



NARST 94<sup>TH</sup> ANNUAL INTERNATIONAL CONFERENCE



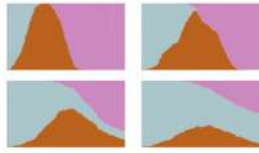
ALMA MATER STUDIORUM  
UNIVERSITÀ DI BOLOGNA

# Scientific simulations as educational tools for the post-pandemic era: the case of the Susceptible-Infectious-Removed model

**Eleonora Barelli**, Olivia Levrini

[eleonora.barelli2@unibo.it](mailto:eleonora.barelli2@unibo.it)

Department of Physics and Astronomy  
University of Bologna



Health

## Why outbreaks like coronavirus spread exponentially, and how to “flatten the curve”

By [Harry Stevens](#) March 14, 2020

NEWS

### **New COVID-19 Modelling Suggests Slower Spread, But More Deaths Expected**

Ontario and Quebec account for about 80 per cent of cases in Canada.

Laura Osman, Canadian Press

[Health](#) | [Local News](#) | [Northwest](#) | [Science](#)

#### **Mathematical models help predict the trajectory of the coronavirus outbreak. But can they be believed?**

May 3, 2020 at 6:00 am | Updated May 3, 2020 at 12:42 pm



How modelling Covid has changed the way we think about epidemics

*Adam Kucharski*

- During the COVID-19 pandemic, the whole society has been exposed to models and computational simulations
- Tools that were routine for experts became part of the popular vocabulary and were used to make decisions
- Simulations embody epistemic and methodological knowledge that deeply changed the ways in which the scientific community today addresses complex problems (Galison, 1996; Winsberg, 1999)



ALMA MATER STUDIORUM  
UNIVERSITÀ DI BOLOGNA

# The goal of the study

We want to show

- how to exploit in teaching (from high-school students to higher levels) the epistemic and societal value of scientific simulations to tackle with the issues raised by COVID-19 pandemic
- how scientific simulations can become an educational tool to implement the policy framework “OECD Future of Education and Skills” (OECD, 2018)



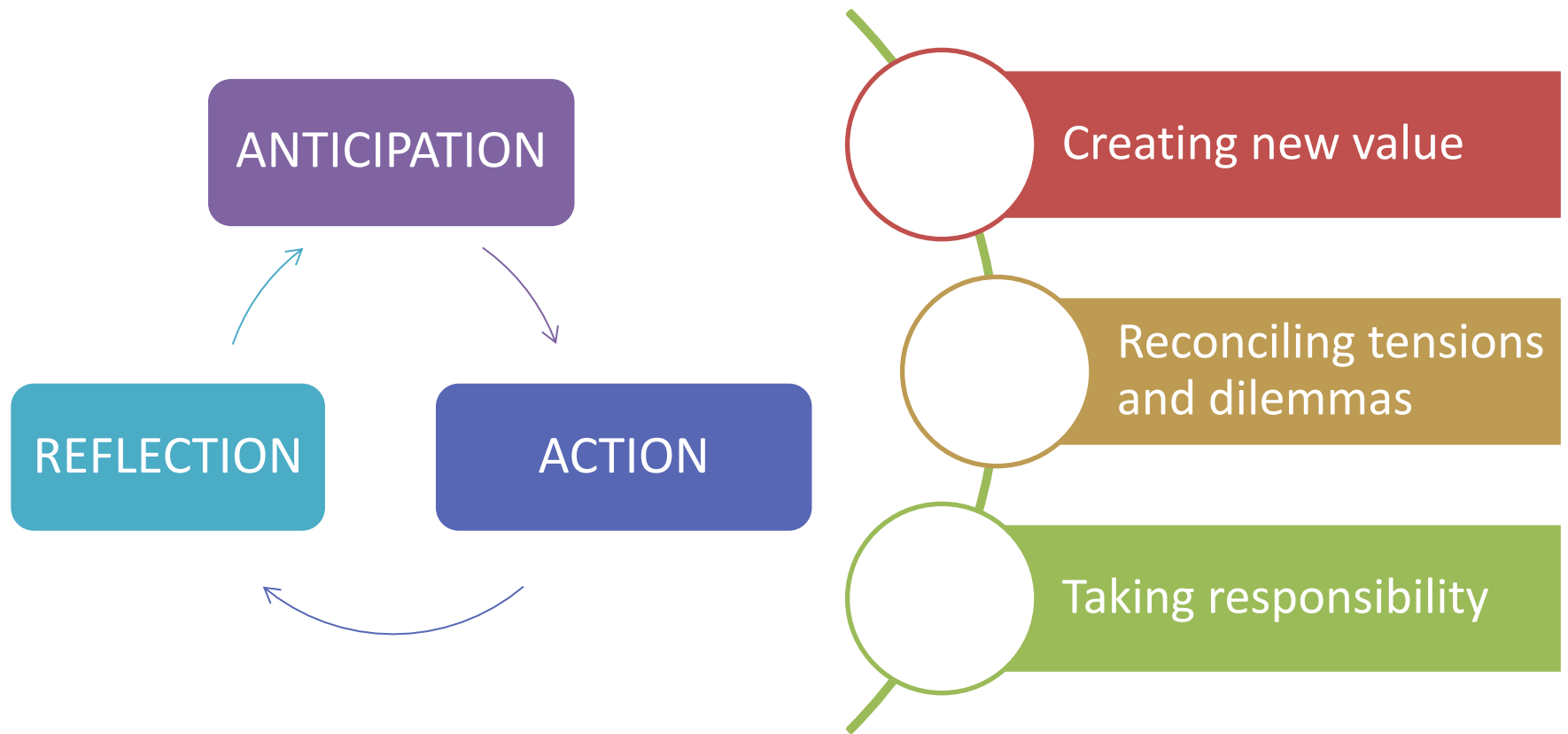
# The framework: OECD Learning Compass 2030



- A non-educational framework that includes in a comprehensive picture outstanding issues that educational research is committed to investigate (skills for sustainable development, future-thinking and agency)
- Similarity with the framework of future-oriented science education and future-scaffolding skills (Levrini et al., 2019, 2021), elaborated within science education research



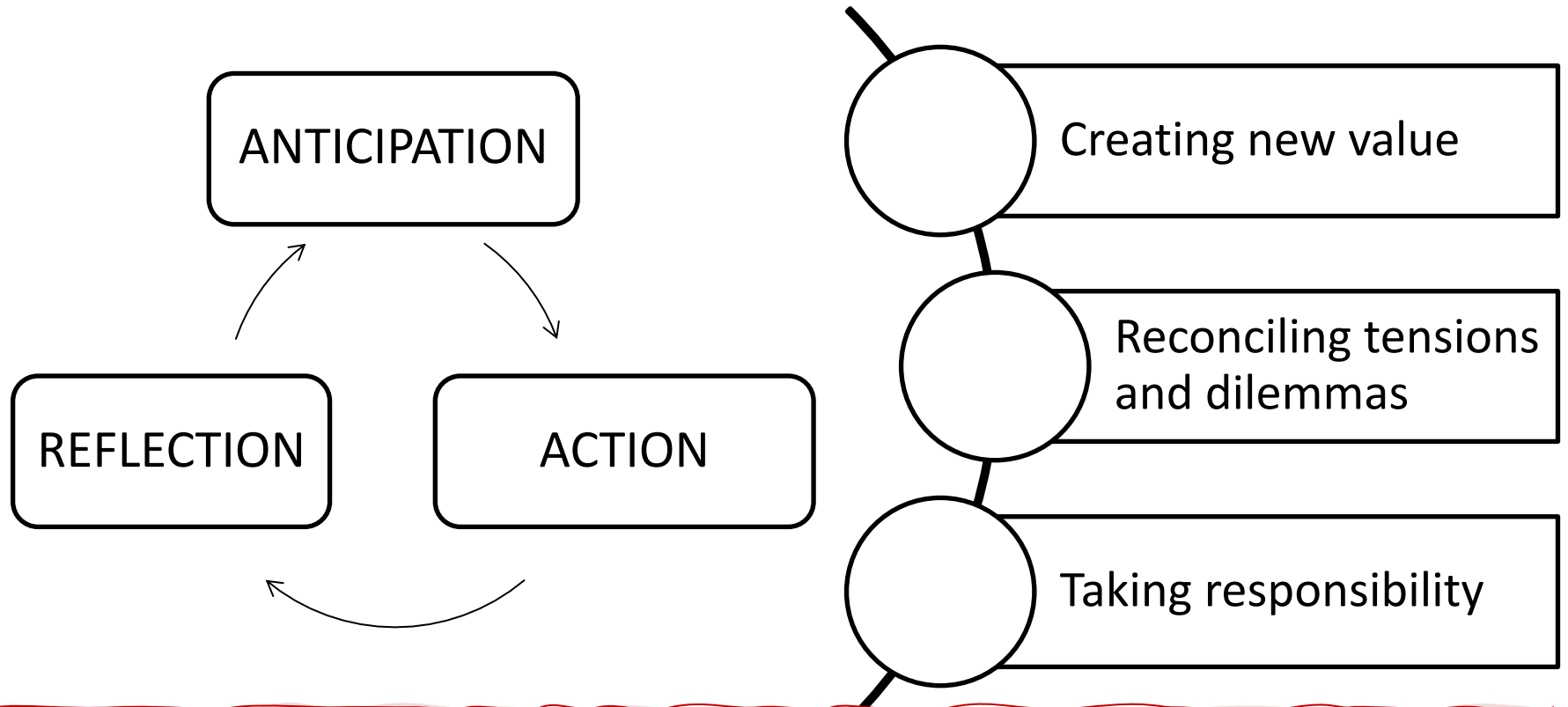
# The framework: OECD Learning Compass 2030



AAR cycle and Transformative Competences



# The framework: OECD Learning Compass 2030



To show how simulations can be related to the elements of the framework, we examine the **epistemic details** of SIR simulations as they emerge from the comparison between **equation-** and **agent-based** approaches

# Equation-based vs agent-based approaches

## Equation-based simulations

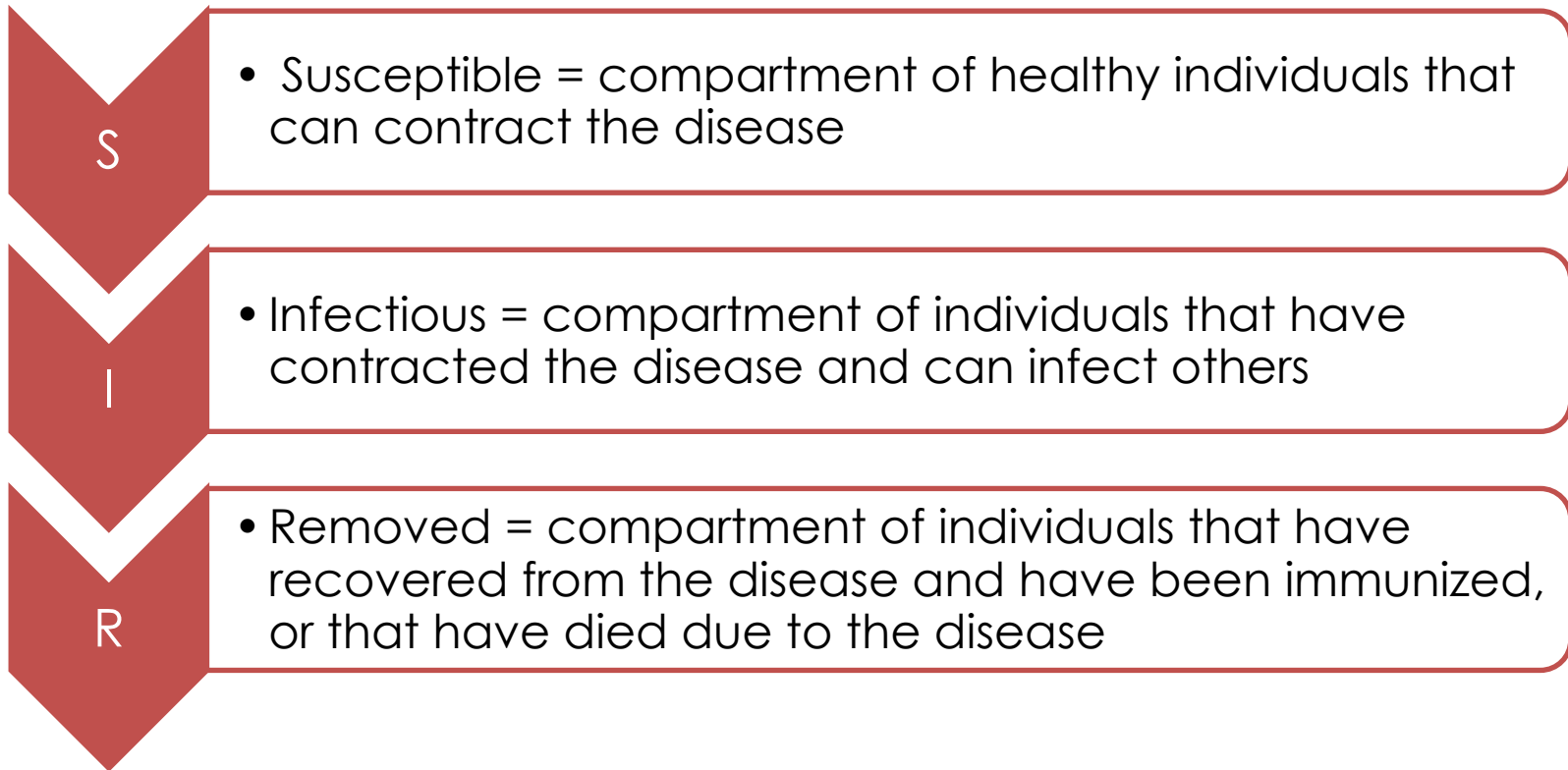
- The evolution of the system is described by **differential equations**
- **Deterministic** solutions
- **Macroscopic** model

## Agent-based simulations

- The dynamics of the system is generated making the individual agents evolve according to **behavioral rules**
- **Probabilistic** in nature
- **Microscopic** model

# The SIR model (Kermack & McKendrick, 1927)

The population is divided in three compartments





# The SIR model (Kermack & McKendrick, 1927)

The temporal evolution of the three compartments is described by a set of three ordinary non-linear differential equations

$$\frac{dS(t)}{dt} = -\frac{\beta I(t)S(t)}{N}$$

$\beta$  = rate of infection

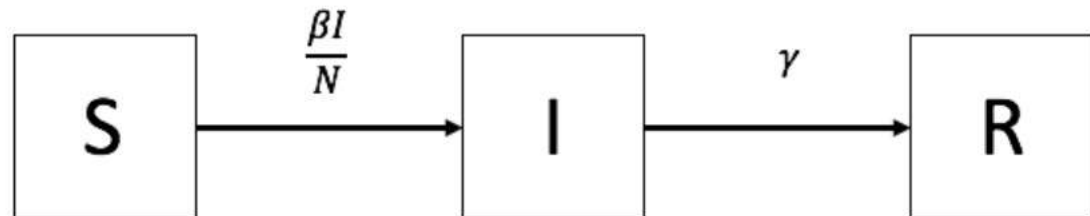
$$\frac{dI(t)}{dt} = \frac{\beta I(t)S(t)}{N} - \gamma I(t)$$

$\gamma$  = rate of recovery or death

$$\frac{dR(t)}{dt} = \gamma I(t)$$

$$S(t) + I(t) + R(t) = N = \text{const}$$

$$\frac{dS}{dt} + \frac{dI}{dt} + \frac{dR}{dt} = 0$$



# Equation-based simulation of the SIR model

- The model is already expressed as a system of differential equations
- It is usually simulated with an equation-based approach
- **Numerical integration** of the system of equations

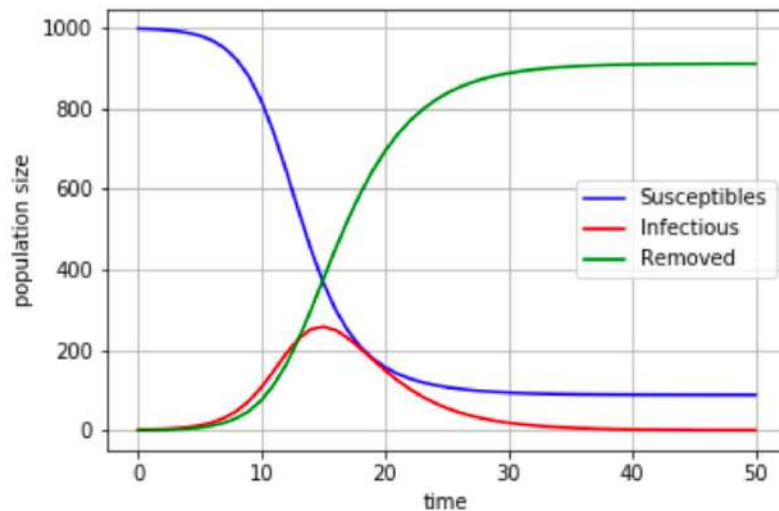


Figure 3. Results of the simulation of the computational equation-based model with odeint integration method: in blue the evolution of susceptible population size, in red the infectious, in green the removed ( $\beta = 0.8$ ,  $\gamma = 0.3$ ,  $N = 1000$ ,  $S(0) = 999$ ,  $I(0) = 1$ ,  $R(0) = 0$ ,  $time = [0, 1, \dots, 50]$ ).

```
def dX_dt(X, t=0):  
    return array([ - beta*X[0]*X[1]/N ,  
                  beta*X[0]*X[1]/N - gamma*X[1] ,  
                  gamma*X[1] ])
```

```
from scipy import integrate  
t = linspace(0, 50, 50)
```

```
X0 = array([N-1, 1, 0])
```

```
X = integrate.odeint(dX_dt, X0, t)
```



# Reformulation of the equations

$$\frac{dS(t)}{dt} = - \frac{\beta I(t) S(t)}{N}$$

Probability that a susceptible becomes infectious

$\beta_c$  = contact rate

Mean number of contacts for each individual

Probability that the contact is infectious

Probability that a susceptible meets an infectious

=

x

Probability that the infectious infects the susceptible

$\beta_i$  = infectivity

$$\frac{I(t)}{N}$$

# Reformulation of the equations

$$\frac{dS(t)}{dt} = - \frac{\beta I(t) S(t)}{N}$$

Probability that a susceptible becomes infectious

=

Mean number of contacts for each individual

x

Probability that the contact is infectious

Probability that a susceptible meets an infectious

x

Probability that the infectious infects the susceptible

$\beta_i$  = infectivity

$\beta_c$  = contact rate

$\frac{I(t)}{N}$

The probability that a susceptible becomes infectious now depends on the probability that a susceptible meets an infectious and on the probability that a susceptible actually becomes infectious



# Agent-based simulation of the SIR model

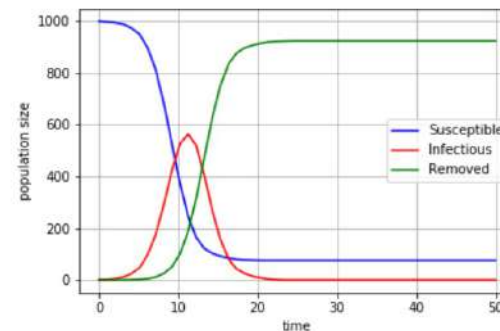
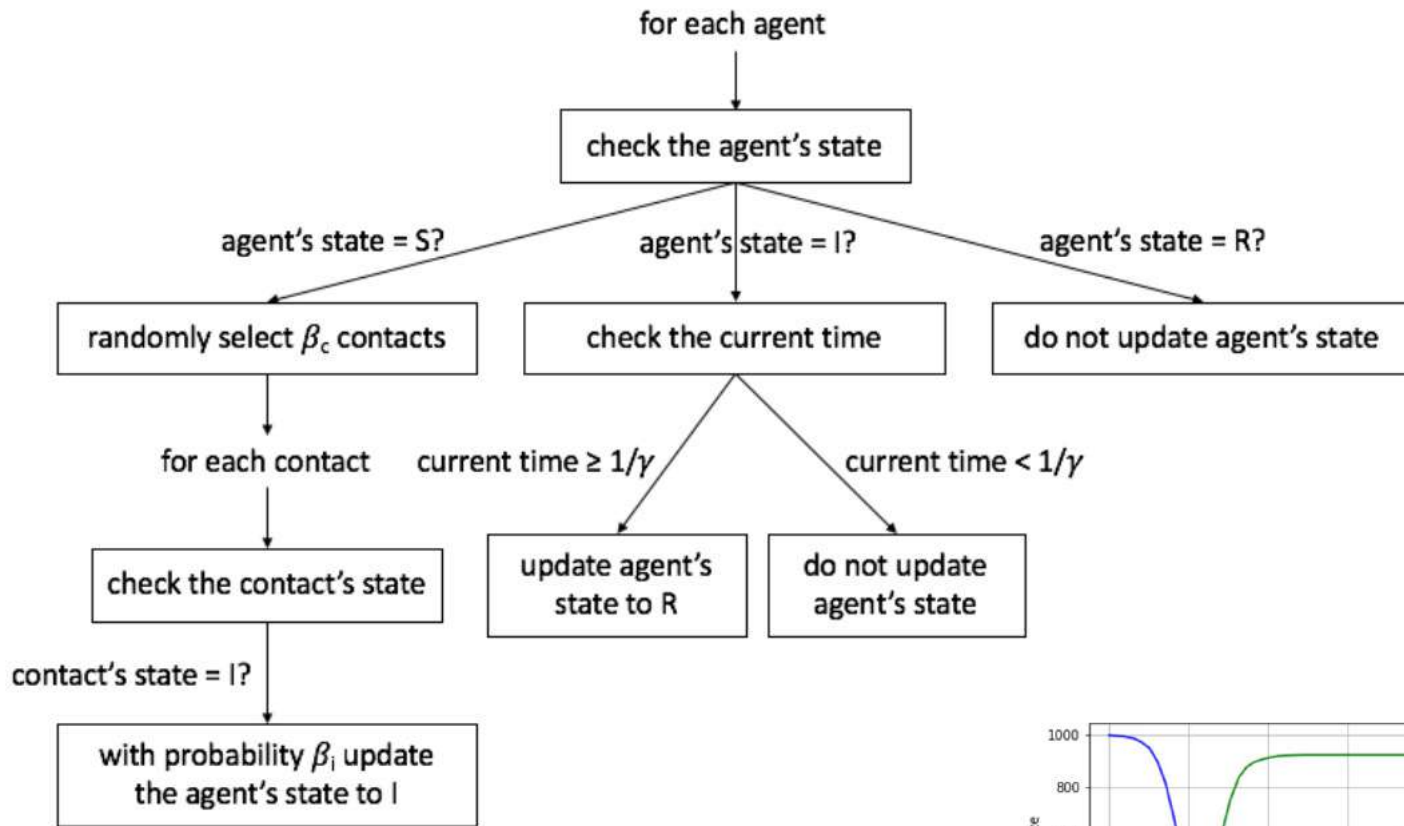
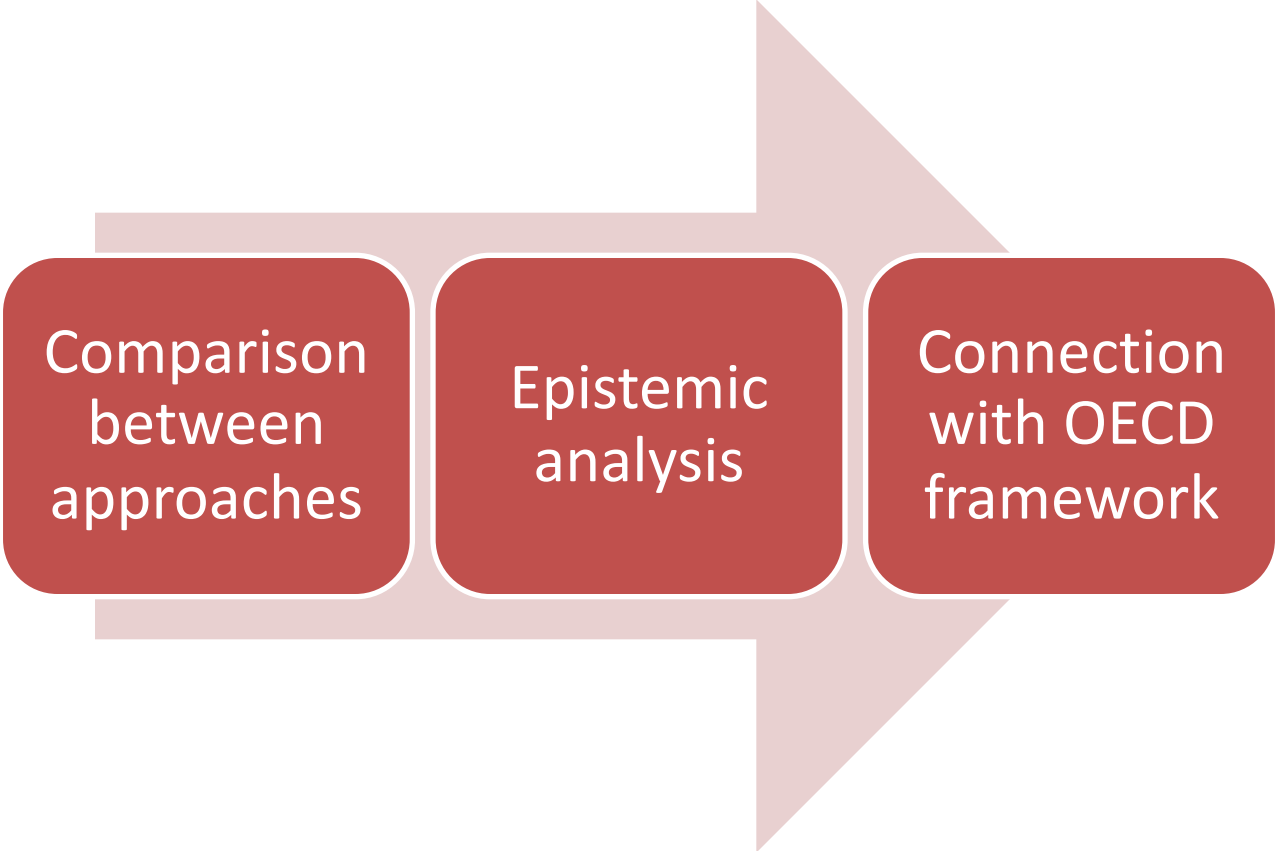


Figure 9. Average results of 100 simulation runs of the computational agent-based model: in blue the evolution of susceptible population size, in red the infectious, in green the removed ( $\beta_i = 0.8$ ,  $\beta_c = 1$ ,  $\gamma = 0.3$ ,  $N = 1000$ ,  $S(0) = 999$ ,  $I(0) = 1$ ,  $R(0) = 0$ ,  $time = [0, 1, \dots, 50]$ ).

(Adapted from Macal, 2011)



Comparison  
between  
approaches

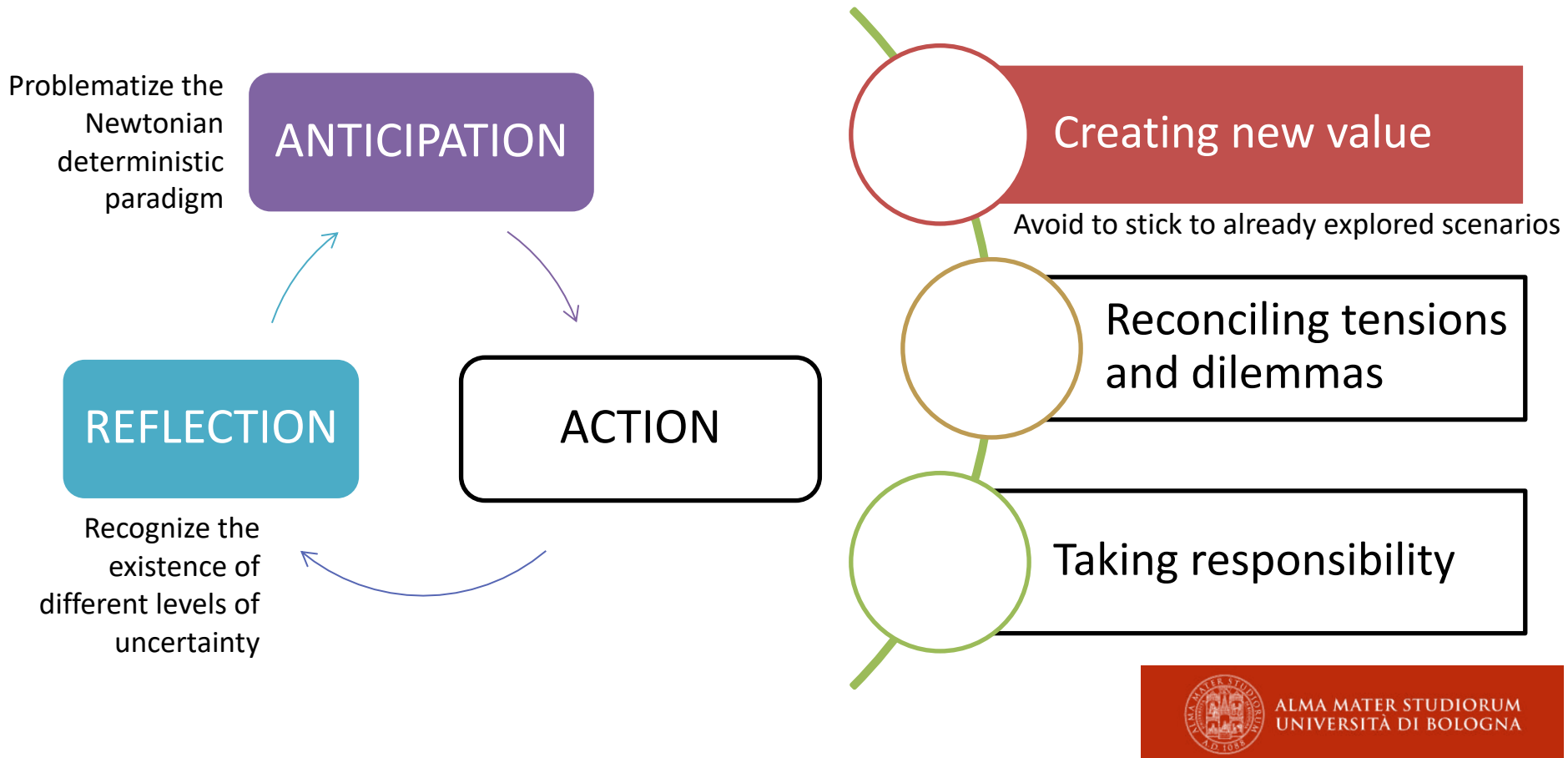
Epistemic  
analysis

Connection  
with OECD  
framework



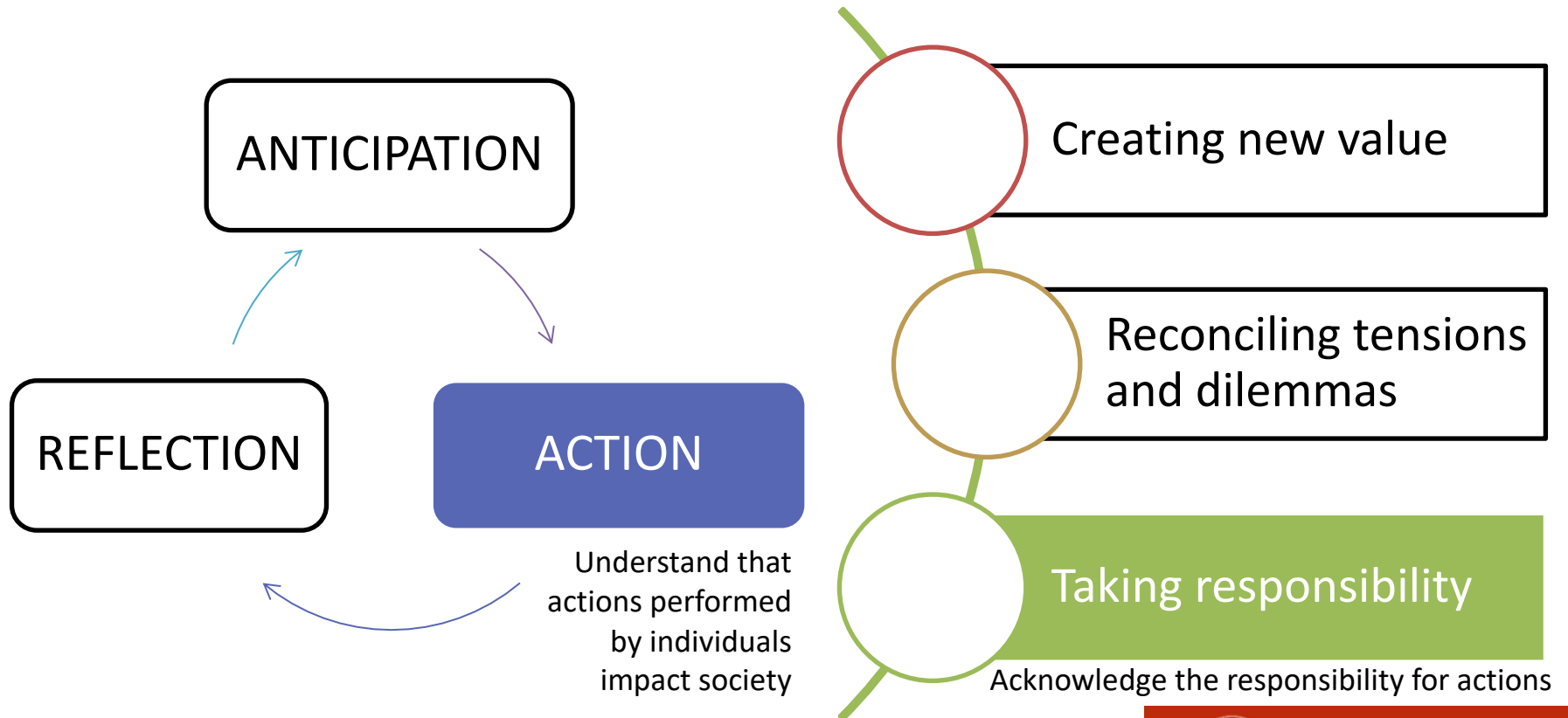
# Epistemic analysis of the approaches to exploit AAR cycle and transformative competence

The nature of agent-based simulations is intrinsically probabilistic



# Epistemic analysis of the approaches to exploit AAR cycle and transformative competence

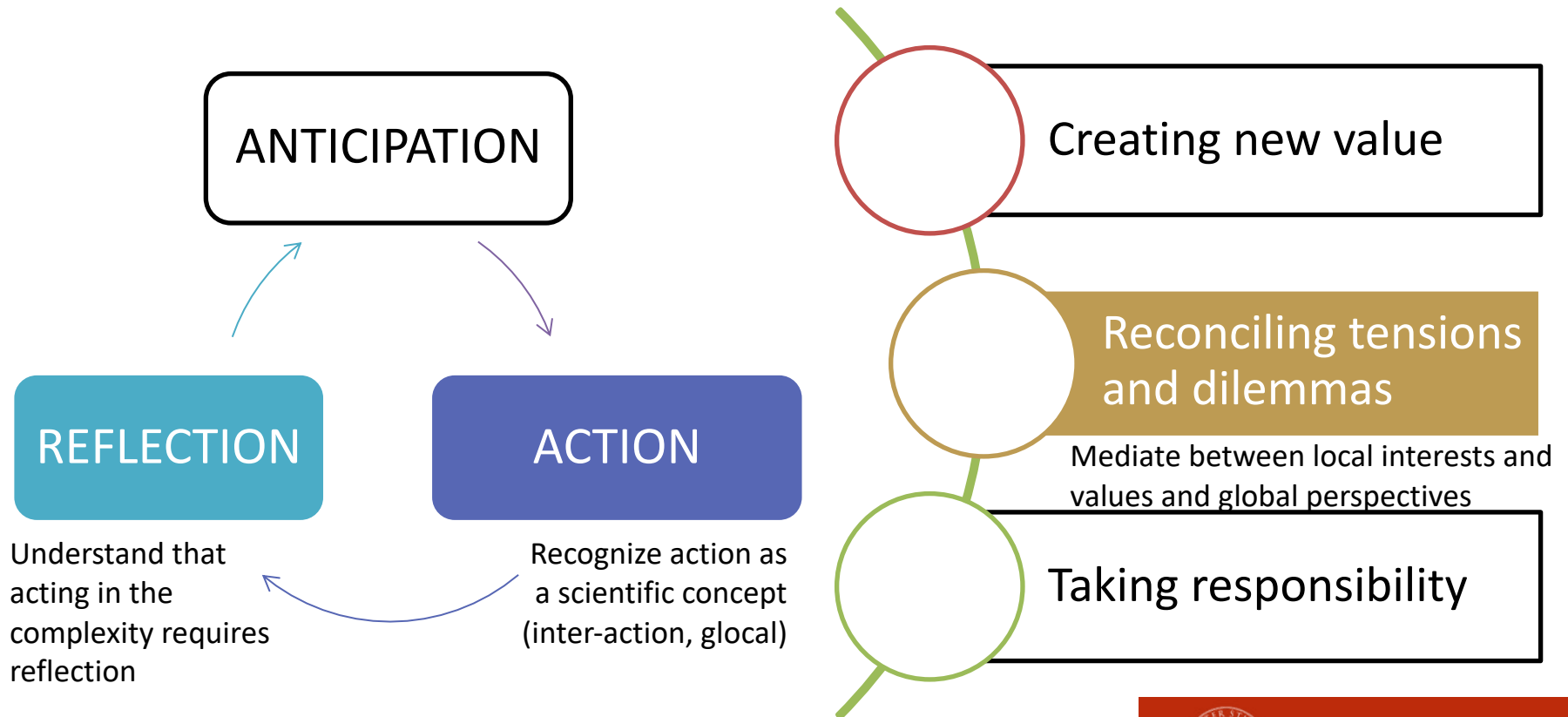
In the agent-based approach, the population consists of individuals, the agents





# Epistemic analysis of the approaches to exploit AAR cycle and transformative competence

The global behaviour is an emergent property that depends non-linearly from local agents' interactions



# Conclusions

- We have addressed a very concrete teaching issue, related to the problem of exploiting scientific simulations in teaching and equipping students with knowledge and skills needed to navigate a complex society
- In particular, we examined the case study of the SIR epidemiological model comparing equation- and agent-based approaches to simulation
- Framing the analysis within the OECD recommendations allowed us to:
  - contribute to find concrete ways to implement visionary educational frameworks and to bridge the gap between policy recommendations and classroom practices
  - point out the epistemological, societal value of a scientific tool like simulations and flesh out, in education, their transformative potential



## Next steps

- Characterize disciplinary and interdisciplinary the equation- and agent-based approaches to SIR model

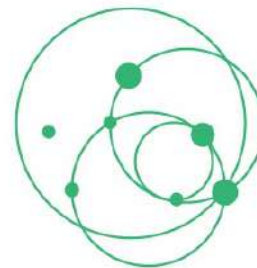
<https://identitiesproject.eu/>

IDENTITIES

Enlightening  
Interdisciplinarity  
in STEM for Teaching

- Study the role of future-oriented science education and future-scaffolding skills to facilitate the transition from the conceptual and epistemic analysis to AAR and transformative competences

<https://www.fedora-project.eu/>



FEDORA



Thank You  
For Your  
Attention