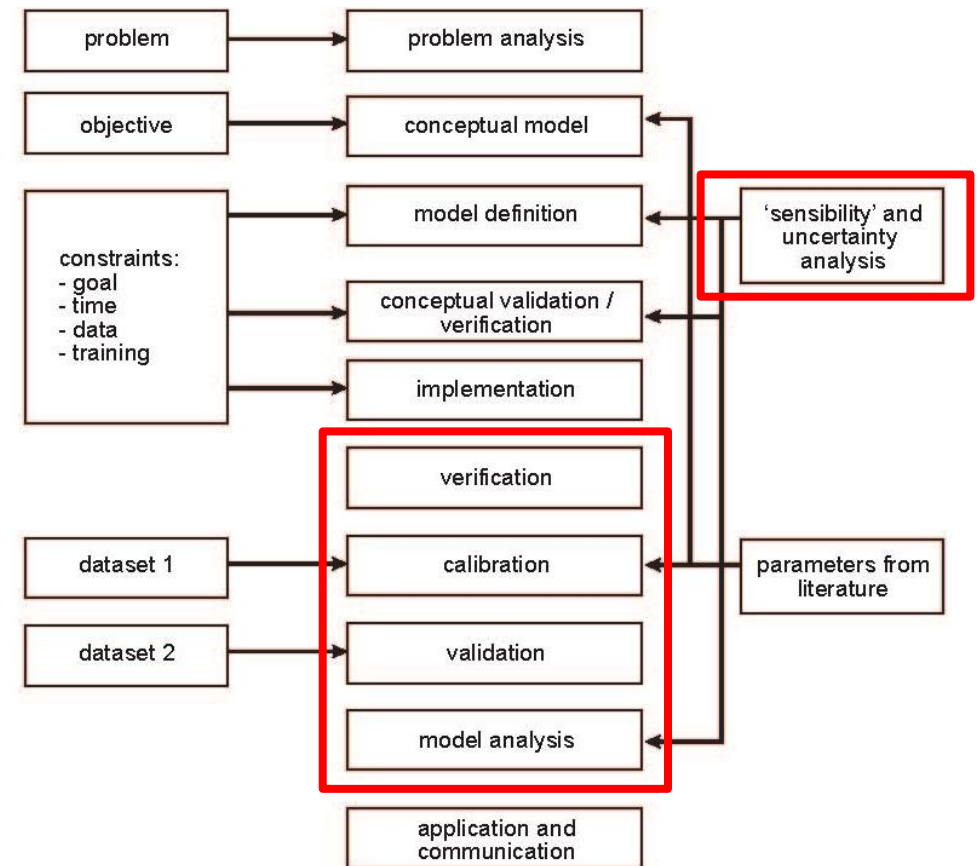


ABM3
ELLEN-WIEN AUGUSTIJN

OVERVIEW

- Verification
- Sensitivity Analysis
- Parameterization
 - Direct parameterization
 - Indirect Parameterization (Calibration)
 - One at a Time
 - All at a Time
- Validation
 - Input validation
 - Process validation
 - Descriptive output validation
 - Predictive output validation

- What about pattern-oriented Modelling?



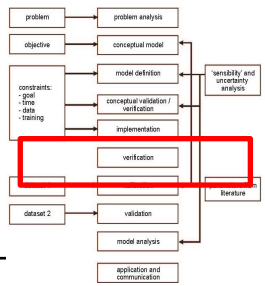


- **Verification** is the task of ensuring that a model satisfies the specifications

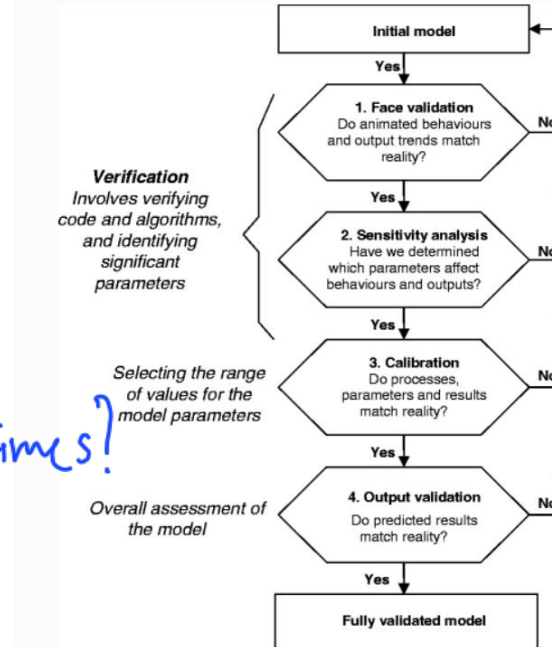
- check agent behavior, agent's trajectory
- aggregated level

VERIFICATION – FACE VALIDATION

if it's sensitive to some params, we might need "calibration"



- History of an individual agent can be “misleading” especially when the simulation contains random effects.
- To determine if the results are typical, it is necessary to repeat the simulation. → how many times?
- Statistical analysis of the results is necessary.

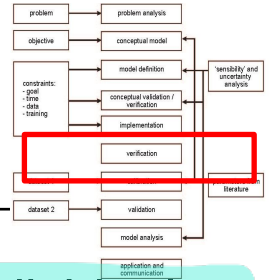


- Sensitivity analysis can proof if the output is sensitive to variation in initial conditions and parameters.
- The effect of different model versions can also be assessed by running controlled experiments.
- Difference in the logic of the model can be studied by comparison of different versions.

Ngo, T.A., See, L. (2012). Calibration and Validation of Agent-Based Models of Land Cover Change. In: Heppenstall, A., Crooks, A., See, L., Batty, M. (eds) Agent-Based Models of Geographical Systems. Springer, Dordrecht. https://doi.org/10.1007/978-90-481-8927-4_10

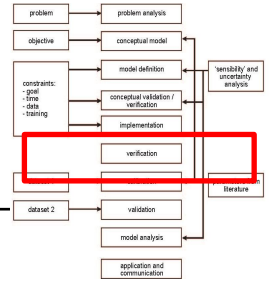
Castle, C. J. E. and A. T. Crooks (2006). Principles and Concepts of Agent-Based Modelling for Developing Geospatial Simulations. [Working paper series UCL - paper 110 - Sep 06.](#)

Local or global sensitivity analysis



- **Local sensitivity analysis:** test how sensitive the model is to the value of each individual parameter.
 - Does not allow us to capture parameter interactions: how the model's sensitivity to one parameter might change as other parameters change
- **Global sensitivity analysis:** test how sensitive the model is when varying all parameters at the same time.
 - Problem is that many different combinations are possible

Sensitivity analysis, uncertainty, robustness



- Does the model reproduce patterns *robustly*, or are these results *sensitive* to changes in model *parameters*?
- How uncertain are the model's outputs? (would it produce the same results if different plausible parameter values are used)
- Uncertainty analysis (UA) looks at how uncertainty in parameter values affects the reliability of model results
- Robustness analysis (RA) explores the robustness of results and conclusions of a model to changes in its structure.

ex. wolf-sheep model

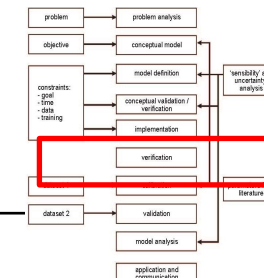


if we change number of sheeps/
wolf at initial, model isn't able
to reproduce same pattern

Stability

stochastic elements

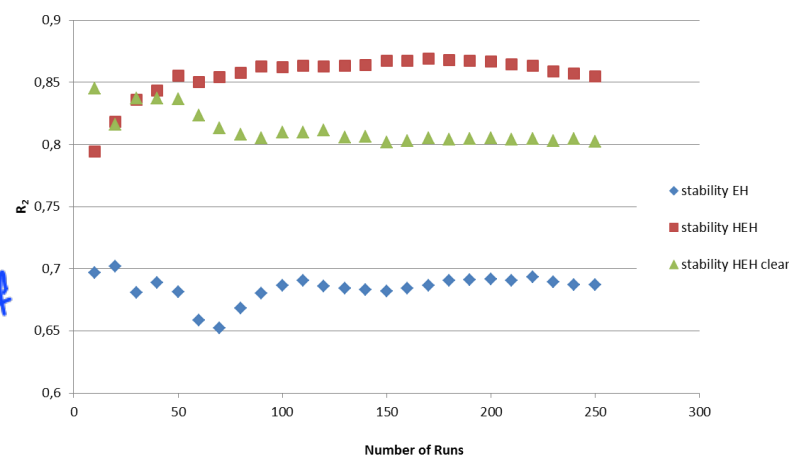
(stoch)



Two methods to check stability (robustness):

- Plotting the **accumulative average** of the state variable (output) over an increasing number of runs.
- The **coefficient of variation** is defined as the ratio between the standard deviation of a sample and the mean of that sample resulting in the following formula: $Cv = \frac{\sigma}{\mu}$

in which Cv is the coefficient of variation, σ the standard deviation of the sample and μ the mean of the sample



(average R^2 depends on number of runs)

N	Cv
5	0.0022
10	0.0024
25	0.0024
50	0.0021
100	0.0021

Determine how many times a new population should be constructed

- Agent-based models use a re-created “synthetic population”
- The synthetic population is normally generated based on statistical data (CBS).

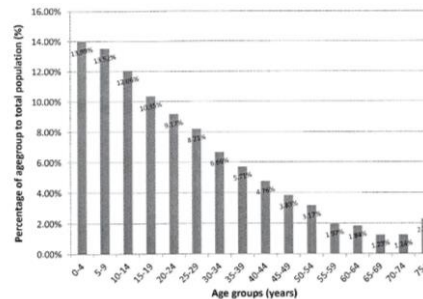
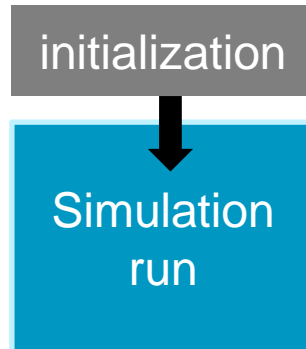


Figure D.4: Ashanti region population divided by age and group (GSS, 2012)

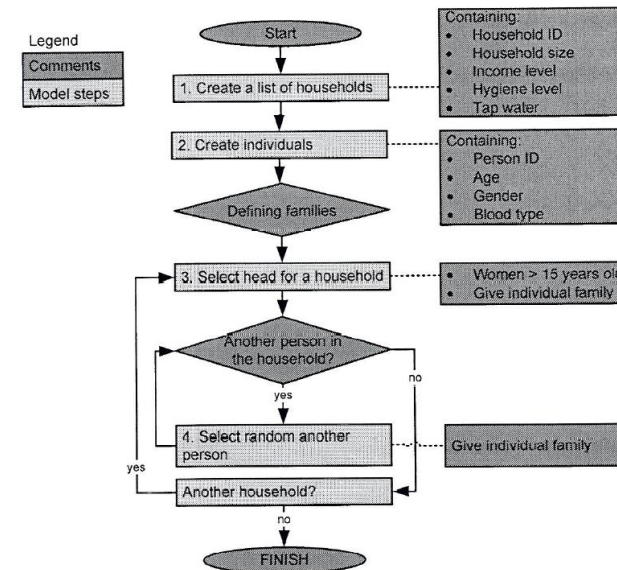
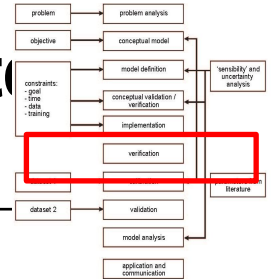
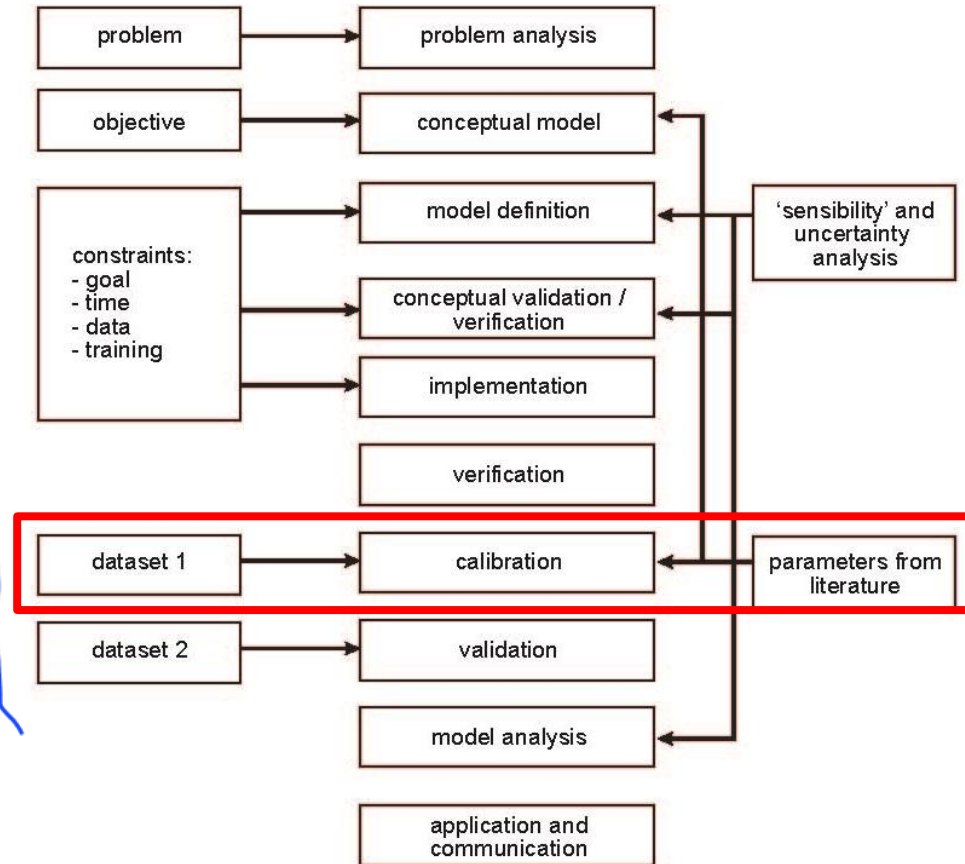
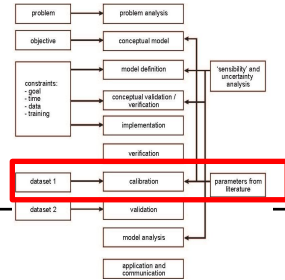


Figure 6.2: Synthetic population generator after Moeckel (2003)



THE MODELLING PROCESS



Parameters are the constants in the equations and algorithms that are used in your model

Parameterization: selection of values for a model's parameters

Calibration: specific type of parameterization in which we try to find a set of values for important parameters.



Purposes of model calibration

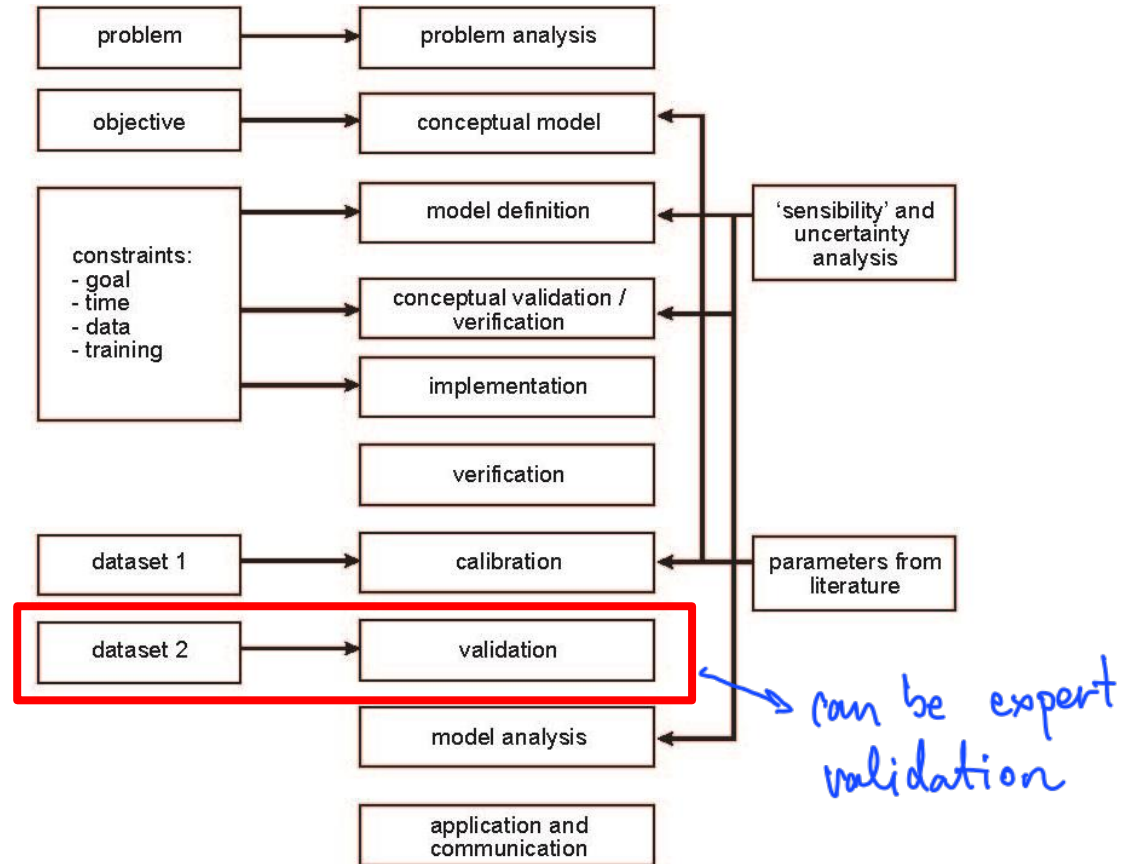
- Model calibration serves the following purposes:
 - Force the model to match empirical data
 - Estimate the value of parameters that cannot be evaluated (measured) directly
 - To test the model's structural realism: can we calibrate it to match the observations within a reasonable range?
- Calibrate each sub-model separately



Categorical versus Best-fit Calibration

- *Categorical Calibration*: ^{range} search for parameter values that produce model results within a category or range you defined as acceptable (mean number of agents between 120 and 150)
- *Best-fit Calibration*: ^{exact no.} you search for one set of parameters that cause the model to best match some exact criteria (mean 135 agents)

THE MODELLING PROCESS

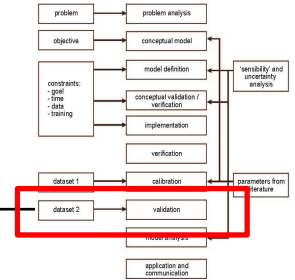


- **Validation** is checking if the model is a good model of the simulated phenomenon
- *Validation – the model shows the macro-level regularities (pattern) that the research is seeking to explain. If so, this is evidence that the interaction and behavior of the agents is the cause of the regularities (pattern).*
- After comparing the macro behavior it is desirable to compare the output of the model with empirical data.

VALIDATION

- A model has a degree of validity (Law and Kelton, 1991)

A model is valid to the extent that it adequately represents the system being modelled (Casti, 1997)



Casti, J.L. (1997) *Would-Be-Worlds: How Simulation is Changing the Frontiers of Science*, John Wiley & Sons, New York, USA.

Law A.M., and W.D. Kelton (1991), *Simulation modeling and analysis*; Second Edition, McGraw-Hill, New York

Axelrod, R. (1997). *Advancing the Art of Simulation in the Social Sciences. Simulating Social Phenomena*, Berlin, Heidelberg, Springer Berlin Heidelberg.

VALIDATION – POSSIBLE PROBLEMS

- Both model and system under analysis are likely to be stochastic. → random
- A model might be able to produce plausible future predictions but may not be able to recreate known past system states. → can't recreate past bec. there weren't there yet.
- Model could be correct but data from the real-world system may not.
- Many simulations are path dependant (i.e. the outcome of a simulation is dependant on the exact initial setup chosen) – history of a simulation is highly significant.

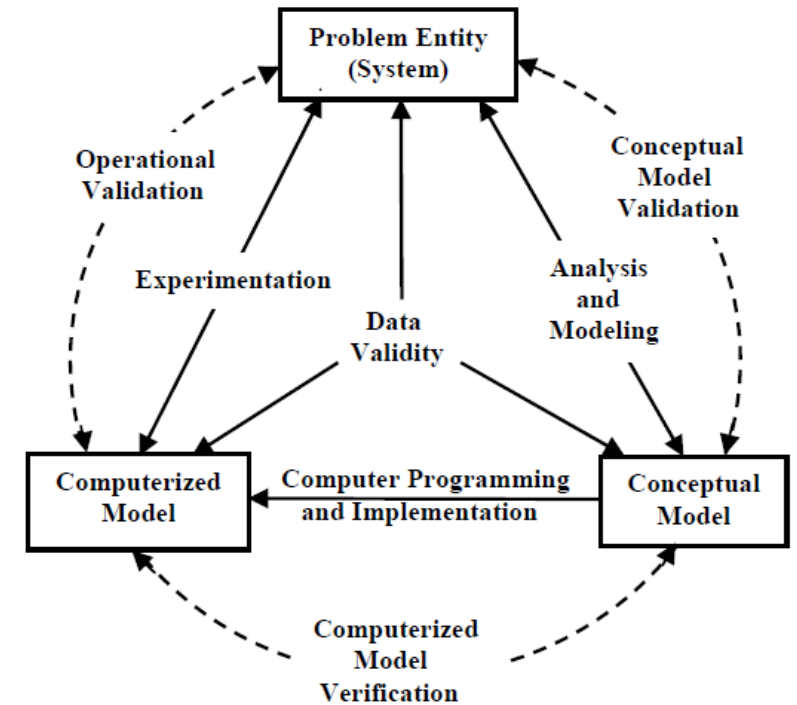


Figure 2: Simplified Version of the Modeling Process

Sargent, R. (2011). [Verification and validation of simulation models.](#)

January 2011, Proceedings - Winter Simulation Conference 37(2):166 - 183

DOI: [10.1109/WSC.2010.5679166](https://doi.org/10.1109/WSC.2010.5679166)

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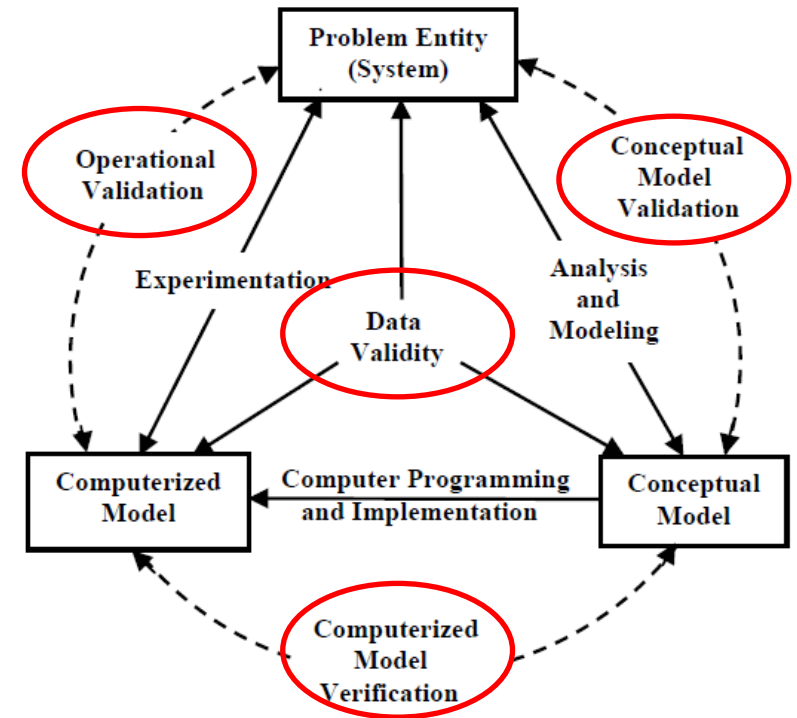


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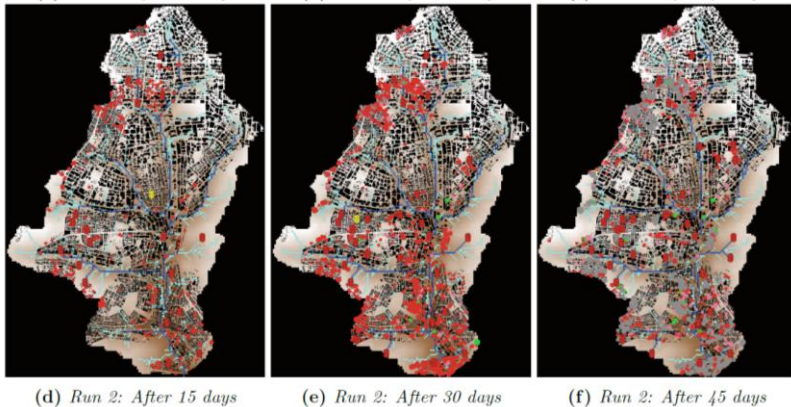
DOI: [10.1109/WSC.2010.5679166](https://doi.org/10.1109/WSC.2010.5679166)



VALIDATION

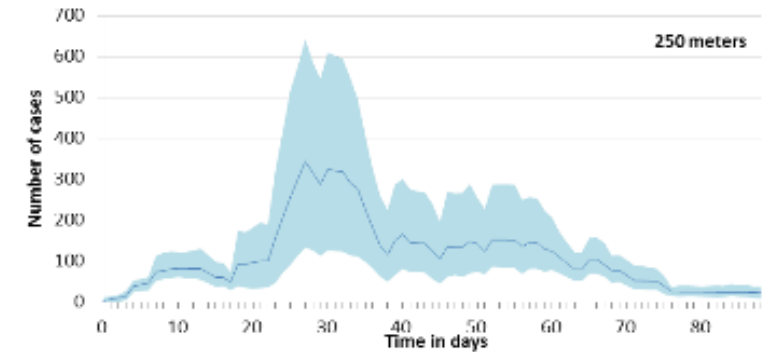
- Validation
 - Input validation
 - Process validation *Conceptual model*
 - Descriptive output validation *→ pattern*
 - Predictive output validation *→ data from the model match with real-world*
- **Macro validation** (at an aggregation level)
- **Micro validation** comparing individual rules/agents
- **Face validation** (do the general ideas about the behavior and properties compare to the real-world) versus **empirical validations** (data validation) *???*
- Validity of a model is always related to the purpose of this model
- All models are simplifications, and all models are wrong

An Example: Using time series and spatial patterns

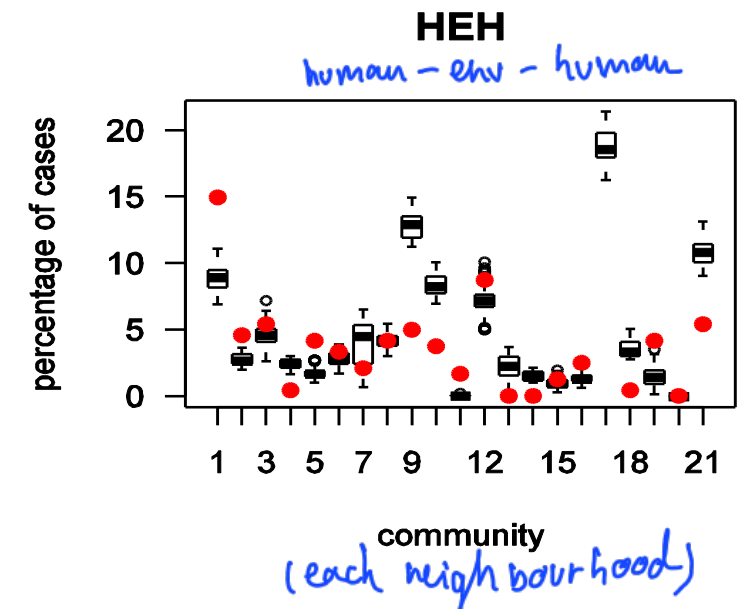


Augustijn, E.-W., et al. (2016). "Agent-based modelling of cholera diffusion." [Stochastic Environmental Research and Risk Assessment](#) **30**(8): 2079-2095.

Should your model be able to reproduce patterns of change over time?



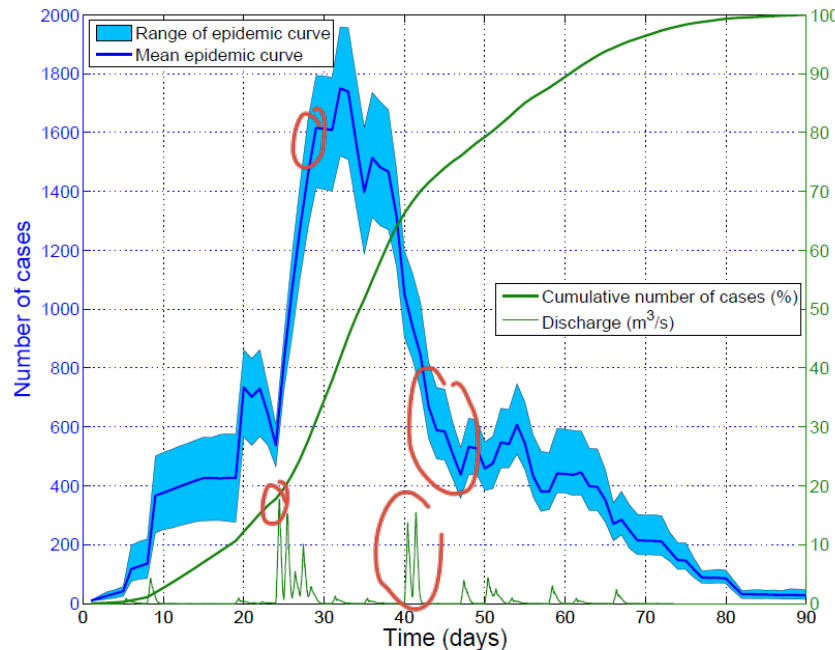
Should your model be able to reproduce patterns in space?



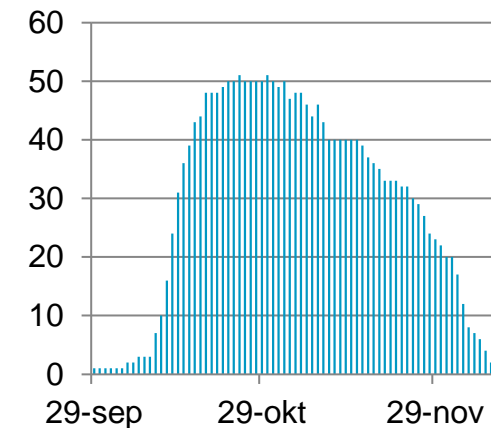
INTEGRATED MODEL

RESULTS

chronological order of events



Range of epidemic curves representing the minimum and maximum number of cases within a set of 90 runs.



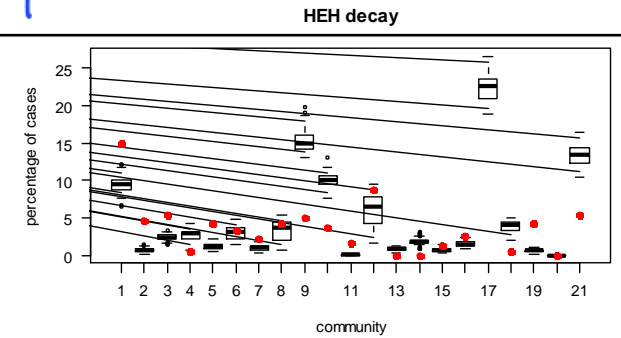
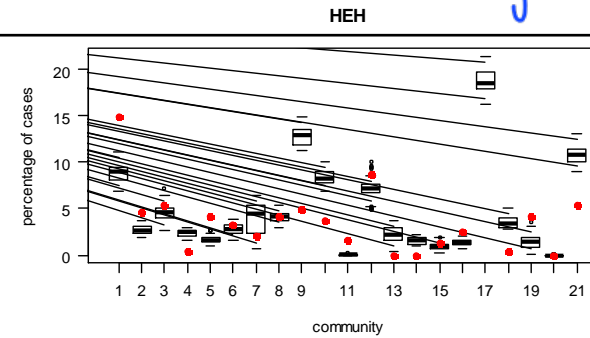
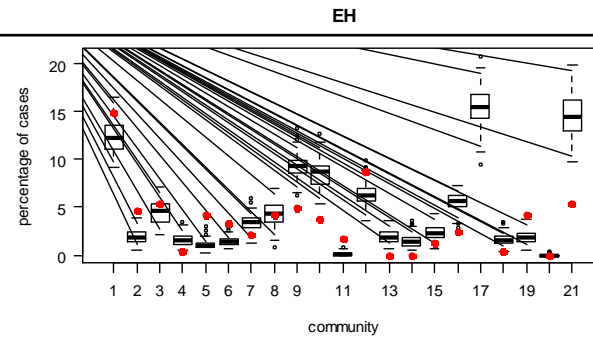
Transmission mechanism	HH	HEH	EH	VT
Average number of cases	80	2461	683	22
Minimum and maximum number of cases	60-104	2237-2608	595-786	10-38
Contribution to total number of cases (%)	2.5	75.8	21.0	0.7

Augustijn, E.-W., et al. (2016). "Agent-based modelling of cholera diffusion." [Stochastic Environmental Research and Risk Assessment](#) **30**(8): 2079-2095.

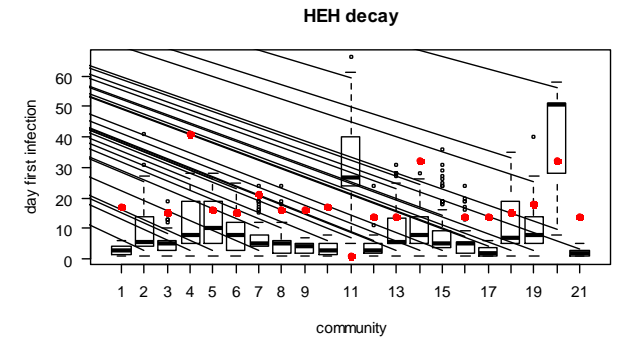
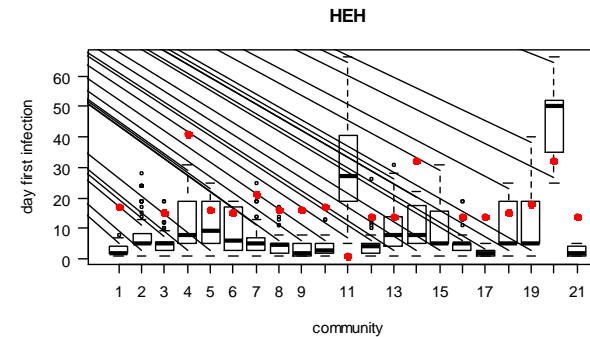
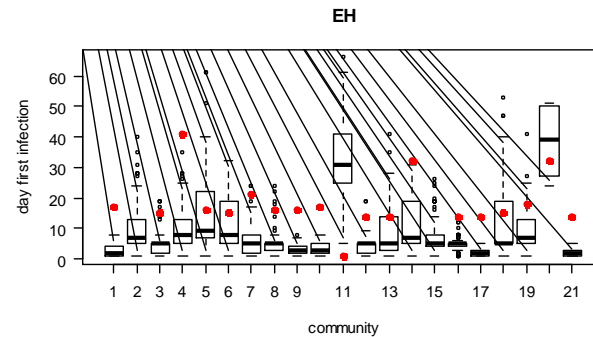
we can integrate ML to adjust the behavior of human → lead to good prediction of cases

EXPERIMENT 1: EVALUATION SPATIAL PATTERNS

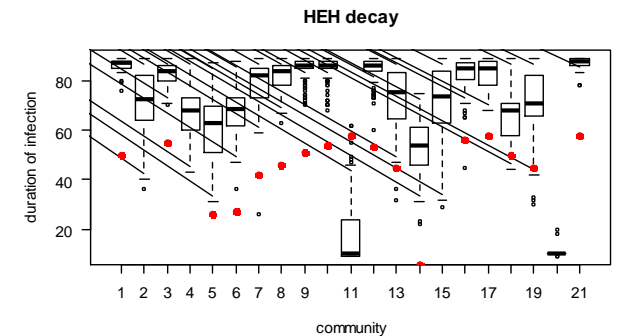
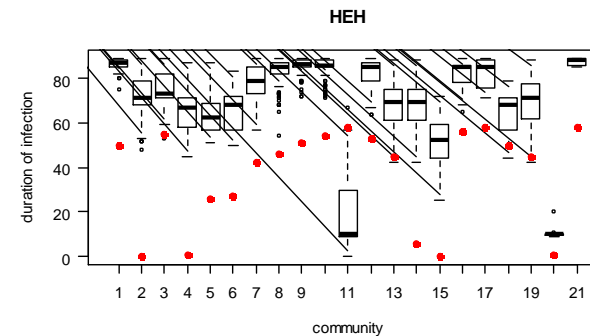
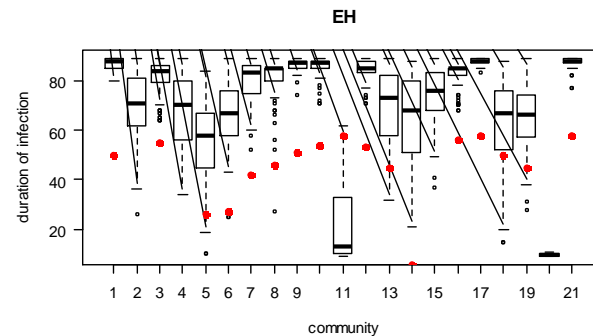
Disease cases



Day first infection



Duration of infection



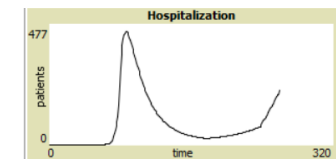
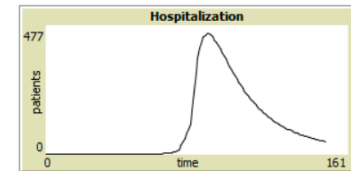
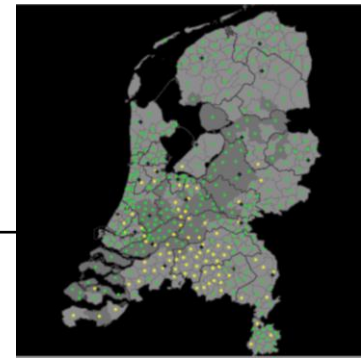
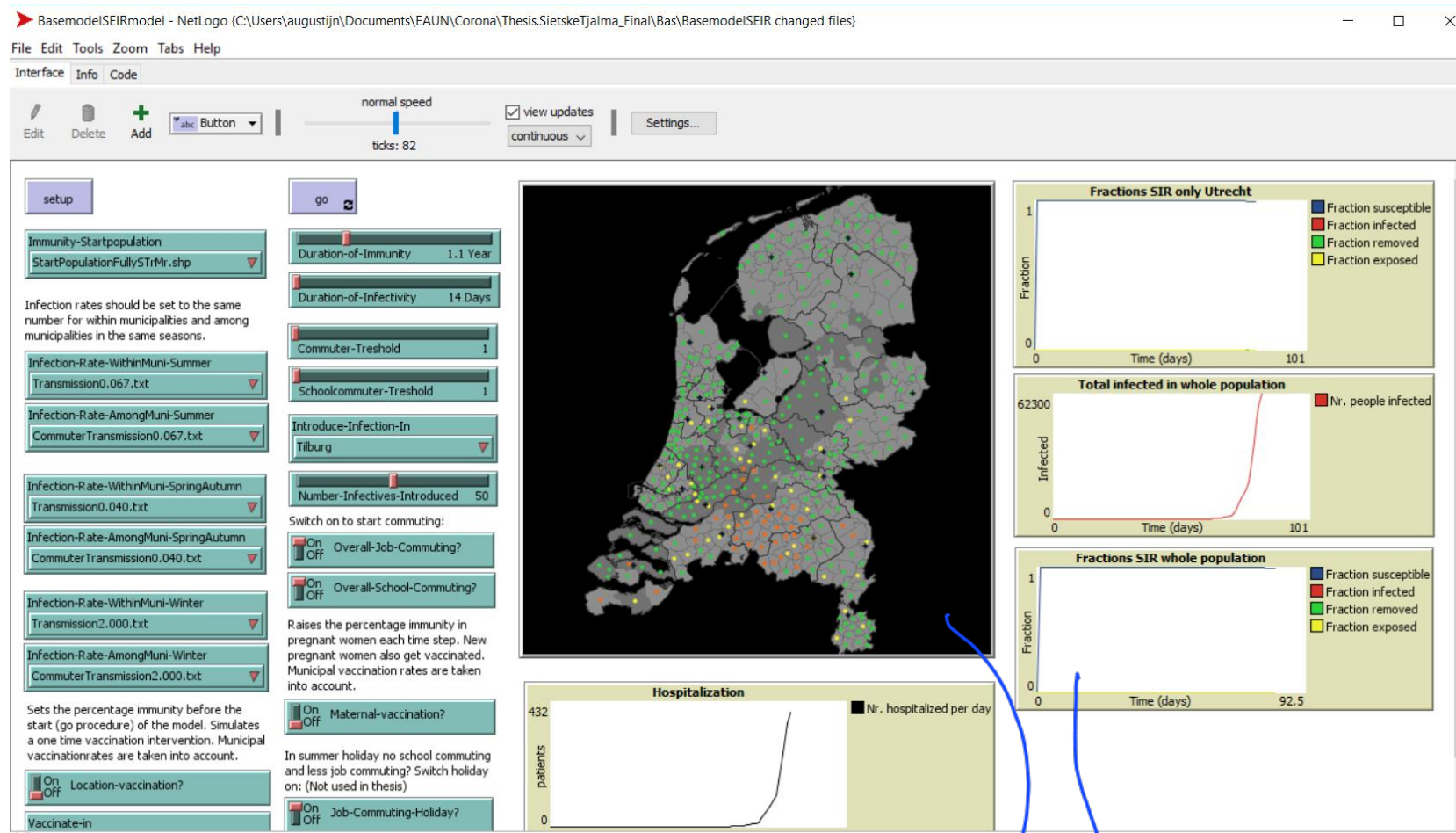
PATTERN ORIENTED MODELING

- Two different/alternative hypothesis
- **Extension:** use the complete area to find the optimal location
- **Infilling:** try to align to existing buildings



Augustijn-Beckers, P., Flacke, J., & Retsios, V. (2011). Simulating informal settlement growth in Dar es Salaam, Tanzania : an agent - based housing model. *Computers, environment and urban systems*, 35(2), 93-103. <https://doi.org/10.1016/j.compenvurbsys.2011.01.001>

Covid-19 model



spatial pattern ↙ ↘ pattern in time

Positive tests per 100.000 inhabitants per week

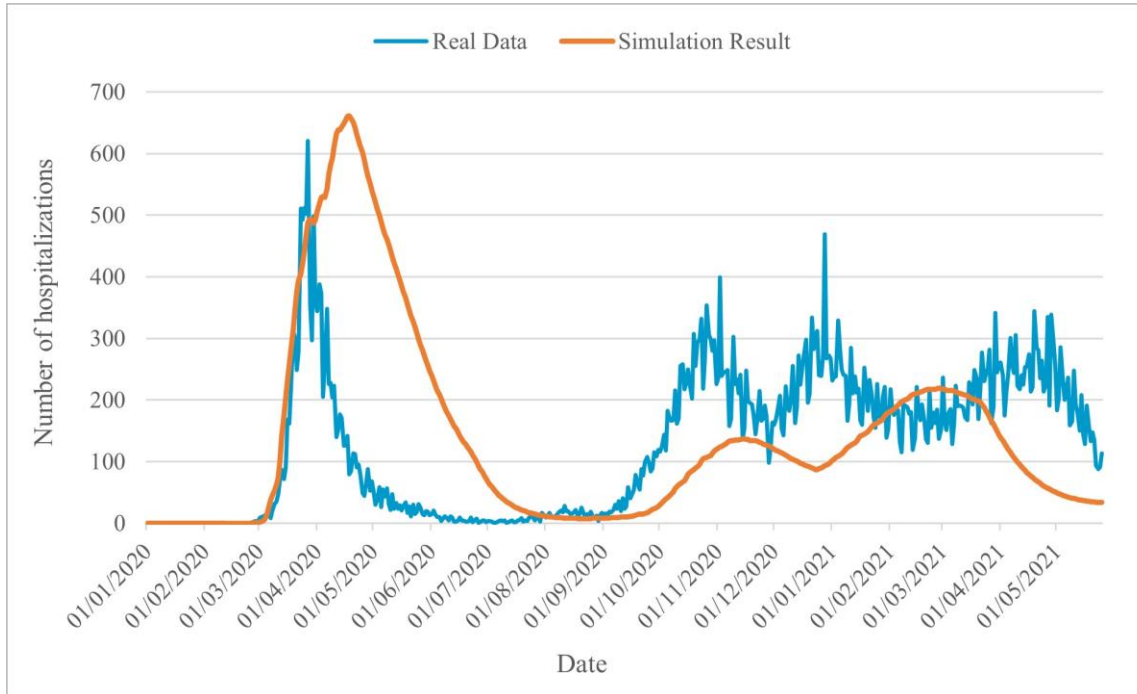
Hospitalized individual (incl. IC) - nationwide per day

Risk level	Positive tests per 100.000 inhabitants per week	Hospitalized individual (incl. IC) - nationwide per day
Risk level 1 Caution	<50	<40
Risk level 2 Concern	≥50	<40
Risk level 3 Serious	≥150	≥40
Risk level 4 Severe	≥250	≥80
Risk level 5 Lockdown	≥350*	≥100*

Coping Strategies

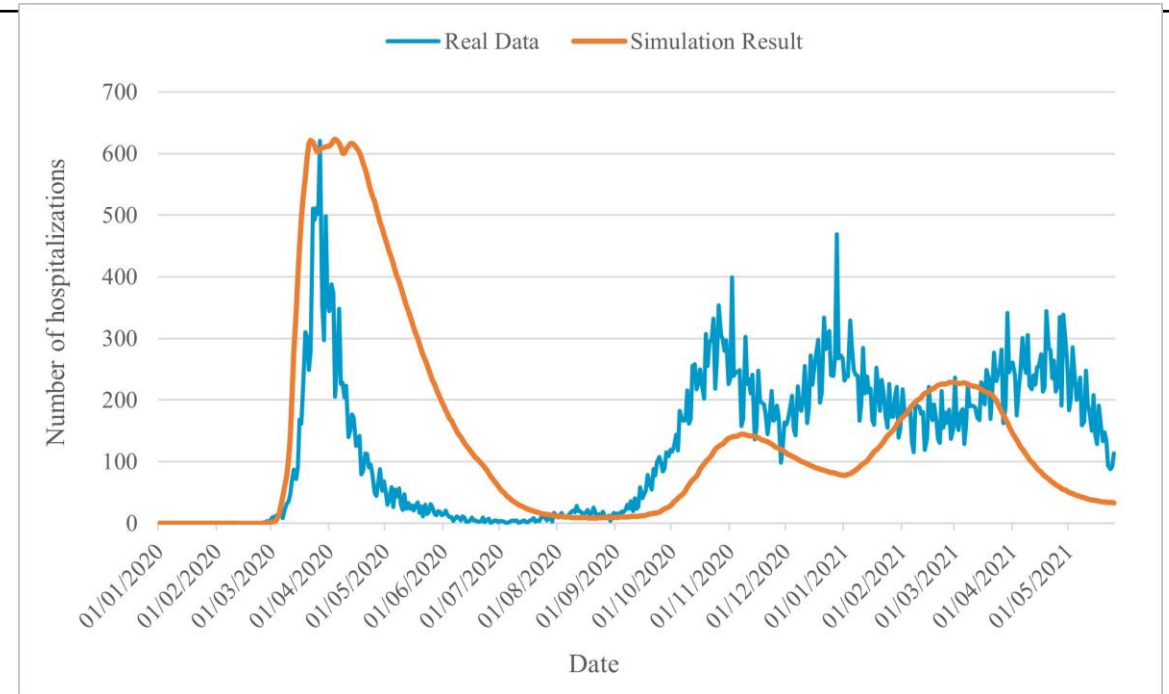
	Level 1	Level 2	Level 3	Level 4	Level 5
Job Commuting	Normal level 100%	Normal Level 80%	Reduced Level 60%	Reduced 40%	Further Reduced 20%
School Commuting	level 100%	Level 100%	Level 100%	100%	0%
GAET commuting	level 100%	Level 80%	Level 60%	40%	20%
VT commuting	level 100%	Level 90%	Level 70%	50%	30%

With closing of schools



Number of hospitalized cases for the RoadMap Scenario

Without closing of schools



Number of hospitalized cases for the age-specific

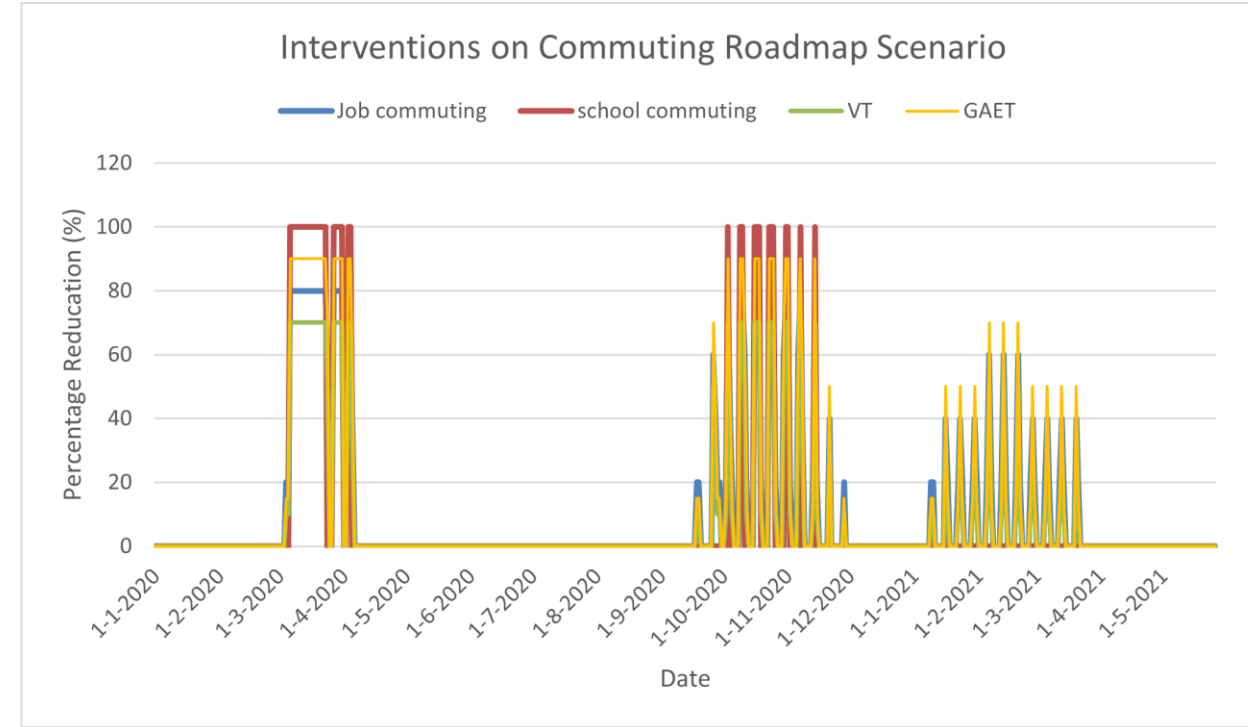
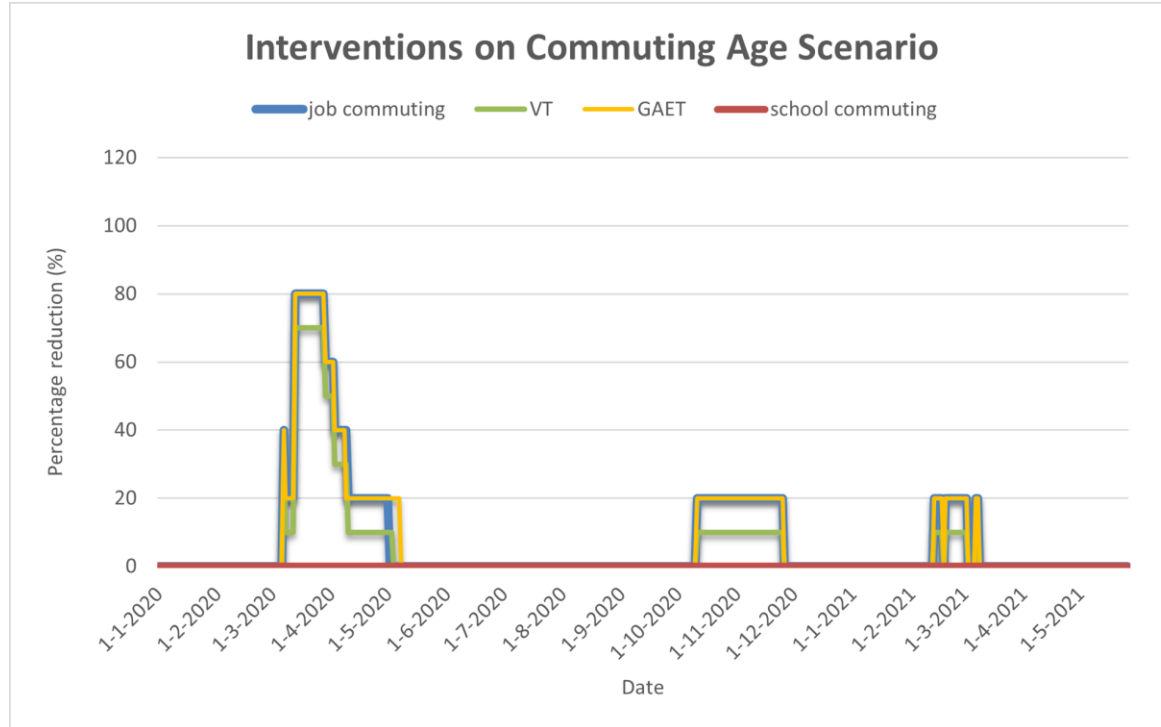
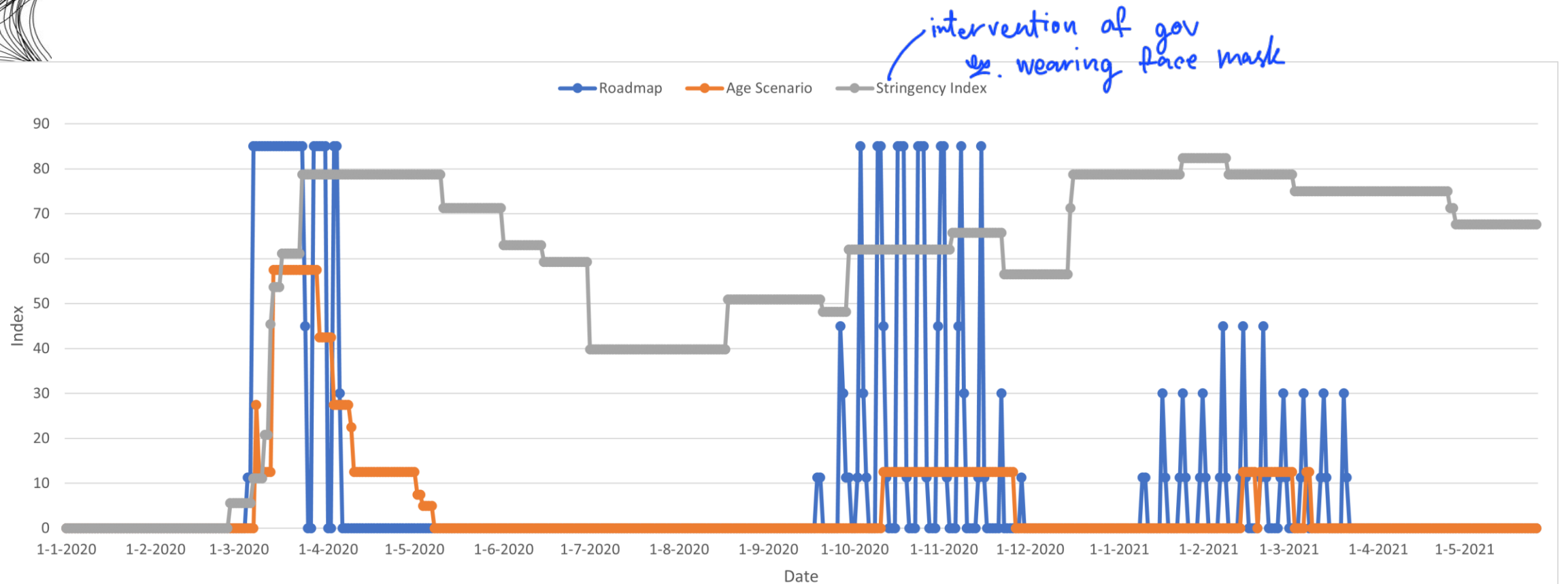


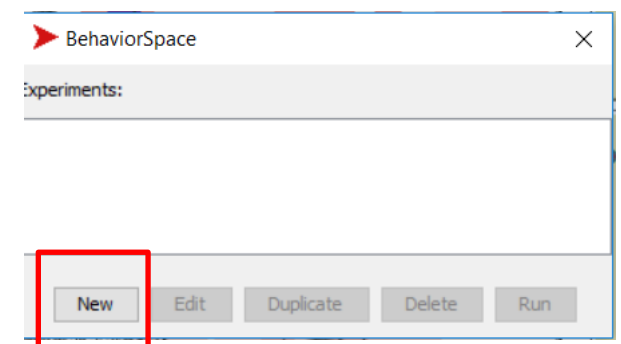
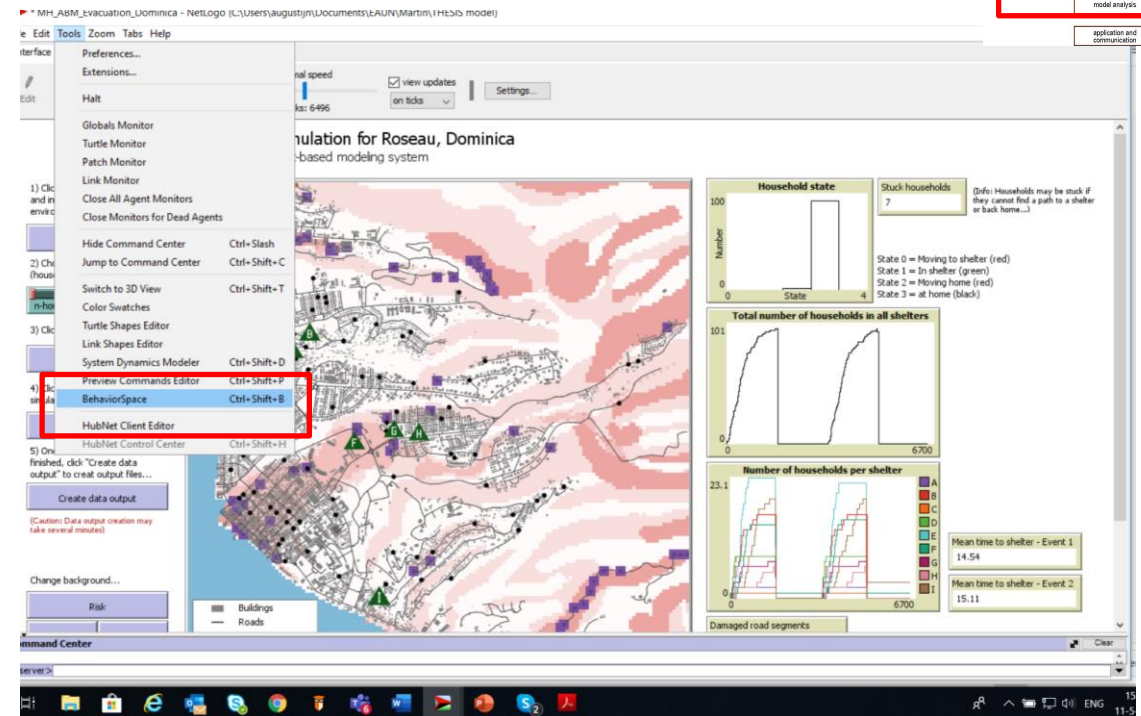
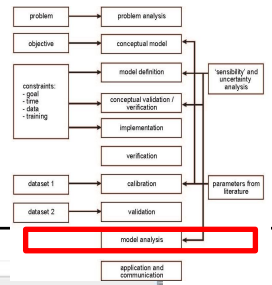
Figure 7: Interventions on Commuting for the RoadMap Scenario (a) and the Age Scenario (b).

MODELING COVID-19



BEHAVIOR SPACE

- Open behavior space via the **Tools** menu
- Create a new experiment
- Edit an existing experiment
- Run an experiment



BEHAVIOR SPACE

- Variables that are automatically added come from the user interface
- For models that are stochastic, increase the number of repetitions
- Carefully check the outputs → number of households per shelter
- Make sure your model does not keep on running for ever

names
variables

repetitions

outputs

Stopping the
model

Experiment

Experiment name

Vary variables as follows (note brackets and quotation marks):

Either list values to use, for example:
 ["my-slider" 1 2 7 8]
 or specify start, increment, and end, for example:
 ["my-slider" [0 1 10]] (note additional brackets)
 to go from 0, 1 at a time, to 10.
 You may also vary max-pcor, min-pcor, max-pycor, min-pycor, random-seed.

Repetitions
 run each combination this many times

☒ Run combinations in sequential order
 For example, having ["var" 1 2 3] with 2 repetitions, the experiments' "var" values will be:
 sequential order: 1, 1, 2, 2, 3, 3
 alternating order: 1, 2, 3, 1, 2, 3

Measure runs using these reporters:

one reporter per line; you may not split a reporter across multiple lines

☒ Measure runs at every step
 if unchecked, runs are measured only when they are over

Setup commands:

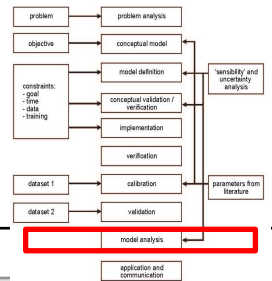
Go commands:

Stop condition:
 the run stops if this reporter becomes true

Final commands:
 run at the end of each run

Time limit
 stop after this many steps (0 = no limit)

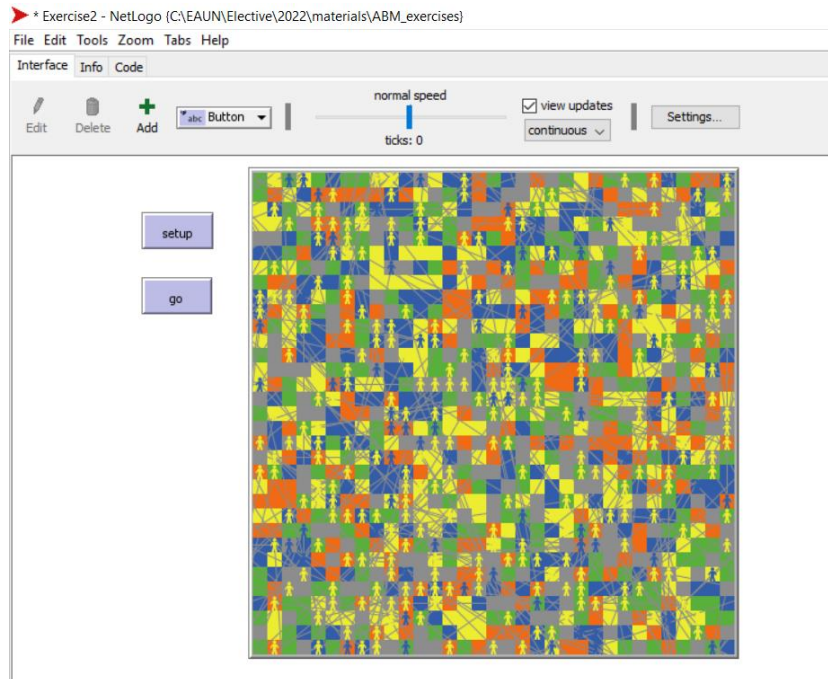
OK Cancel



CODE EXAMPLES

CREATING LINKS

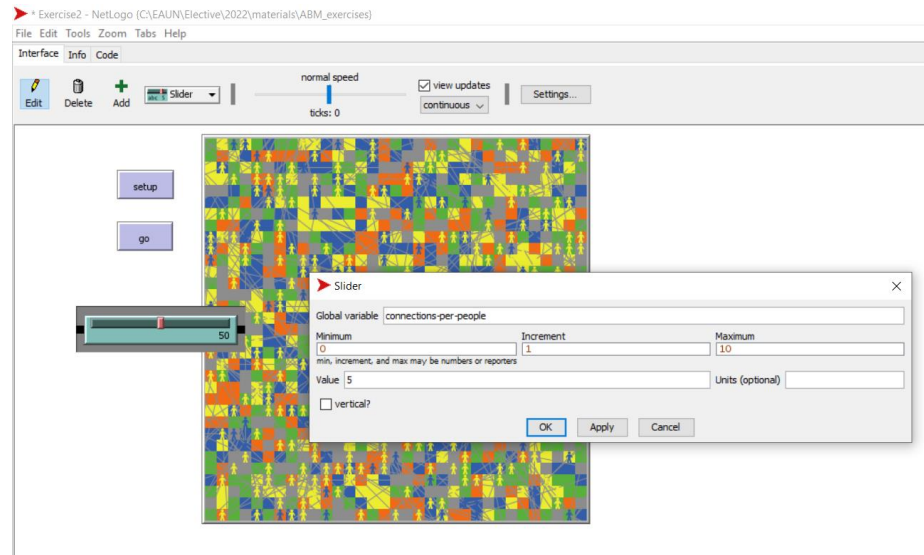
```
to setup-agents
  create-residents 100
  ask residents [move-to one-of patches set color blue set shape "person"]
  create-tourists 300
  ask tourists [move-to one-of patches set color yellow set shape "person" set stay-duration (random 10 + 1) create-link-with one-of residents]
end
```



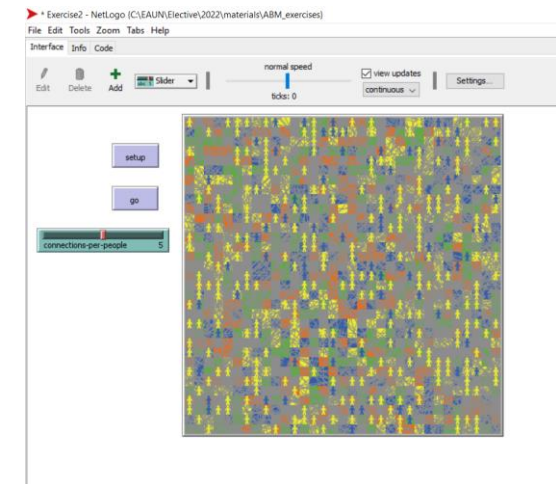
- You now see lines between the tourists and the residents when you run the setup

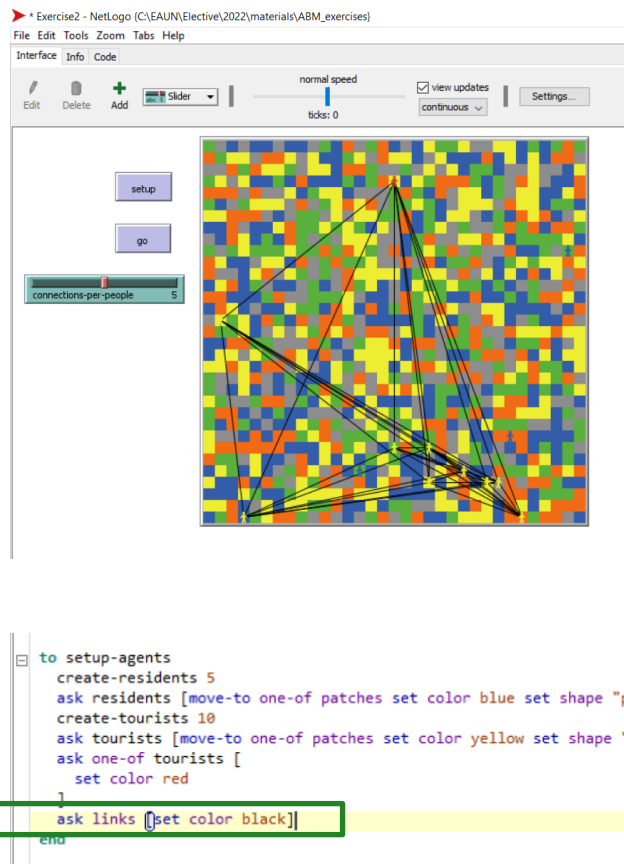
CREATE MORE LINKS

- Create a slider to define the number of links to create



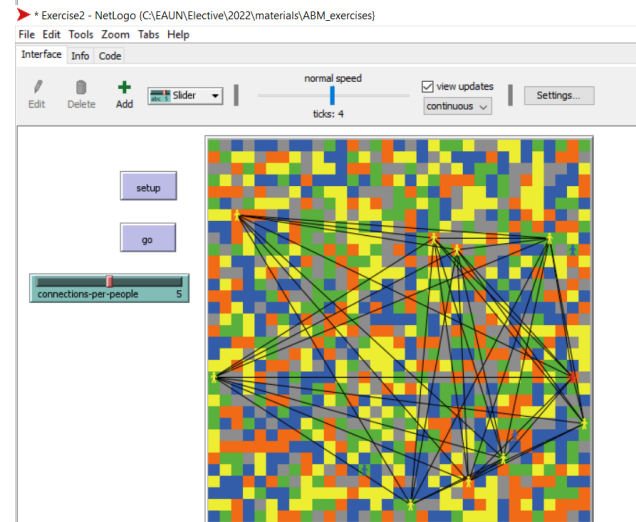
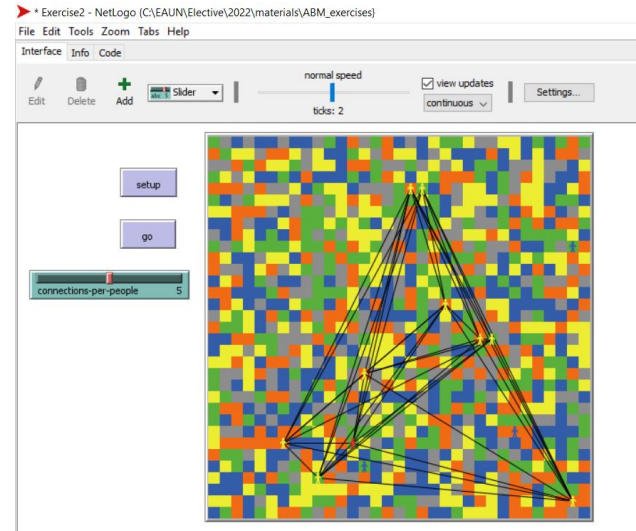
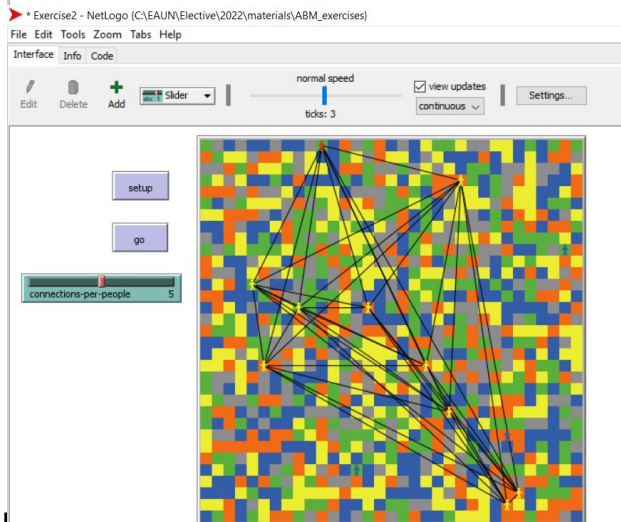
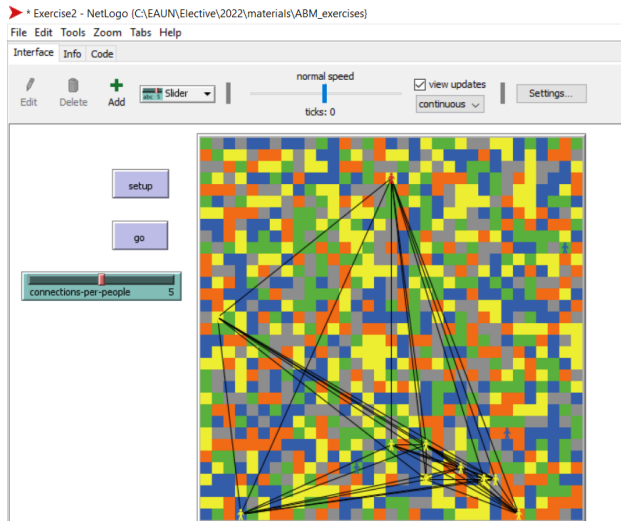
```
to setup-agents
  create-residents 100
  ask residents [move-to one-of patches set color blue set shape "person"]
  create-tourists 300
  ask tourists [move-to one-of patches set color yellow set shape "person" set stay-duration (random 10 + 1) create-links-with n-of connections-per-people other tourists]
end
```





- Give the links another color
- Reduce the number of agents to see the links

MOVING THE TURTLES



- When we run the go, and our turtles move, the network will remain as is (the links will move with the turtles)

SPREAD MESSAGE

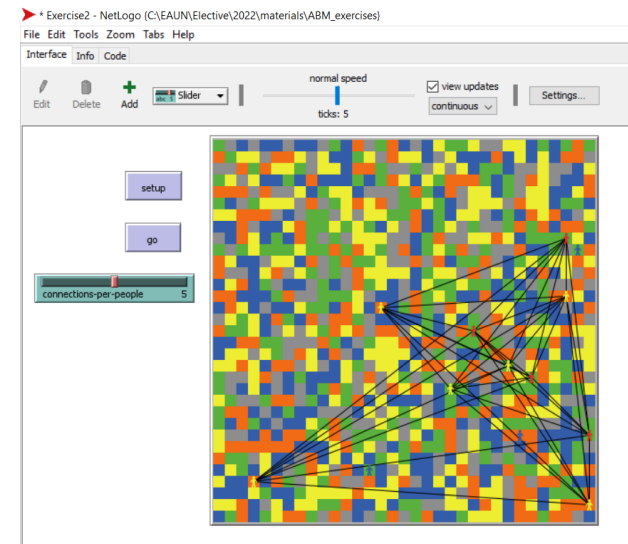
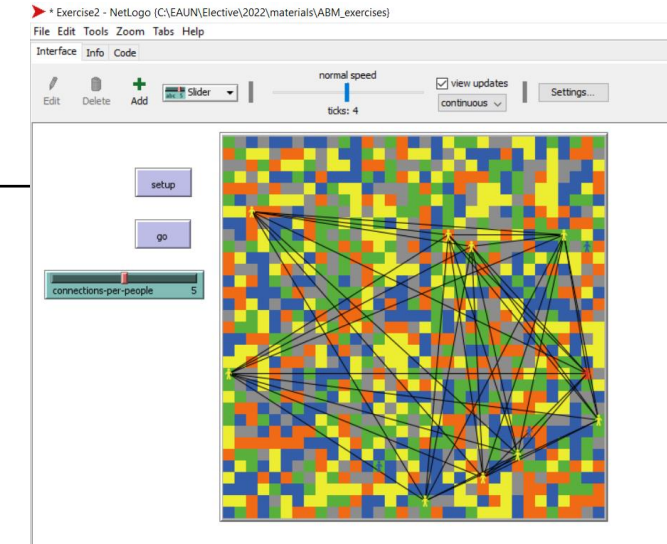
```
* Exercise2 - NetLogo (C:\EAUN\Elective\2022\materials\ABM_exercises)
File Edit Tools Zoom Tabs Help
Interface Info Code
Find... Check Procedures Indent automatically Code Tab in separate window

breed [residents resident]
breed [tourists tourist]

turtles-own [stay-duration activity]
patches-own [landuse]
globals [landuse-list]

to setup
  clear-all
  setup-environment
  setup-agents
  reset-ticks
end

to go
  if ticks mod 7 != 6 or ticks mod 7 != 0 [
    ask residents [set activity 1]
    ask tourists [set activity one-of [2 3]]
    move2
    ask tourists [
      if color = red [
        ask one-of link-neighbors [
          set color red
        ]
      ]
    ]
  ]
  tick
end
```



- We see all turtles turn red
- Spread the fact that you should be aware of tick risk

TEAM BASED LEARNING 2
QUESTIONS VALIDATION



Validation is regarded to be the most difficult part of ABMs. Which step in the validation process is the most difficult part of validating the Evacuation model? You can select multiple answers.

- a. **Input validation**, as we cannot set a building to fire to collect data that is correct/valid.
- b. **Process validation**, as we do not know what people in a building are doing at the time an evacuation starts.
- c. **Descriptive validation**, as there are no patterns that we can replicate.
- d. **Predictive validation**, as there is no independent data available



To check the validity of a model, you should know the purpose of the model. Which of the statements below about the Wolf-Sheep-Grass model is correct?

- a. This model does not have a purpose, and therefore, it cannot be validated.
- b. The purpose of this model is to show that wolf-sheep dynamics is a complex system. Therefore, descriptive output validation is the most important aspect of the validation process.
- c. The purpose of this model is to predict how many sheep can survive with a given number of wolves in a neighborhood. Therefore, predictive output validation is the most important aspect of the validation process.
- d. The problem with this model is that not all processes, like flocking of sheep, are implemented, and therefore, the model cannot be validated.



In the Living Textbook, you find the concept “validation” as one of the steps in the ABM design steps. Under challenges, various issues are listed that might apply to the evacuation model. Select all correct statements below.

- a. The stochastic nature
- b. Predictive versus retrodictive capability
- c. Data Quality
- d. Path Dependency