

New models	The virus model	How the model works	Exploring cases	Factual/counterfactual analyses	Planning vaccination campaigns	A new model	Agricultural field
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Nuovi modelli, dalla diffusione della Covid-19 all’agricoltura di precisione

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Accademia di Agricoltura—10 novembre 2021, slides at
<https://terna.to.it/aa.pdf>



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An experiment with GA

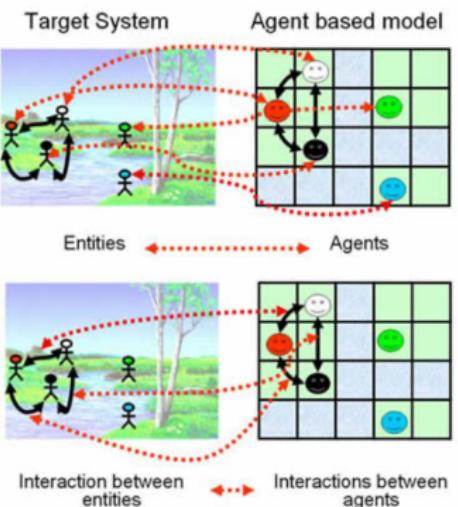
What if

A new model

Agricultural field

Introduction

Observer & Model: two levels



From M. Galán, L.R. Izquierdo, S.S. Izquierdo, J.I. Santos, R. del Olmo, A. López- Paredes, B. Edmonds: Errors and artefacts in agent-based modelling. *Journal of Artificial Societies and Social Simulation*, 12 (1):1, 2009. ISSN 1460-7425.

<http://jasss.soc.surrey.ac.uk/12/1/1.html>



Tools



*From Santa Fe Swarm(1995,
<http://www.swarm.org>) to SLAPP, in
 Python and with a better time
 management*

SLAPP

Swarm-Like Agent Protocol in Python

Here you have SLAPP v.0.91 (in the [SLAPP repository](#) you have related material and old versions).

A reference manual is coming (expected with version 1.0, in June 2015).

Five chapters of a forthcoming book will be related to SLAPP.

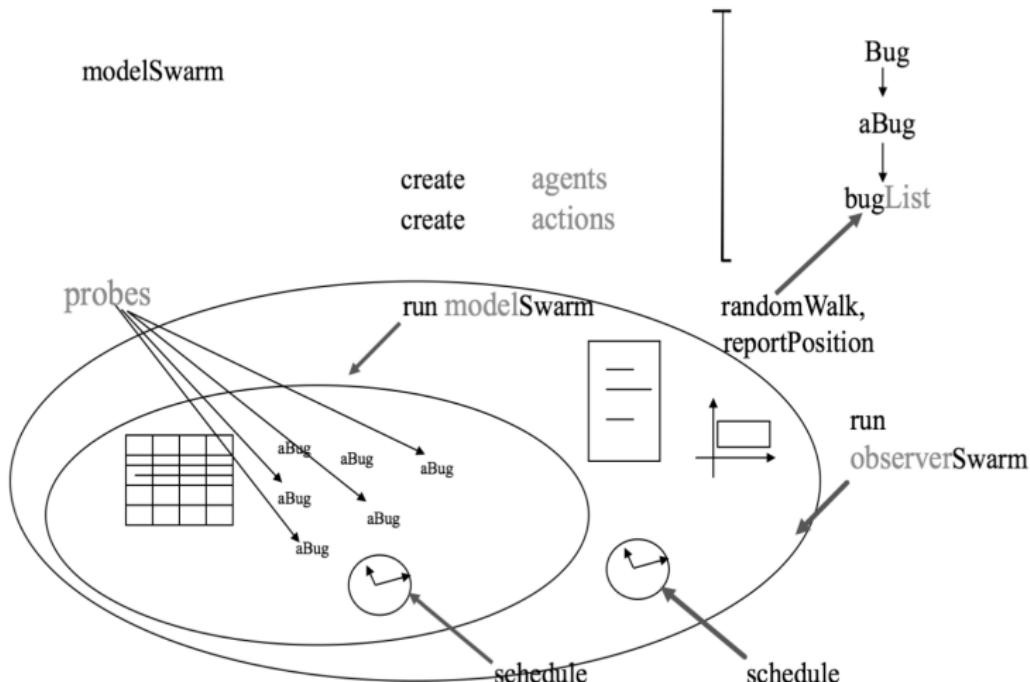
SLAPP logo: credits to [Steve Rogers](#).



<https://github.com/terna/SLAPP3/>

Scheme

Swarm—a library of functions and a protocol





An ABM on virus diffusion: a paper describing the model

G. Pescarmona, P. Terna, A. Acquadro, P. Pescarmona, G. Russo, E. Sulis, and S. Terna. *An Agent-Based Model of COVID-19 Diffusion to Plan and Evaluate Intervention Policies*, 2021. <https://arxiv.org/abs/2108.08885>.

Introduction

- A micro-based model of interacting agents, following plausible behavioral rules into a world where the Covid-19 epidemic is affecting the actions of everyone.
- The model works with:
 - i infected agents categorized as symptomatic or asymptomatic and
 - ii the places of contagion specified in a detailed way, thanks to agent-based modeling capabilities.
- The **infection transmission** is related to three factors: the infected person's characteristics and those of the susceptible one, plus those of the space in which a contact occurs.

- The micro-based structure of the model allows factual, counterfactual, and conditional simulations to investigate both the spontaneous or controlled development of the epidemic. Examples of counterfactual situations are those considering:
 - i different timing in the adoption of the non-pharmaceutical containment measures;
 - ii alternative strategies focusing exclusively on the defense of fragile people.
- The model generates complex epidemic dynamics, emerging from the consequences of agents' actions and interactions, with high variability in outcomes, but frequently with a stunning realistic reproduction of the contagion waves that occurred in the reference region.
- We take charge of the variability of the epidemic paths within the simulation, running batches of executions with 10,000 occurrences for each experiment.



- The **AI and inverse generative sides of the model** come from constructing a meta-agent optimizing the vaccine distribution among people groups—characterized by age, fragility, work conditions—to minimize the number of symptomatic people (deceased persons come from there).
- We can characterize the action of the planner both:
 - i introducing ex-ante rules following “plain” or “wise” strategies that we imagine as observers or
 - ii **evolving those strategies via the application of a genetic algorithm.**
- The genome is a matrix of vaccination quotas by people groups, with their time range of adoption.

The model

- As the agents can be Susceptible, Infected, symptomatic, asymptomatic, and Recovered, the name of the model is S.I.s.a.R., with the capital letters recalling the S.I.R. scheme.
- We use NetLogo, at <https://ccl.northwestern.edu/netlogo/>.
- S.I.s.a.R. is at <https://terna.to.it/simul/SIsaR.html> with information on model construction, and an online executable version.
- A paper is published at <https://arxiv.org/abs/2108.08885>
- The model includes the structural data of Piedmont, an Italian region, but we can easily calibrate it for other areas. The simulation reproduces a realistic calendar (e.g., national or local government decisions) via a dedicated script interpreter.



The scale and the items

- 1 : 1000, for a population of 4,350,000 people.
- Houses.
- Schools.
- Hospitals.
- Nursing homes,
- Factories.

The world 3D

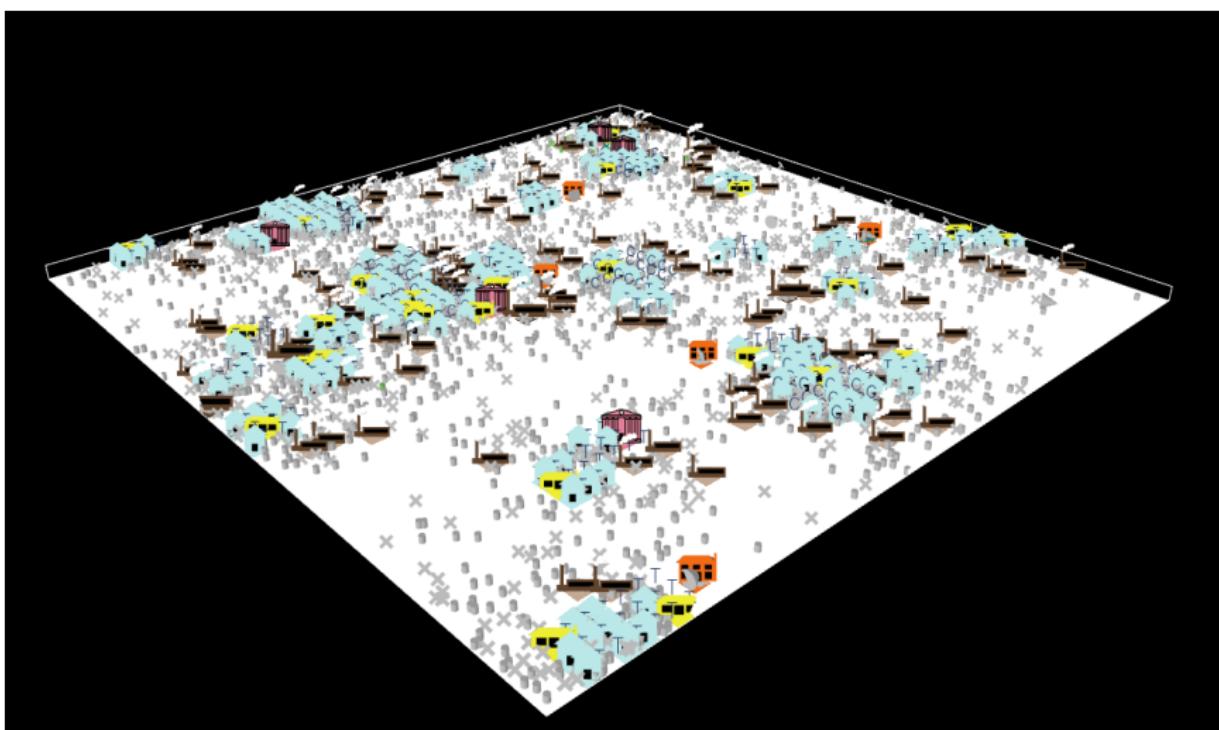


Figure 1: The world 3D

A circular scheme

S.I.s.a.R. outline

Sending people at home: the day starts at midnight, sending people at home (or holding them in hospitals or in nursing homes).

Diffusing contagion at home, in hospitals, nursing homes
contagion \Rightarrow probabilityOfGettingInfection * intrinsicSusceptibility

interpersonal distance are not relevant here, as people are in an enclosed space

$K = \#movementCyclesPerTick$ determines the number of actions/movements in a day
 K steps, with some action uniquely in step 1 (go to work or school)
 Going to: work, school, generically out of the home to visit persons (n . homes), shop, etc.
ablate

```
ndition I  
ndition II using  
PeopleNotFragileNotSymptomaticLeavingHome  
PeopleAnyTypeNotSymptomaticLeavingHome  
openFactoriesWhenLimitationsOn  
opFragileWorkers  
ctivateSchool with %Students
```

- People that cannot move stay at home
- Other steps, if any ($K > 1$)
 - Going out of the home to walk, shop, etc.
 - Diffusing infection, among any people moving in any place,
 - but school (only students and teachers enter there); here interpersonal distance is relevant,
 - with radiusOfInfection 

radiusOfInfection

Emerging immunity in vaccinated people (if any)

Managing infections: emerging after incubation or recovering

Diffusing infection in factories and school, among stable people there

(interpersonal distance are not relevant here, as people are in an enclosed space).

Diffusing infection, among any people moving in any place, but school

(only students and teachers enter there); here interpersonal distance is relevant, with `radiusOfInfection`

Figure 2: The scheme: def. and values of the parameters at <https://terna.to.it/simul/howSIsaRworks.pdf>

Contagion representation

- The model allows analyzing the sequences of contagions in simulated epidemics, reporting the places where the contagion occur.
- We represent each infected agent as a horizontal segment (from the starting date to the final date of the infection) with vertical connections to other agents if they receive the disease.
We represent the new infected agents via further segments at an upper level.
- With colors, line thickness, and styles, we display multiple information.
- This enables understanding at a glance how an epidemic episode is developing. In this way, it is easier to reason about countermeasures and, thus, to develop intervention policies.

Examples (1/2)

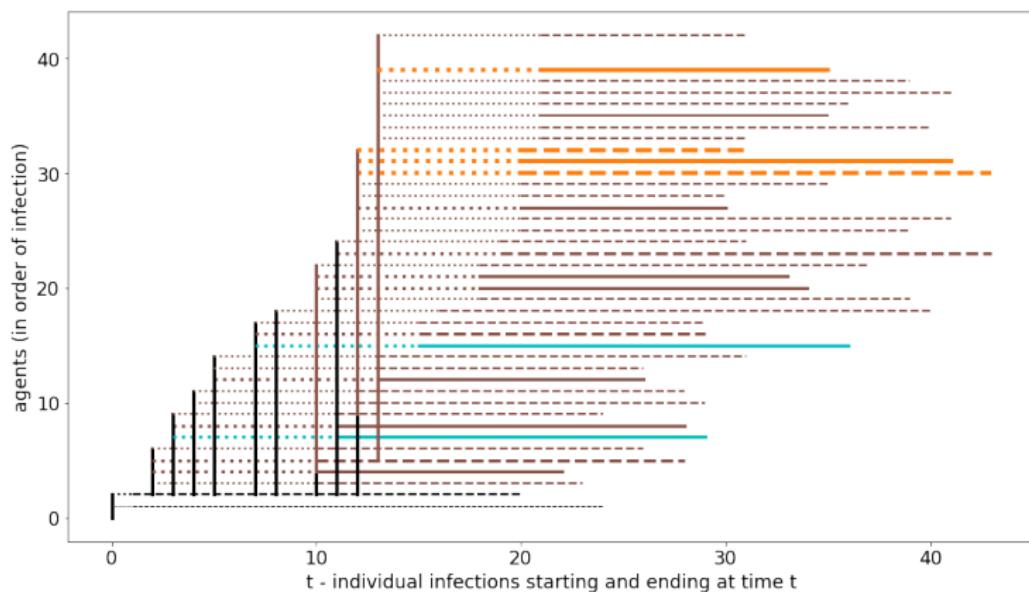
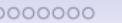
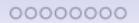
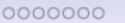


Figure 3: A case with containment measures, first 40 infections: workplaces (brown) and nursing homes (orange) strictly interweaving



Examples (2/2), whole epidemic

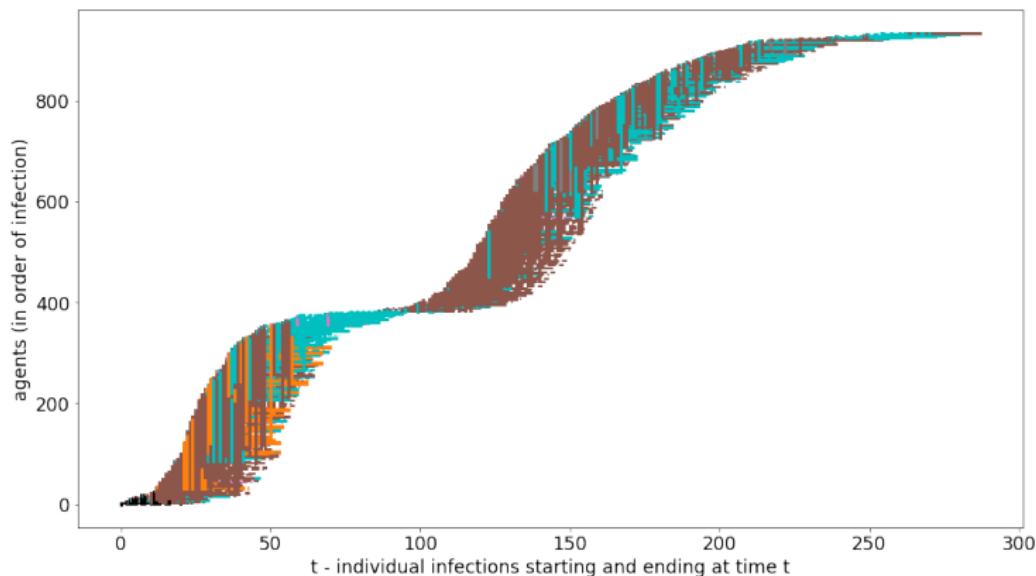


Figure 4: A Case with containment measures, the whole epidemics: workplaces (brown) and nursing homes (orange) and then houses (cyan), with a bridge connecting two waves



Simulation batches

- We explore systematically the introduction of factual, counterfactual, and prospective interventions to control the spread of the contagions.
- Each simulation run—whose length corresponds to the disappearance of symptomatic or asymptomatic contagion cases—is a datum in a wide scenario of variability in time and effects.
- We need to represent compactly the results emerging from batches of simulation repetitions, to compare the consequences of the basic assumptions adopted for each specific batch.
- Besides summarizing the results with the usual statistical indicators, we adopt the technique of the heat-maps.
- Each heat-map reports the duration of each simulated epidemic in the x axis and the number of the symptomatic, asymptomatic, and deceased agents in the y axis. The z axis is represented by the colors, as in the logarithmic scale on the right of each picture.
- In our batches we have 10,000 runs.

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10,000 epidemics without control in Piedmont

	symptomatic	totalInfected&Deceased	duration
count	10000.00	10000.00	10000.00
mean	969.46	2500.45	303.10
std	308.80	802.88	93.50

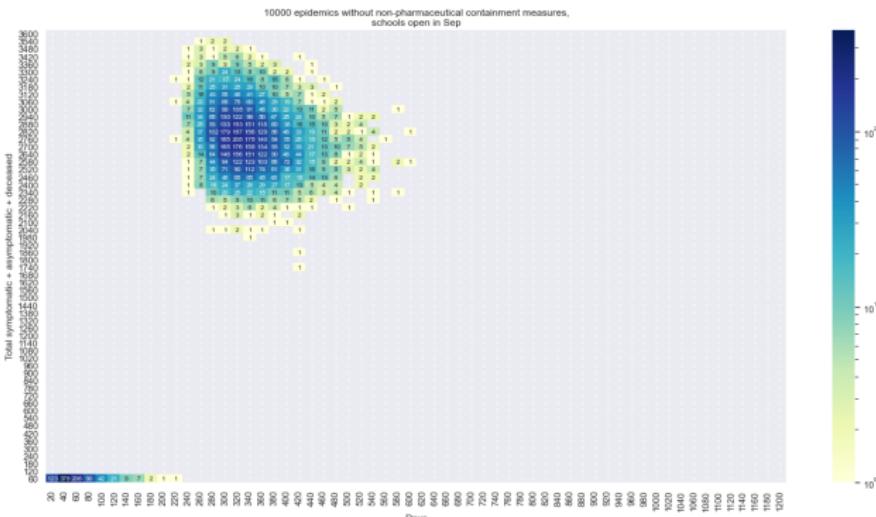


Figure 5: Without non-pharmaceutical containment measures

10,000 epidemic with basic control in Piedmont

	symptomatic	totalInfected&Deceased	duration
count	10000.00	10000.00	10000.00
mean	344.22	851.64	277.93
std	368.49	916.41	213.48

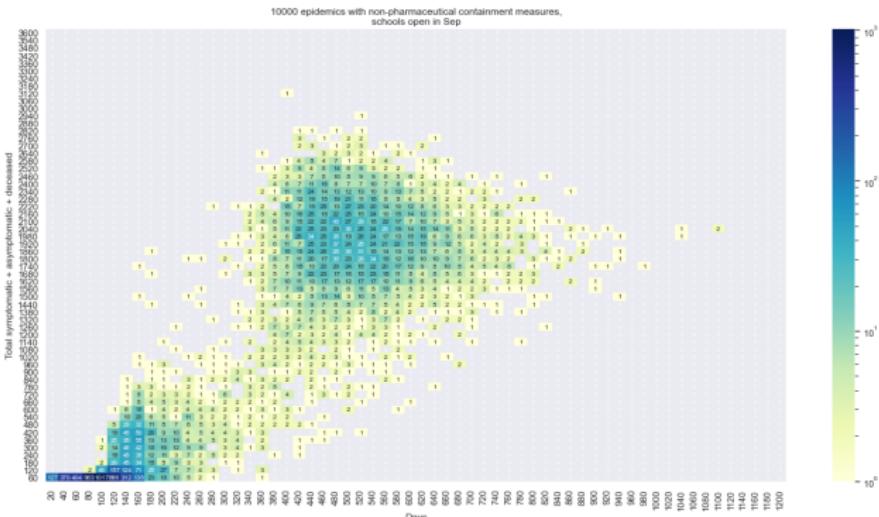


Figure 6: First wave with non-pharmaceutical containment measures

Key points (Summer and Fall 2020)

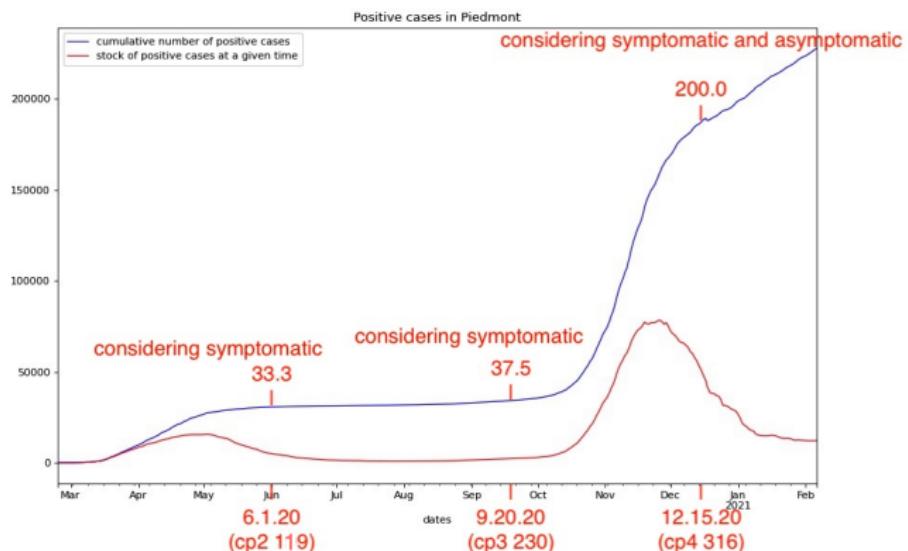


Figure 7: key points in epidemic dynamic in Summe and Fall 2020



Non homogeneous data

- Following the Civil Protection Department web site <http://www.protezionecivile.it/web/guest/department>, we find the repository <https://github.com/pcm-dpc/COVID-19>.
- In the first wave we had uniquely data about symptomatic infected people, but from October 2020 data are mixed.
- In the above *git* repository, in October and November we had “Positive cases emerged from clinical activity”, unfortunately then reported as “No longer populated” (from the end of November, my observation) and “Positive cases emerging from surveys and tests, planned at national or regional level”, again “No longer populated” (from the end of November, my observation).
- Using those two series, it was possible to estimate a subdivision between symptomatic and asymptomatic cases, which is no longer possible.

Updated series, with a third wave (data until the beginning of June)

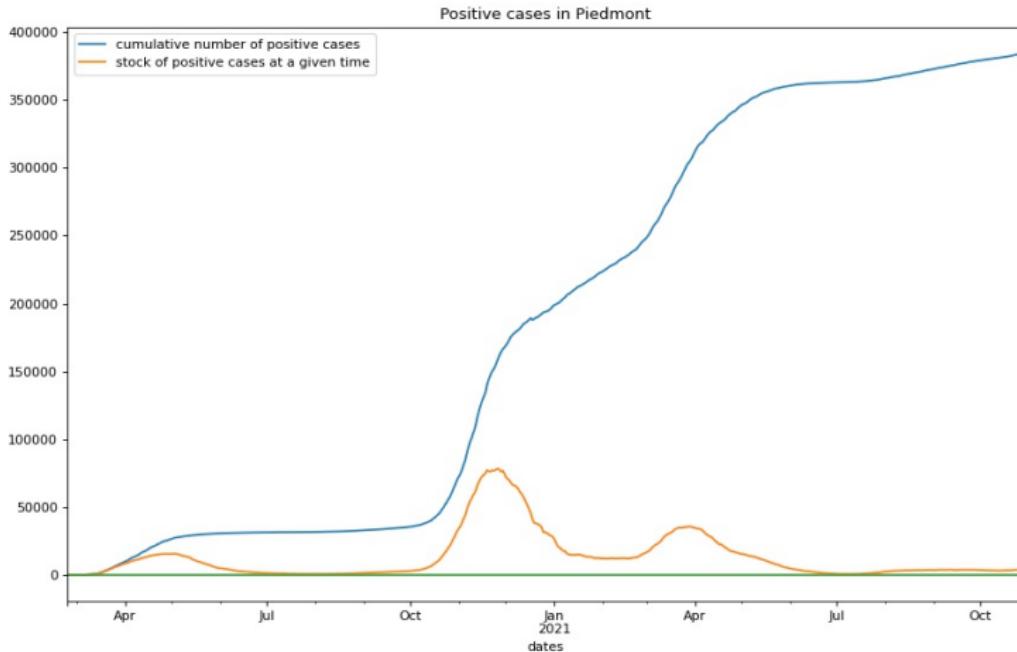


Figure 8: Data for Piedmont

Spontaneous second wave, without specific measures

170 epidemics stable in Summer 2020 out of 10,000, rule: at Jun 1, 20 select if sym. (10, 70], actual v. 33.3 & at Sep 20, 20 select if sym. (20, 90], actual value 37.5; **140** at Dec 15, 20, rule: sym.+asym.>Sep 20, 20, actual value: 200.0.

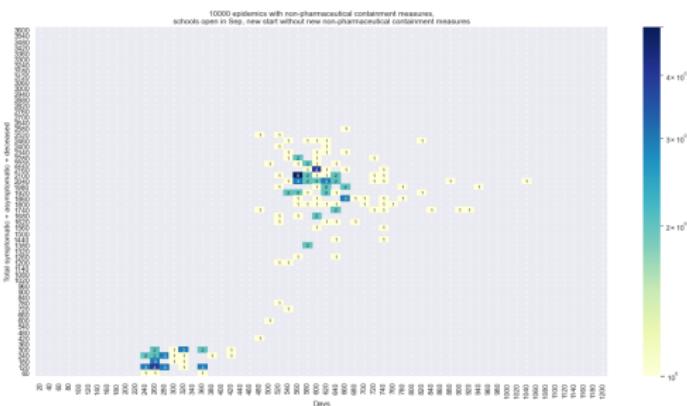


Figure 9: First wave with non-pharmaceutical containment measures, spontaneous second wave, without specific measures

Second w., new infections from outside, without specific measures

1407 epidemics stable in Summer 2020 out of 10,000, rule: at Jun 1, 20 select if sym. (10, 70], actual v. 33.3 & at Sep 20, 20 select if sym. (20, 90], actual value 37.5; **1044** at Dec 15, 20, rule: sym.+asym.>Sep 20, 20, actual value: 200.0.

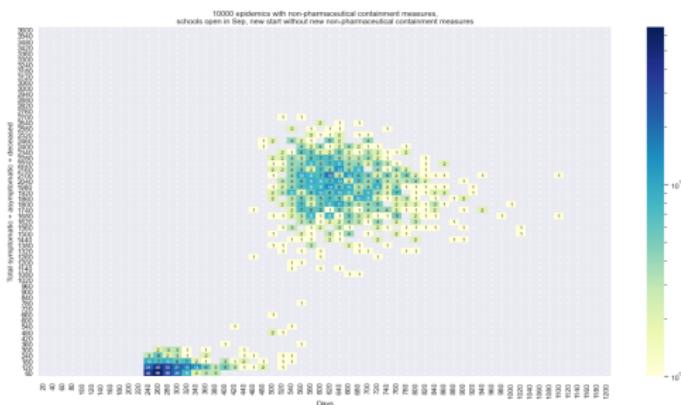


Figure 10: First wave with non-pharmaceutical containment measures, forcing the second wave, without specific measures

(1000) cum. v.	Jun 1, 20 sym.	Jun 1, 20 all	Sep 9, 20 symp.	Sep 9, 20 totalInf.	Dec 15, 20 symp.	Dec 15, 20 totalInf.	Feb 1, 21 symp.	Feb 1, 21 totalInf.	May 1, 21 symp.	May 1, 21 totalInf.	Dec 15, 20 symp.	Dec 15, 20 totalInf.	to end days
count	1407.0	1407.0	1407.0	1407.0	1044.0	1044.0	1005.0	1005.0	980.0	980.0	1044.0	1044.0	1044.0
mean	35.6	72.7	40.0	84.1	180.4	462.1	354.1	900.4	623.8	1563.3	726.6	1810.9	620.9
std	14.1	42.6	16.7	52.8	134.6	354.6	213.8	535.4	217.9	527.0	221.9	544.0	110.8

Second w., new infections from outside, with new specific measures

1407 epidemics stable in Summer 2020 out of 10,000, rule: at Jun 1, 20 select if sym. (10, 70], actual v. 33.3 & at Sep 20, 20 select if sym. (20, 90], actual value 37.5; **874** at Dec 15, 20, rule: sym.+asym.>Sep 20, 20, actual value: 200.0.

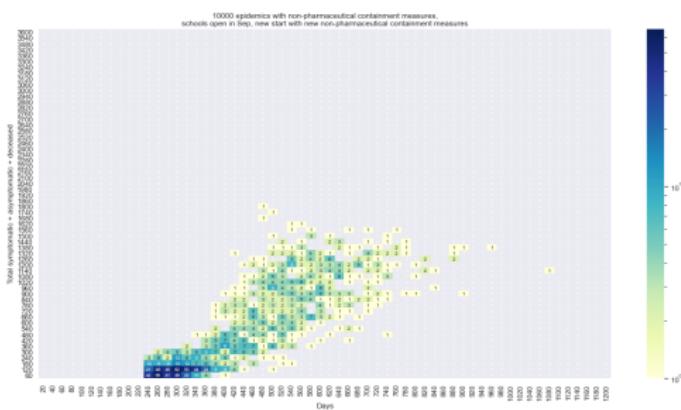


Figure 11: First wave with non-ph. containment measures, forcing the second wave, **with new specific non-ph. containment measures**

(1000) cum. v.	Jun 1, 20		Sep 9, 20		Dec 15, 20		Feb 1, 21		May 1, 21		Dec 15, 20 to end	
	sym.	all	symp.	totalInf.	symp.	totalInf.	symp.	totalInf.	symp.	totalInf.	symp.	totalInf.
count	1407.0	1407.0	1407.0	1407.0	874.0	874.0	719.0	719.0	523.0	523.0	874.0	874.0
mean	35.6	72.7	40.0	84.1	130.0	340.6	194.4	512.8	295.7	791.2	252.7	666.4
std	14.1	42.6	16.7	52.8	83.9	232.6	104.1	276.9	119.1	300.6	156.8	416.4
												122.7

Time factor

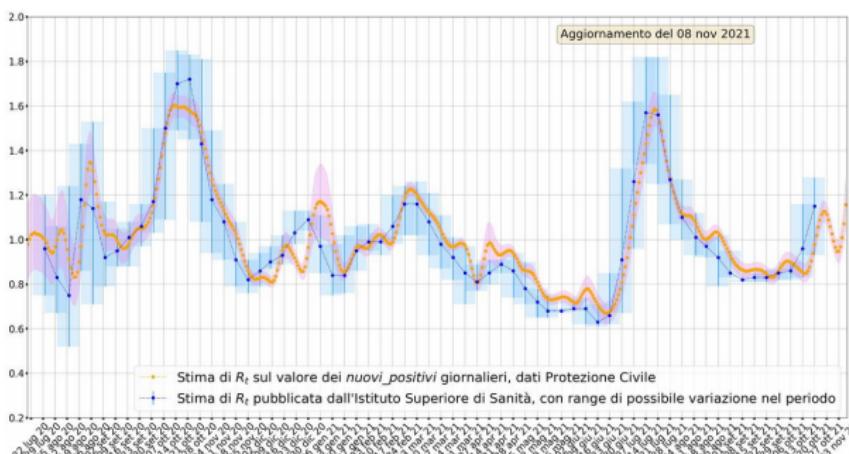


Figure 12: In blue the R_t values as reported by the Istituto Superiore di Sanità and in red the calculation published regularly at <https://mondoeconomico.eu> by Stefano Terna¹.

¹Methodology: https://github.com/tomorrowdata/COVID-19/blob/main/notebooks/Rt_on_italian_national_data.ipynb and Section 5.4 at <https://arxiv.org/abs/2108.08885>



Second w., new infect. from outside, with new specific meas. -20 days²

1407 epidemics stable in Summer 2020 out of 10,000, rule: at Jun 1, 20 select if sym. (10, 70], actual v. 33.3 & at Sep 20, 20 select if sym. (20, 90], actual value 37.5; **769** at Dec 15, 20, rule: sym.+asym.>Sep 20, 20, actual value: 200.0.

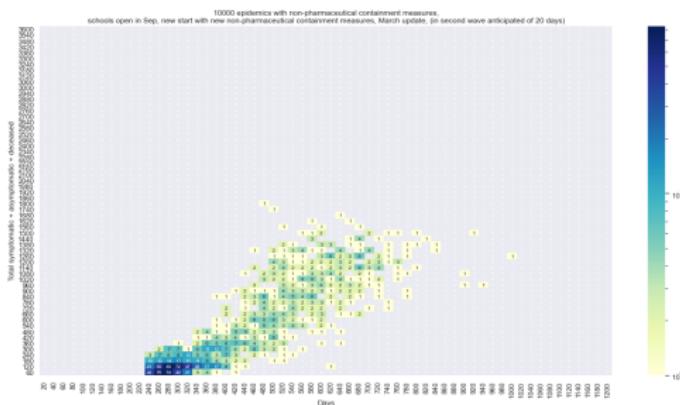
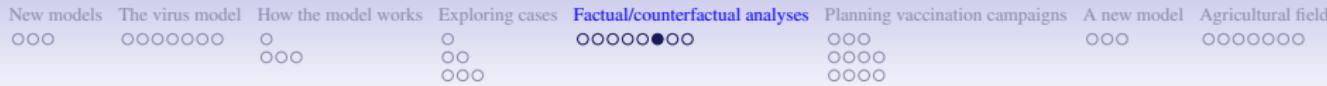


Figure 13: First wave with non-ph. cont. meas., forcing the second wave, with new specific non-ph. cont. meas., 20 day anticipation

(1000) cum. v.	Jun 1, 20		Sep 9, 20		Dec 15, 20		Feb 1, 21		May 1, 21		Dec 15, 20 to end		
	sym.	all	symp.	totalInf.	symp.	totalInf.	symp.	totalInf.	symp.	totalInf.	symp.	totalInf.	days
count	1407.0	1407.0	1407.0	1407.0	769.0	769.0	637.0	637.0	471.0	471.0	769.0	769.0	769.0
mean	35.6	72.7	40.0	84.1	112.2	294.2	172.0	467.9	276.5	748.6	248.9	663.4	499.3
std	14.1	42.6	16.7	52.8	66.8	188.4	91.5	251.3	112.9	286.9	158.0	417.5	124.1

²N.B.: (i) anticipation limit Oct 5.; (ii) also the ending date of each measure is anticipated of 20 days.





Fragile persons

- A possible strategy is to stop all fragile people for a given period when R_t starts increasing (also with fragile workers in sick leave, if unable to work remotely).
- We have also relevant social benefits, e.g., schooling, and economic benefits, as activities do not stop

Sec. w., new infect. from outs., stop fragile people. 60 days from Oct. 5³

1407 epidemics stable in Summer 2020 out of 10,000, rule: at Jun 1, 20 select if sym. (10, 70], actual v. 33.3 & at Sep 20, 20 select if sym. (20, 90], actual value 37.5; **886** at Dec 15, 20, rule: sym.+asym.>Sep 20, 20, actual value: 200.0.

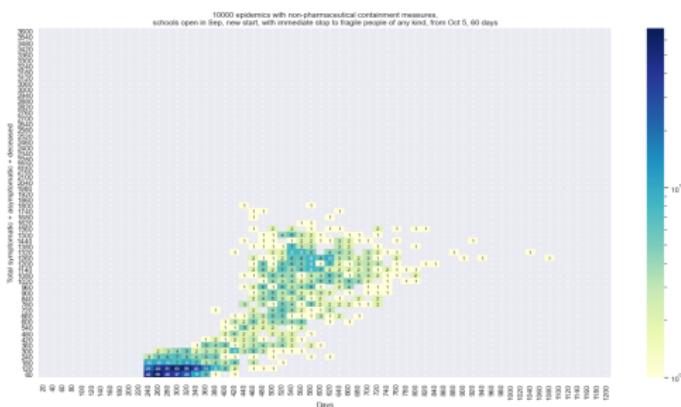


Figure 14: First wave with non-ph. cont. meas., forcing the sec. w.; **in sec. w., uniquely stop fragile people, including fragile workers**

(1000)	Jun 1, 20 sym.	Sep 9, 20 all	Sep 9, 20 symp.	Dec 15, 20 totalInf.	Feb 1, 21 symp.	May 1, 21 symp.	Dec 15, 20 symp.	to end totalInf.	days
count	1407.0	1407.0	1407.0	1407.0	886.0	886.0	761.0	761.0	886.0
mean	35.6	72.7	40.0	84.1	128.1	326.3	211.0	555.1	323.3
std	14.1	42.6	16.7	52.8	89.6	234.2	118.1	306.7	170.7

³ Schools are always working 100% in this case.

To recap (all waves)

Scenarios			Dec 15, 20 symp.	Dec 15, 20 totalInf.	to end symp.	to end totalInf.	days
no							
containments in spontaneous second wave	count	140.0	140.0	140.0	140.0	140.0	
	mean	248.4	648.7	701.1	1757.9	594.2	
	std	167.4	424.3	246.4	599.7	118.9	
no							
containments in forced second wave	count	1044.0	1044.0	1044.0	1044.0	1044.0	
	mean	180.4	462.1	726.6	1810.9	620.9	
	std	134.6	354.6	221.9	544.0	110.8	
basic							
containments in forced second wave	count	874.0	874.0	874.0	874.0	874.0	
	mean	130.0	340.6	252.7	666.4	494.1	
	std	83.9	232.6	156.8	416.4	122.7	
-20 days							
containments in forced second wave	count	769.0	769.0	769.0	769.0	769.0	
	mean	112.2	294.2	248.9	663.4	499.3	
	std	66.8	188.4	158.0	417.5	124.1	
frag. p. & workers							
control in forced second wave	count	886.0	886.0	886.0	886.0	886.0	
	mean	128.1	326.3	301.1	792.3	515.5	
	std	89.6	234.2	170.7	450.2	116.9	

Table 1: Report of the key results, with count, mean, and std

Planning a vaccination campaign using GAs (with non-pharmaceutical containment measures in action)

- Exploring vaccination sequences, using *genetic algorithms*. A detailed note, frequently updated, is at
<https://terna.to.it/simul/GAresultPresentation.pdf>.
- We compare the effect of choosing the vaccination quotas via GAs with two predetermined strategies, considering three hypotheses (vaccinated people: still spread the contagion; do not spread the contagion; do it in the 50% of the case); we show here only the first case results.
- Key dates:
 - in the internal calendar of the model, day 373 is Feb. 12th, 2021, which is effectively the starting point of the vaccinations in the region;
 - the day of the effectiveness of the initial vaccinations, 40 days later, is day 413 (Mar. 22nd, 2021).

Vaccination groups

We take into consideration seven groups in order of decreasing fragility but also considering the exposure to contagion:

g1 extra fragile people with three components;

- due to intrinsic characteristics: people in nursing homes;
- due to risk exposure:
 - nursing homes operators;
 - healthcare operators;

g2 teachers;

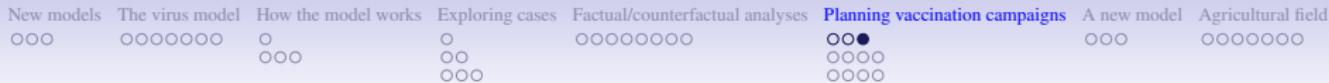
g3 workers with medical fragility;

g4 regular workers;

g5 fragile people without special characteristics;

g6 regular people, not young, not worker, and not teacher;

g7 young people excluding special activity cases (a limited number in *g1*).



A specific realistic case

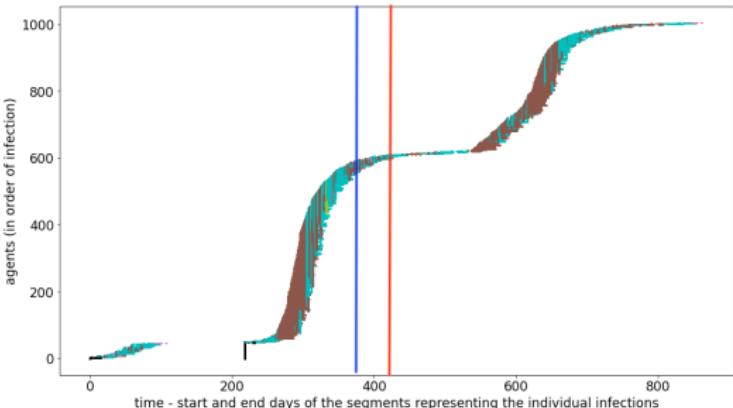


Figure 15: Crucial dates: blue line for the starting point of the vaccination campaign and red line for the start of the effectiveness of the initial vaccinations

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Time dynamics without vaccinations

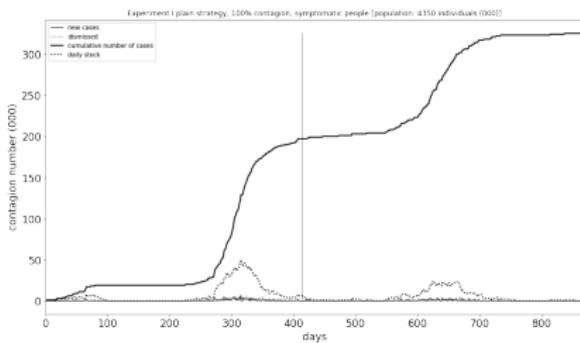


Figure 16: Experiment I, 'base symptomatic series; the vertical line is at day 413 is not relevant here'



Time dynamics with *plain* vac. strategy, vac. people still spreading the infection

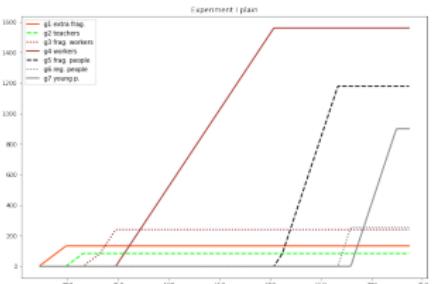


Figure 17: “Plain” vaccination sequence; on the y axis the number of vaccinated subjects of each group (if vaccination is complete, the line is horizontal)

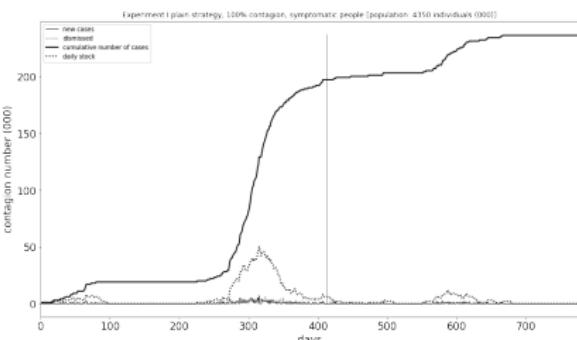


Figure 18: “Plain” vaccination symptomatic series; the vertical line is at day 413, when the effectiveness of first vaccination starts

Time dynamics with *wise* vac. strategy, vac. people still spreading the infection

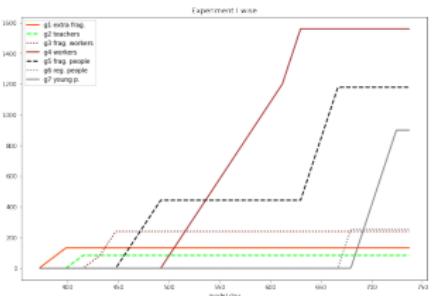


Figure 19: “Wise” vaccination sequence; on the y axis the number of vaccinated subjects of each group (if vaccination is complete, the line is horizontal)

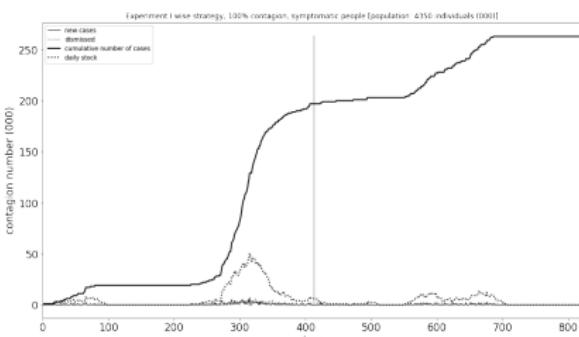


Figure 20: “Wise” vaccination symptomatic series; the vertical line is at day 413, when the effectiveness of first vaccination starts

Time dynamics with best GAs strategy, vac, people still spreading the infection

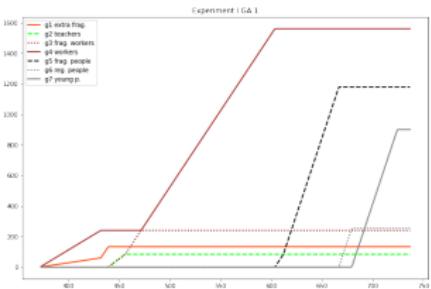


Figure 21: GA 1 vaccination sequence; on the y axis the number of vaccinated subjects of each group (if vaccination is complete, the line is horizontal)

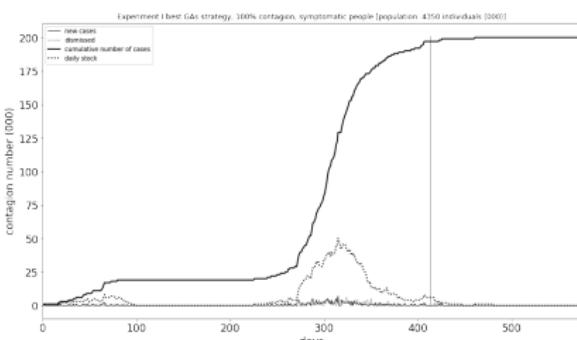


Figure 22: GAs vaccination symptomatic series; the vertical line is at day 413, when the effectiveness of first vaccination starts



What if

What if we increase quantities in *plain* and *wise* strategies?

From day	Q.	Q. +
373	5	10
433	10	15
493	10	25
553	10	25
613	20	25
738	end	

Table 2: New daily quantities



Time dynamics with *plain+* vac. strategy, v. people still spreading the infection

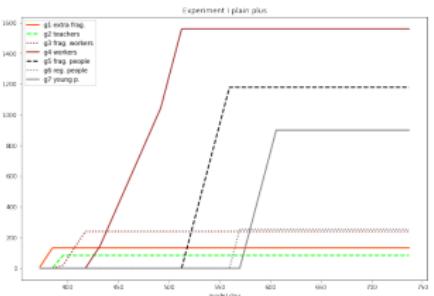


Figure 23: “Plain+” vaccination sequence; on the y axis the number of vaccinated subjects of each group (if vaccination is complete, the line is horizontal)

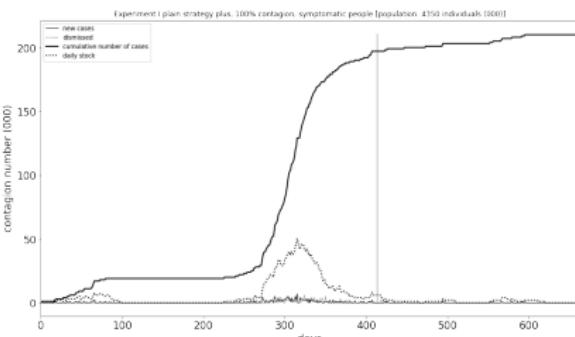


Figure 24: “Plain” vaccination symptomatic series; the vertical line is at day 413, when the effectiveness of first vaccination starts



Time dynamics with *wise+* vac. strategy, v. people still spreading the infection

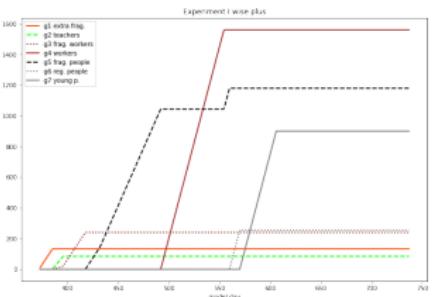


Figure 25: “Wise+” vaccination sequence; on the y axis the number of vaccinated subjects of each group (if vaccination is complete, the line is horizontal)

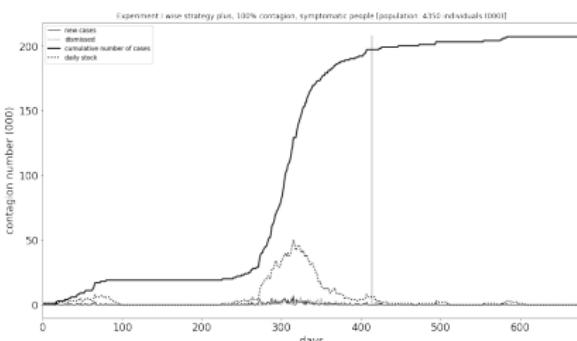


Figure 26: “Wise+” vaccination symptomatic series; the vertical line is at day 413, when the effectiveness of first vaccination starts



Synopsis

Hypothesis: vaccinated people, if infected, are diffusing the contagion.

Case day 413 (1000)	At no vaccin.	Final plain vaccin.	Final wise vaccin.	Final GAs vaccin.	Final plain + vaccin.	Final wise + vaccin.
I	197	325	236	263	200	210
-	128	39	66	3	13	10

Table 3: Results of the vaccination campaigns: only symptomatic people (second row: minus day 413)

A new model: the map

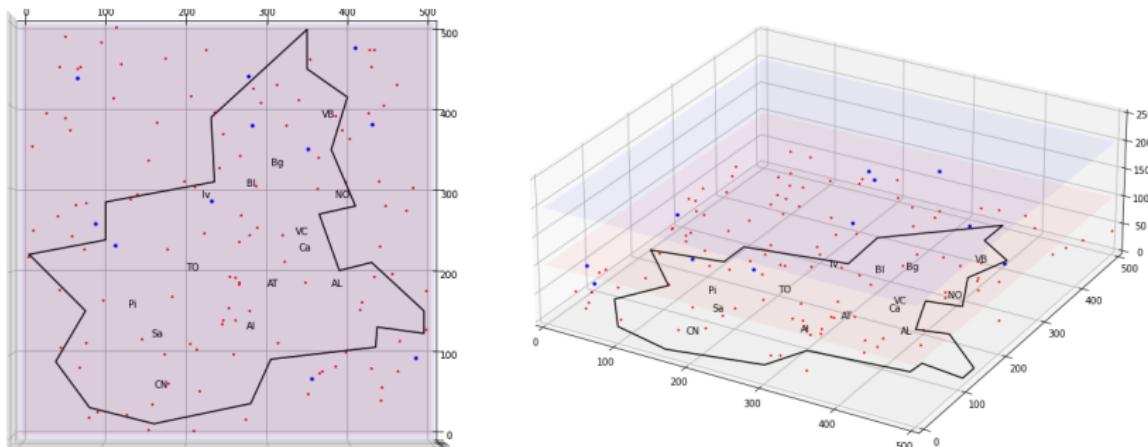


Figure 27: 3D Piedmont



A new model: the scale and the items

- 1 : 100.
- *Infection engine*,
<https://terna.to.it/simul/InfectionEngine.pdf>.
- Houses.
- Schools.
- Hospitals.
- Nursing homes,
- Factories.
- Transportations.
- Aggregation places: happy hours, night life, sport stadiums, discotheques, ...
- Networks (family networks, professional networks, high-contact individuals,⁴ ...)

⁴G. Manzo and A. van de Rijt. Halting sars-cov-2 by targeting high-contact individuals. Journal of Artificial Societies and Social Simulation, 23(4):10, 2020. ISSN 1460-7425. doi: 10.18564/jasss.4435. URL <http://jasss.soc.surrey.ac.uk/23/4/10.html>.



The tool: S.L.A.P.P.

Scientific advertising: <https://terna.github.io/SLAPP/>

SLAPP

Swarm-Like Agent Protocol in Python

[View the Project on GitHub](#)
terna/SLAPP



SLAPP

What version of Python do you use?

SLAPP3 uses Python 3

SLAPP2 uses Python 2

Swarm-Like Agent Protocol in Python

At **SLAPP 3**, you have SLAPP running in Python 3 (in the [SLAPP repository](#) you have a lot related material and a large set of old versions; the 2.0.x version is the last one related to Python 2).

We have here also a [Reference Handbook](#) (it is still a draft and has to be improved).

Five chapters of the book of Boero, R., Morini, M., Sonnessa, M., and Terna, P., [Agent-based Models of the Economy - From Theories to Applications](#), are related to SLAPP.

This project is maintained by [terna](#)

Hosted on GitHub Pages — Theme by [orderedlist](#)

The new book of Mazzoli, M., Morini, M., and Terna, P., [Rethinking Macroeconomics with Endogenous Market Structure](#), is deeply based on SLAPP.

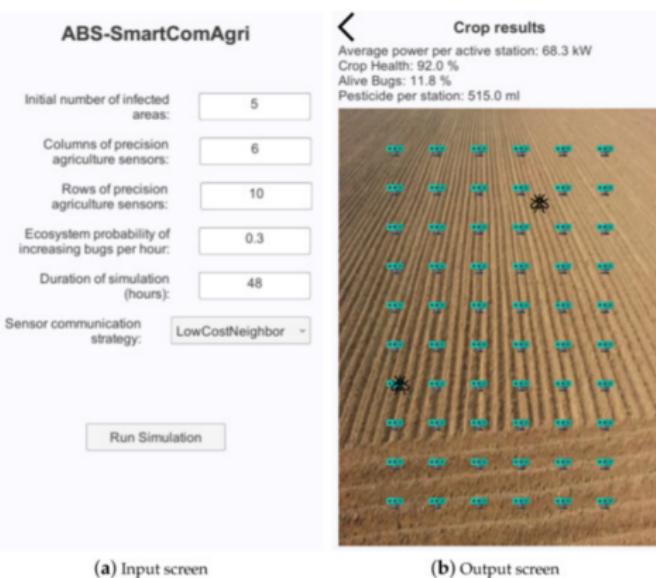
Figure 28: Swarm-Like Agent Protocol in Python



ABMs in agricultural field

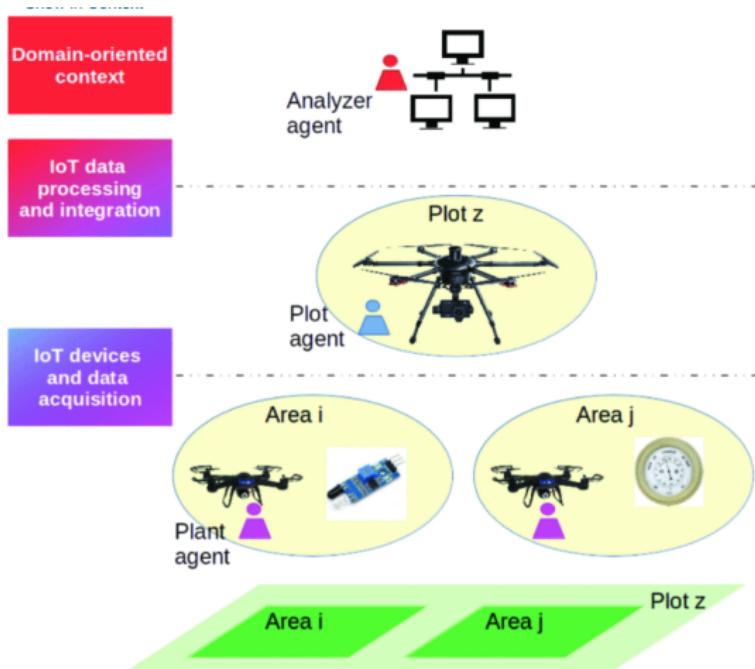
An offer of cooperation.

Examples in agricultural field (1/5)



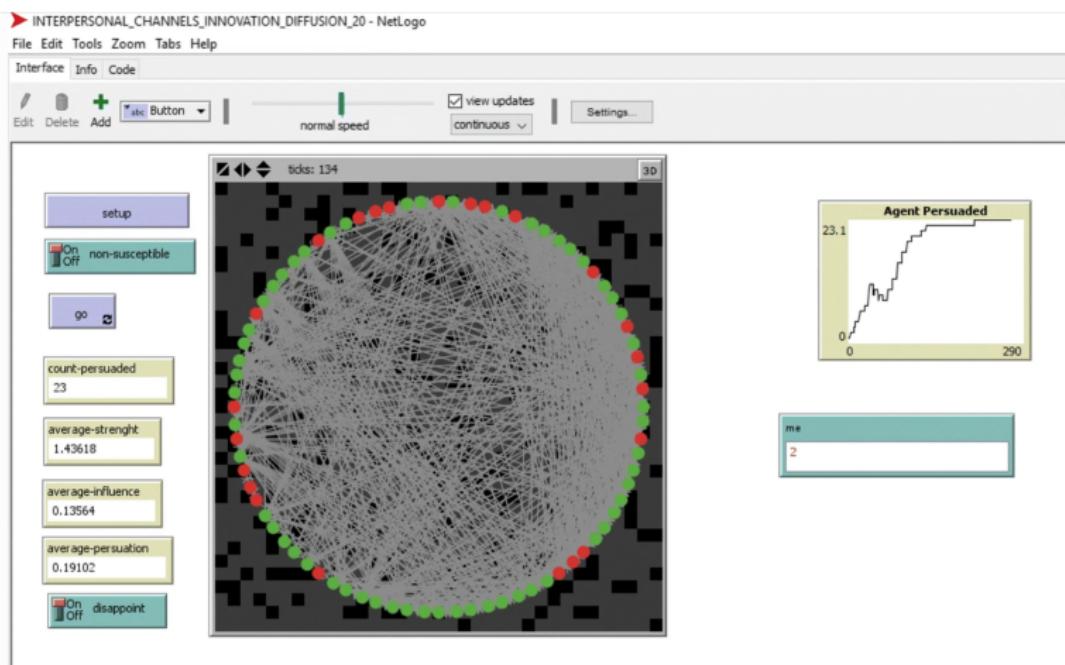
ABS-SmartComAgri: An Agent-Based Simulator of Smart Communication Protocols in Wireless Sensor Networks for Debugging in Precision Agriculture

Examples in agricultural field (2/5)



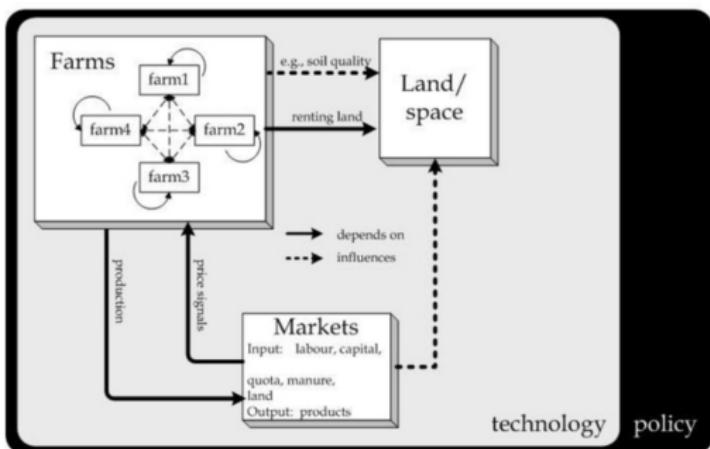
Towards a layered agent-modeling of IoT devices to precision agriculture

Examples in agricultural field (3/5)



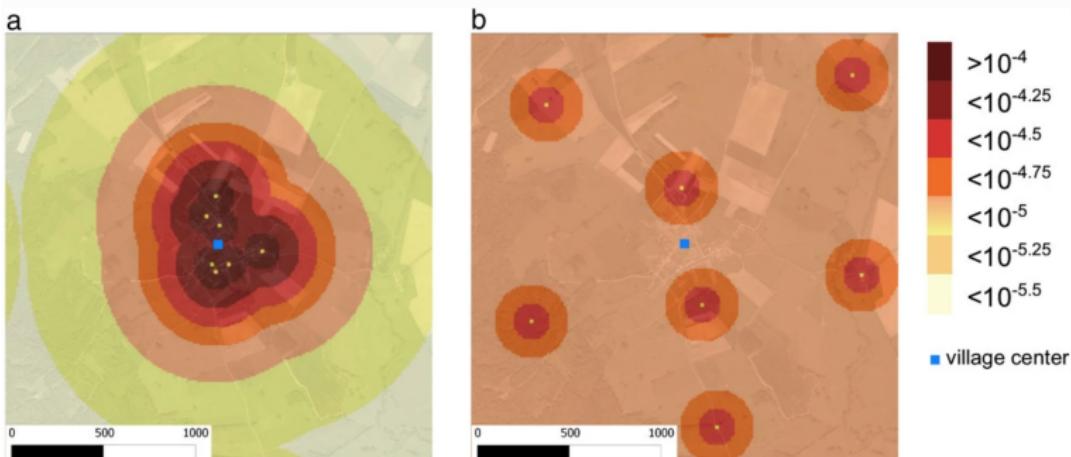
Improving diffusion in agriculture: an agent-based model to find the predictors for efficient early adopters

Examples in agricultural field (5/5)



Agent-based Analysis of Agricultural Policies: an Illustration of the Agricultural Policy Simulator AgriPoliS, its Adaptation and Behavior

Examples in agricultural field (5/5)



Maps of the average level of environmental contamination predicted in the site by the ABM for each setting (a setting 1, b setting 2). Farms are represented by yellow squares and the average levels of contamination of each distance class are represented by the different colors. Levels of contamination ranged from 10^{-4} to $10^{-5.5}$. The center of the village is for both settings the geometric center of the modeled area.

Agricultural landscape and spatial distribution of *Toxoplasma gondii* in rural environment: an agent-based model



Many thanks.

Slides at <https://terna.to.it/aa.pdf>.

Home page at <https://terna.to.it>.