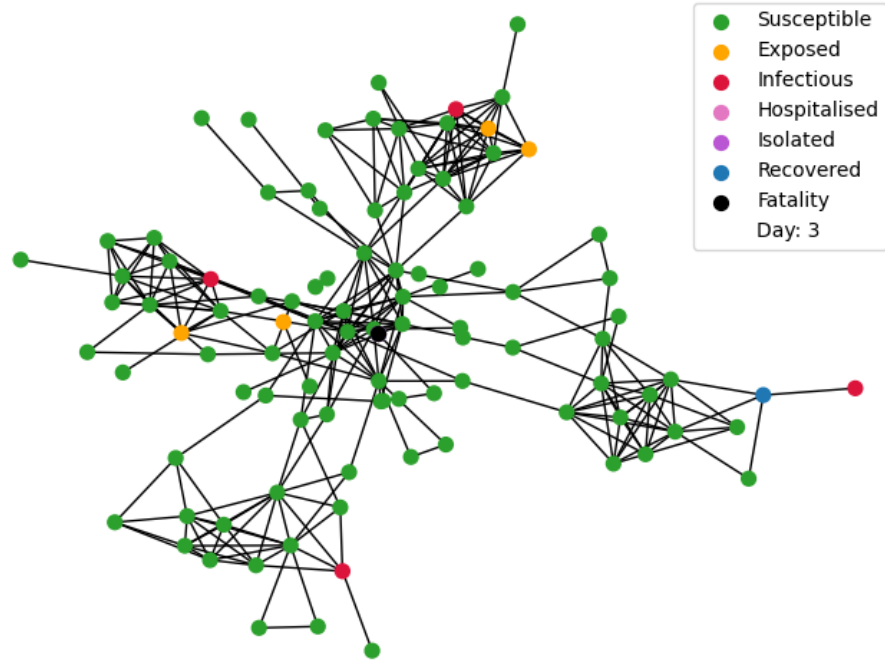


Figure 2: An Example Netowrk

Example of the network of 100 agents To simulate the childcare scenario they are split into 5 groups of about 20 agents each with high connectivity inside the group and low connectivity between the groups The agents current state is shown by the nodes colour and the edges are that agents close contacts which the disease can spread easiest though the population

2.2 Description of Model Parameters

The Current Modifications of the model relate to allowing 10 of the Model Parameters that relate to Compliance to be dynamically updated each day dependent on a given rule. The 5 most relevent of these being.

- What proportion of agents take a test immediately as a result of having symptoms
- What proportion of agents will do surveillance
- What proportion of agents will isolate given they have a symptomatic case
- What proportion of agents will isolate given a positive result from a test
- What proportion of agents in a group isolate given one of them has a positive result from a test

A simple model might only use a few of these compliance parameters such as and is how the parameters work in the base model

Symtomatic Testing Rate Compliance	50%
Surveillance Testing Rate Compliance	50%
Isolation from Positive Test Compliance	100%

2.3 Description of Compliance In the Model

Compliance can be set a variety of ways in the modified model but in this paper 3 are used

- The default setting where agents are given an initial value for compliance and is unchanging e.g. 50% are set to comply and will always do so
- The non strategic model where agents are given an initial compliance and can become more compliant depending on the network situation , their local situation or a mix
- The strategic model where agents utilise knowledge of the amount of agents in the network who will comply to make a decision to comply or not

3 Childcare Model and Benchmark Tests

- Parameter justification
- test false negative rate 0.36 [6]
- R0 mean of Omicron is 9.5, delta is 5.4 [2]
- base compliance for behaviour is set at levels of 0.5 and 1, to represent 50% of agents and all agents complying
- 100 agents in model split into 5 groups with high interconnectedness within the group and low connections between groups
- compliance increases for 2 actions, symptomatic test and regular interval surveillance tests
- The cost of compliance is fixed and there is a reward based on if connecting agents are isolated, hospitalised or a fatality
- from looking at early results since only 2-3 tests a week there is a delay and the higher the r0 the fewer chances they can have to change their mind about compliance

- the R_0 of 9.5 or 5.4 may be too high as it assumes children spread the disease at the same rate as the general population as well as all disease spread occurring in childcare which it probably is not. Therefore 2 other R_0 cases of 3 and 2 are used as they may more accurately capture a real world scenario where not all transmission is through the 1 childcare setting

The benchmark model is associated with one single large peak of disease spread

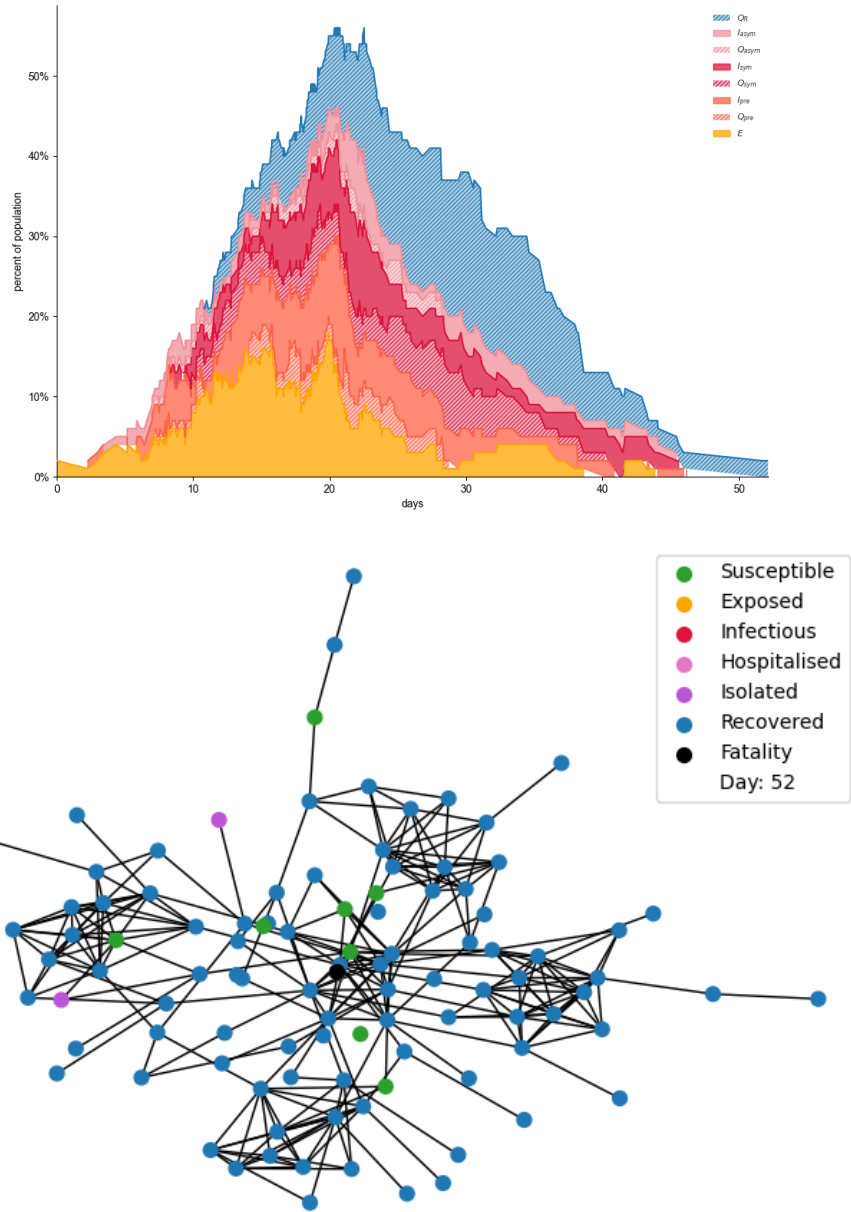


Figure 3: Benchmark Run with Unchanging Compliance at 50% for regular twice weekly surveillance testing and testing if symptomatic with the resulting final network. 92 of the 100 agents received the infection

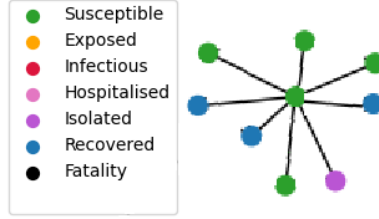


Figure 4: An Example Simplified Network

4 Non-Strategic Model

We have a fixed cost of compliance to an action and varying factors that can raise it. A mix of global and local behavioural factors can be added to make an agent more compliant. In the current build these are the known positive cases in the network in the past 14 days. The proportion of contact agents (those which share an edge) which are a fatality, hospitalised or in isolation. These values are taken away from the base cost and if the result is lower from a specified value the agent will be compliant to that action, otherwise they are not

For a simple example we have a situation for this central agent

Symtomatic Testing Rate Compliance	50%
Surveillance Testing Rate Compliance	50%
This agent's base aptitude	uniformly distributed (-0.1,0.1)
Base Cost of Compliance	0.5

$$\text{Compliance} = 0.5 - (5 * 1/10) - (4 * 0.02) + 0.3 = 0.22$$

In this case the agent is compliant it will test if they develop a symptomatic case immediately and will do surveillance testing as the compliance is now less than the threshold of 0.5 if the neighbouring agent that entered hospital returns a positive test, our agent will enter isolation if a positive test is returned the reward for agents can be based on a mix of both a local and global network situation as in the example runs

Of the 10 possible compliance parameters built into the existing mode, 3 are used these are

- immediately testing if the agent has a symptomatic case
- using regular surveillance testing
- isolating if a positive test is returned - set to always be true

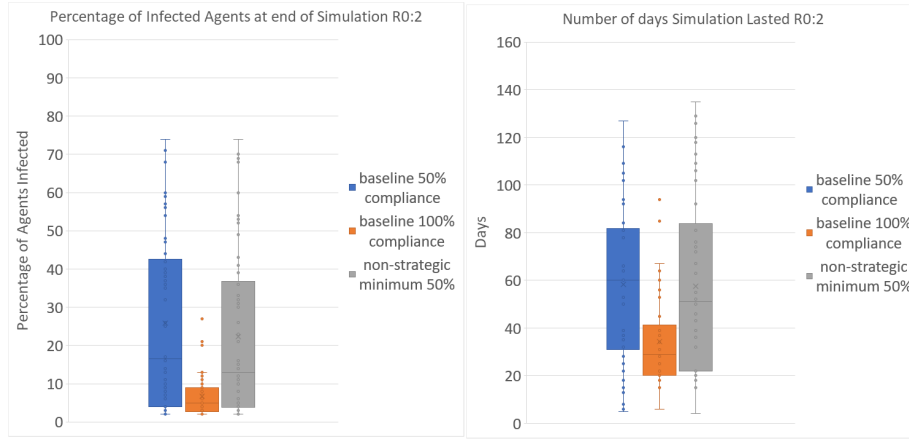


Figure 5: Test Results over R0 of 2

5 Analysis of Testing Compliance

To evaluate the result of the model we take the average of these two values over many runs

- What percentage of the population caught the contagion?
- How many days did it take for the outbreak to stop with 0 actual active cases?

We can analyse compliance by comparing the effect varying levels have on the length a contagion actively spreads and what proportion of the population becomes infected. Firstly a baseline can be set to explore the parameter space, then further tests done to see the effect having compliance change as a result of the current known spread of the contagion in the network.

Over 4 different R0 virus base reproduction rate was the model run with, over the different R0 levels the reduction in total cases reduced from a -2% improvement with an R0 of 9.5 to a 28% improvement with an R0 of 2. This is probably due to one of the aspects the model relating to the 14 day known positive cases recorded. With a low R0 in the system there are more chances for agents to test themselves before they pass on the disease and these reported cases incentivise even more agents to do symptomatic and surveillance testing. With extremely high R0 numbers it is observed that the non-strategic behavioural model has little effect on the total number infected. Across all cases it can be seen that the amount of time the simulation runs for is tied with how many agents get infected. Both the static values of compliance and non-strategic behavioural model follow this trend.

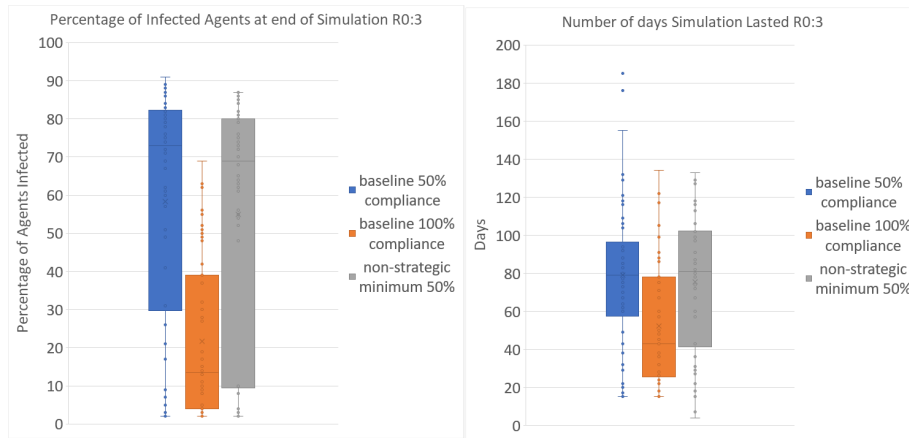


Figure 6: Test Results over R0 of 3

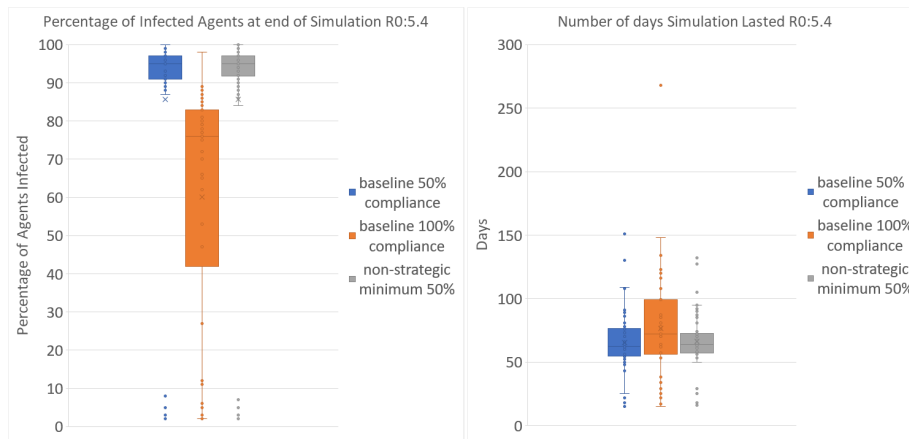


Figure 7: Test Results over R0 of 5.4

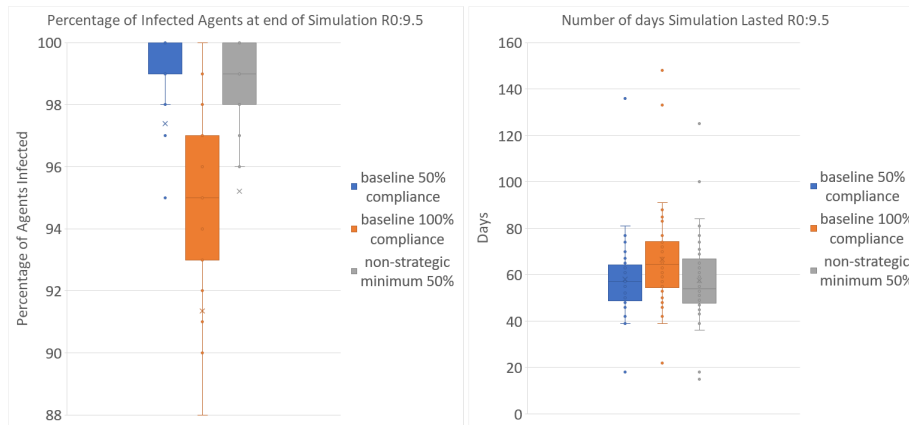


Figure 8: Test Results over R0 of 9.5

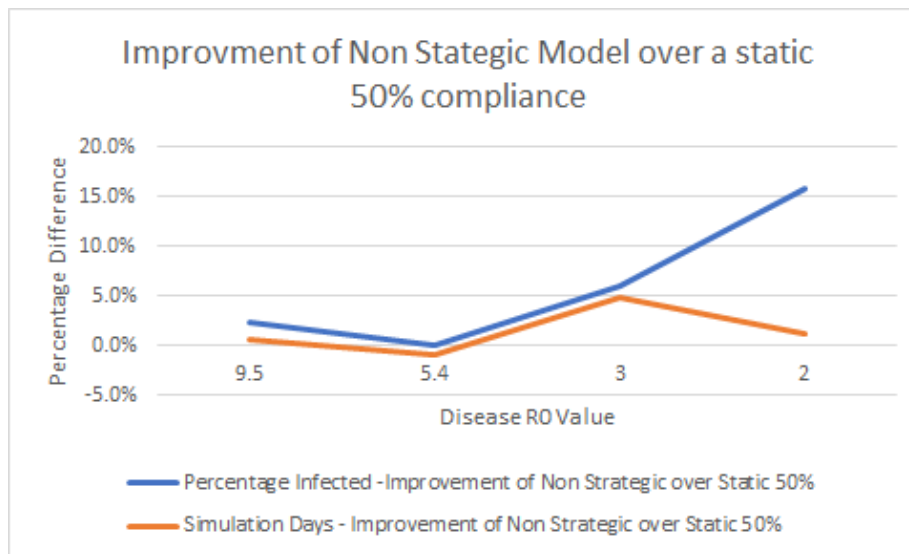


Figure 9: Test Results Comparing the Results of R0 values

6 Strategic Model

The view of benefit and cost can be grouped into 3 categories. There is a level at which people will not contribute as they find the act pointless as they know near no agents will comply

- 1. Global Situation of the model e.g. changes over time but is identical across all agents e.g. number of active cases or rate of change in spread rate, could place 2 infections in first week and compare each week to the last week, rate of change week on week $[0,0,0,0,0,2]$ if the next week 4 cases are spread, the rate would be 2
- 2. Structural Layout of the model, identical over time but varies between each agents e.g. number of close contacts or number of agents one can reach in 1 step, or could be 2 steps, the more agents one is in contact with the greater the benefit
- 3. Structural Situation of the model, changes over time and varies for each agent e.g. number of close contacts in a particular state or group of states like hospitalised, fatality, isolated etc, if a close contact is in one of those states the benefit to compliance is greater

for the following tests these can be used to create a personal benefit for each agent

it could be possible to use of a mix of these situations and in real life that may be the case with peoples decision making process

Structural layout of the network, for example number of close contacts an agent has

global situation for example the number of reported cases in the last 2 weeks, the number of actual cases in the last 2 weeks, the growth of the disease spread rate

local situation for example the number of neighbouring agents that are a fatality, hospitalised or isolated

benefit depends on change on rate of disease, number of agents around you that are hospitalised, fatality or isolated

one model is exclusively global, one is exclusively local

number of infections of those connected to me, 1 or 2 degrees, that are exposed or infectious

number of infections globally

agents will know the total percent of agents that are compliant and use that to work out how compliant they are

one universal curve for compliance, would this mean it has to be based off global factors?

do I actually need to mathematically work out a curve and its intercepts

every day am I updating this curve?, can agents move to exactly where they want, or is it a slow process, they can only move X amount every day?, a limited amount makes sense as that would match the curves in other papers

like thomas? Or would you see where the maxima is on that day and instantly assign the end that value?

come up with a equation (payoff function), x parameter to produce a curve like in the papers parameters go in , map plot, make it a notebook

does it change based on rate of change or on number of cases - if its global every curve would be the same if its local every curve is different for each agent notebook

how do curves change with infection numbers/rate

keep adding to latex doc.

once its a good function integrate it

Childcare first introduces compliance Describe the model

When you comply you pay a cost

Growth of disease rate

7 Results

8 Discussion

9 Conclusion

A Appendix Section

Removed parameter section

- TRACING_COMPLIANCE_RATE (what proportion of agents comply with contact tracing)
- ISOLATION_COMPLIANCE_RATE_POSITIVE_CONTACT (what proportion of agents isolate given they are a close contact)
- TESTING_COMPLIANCE_RATE_TRACED (what proportion of agents take a test immediately as a result of being informed they are a close contact)
- ISOLATION_COMPLIANCE_RATE_SYMPOMATIC_GROUPMATE (what proportion of agents will isolate given one of their group mates has a symptomatic case)
- ISOLATION_COMPLIANCE_RATE_POSITIVE_CONTACT_GROUPMATE (what proportion of agents in a group isolate given one of them is a close contact)

References

- [1] Alexander Karaivanov. A social network model of covid-19. *PLOS ONE*, 15(10), 2020.
- [2] Ying Liu and Joacim Rocklöv. The effective reproductive number of the omicron variant of sars-cov-2 is several times relative to delta. *Journal of Travel Medicine*, 29(3), 2022.
- [3] Ryan McGee. seirplus, Jan 2021.
- [4] Ryan S. McGee, Julian R. Homburger, Hannah E. Williams, Carl T. Bergstrom, and Alicia Y. Zhou. Model-driven mitigation measures for re-opening schools during the covid-19 pandemic. *Proceedings of the National Academy of Sciences*, 118, Jul 2021.
- [5] Ryan S. McGee, Julian R. Homburger, Hannah E. Williams, Carl T. Bergstrom, and Alicia Y. Zhou. Proactive covid-19 testing in a partially vaccinated population. Aug 2021.
- [6] Thea van de Mortel. Closed centres, anxious parents and infected kids: What’s the plan for childcare during omicron?, Feb 2022.