

Survey of Agent-Based Simulations for Modelling COVID-19 Pandemic

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ABSTRACT

On the 11th of March 2020, the World Health Organization (WHO) declared COVID-19 as a pandemic. Part of controlling measures of the pandemic is to understand the disease's trajectories. There are several possible interventions that can prevent and control its spread. Determining an optimal strategy is critical for policymakers to understand the impact of different scenarios. Many modelers are using agent-based simulations to virtually examine the efficiency of these scenarios. This paper aims to review published papers that discussed Agent-based simulation (ABS) for modelling the COVID-19 pandemic. Major databases were searched for published articles in 2020 from top-ranking journals. Ten published papers were carefully chosen, and their findings were summarized and discussed. Among the methods used, three ABS models were shared as open source. Major findings included mask-wearing and working/studying from home as the optimal strategies, whereas airport screening is insufficient, and vertical isolation is similar to 'doing nothing' scenarios. Finally, one paper discussed the gaps in ABS and proposed a call of actions to the scientific community and guidelines to responsibly improve the ABS modelling's quality. This paper can contribute to understanding the current landscape of the COVID-19 pandemic simulation models and their limitations. It is proposed to access selected open-sourced agent-based models to evaluate, utilize, customize or learn from, to help conduct more accurate simulations.

1. Introduction

The city of Wuhan witnessed a viral outbreak for the first time in December 2019. Since then, the virus was classified as a zoonotic coronavirus, belonging to the same family of the Severe Acute Respiratory Syndrome (SARS) coronavirus and the Middle East Respiratory Syndrome (MERS) coronavirus. The novel virus was later named SARS-CoV-2. By February 2020, 33,738 infections were reported in China with 811 deaths. On the 30th of January 2020, The World Health Organization (WHO) confirmed the outbreak as a Public Health Emergency of International Concern. On the 11th of March, the same organization declared COVID-19 as a pandemic [1].

COVID-19 has brought many challenges to humanity, including a demand on novel medical treatments, social policies and economic initiatives. The fast response from the scientific community included dedicating studies from two perspectives to deal with the pandemic: researches focusing on treatment and diagnostic tools, and another dealing with the viral spread

forecasting models [2]. While tight social distancing measures have proven effective in slowing down the viral spread; it was expected that relaxing some of these constraints will result in a second wave of spread [3]. Computational models provide quantitative support to public health teams and policy makers' in their readiness [4]. One of the most useful insights provided by computational model simulations is its ability to virtually explain natural phenomena, which are either not possible to simulate in a real-world, or too costly to be simulated. These simulations can support decision-makers in studying the current plans' efficiency and assist them in developing necessary policies and strategies to limit and control the COVID-19 pandemic [5].

The fast-paced growth of computer processing power allowed the utilization of Agent-Based Models (ABMs) to widely build simulation models for outbreaks [6]. Although experts are building many models to simulate and forecast the infection's trajectories, they do not cover the sophisticated behavioral and social aspects of societies that are exposed to this outbreak [7]. In this research, a survey was conducted on several published papers that used ABM to simulate infectious diseases outbreaks in different scenarios and environments. This review aims to understand the

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various uses and limitations of the ABS in building diseases outbreak models.

2. Literature review

2.1. Modelling pandemics

To better understand how a pandemic grows and how human intuition fails, it is first essential to understand the concept of exponential growth (1). It describes how quantities increase over time. In the early days of a pandemic, the growth rate looks like a linear growth, until the window of action had passed. This was particularly true with COVID-19 (Figure 1).

$$X_t = X_0 (1 + R)^t \quad (1)$$

Moreover, another feature of the growth of the positive cases was that the numbers were time-latent. This means that the positive cases are not identified until infected individuals had already been sick for several days and have likely infected other individuals [8]. Therefore, humanity realized the importance of finding new ways to help understand trajectories of pandemics and implement necessary policies to limit and prevent outbreaks such as COVID-19.

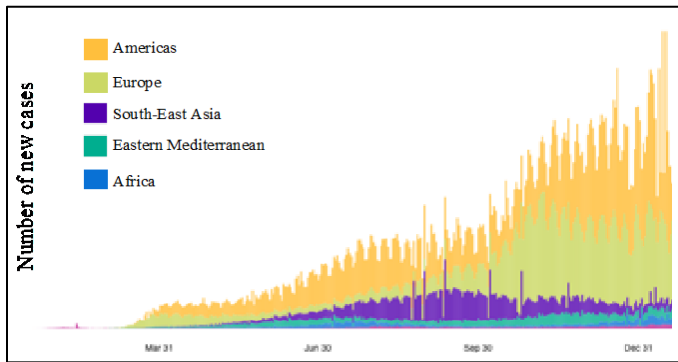


Figure 1: Growth of COVID-19 cases in the world in 2020 [1].

Modelling pandemics are rooted in traditions that go back to the 1920s, when mathematical equations were applied to simulate and model the distribution of the disease in populations, including positive cases, recovered cases and fatality rates. While these mathematical-derived approaches helped experts to understand the threshold, nature and herd immunity of plagues; it did not explain social and behavioral factors such as an individual’s behavioral response towards public policies, measures, or the impact of mixed social contact on spreading trajectories [7].

2.2. Agent-based Models (ABMs)

In the early 1990s, the agent-based simulation was known as a third way of doing science. The field has substantially transformed since and found its way to the primary science field. Moreover, many ABS frameworks have been developed, such as RePast, AnyLogic and NetLogo. In addition, the open database of ABS models allowed swift and decisive prototyping of model developments [9]. For this paper, we can define the ABM as a computational simulation model, where agents are behaving independently without an external force. The agents’ actions are usually a response to the situation they are exposed to during the process of the simulation [10]. Unlike other simulation models, ABMs are described as a simpler one. This type of models do not

use a complicated set of rules or advanced architectures. Nevertheless, they can generate a variety of complicated patterns (behaviors) based on the modelling attributes caused by its simple agent interaction [11].

2.3. Gaps in Agent-Based Models

While mathematical concepts can be concisely defined and described, the same cannot be said of simulation models. Describing early ABMs was challenging to write or understand. No one knew where to locate the different types of information. As a result, the descriptions were often incomplete, and reproducing the model was ultimately impossible [12].

Although many signs of progress were being made in modelling pandemics through agent-based simulations, whenever a new pandemic occurs, similar modelling issues repeatedly arise and require precise attention. These issues include: (1) predicting sophisticated results when the vital data are not available nor reliable (2) reusing the model in a different setting that it was not designed for, and (3) not carefully maintaining good standards, practices or procedures, either for a race to academic recognition or for an immediate response to political pressure [7]. In contrast, [11] proposed a standard format for describing ABMs, known as the “Overview, Design, Details” concept (ODD). ODD was one of the initiatives to counter reoccurring challenges faced, in what has been known as “replication crisis”. ODD was developed to make it easier to write, read, implement and replicate ABM. It can include mathematical and short algorithms and is structured into seven steps, as shown in Figure 2. These steps are grouped into three categories, Overview, Design and Details.

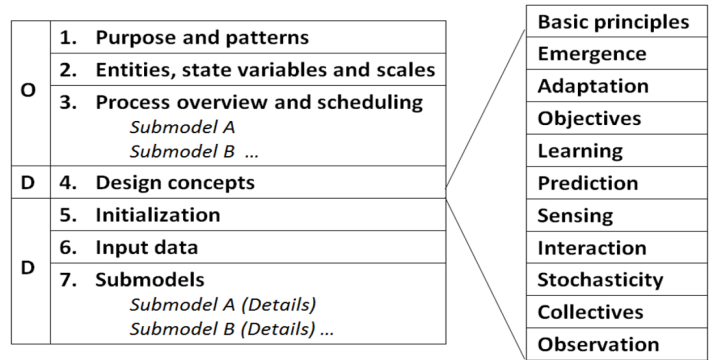


Figure 2: Overview, Design, Details (ODD) concept

3. Methods

3.1. Methodological approach

Document analysis is one of the effective and efficient ways to start the collection of data on a certain topic. Published articles can supply recent and broad data coverage [13]. Search Engines can facilitate faster data analysis. For this research, the critical survey approach was chosen to conduct a primary research, to understand the various uses and limitations of ABS, in building diseases outbreak models.

3.2. Data Collection Plan

Major databases were searched including ScienceDirect, Google Scholar and Crossref for (“Agent-based simulation”) OR

(“Agent-based Model”) OR (“Individual-Based Model”) AND (“COVID-19”). Since the focus was COVID-19, and the pandemic only started spreading in early 2020; the time frame was set for all published journal articles in 2020 up to the 19th of October 2020. The search was conducted without language restrictions; finally, all non-mathematical computing models articles were excluded.

3.3. *Quality of papers*

The focus subject and the time frame of the surveyed papers are relatively new; therefore, high citations was not a good indicator. Alternatively, Scimago Journal & Country Rank ranking (SJR) of [14] was used. Most of the examined papers are ranked between Q1 & Q3. Afterwards, if a paper was not published in a journal yet, the authors’ credibility were examined via the ORCID database [15] to check if they have previously published in high ranked journals.

Table 1: Quality assessment checklist

#	Question
1	Is the paper contained in a high ranked journal?
2	Is the paper cited by more than 5?
3	Have the Authors published before in a high-ranking journal?
4	Is the paper providing clear purpose, method and findings?

The papers’ quality was evaluated against the quality assessment checklist (Table 1). Each answer has a value, and three different values determine the answer to each question. Questions answered with ‘Yes’ have been assigned with a value of (1), questions answered with ‘No’ have been assigned with a value of (0), while questions partly answered have been assigned with a value of (0.5). All papers that scored above 60% were included in the analysis (Table 2).

Table 2: Quality assessment result

Source	Q1	Q2	Q3	Q4	Total	%
R01	1	1	1	1	4	100%
R02	1	1	1	1	4	100%
R03	1	1	1	1	4	100%
R04	0	0.5	1	1	2.5	63%
R05	0.5	0.5	1	1	3	75%
R06	1	1	1	0.5	3.5	88%
R07	1	1	1	1	4	100%
R08	1	0	1	1	3	75%
R09	1	1	1	1	4	100%
R10	1	0.5	1	1	3.5	88%

3.4. *Selection of papers*

Ten journal articles that matched the research criteria were identified. Table 3 shows each study along with the journal’s name, ranking, model used, and its availability. The following table (Table 3) shows the main qualified articles for the further analysis.

Table 3: Names of authors, source and journal and their SJR Quartiles rank

N o.	Journal Name	SJR Quartiles	Model Availability	Country
R01	Chaos, Solitons and Fractals	Q1	https://bit.ly/COVID19_ABSsystem	Brazil
R02	Nature Human Behaviour	Q1	Available from the corresponding author upon request.	USA
R03	Chaos, Solitons and Fractals	Q1	https://github.com/InstituteforDiseaseModeling/covasim . https://github.com/Jasminapg/Covid-19-Analysis .	UK
R04	arXiv.org	-	No	Germany
R05	Informatics in Medicine Unlocked	Q3	No	Iran
R06	JASSS	Q1	No	Italy
R07	Computers in Biology and Medicine	Q2	Pseudo-code	Mexico
R08	Safety Science	Q1	No	Chile
R09	Mathematical Biosciences	Q2	https://www.github.com/gressman/covid_university	USA
R10	Building and Environment	Q1	No	China

4. **Results**

After analyzing the main contribution of each article, the following are the main identified themes of the studies.

4.1. *Agent-Based model of health and economic effect simulation*

The authors of [2] pointed out that it is a challenge for decision and policy makers with only limited studies available on the pandemic.. This paper proposed an ABM simulating the viral and economic impact of COVID-19 in a closed environment. The results can be widely used by decision and policy makers to evaluate the effectiveness of different social policies in a real-life scenario.

The proposed ABM is simulating a closed society living in a shared limited environment. The community consists of individuals and groups of families, businesses and a government. Each agent is interacting with each other. The given characterization by the model is supposed to emulate the character of a community.

The authors of [6] are suggesting another Agent-Based simulation based on pedestrian dynamics. This paper analyzed how pedestrians behave in open areas in conditions with viral spreading, such as with COVID-19. This evaluation was done to gather useful insights about exposure time, and the effectiveness of social distancing measures, as stated by the German government (1:5m), at “an infection rate of 2%”.

The paper provided a different scenario of pedestrian dynamics in realistic situations while focusing on contagious diseases. The simulation was based on ABM works including “A Mathematical model for the behavior of pedestrians” (1991), “Simulating dynamical features of escape panic” (200), “Self-organizing pedestrian movement” (2001) and “Simulation of pedestrian crowds in normal and evacuation situations” (2002).

The simulation conducted a pedestrian scenario in an ideally sized German supermarket covering 4,800 square meters (80 x 60 meters), including shelves and counters as shown in Figure 5. There were 34 destinations identified as “point of interest” with a set of agents being defined as “infected”.

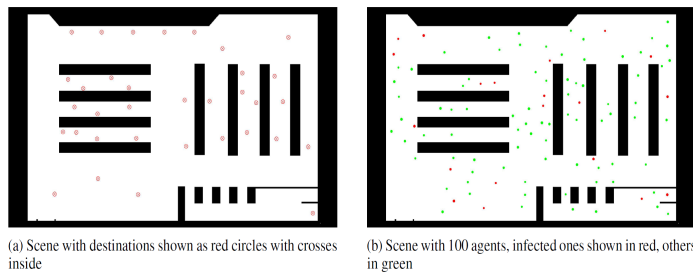


Figure 5: Supermarket layout with walking spaces in white and obstacles in black

The experiment conducted using ABS revealed promising and valuable results for policy makers. The experiment assumed that if the infection rate was 2%, a minimum distance of 1:5 m can be kept if the density is not exceeding one individual for every 16 square meters. The proposed model can be a foundation to better assess the risks in public places within the circumstances of contagious diseases [6].

4.5. Spatio-temporal Agent-Based simulation

The study of [5] investigated the impact of different closure scenarios to control and limit the COVID-19’s spread, including schools, educational institutes and the workplace, in addition to social distancing practices in the city of Urmia, Iran. In this paper, the COVID-19 outbreak was simulated with an ABM using a described set of behaviors in a defined environment. The SEIRD model (Susceptible Infectious Recovered Deceased Model) was used to mimic the infection of a human agent. Then, an all-controlling plan was applied to the model.

The attribute of the ABM’s environment cells was defined based on the spatial data. The related value was defined based on the spital data layers, as shown in Figure 6.

The simulation results indicated that closure of schools and educational institutes (for 50 days between the 21st of February – the 10th of May 2020) decreased the infection rate by an average of 4.94 per cent weekly and a total of 49 per cent. While working from home at 30 & 70 per cent of Urmia city’s population (from

the 21st of February 2020 – the 10th of May 2020) decreased the infection rate by 3.30 & 5.25 per cent weekly and 32.98 & 52.48 per cent in total. Consequently, applying social distancing (from the 27th of March 2020 at 30 & 70 per cent of Urmia city’s population), led to a decrease in the infection rate at 5.24 & 10.07 per cent weekly, in addition to 31.46 & 60.44 per cent in total. The main finding recommended applying social distancing to the majority of the population. The paper claimed their novelty was in the model itself, but the model’s source code was not provided or shared.

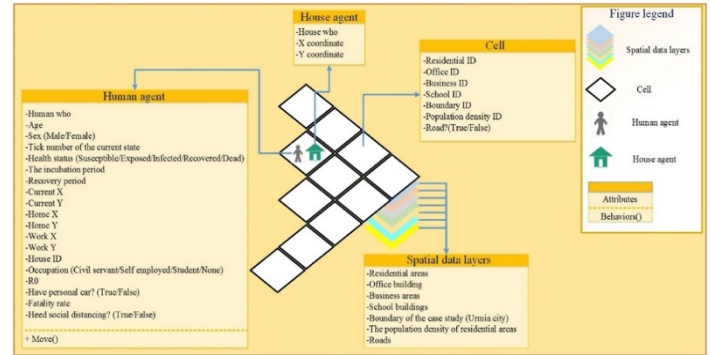


Figure 6: The elements of the Spatio-temporal ABM and their details

4.6. Simulating COVID-19 in facilities

The paper of [11] presented an ABM to evaluate the transmission risks in building environments. The paper tested several scenarios to evaluate and validate the proposed model. The model has been built as an iterative scheme consisting of a Pseudo-code. It was adopted and modified from the work of Professor Hiroki Sayama.

The paper provided the code lines which was considered as simple and can be coded efficiently. According to the study, the model showed useful and promising results that may help decision-makers to determine optimal strategies to reduce infection rates inside facilities and closed environments. The entire process of the model’s operation is summarized in Figure 7.

4.7. Simulation of COVID-19 in the construction sector’s context

The author of [18] examined another vital sector that little is known about. The study focused on the outbreak of COVID-19 among construction workers using an ABM. The paper’s aim was no different from that of other papers, which was examining different intervention strategies. However, there was probable novelty in examining the pandemic’s spread in the context of the construction environment. The approach used was classifying the risk of activities for the workers between low, medium and high, followed by simulation of the spread in a construction project. The result of the simulation recommended a reduction in workforce in construction projects between 30 to 90 per cent. However, it was mentioned that a limitation of the analysis might result in the oversimplification of the realistic condition. Nevertheless, the finding of the study can be used as a basis to simulate the potential of second waves after relaxing the social distancing restrictions in regions that have heavy construction works.

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Algorithm 1. Pseudo-code for the COVID-19 transmission risk model
1. Input: A, B, lin, uin, lcm, ucm, [Lx, Ux, Ly, Uy], α, maxiter, R, S, k=1
2. Prii ← InitializeProbIn(A, lin, uin);
3. PrcmA, PrcmB ← InitializeProbMob(A, B, lcm, ucm);
4. A(k), B(k) ← InitializePos(A, B, [Lx, Ux, Ly, Uy]);
5. while k ≤ maxiter do
6. For each ai(k) ∈ A(k)
7. F ← FindAnInfectedAgentInNeighbor(B(k), R);
8. If (F == 1) then
9. If (rand ≤ Prii) then
10. A(k) ← DeleteFromA(ai);
11. B(k) ← IncludeInB(ai);
12. end If
13. end If
14. end for
15. for each ai(k) ∈ A(k)
16. If (rand ≤ PrcmA) then
17. If (rand ≤ α)
18. ai(k+1) ← LocalMovement(ai(k), S);
19. else
20. ai(k+1) ← LongMovement([Lx, Ux, Ly, Uy]);
21. end If
22. else
23. ai(k+1) = ai(k)
24. end If
25. end for
26. for each bj(k) ∈ B(k)
27. If (rand ≤ PrcmB) then
28. If (rand ≤ α)
29. bj(k+1) ← LocalMovement(bj(k), S);
30. else
31. bj(k+1) ← LongMovement([Lx, Ux, Ly, Uy]);
32. end If
33. else
34. bj(k+1) = bj(k)
35. end If
36. end for
37. k = k + 1
38. end while
    
```

Figure 7: Proposed ABS model algorithm by [11]

4.8. Simulation accuracy in a responsible manner

Within the context of numerous simulations available and under the pressure of polices makers, [7] presented a paper proposing guidelines, urging the scientific community to follow, so they can handle the transparency and speediness of their findings in a responsible manner.

While many experts have provided sound simulations and prediction of the COVID-19 trajectories; their papers might not cover the complete sophisticated behavioral and social parameters.

Therefore, [7] called for urgent steps to improve the quality of ABMs. The first proposed step was to examine previous ABMs of COVID-19 and explore their potentials, identify gaps, and suggest counter measures. The counter measures would aid in the rapid demand of immediate responses, which is a critical step to avoid collective mistakes. The second suggestion was to consider the pressure between modelling efforts and policy makers, and develop a better understanding of the challenges which aroused from the over expectation of the emerged knowledge, due to misinterpreting the science of modelling. Finally, measures to address these gaps and improve the relationship between the sciences and public policies was suggested. These included a call for a wide-scale cooperation between the public and academic stakeholders, where knowledge, models and data can be exchanged and shared.

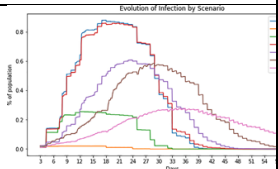
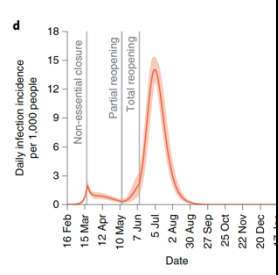
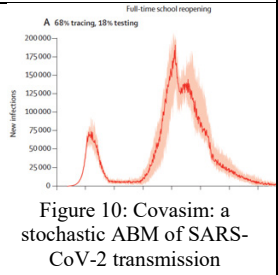
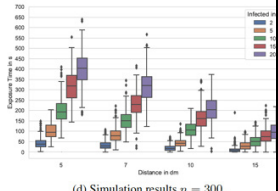
The call to action by [7] can be summarized into two major actions. First was a call to modelers to maintain high standards when building ABMs. The second was for institutional agencies who own useful data that can help calibrate and inform COVID-19 models, to engage with trusted academic associations to make this data available for the rest of the scholars.

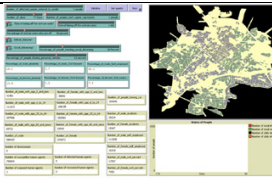
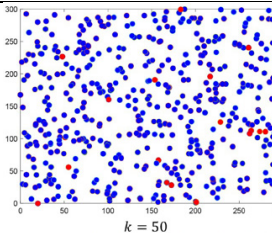
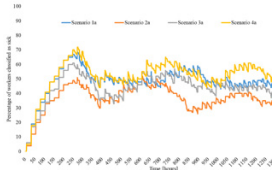
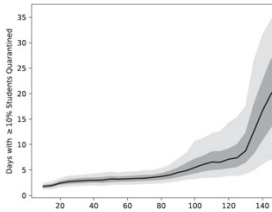
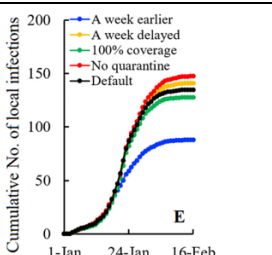
Moreover, the paper recommended to follow three best practices that are already well known among agent-based modelers: (1) building models based on open sources such as MASON and NETLOG, (2) following standing documenting protocols, (3) using long-lasting online content storage tools such as ComSES, to ensure the model is fully documented prior to submission. The paper urged scholar modelers to follow these guidelines, which are very critical during this exceptional times, to enhance their credibility.

4.9. Summary

The following Table 4, summarizes all the examined papers, including article code (N), the main purpose, method used with a model screenshot, the main findings and country (C) of the experiment. The summary excludes the results of [7]. As it is proposing ABM simulation documentation guidelines for the scientific community to follow. However, it has been included for its high importance and relevance to the overall survey.

Table 4: Summary of the analyzed papers

N	Purpose	Method / Model screenshot	Main Finding	C
R01	To simulate health and economic effects of social distancing interventions	 Figure 8: COVID-ABS: a new SEIR ABM	“A useful tool to assist politicians and health authorities to plan their actions against the COVID-19” [2].	Brazil
R02	To simulate the impact of testing, contact tracing and household quarantine on second waves of COVID-19	 Figure 9: R, version 3.4.4	“A response system based on enhanced testing and contact tracing can have a major role in relaxing social-distancing interventions in the absence of herd immunity against Covid-19” [3].	USA
R03	To Determine the optimal strategy for reopening schools	 Figure 10: Covasim: a stochastic ABM of SARS-CoV-2 transmission	“A comprehensive and effective test–trace–isolate strategy would be required to avoid a second COVID-19 wave” [16].	UK
R04	To simulate Pedestrian dynamics for exposure time estimation	 Figure 11: ABM based on the model of Helbing et al.	“Results suggest that a density of one person per 16m ² or below is sufficient” [6].	Germany

R05	To simulate the Spatio-temporal of COVID-19	 <p>Figure 12: Netlog ABM</p>	Demonstrate “how the disease is likely to evolve amongst the society and populations and assess the impacts of control strategies on controlling the outbreak of the disease” [5].	Iran
R07	To simulate the COVID-19 transmission risks in facilities	 <p>Figure 13: Transmission ABM</p>	“Useful information to produce strategies for reducing the transmission risks of COVID-19 within facilities” [11].	Mexico
R08	To simulate COVID-19 on construction workers	 <p>Figure 14: ABM</p>	“Workforce from a construction project may be reduced by 30 to 90% due to the spread of COVID-19” [18].	Chile
R09	Simulating COVID-19 in a university environment	 <p>Figure 15: ABM</p>	“Results indicated that large scale randomized testing, contact-tracing, and quarantining are important components of a successful strategy for containing campus outbreaks” [17].	USA
R10	Simulating intervention methods on COVID-19 transmission in Shenzhen	 <p>Figure 16: Agent-based SEIIR model</p>	“Intervention strategies implemented in Shenzhen were effective. Results may be useful for other cities when choosing their intervention strategies” [19].	China

5. Discussion

This paper examined ten research reports adapting ABMs to simulate the impact of COVID-19. Paper [11] presented an ABM to assess the transmission risk in building environments. However, this model was not the only one. [2] also proposed a new Agent-based simulation with promising results. The presented ABS model was not only simulating the transmission risks, but it also emulated the economic impact of businesses and the government. At the same time, an SEIR ABM simulating pandemic outbreaks

with the help of a society of agents called COVID-ABS was presented.

Furthermore, other researchers tried to combine more than one tool. In [5], the author used Spatio-temporal simulation to present the impact of various pandemic preventing strategies, including closures, social distancing, working and studying from home in Urmia, Iran. Other researchers also examined these strategies in Boston, United States. In [3], the author combined more than one instrument. In order to build a detailed ABS simulating a second wave; anonymized, geolocalized mobility data were integrated with census and demographic data. On the other hand, In [6] the author tried to approach pandemic controlling strategies from a different perspective. They focused on examining data on pedestrians behavior and collective time of exposure to evaluate the efficiency of social distancing measures within COVID-19’s context.

The aims of the remaining researches were not in developing the model itself, but to use existing ones to support decision-makers in identifying the right intervention strategies. Their novelty was the context of the simulation. In [19], the author presented and analyzed different preventing strategy in Shenzhen. In comparison, [16] conducted a simulation to determine an optimal strategy in the education sector. Likewise, [17] also examined different strategies and measures in the education sector, but their focus was university environments. In [18], the author has moved to a different sector, which is the construction sector. In [18], the author attempted to study the pandemic’s impact on construction workers and the best way to lower the infection rate. Alternatively, [7] presented a paper proposing guidelines to ensure transparency and speediness of modelers’ findings being made available, are done responsibly.

Most papers used ABMs in their methods except [7], which method and approach were not clear or defined. Generally, the findings of the proposed models were promising and have some useful insights.

The authors of [5] suggest future research to be conducted in a different region that has not been evaluated yet, using different parameters and attributes.

Among the surveyed papers, two significant findings related to widely shared beliefs were proven wrong: The paper of [19] found out that airport screening was not very useful. Forty-six per cent of infected cases were not detected during airport screenings due to several reasons. Part of the failure was due to (1) the sensitivity of the entry and exit screening, (2) asymptomatic cases, and (3) the incubation period, while [2] found that vertical isolation policies promoted by some countries like Brazil and United States was similar to ‘doing nothing’ scenarios. In contrast, [19] reported that working and studying from home and mask-wearing were found to be the most effective policies. [2] seem to reach the same conclusions but in a more realistic manner. Their finding focused on combining wearing of masks and 50 per cent lockdown with social distancing. However, [16] and [3] have added wide-scaled testing as an additional layer of intervention policy. Although nation-wide testing might be theoretically useful, it does not take into consideration the social and behavioral aspect of individuals in addition to their economic situation, where testing is not free in some regions. In addition, there are the issues of testing accuracy,

as pointed out by [17] and the detectability rate as mentioned by [19].

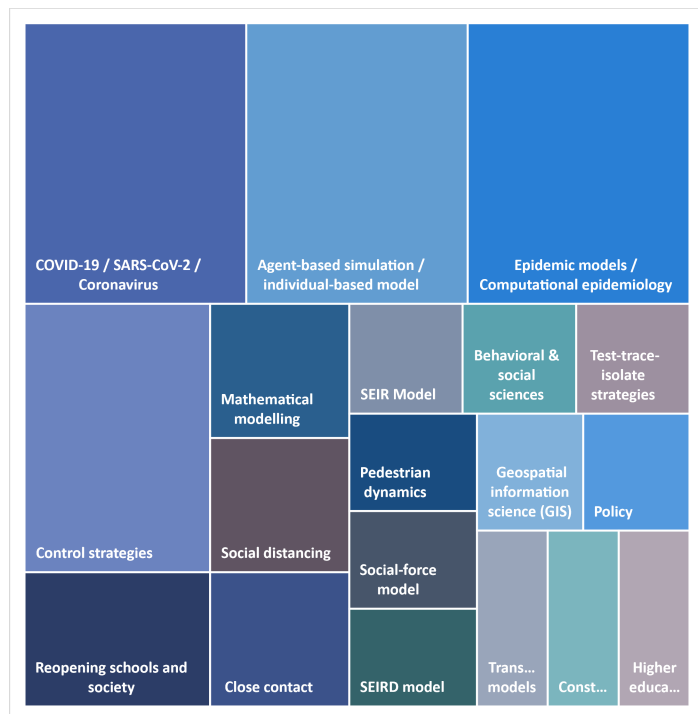


Figure 17: Most discussed subjects based on keywords

6. Limitation and Future work

Many noteworthy articles were missed due to the limited knowledge of different keywords and synonyms used by scholars. Figure 17 shows different keywords commonly used by the examined papers. Moreover, many scholars use arXiv (arXiv.org) e-Print archives as a swift free distribution service with open access. However, papers published in these databases are by invitations only, and not peer reviewed. Therefore, it is not recognized by Scimago's Journal & Country Rank ranking.

7. Conclusion

This paper examined the different discussions on the COVID-19 pandemic's simulations that used ABMs. It was found that agent-based simulation for COVID-19 was used in a variety of contexts, including schools, universities, workplaces, facilities and construction. Methods used varied from simple ones consisting of a few lines of code, to sophisticated models combining mobility data with demographic data or Spatio-temporal data, or from different perspectives such as ABS based on pedestrian dynamics. The main findings of this analysis was in line with works of literature, where mask-wearing and social distancing are among the best strategies, and against some common beliefs, airport screening and vertical isolation were less effective. The main limitation and future suggestions are on improving the quality of papers by following modelling guidelines proposed by ODD, validating the test accuracy, and considering the behavioral and social attributes of individuals in different cultures and regions.

Outbreaks are usually very sophisticated phenomena, and no one specific model can be applied to all cities of the world [5]. The different parameters do not only impact the infection rates, but they

also vary from one place to another. These parameters include but are not limited to culture differences, literacy, awareness, interactions, transportation, the urban context, population density, job diversity, age variations, gender, number of employees, number of students and people with private transportation means. These only few of the parameters that impact the way the outbreak spreads in different regions. The papers recommended that similar simulations be conducted in a different part of the world using different parameters and different controlling and preventing strategies.

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