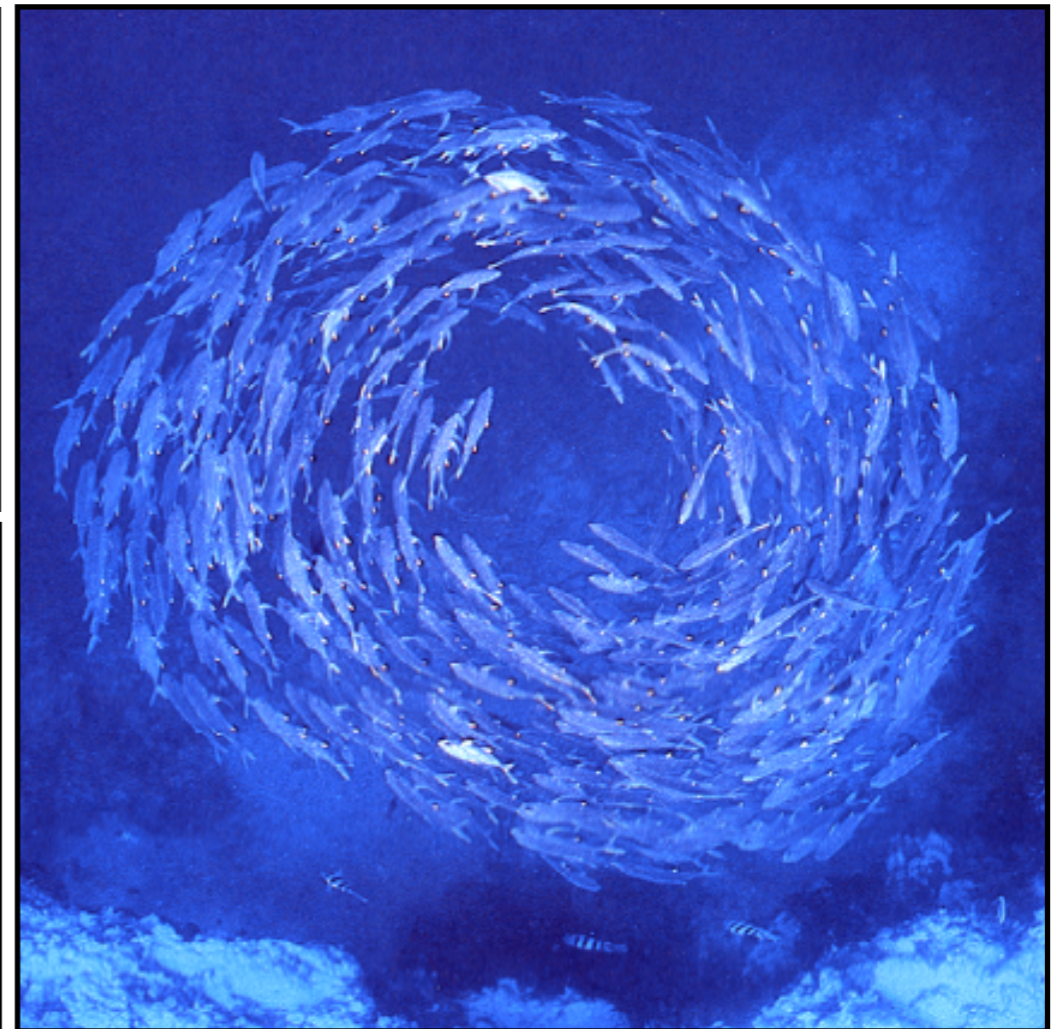
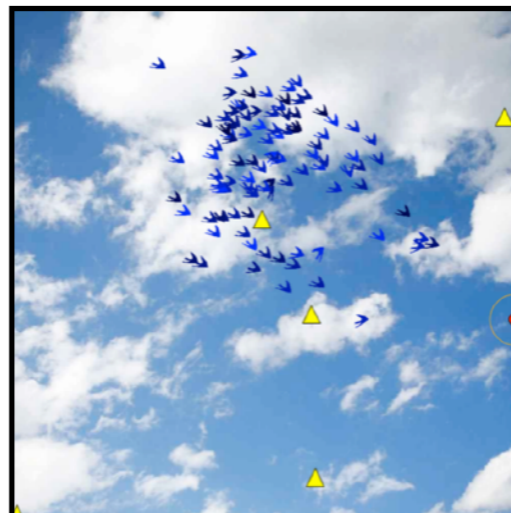
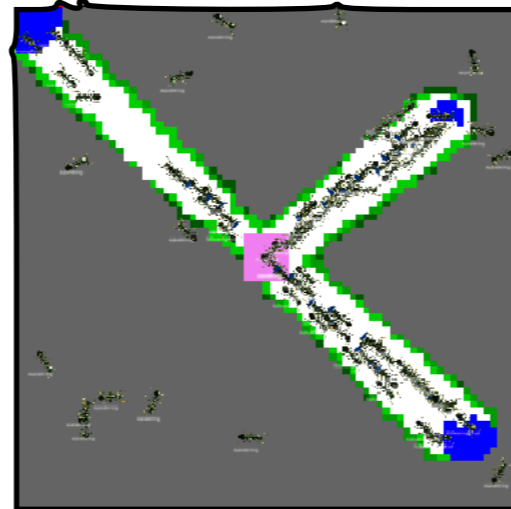
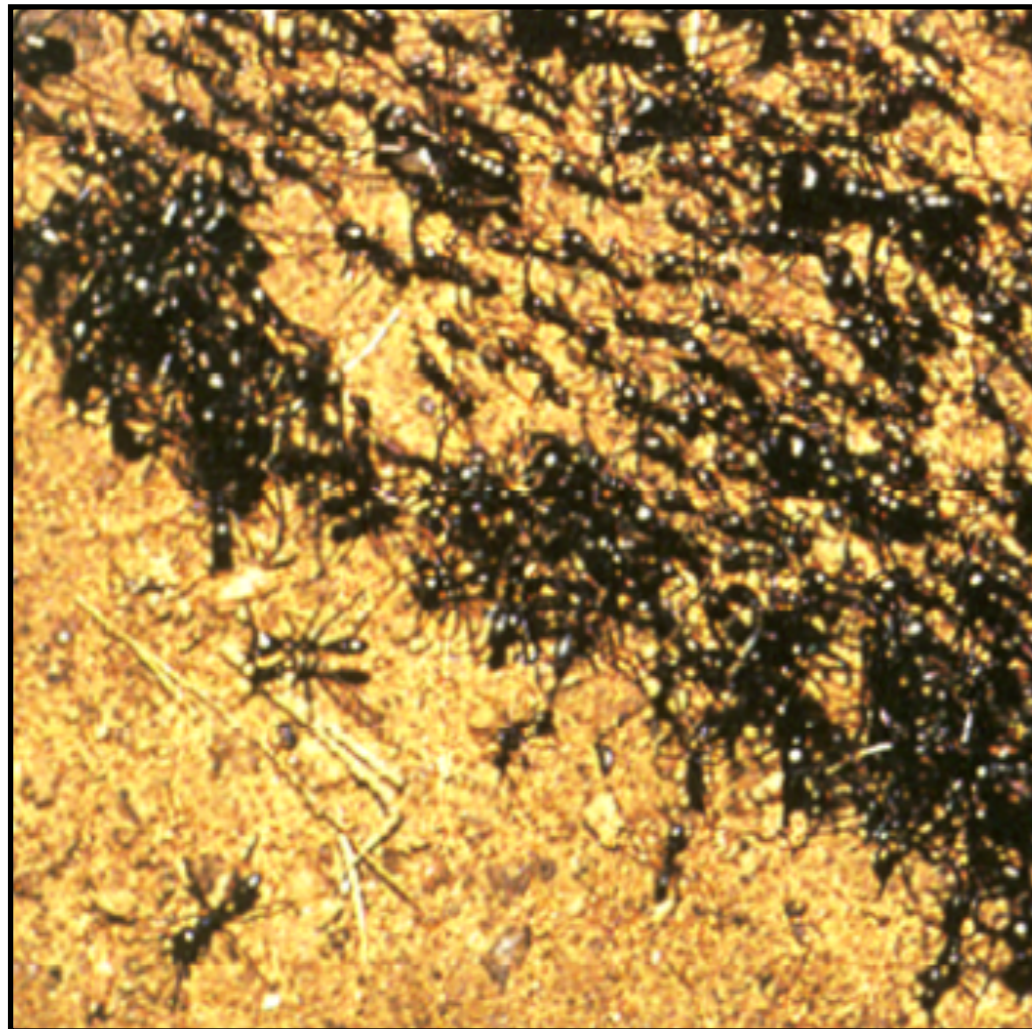


Agent, Multi-Agent Systems and Agent-based Modelling and Simulation

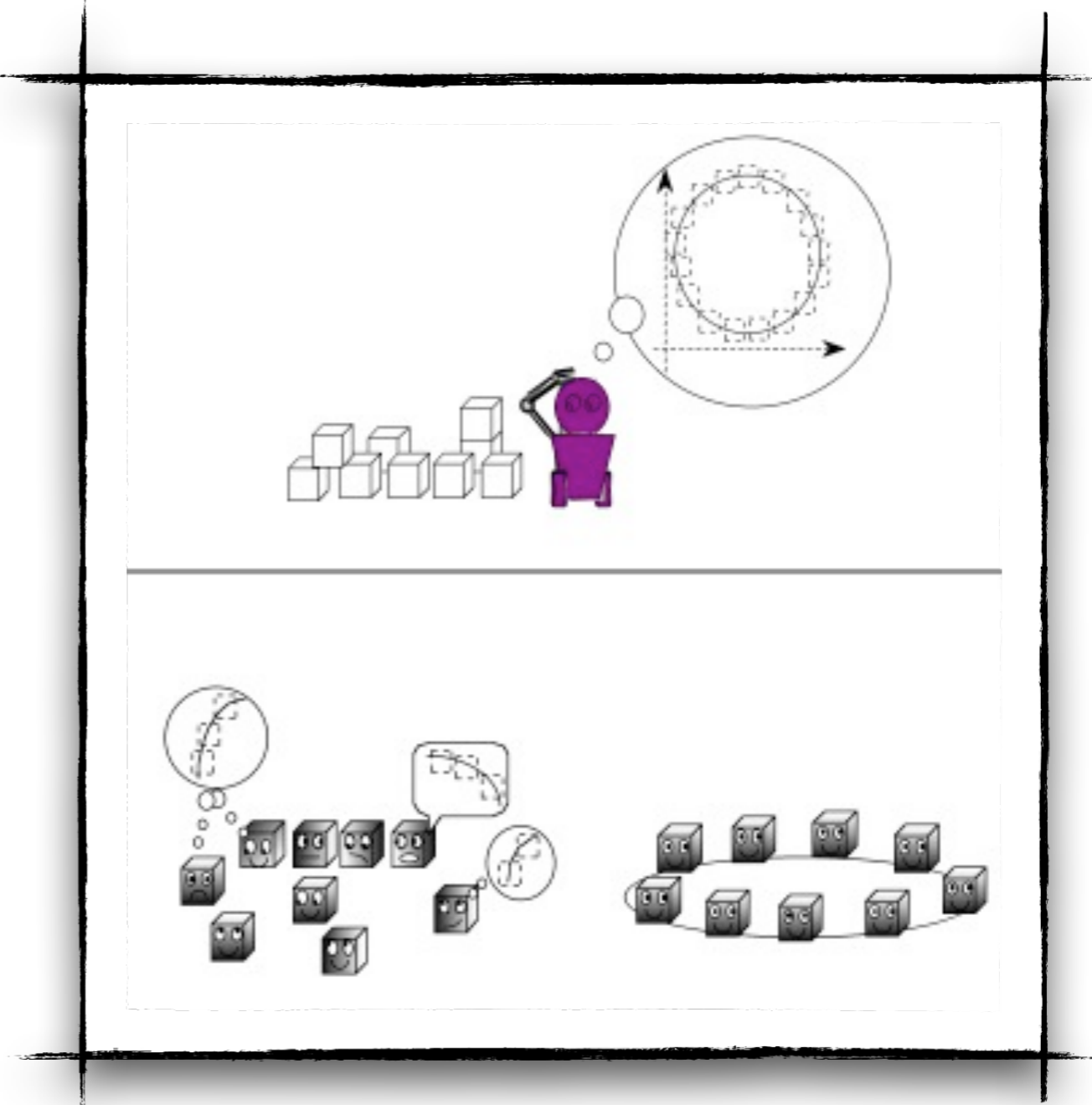


Benoit Gaudou, UMMISCO, IRIT (Univ. Toulouse 1)

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benoit.gaudou@gmail.com

Introduction

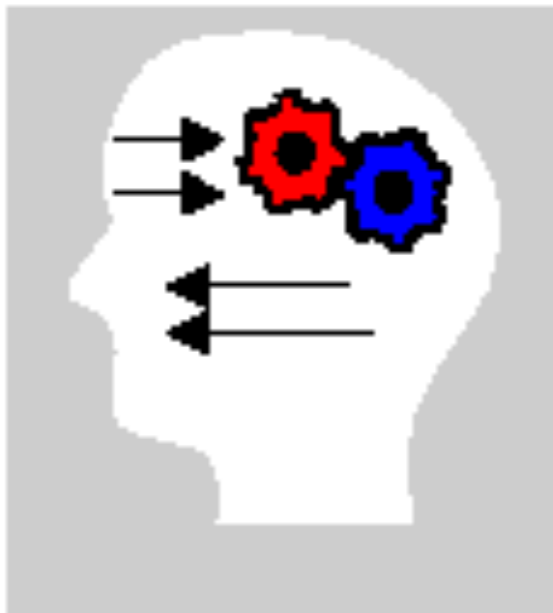
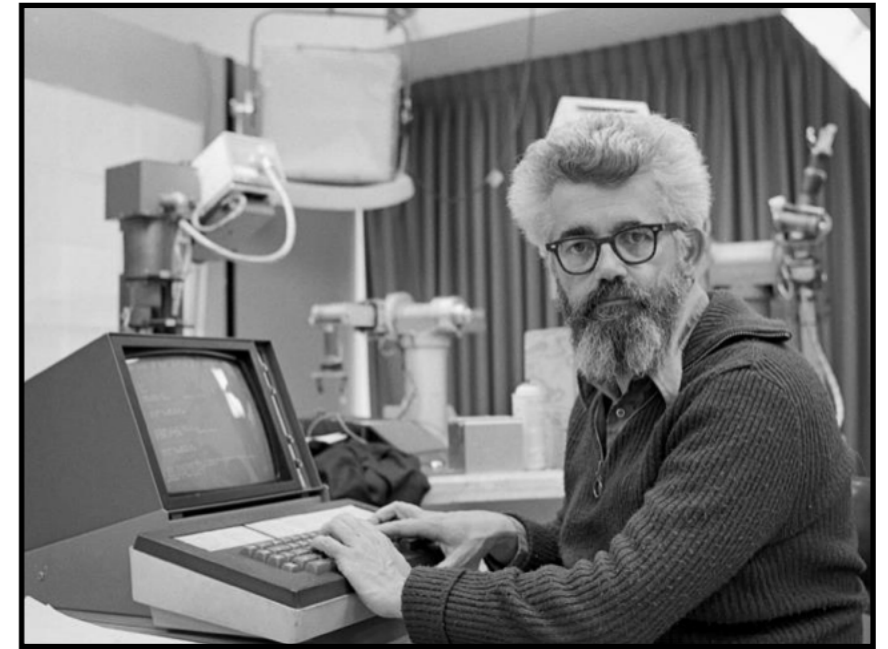


In 1956, John MacCarthy organized the first workshop about Artificial Intelligence.

► In AI, the basic model is the individual intelligence.

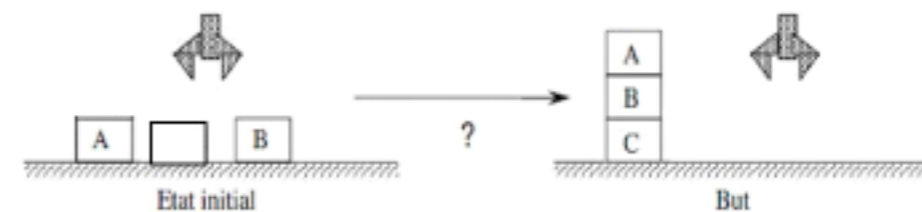
E.g.:

- Planner (STRIPS 1971)
- Expert systems.



The expert

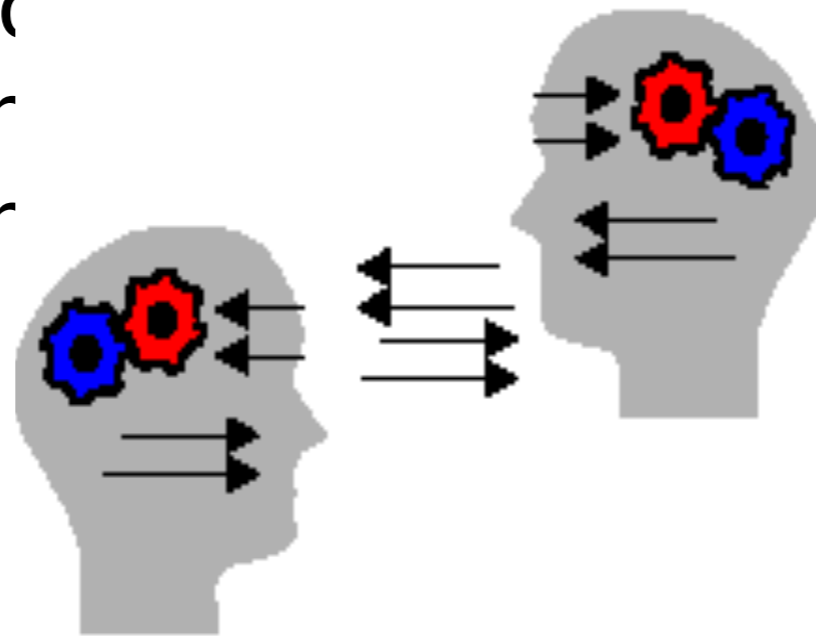
STRIPS planner



```
(:action stack
:parameters (?x – block ?y - block)
:precondition (and (holding ?x) (clear ?y))
:effects (and (not (holding ?x)) (not (clear ?y))
            (on ?x ?y) (clear ?x) (handempty))
```

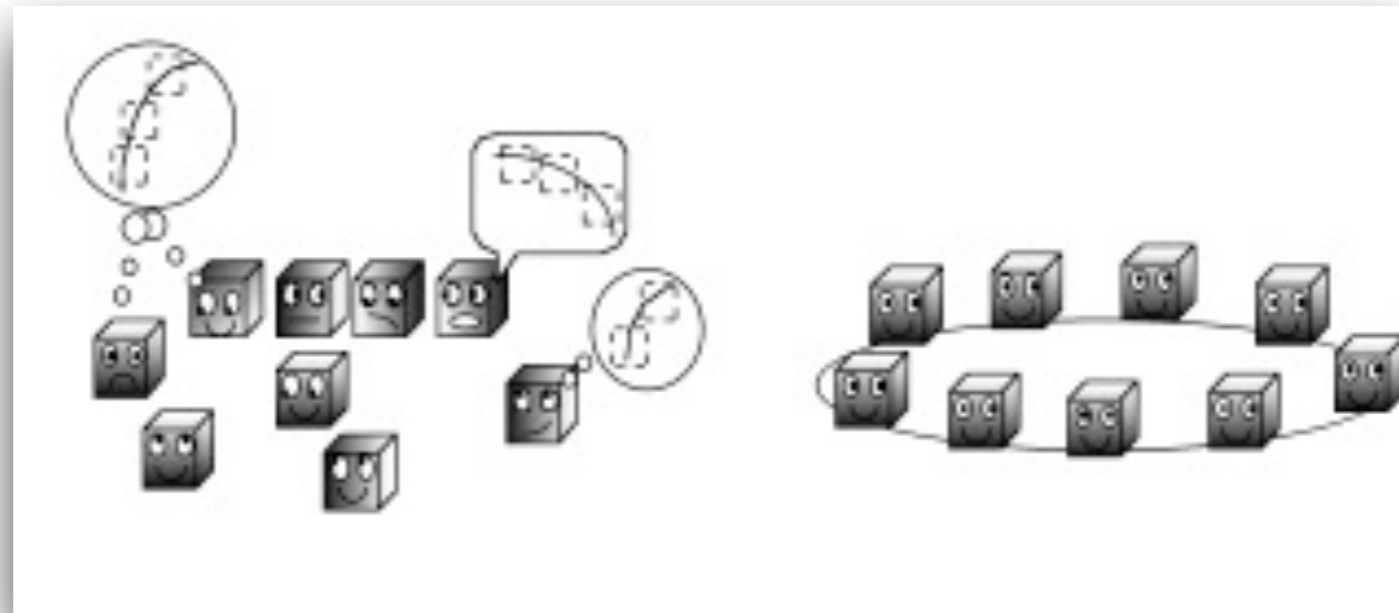
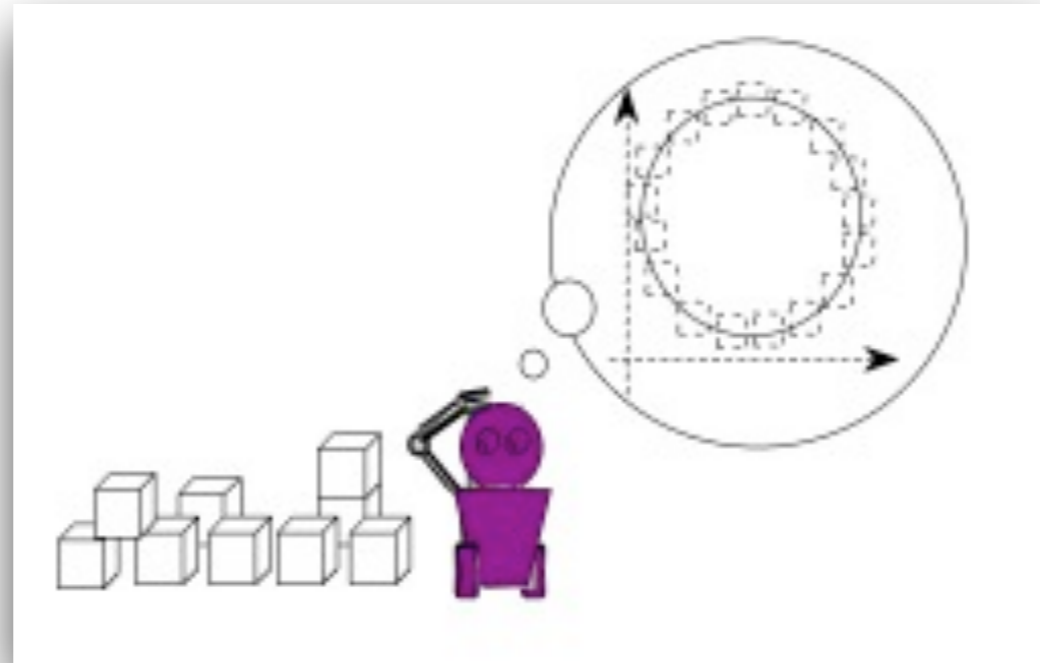
In 80's, Distributed AI (1980) (among others) appear in opposition to AI.

- ▶ Instead of considering only cognitive faculties of an individual, **DAI** postulates that intelligence "**emerges**" from the interaction between agents and between agents and environment.



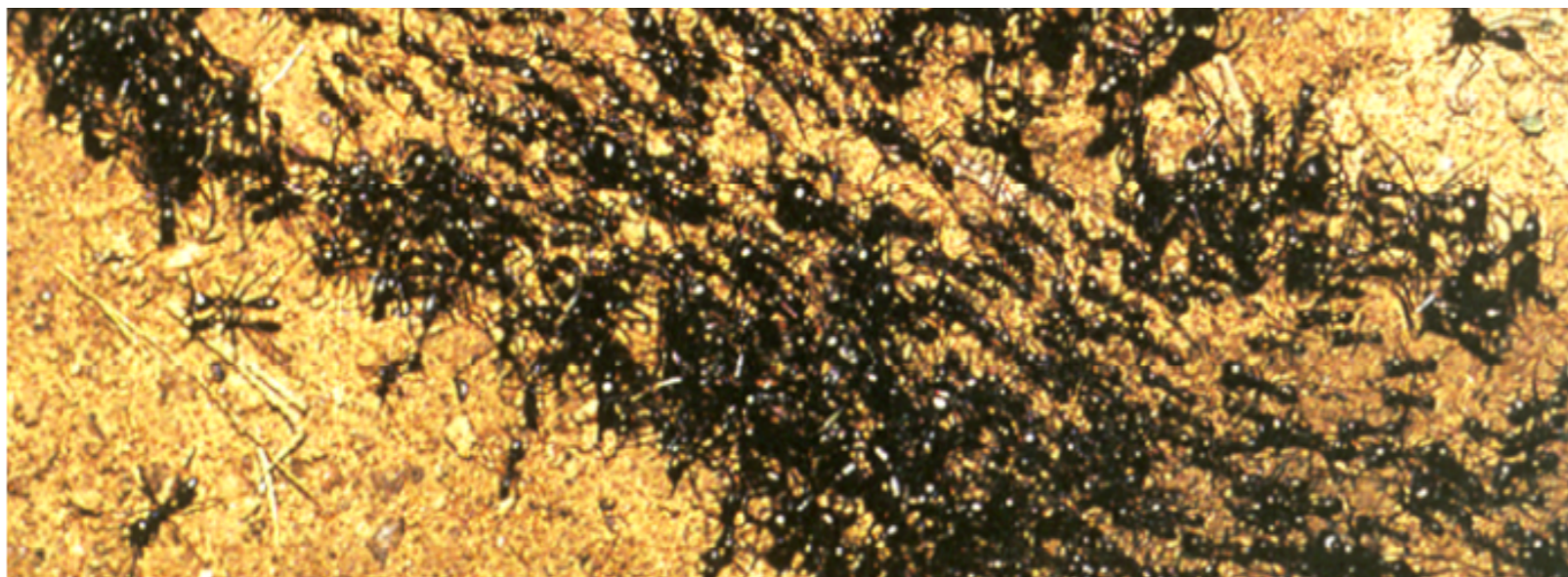
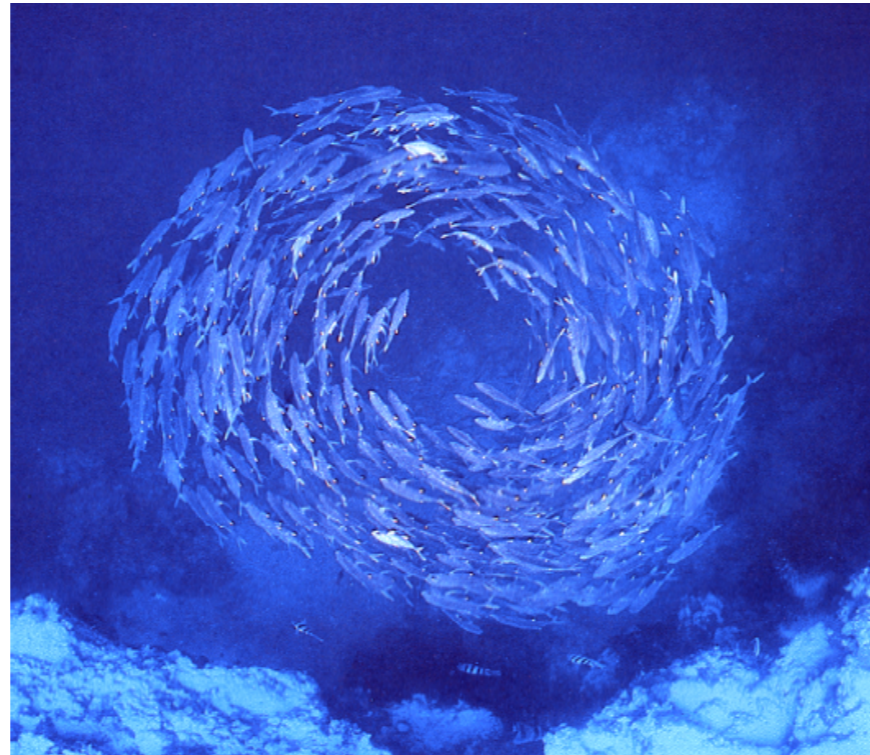
A community

AI vs DAI on a problem: How to draw a circle with boxes ?



Multi-Agent Systems are interested in complex systems where artificial and/or natural entities interact to produce collective behaviors.

- ▶ MAS is interested in complex systems where artificial and/or natural entities interact to produce collective behaviours.



What is an agent?

- ▶ An agent is a computer system that is situated in some **environment**, and that is capable of **autonomous action** in this environment in order to meet its **design objectives**. (Wooldridge, 2008)

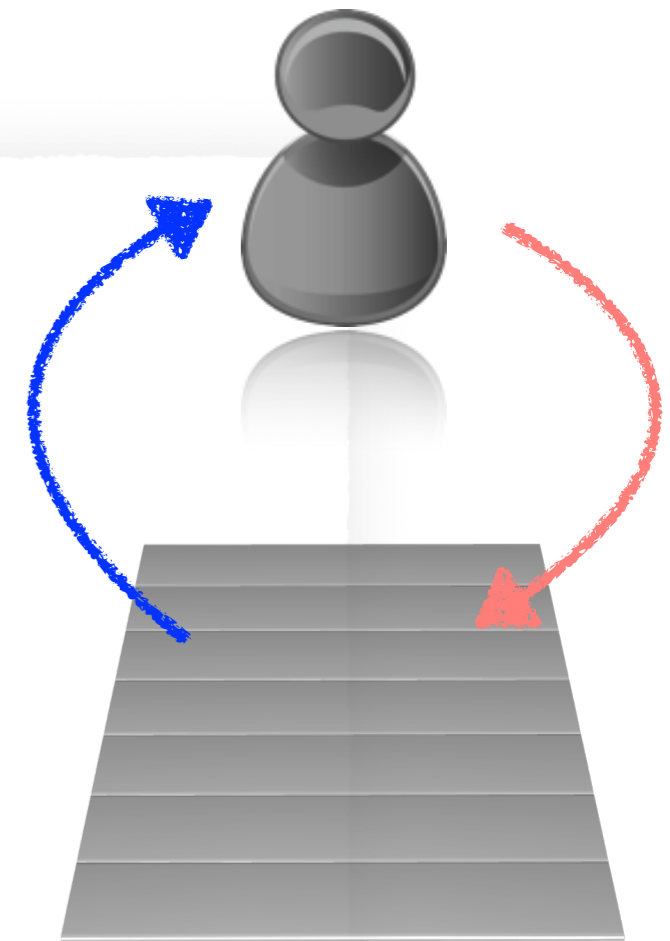
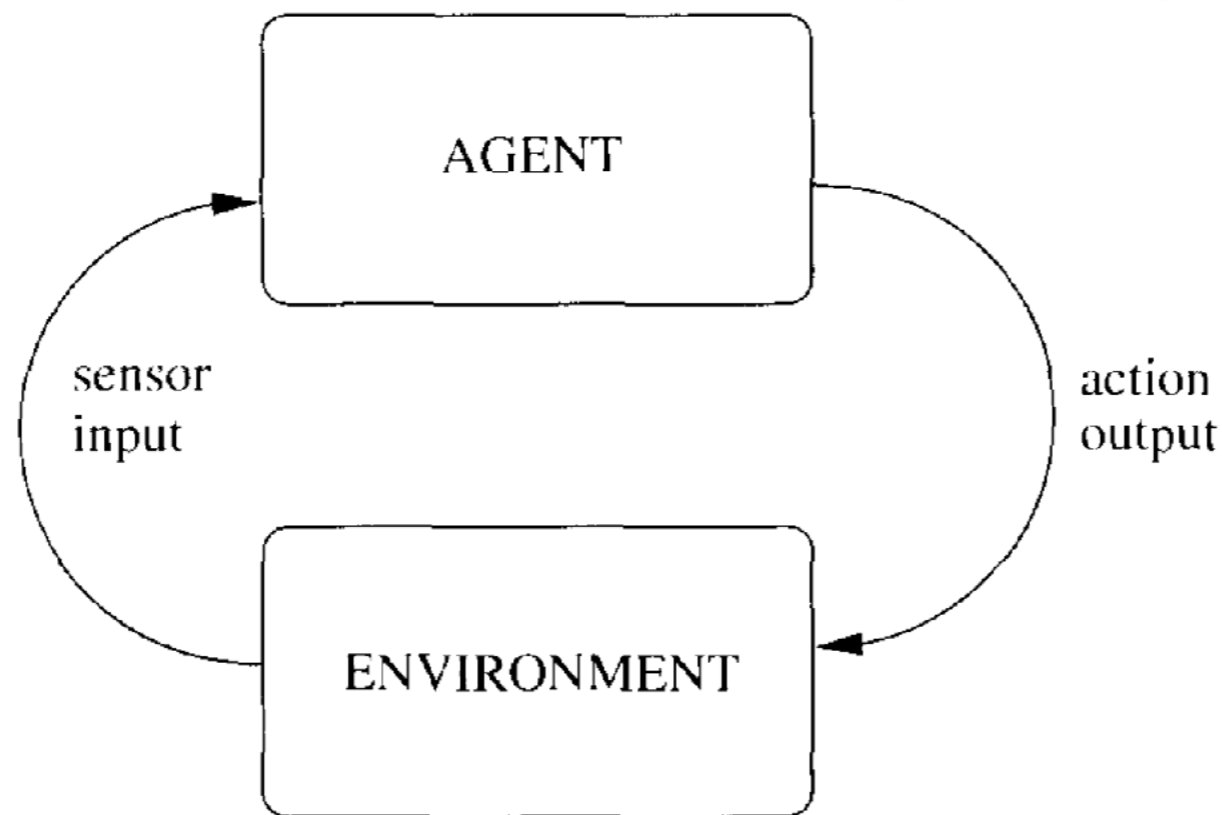
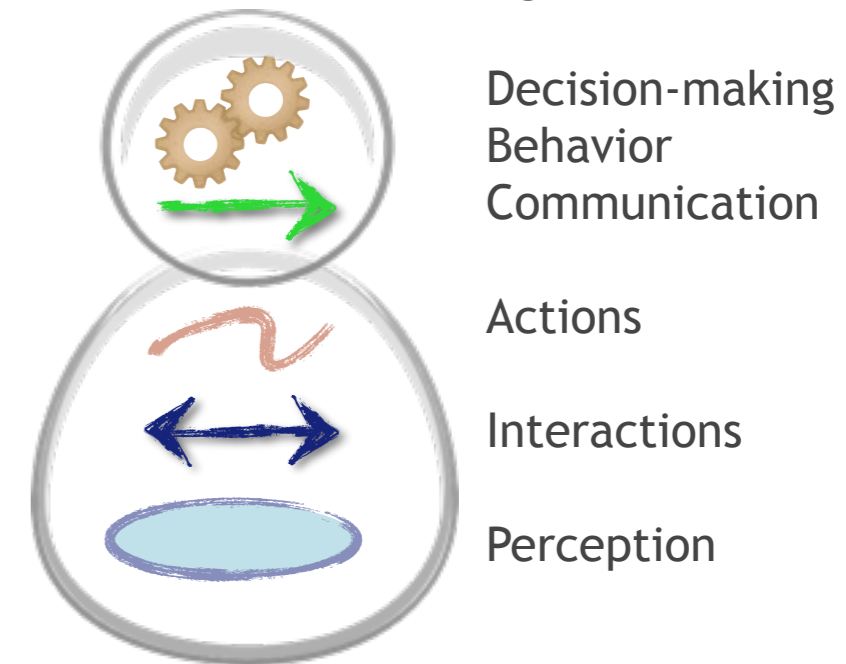


Figure 2.1 An agent in its environment. The agent takes sensory input from the environment, and produces as output actions that affect it. The interaction is usually an ongoing, non-terminating one.

What is an agent?

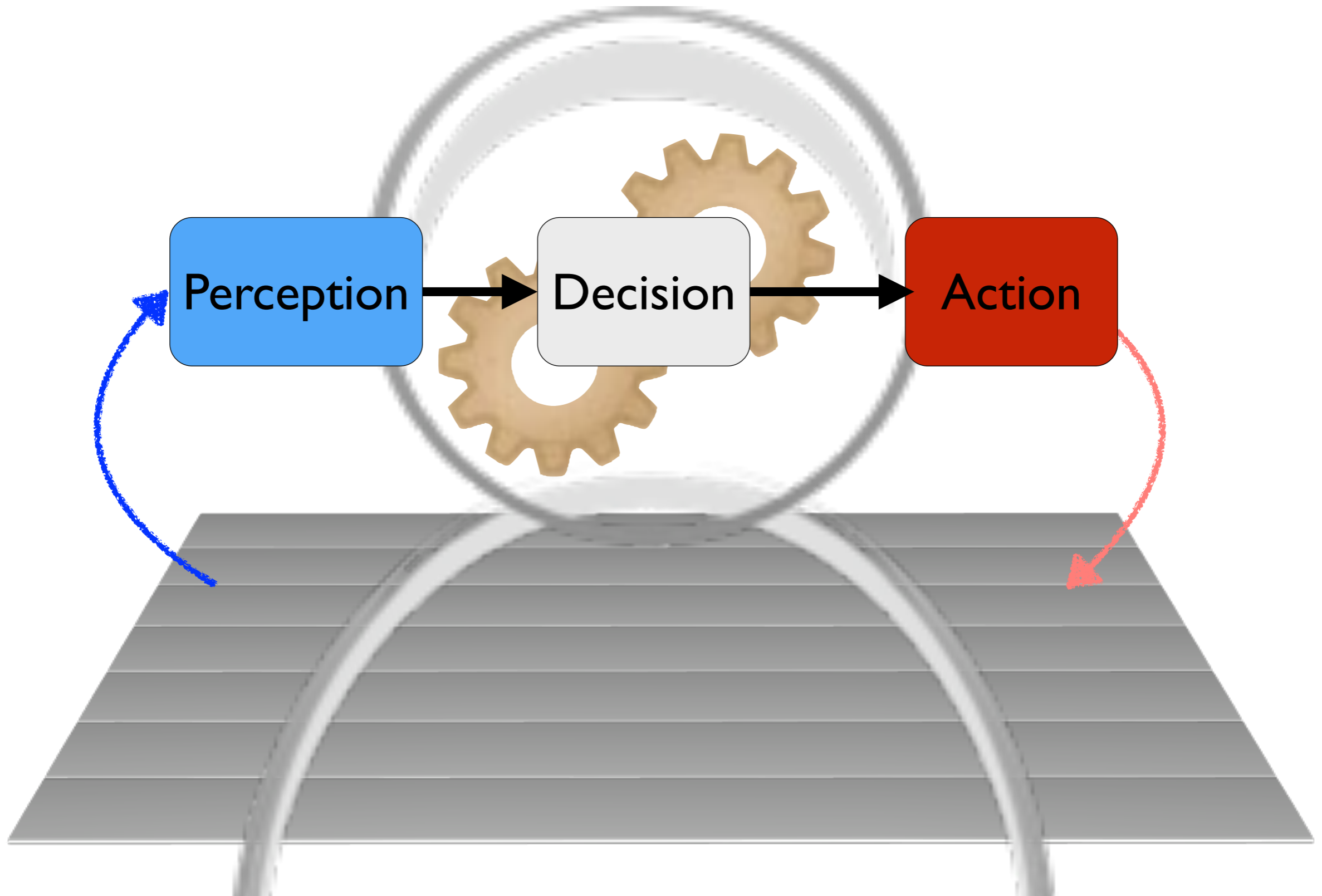
(Ferber, 95) names **agent** a **physical or virtual entity** which:

- a. can **act** in its **environment**,
- b. can **communicate** with other agents
- c. has its own set of **individual objectives**
- d. has its own **ressources**
- e. can **perceive** (partially) its environment
- f. has a **partial representation** of its environment (or even no representation)
- g. has **skills** and can **offer services**
- h. can (eventually) reproduce
- i. has a behavior tending to **satisfy its own objectives**, **taking into account** its **ressources** and **skills**, its **perception**, **representations** and **communication** it gets.



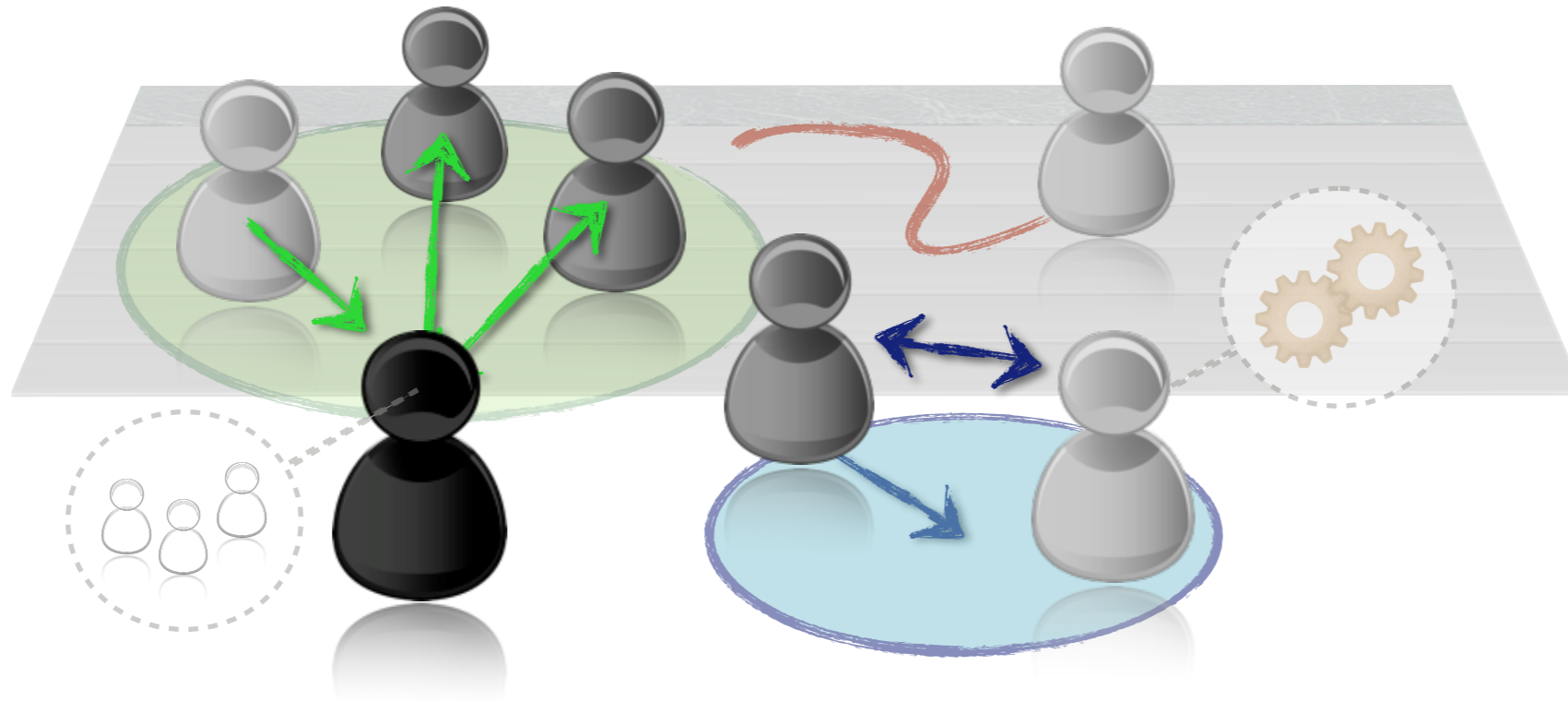
The basic agent life-cycle has 3 steps.

Agents can be reactive or cognitive depending on the decision complexity.



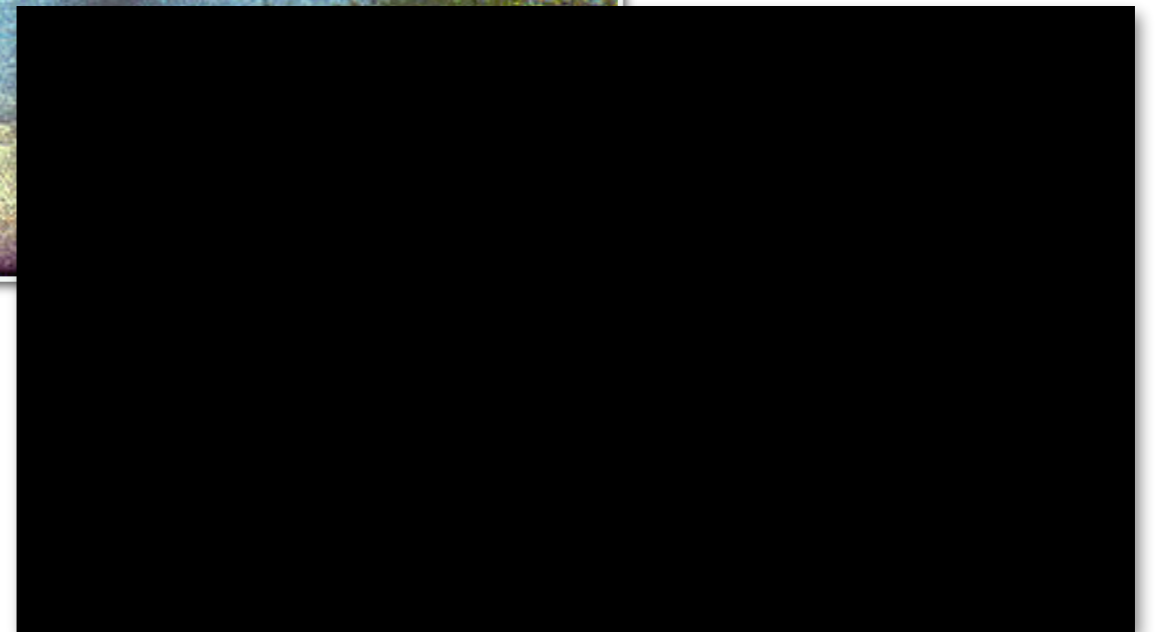
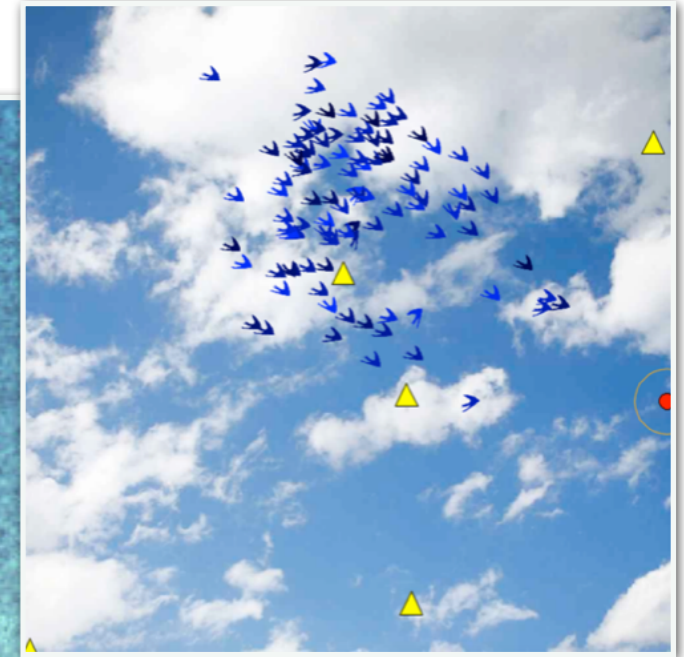
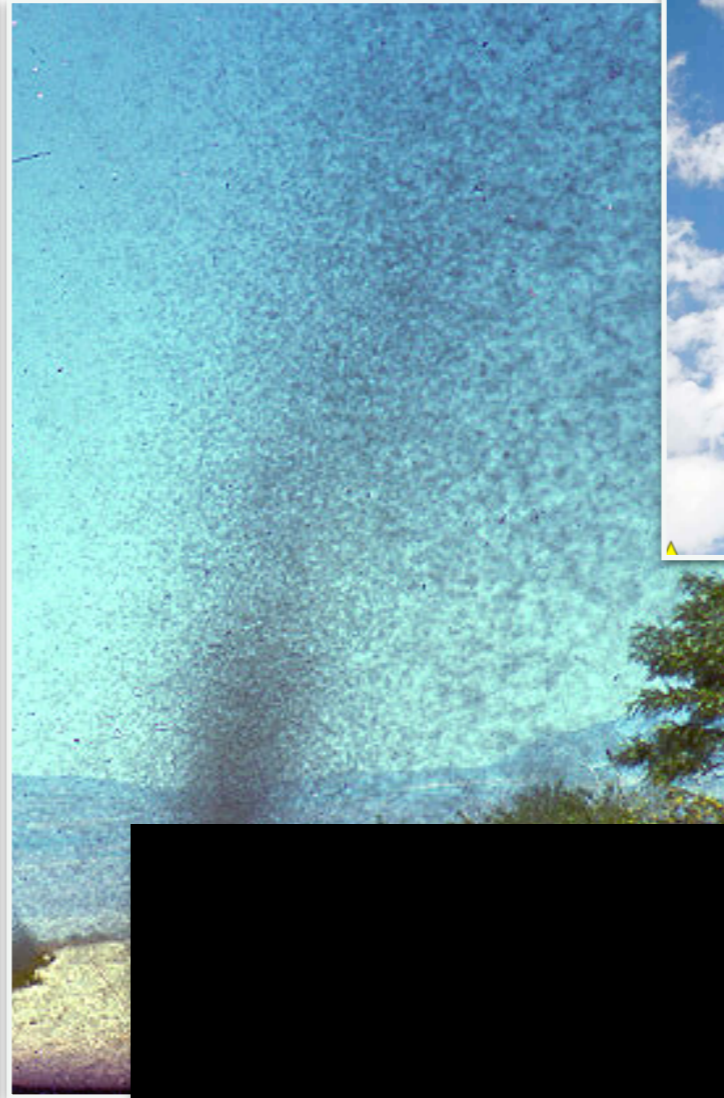
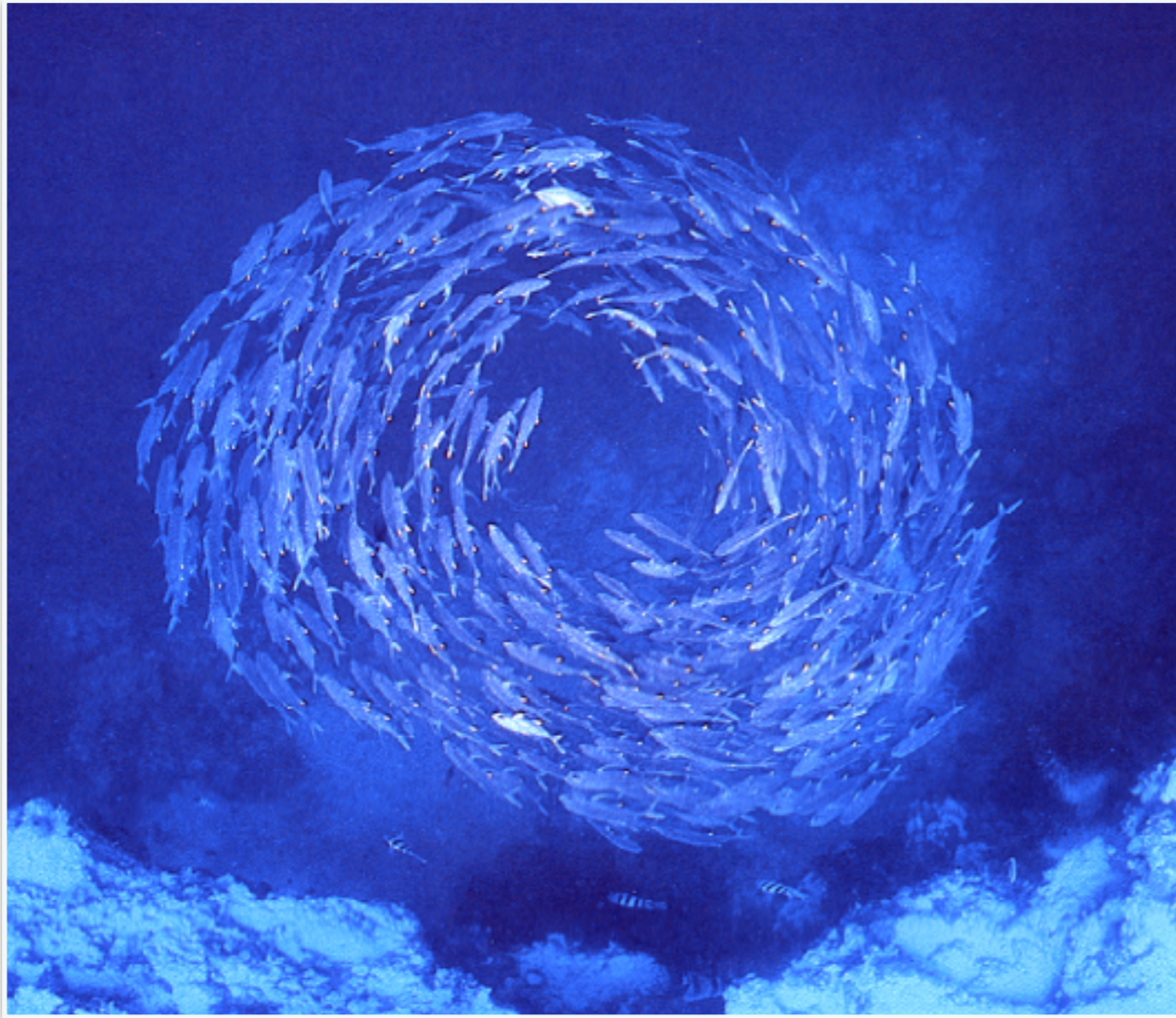
What is a Multi-Agent System ?

- ▶ A Multi-Agent System is a **set of agents acting and communicating in an environment.**



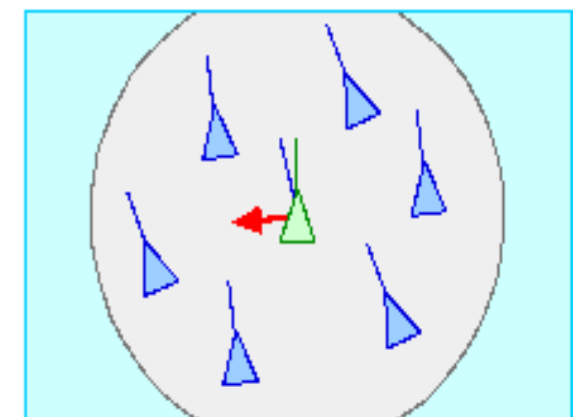
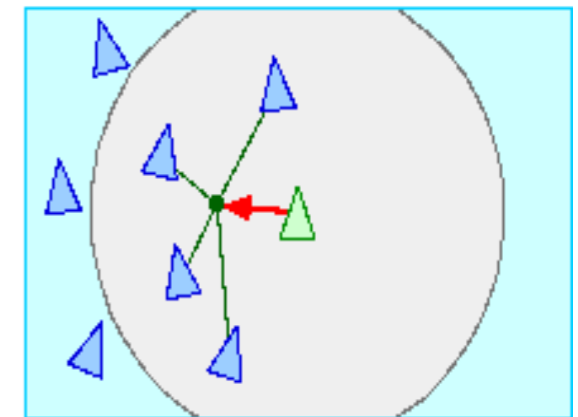
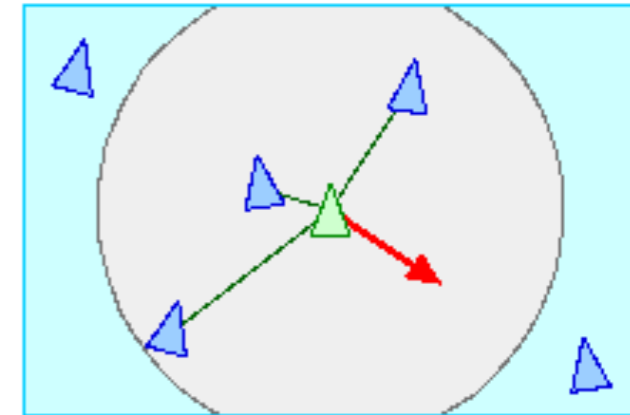
- ▶ In addition some authors add:
Passive objects and an organization or coordination mode.

Example of reactive agents: Reynolds' Boids (Craig Reynolds (1987))

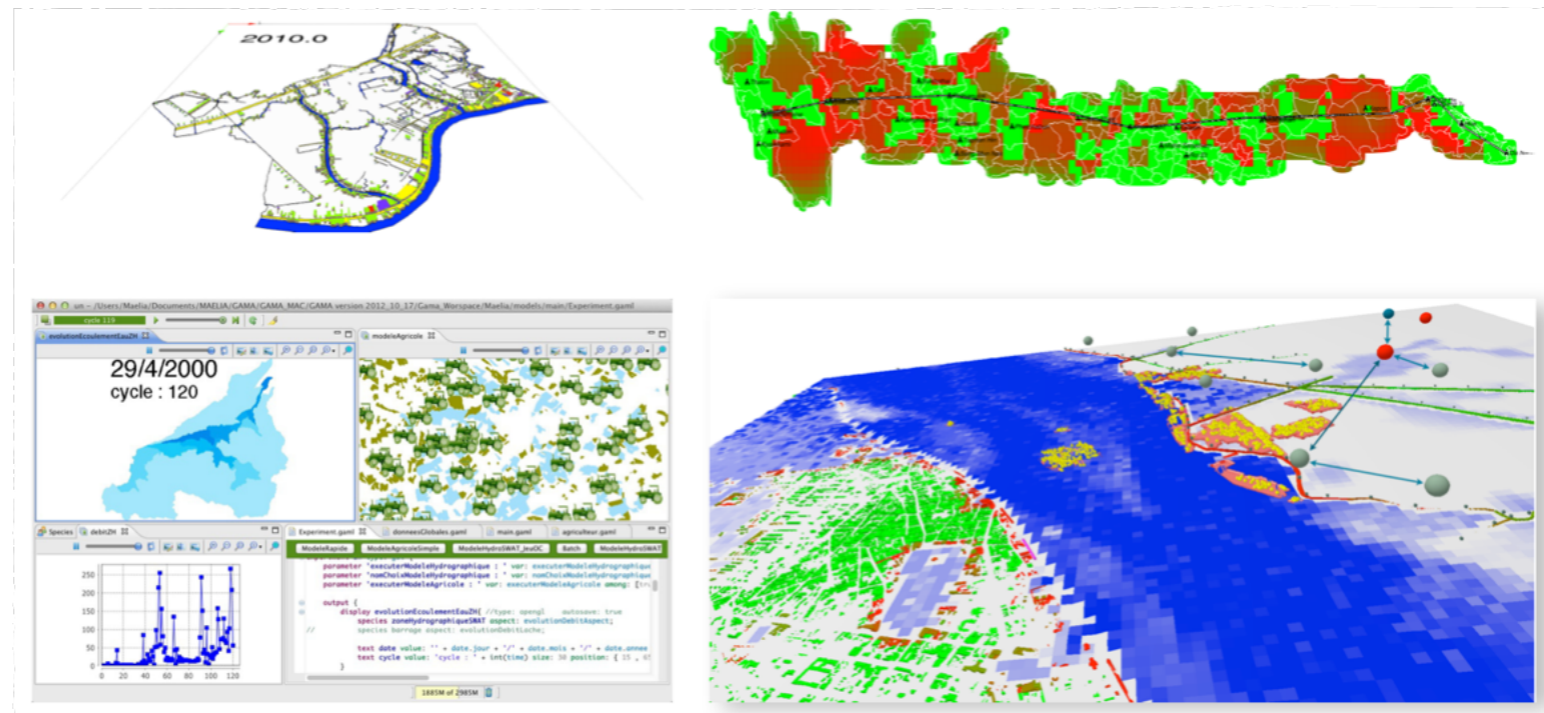


Boids: Interactions by field of forces

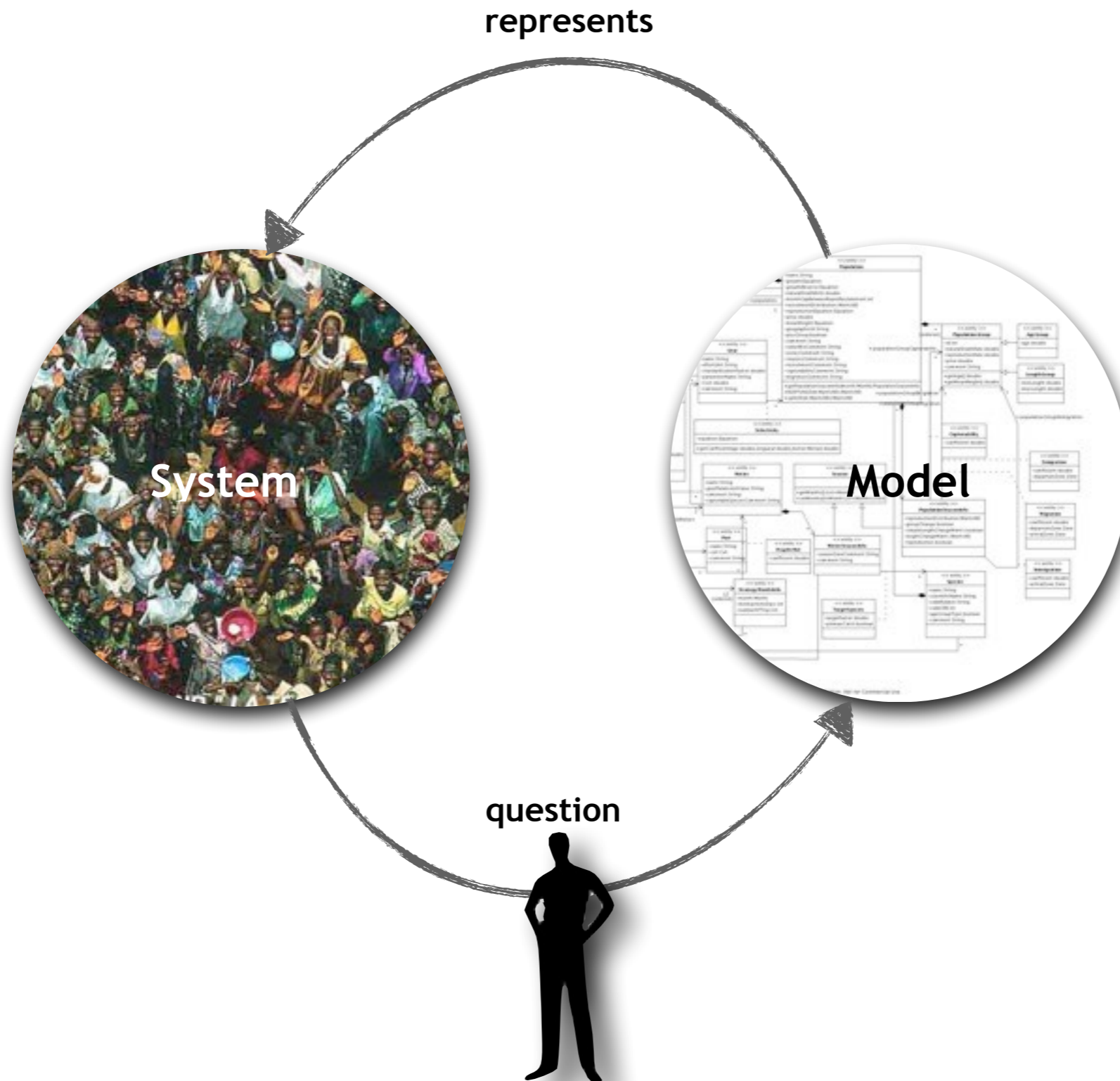
- ▶ **Separation:** the force is calculated so as to push back the Boid from its close neighbours.
- ▶ The purpose of **cohesion** is to bring the Boid closer to the gravity center of the group.
- ▶ The force of **alignment** is calculated as being an average between speeds of the neighbours.



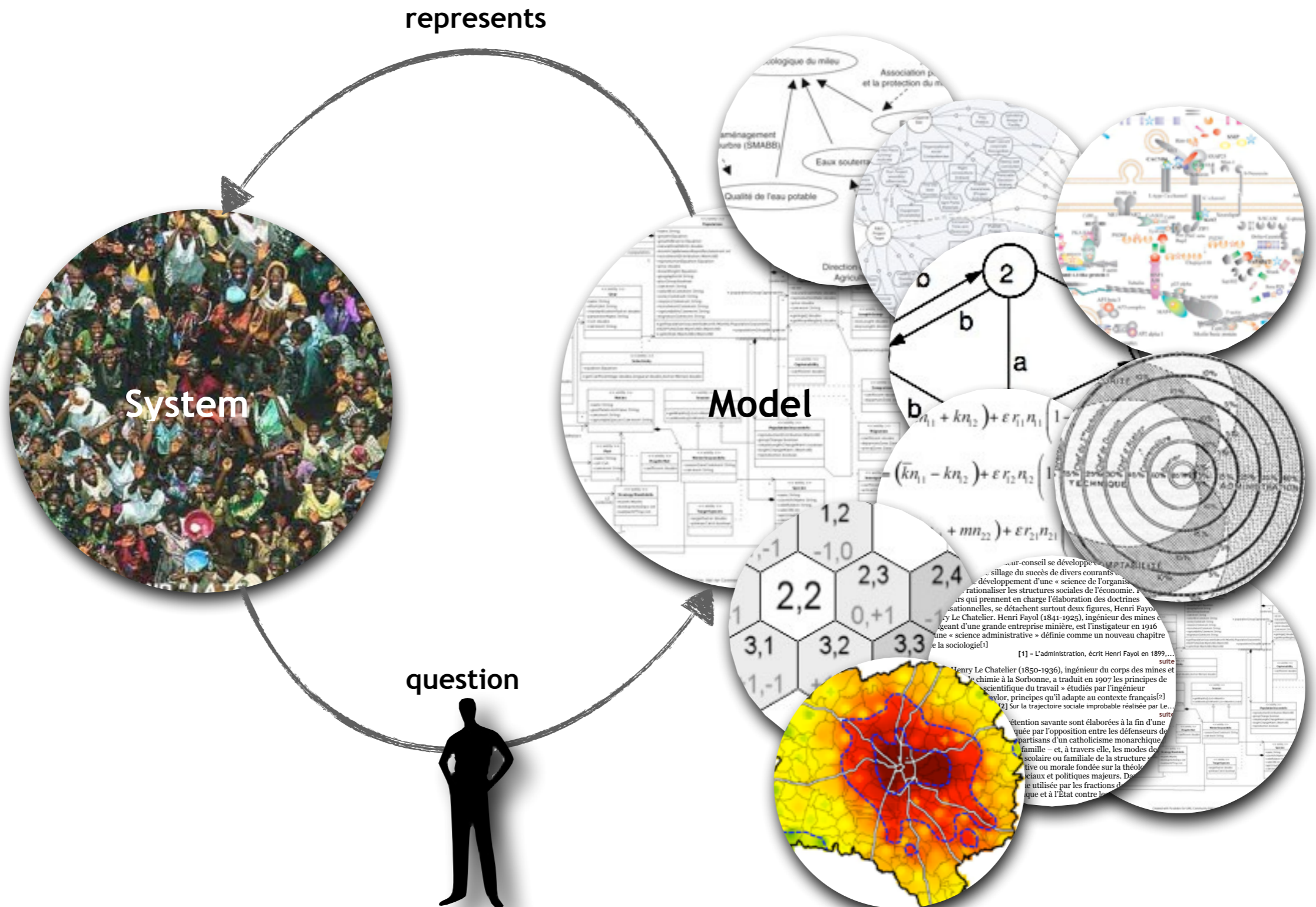
Agent-Based Modelling and Simulation



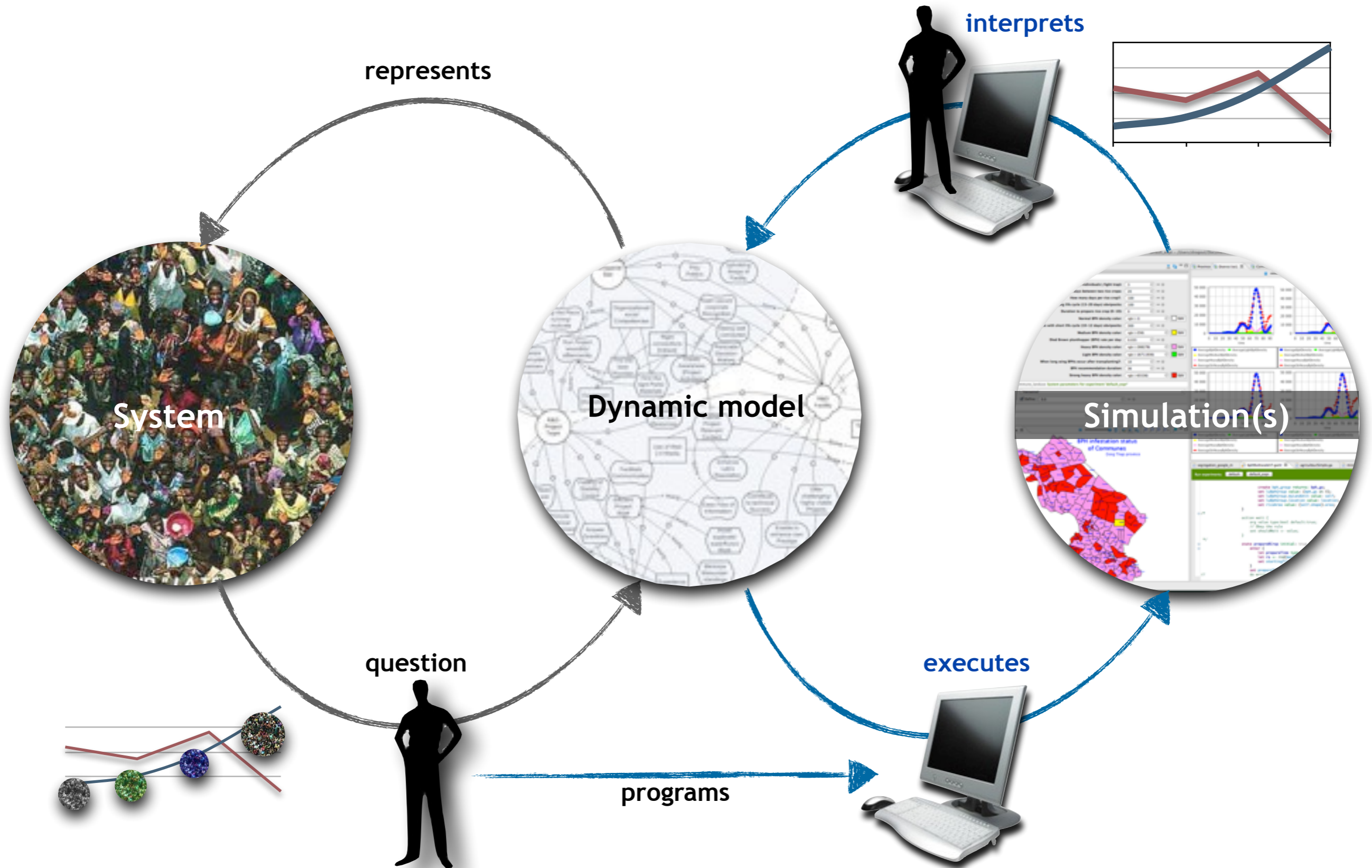
A model is a simplified representation of a reference system, designed to help answering a question on this system.



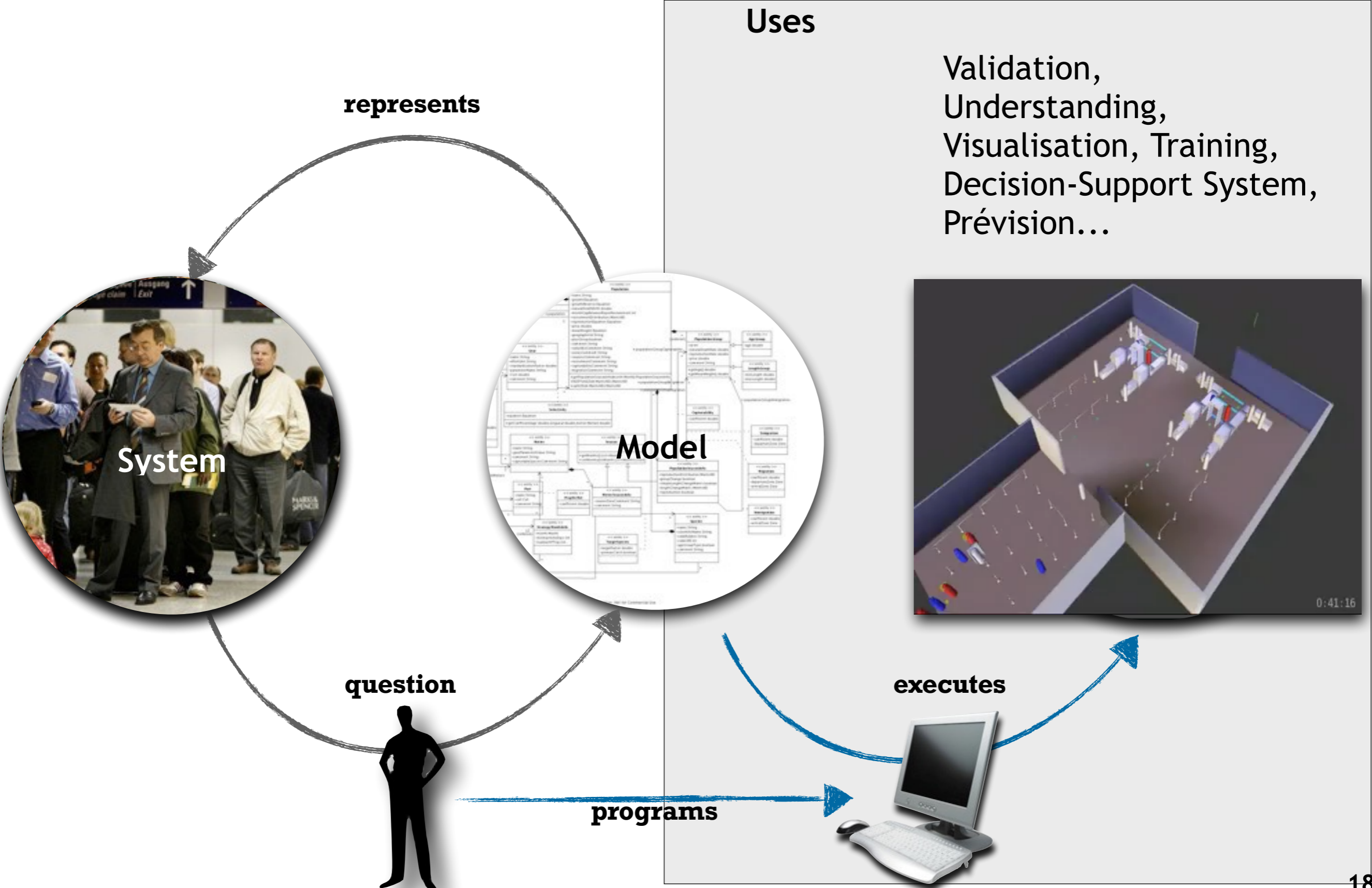
Representation can use multiple supports and languages,
 depending on the question, traditions...
 They are specified by meta-models.



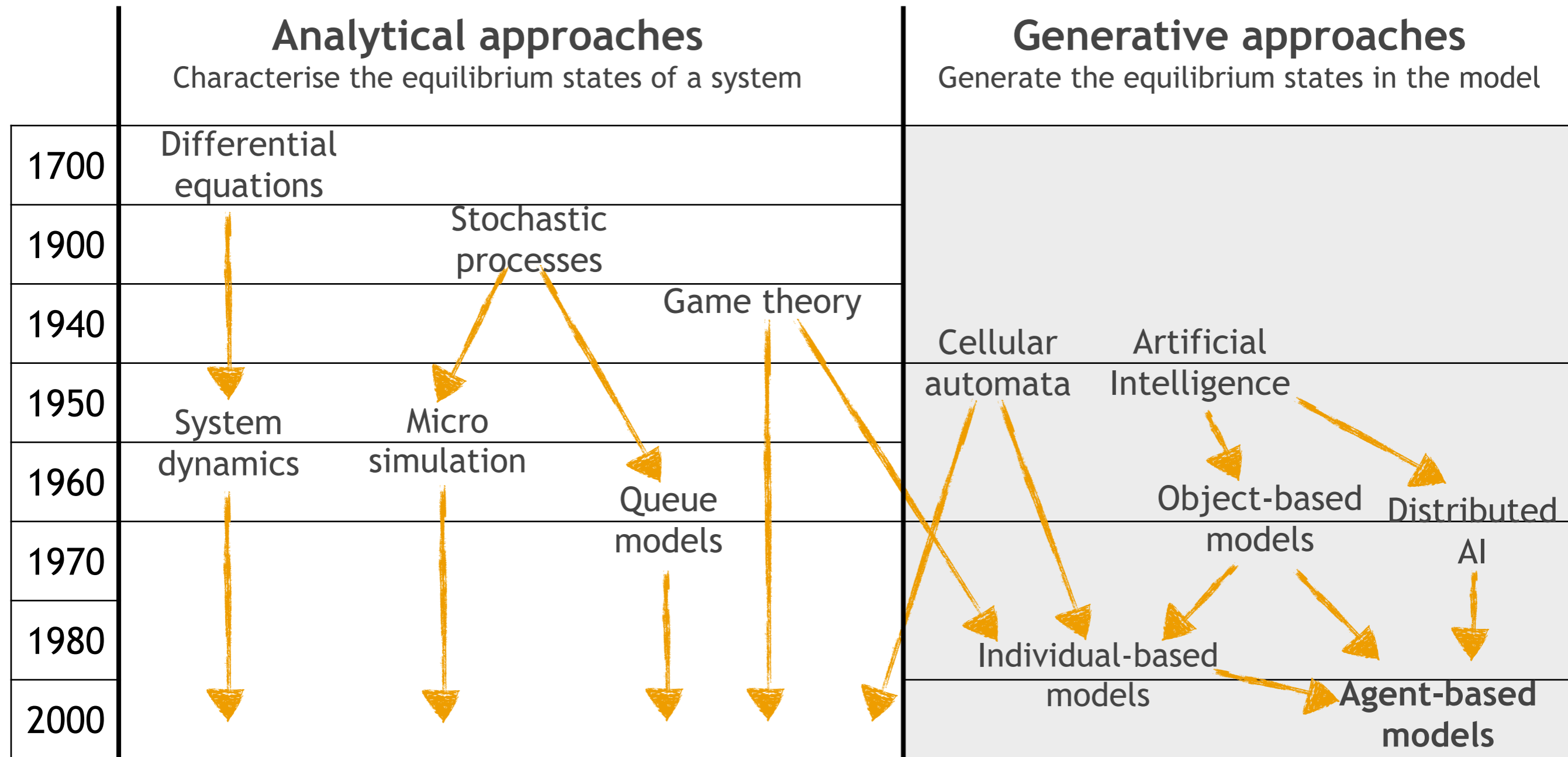
The execution of a (dynamic) computer model is called a simulation.



Uses of the simulation



Dynamic models and simulations meta-models have a long history

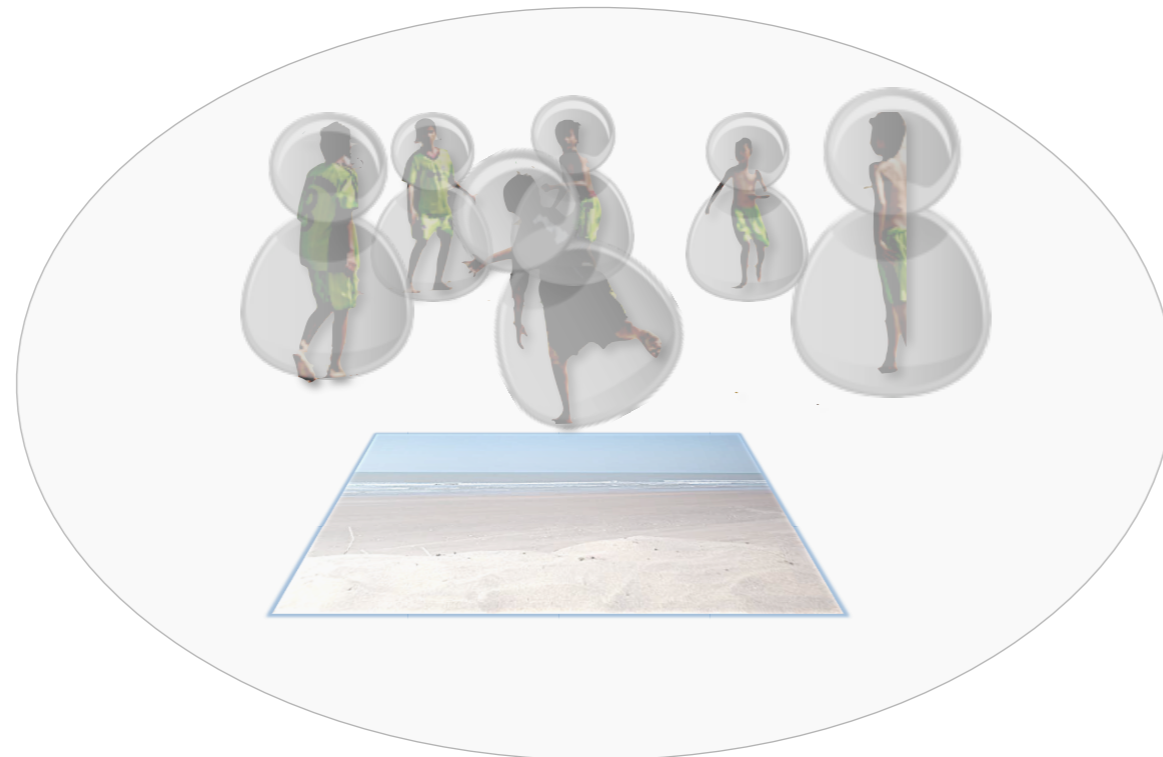


(adapted by Alexis Drogoul from Troitzsch & Gilbert 2001)

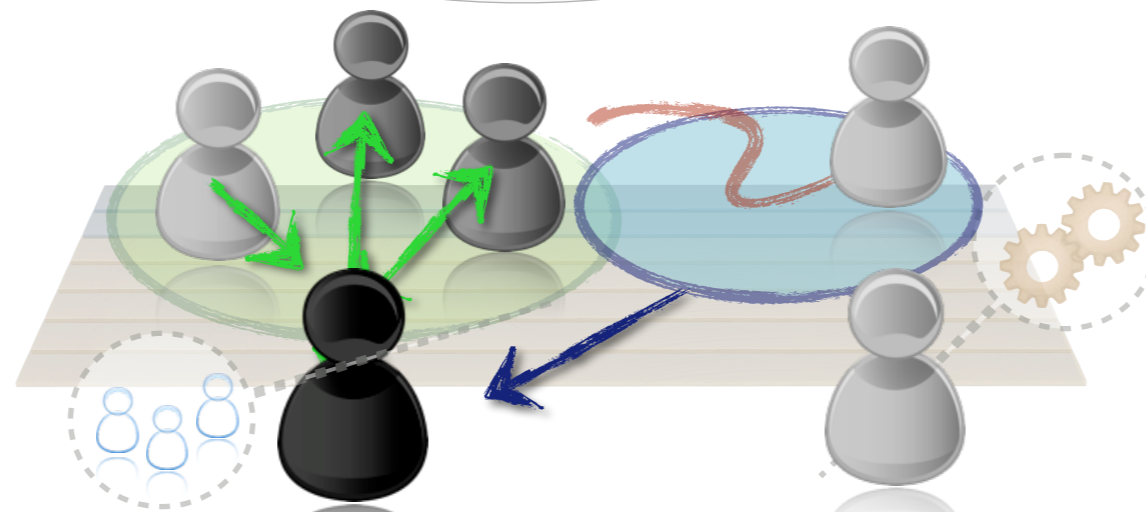
Agent-based models rely on a form of individual-based (1 to 1) representation.



Reference system

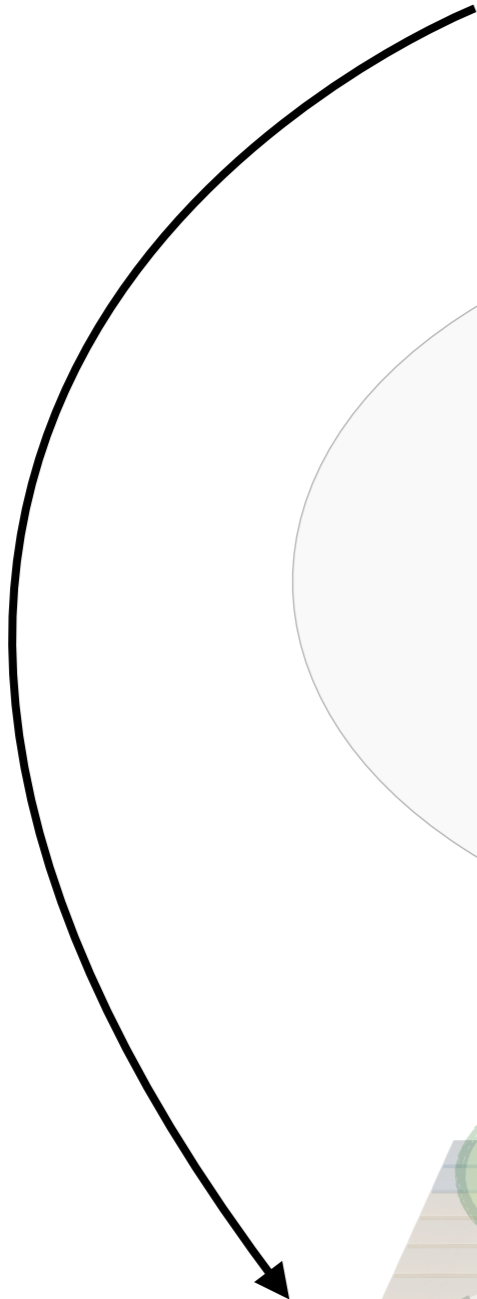


Modeling

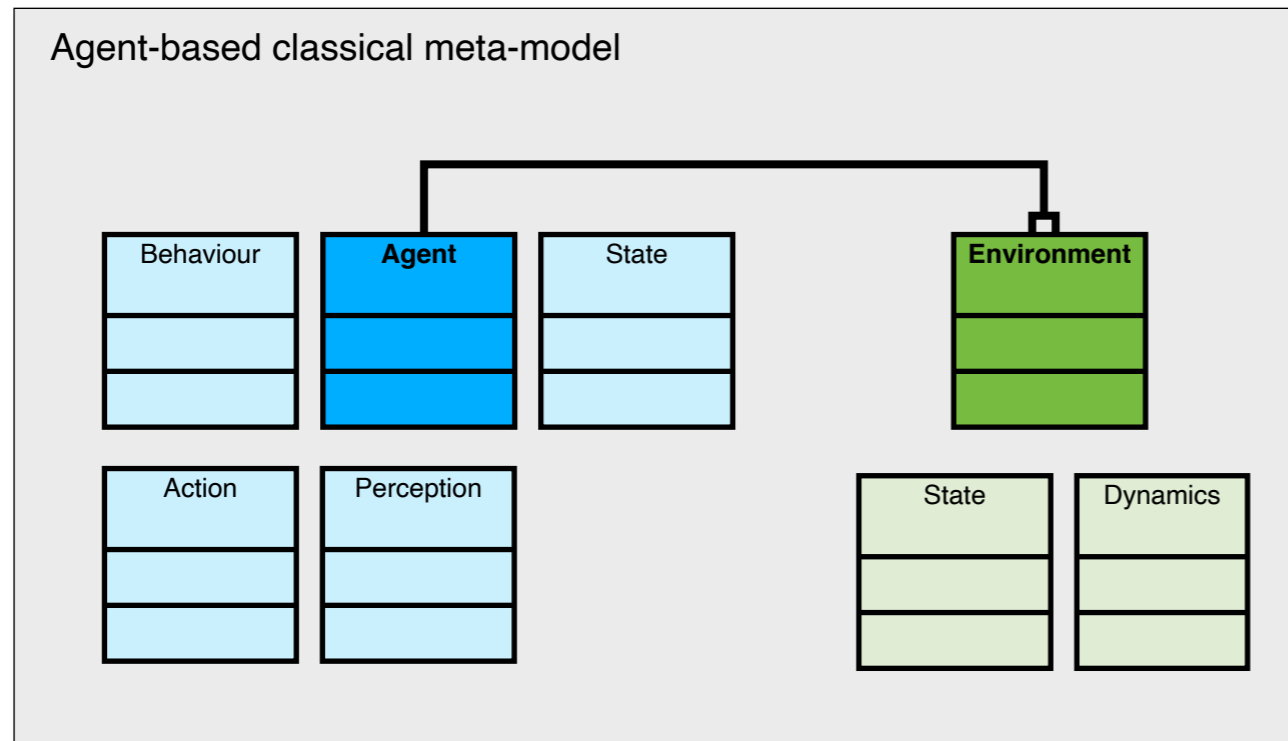


Agent-based model

represents



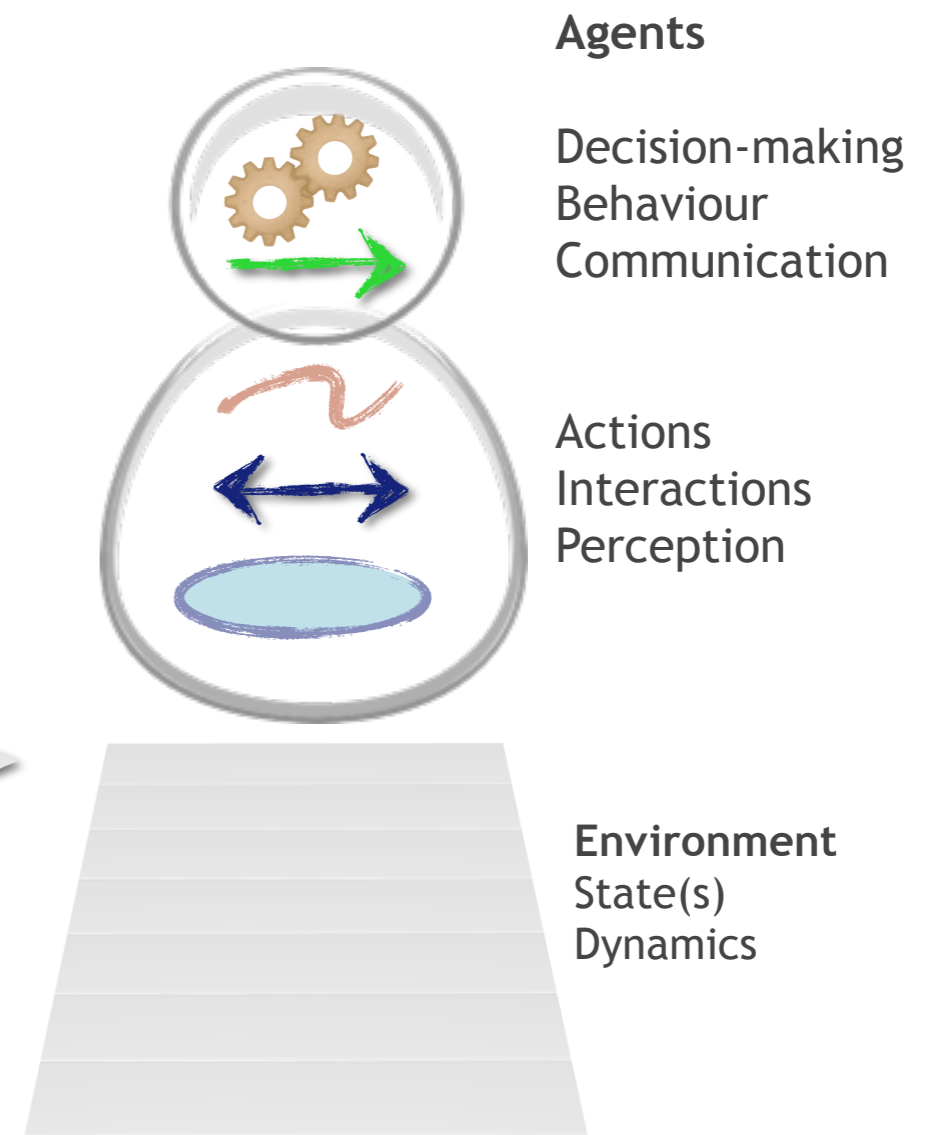
Agent-based models are written using a few key concepts



```

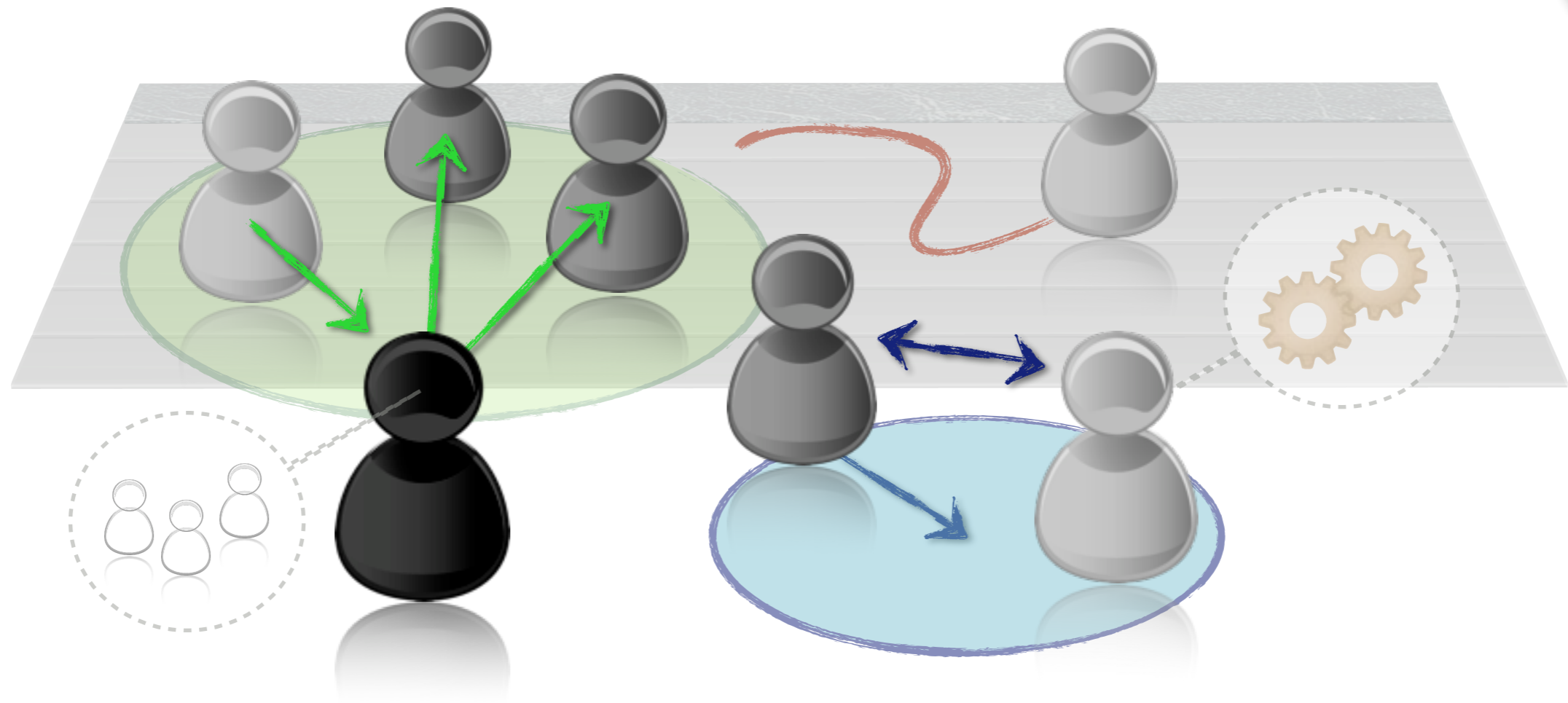
environment width: width_and_height_of_environment height: width_and_height_of_environment
species cells skills: [moving] {
  const speed type: float <- speed_of_agents ;
  rgb color <- [100 + rnd(155), 100 + rnd(155), 100 + rnd(155)] ;
  float size min: 1 max: 10 <- 4;
  int strength <- 0 ;
  float range min: range_of_agents max: width_and_height_of_environment ;
  cells leader <- self ;
  int heading <- rnd(359) update: leader.heading;
  reflex move {
    do move ;
  }

  reflex change_leader when: (leader != self) and (self distance_to_leader > range) {
    if grow_leader {
      set range of mv leader <- (range of mv leader) * 1.1 ;
    }
  }
}
  
```



The agents and their environment constitute a virtual «micro-world» that can be experimented like a real system (with more freedom)

Time



Environment

Illustration: road traffic

- ▶ Traffic jams, delays
 - How to understand them?
 - How to anticipate them?
 - How to control them?
- ▶ Difficulties
 - « Experiments » are difficult
 - Reproduction is impossible
- ▶ Simulation is the only way



Classical macroscopic models

- ▶ based on a (physical) analogy between the flow of traffic and the flow of a fluid in a pipe.

$$\begin{cases} Q(\Delta x, \Delta t) = K(\Delta x, t)V(\Delta x, \Delta t) \\ V(\Delta x, \Delta t) = V_e(K(\Delta x, t)) \\ \frac{\partial Q(\Delta x, \Delta t)}{\partial x} + \frac{\partial K(\Delta x, t)}{\partial t} = 0 \end{cases}$$

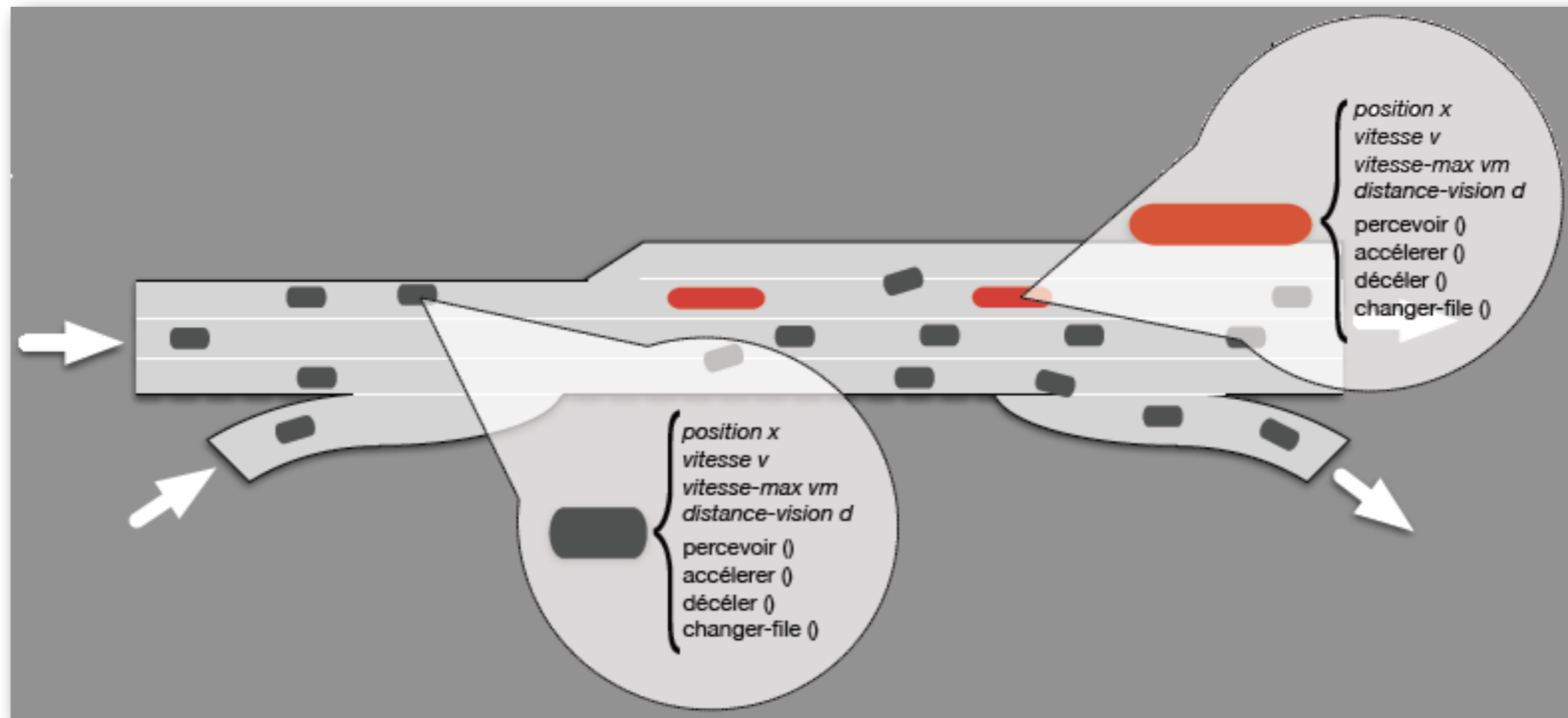
LWR macroscopic model



- ▶ Flow of vehicles characterised by macroscopic attributes, behaviours averaged in global equations.

Agent-based models

- ▶ Attributes and behaviours (ou equations, rules ...) are “attached” to individuals.



- ▶ Allow for heterogeneous vehicles and behaviours and stochasticity.
- ▶ Analytical results impossible to obtain: simulation is the only possibility.

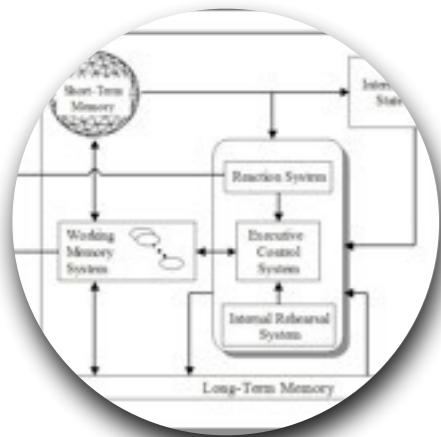
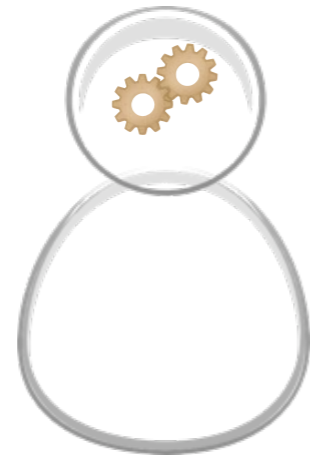
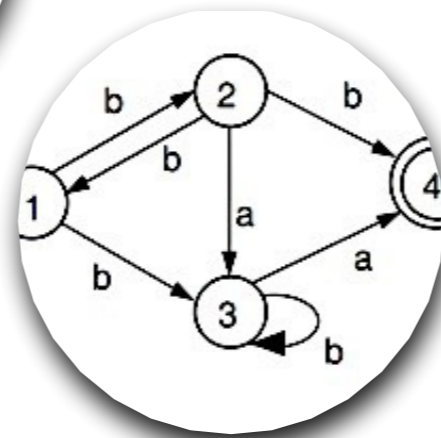
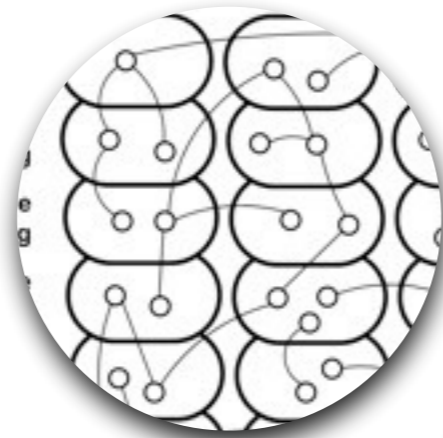
Agent-based models are versatile and heterogeneous: agents can represent any object or aggregation of objects of the reference system.



SimPop model
<http://www.simpop.parisgeo.cnrs.fr/>

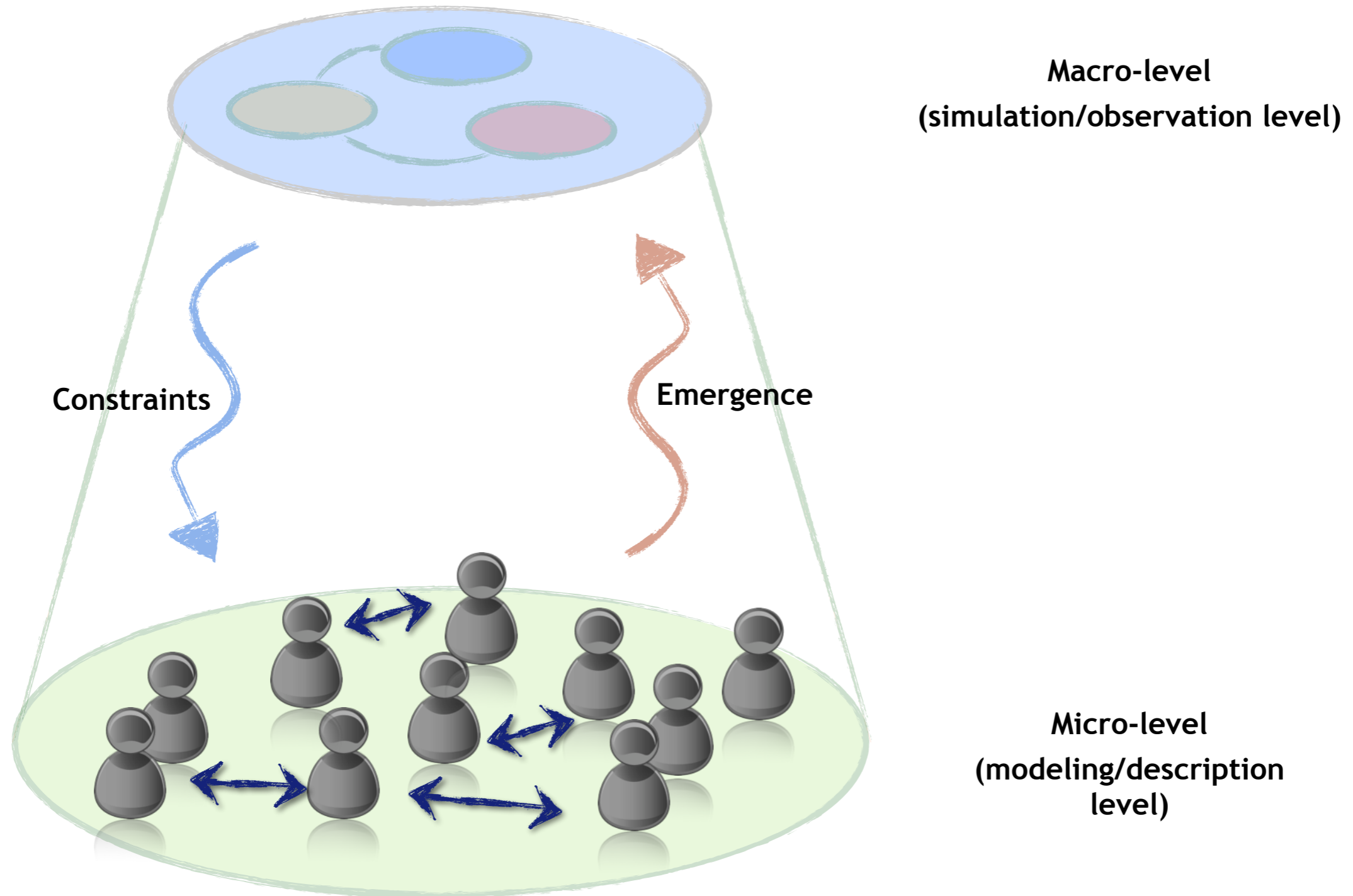
Agent-based models are agnostic : agents can be programmed using any language or decision-making architecture

```
reflex change_leader when  
  if grow_leader {  
    set range of my l  
  }  
  set leader <- self ;  
  set color <- [100 + rn  
  set range <- range_of  
  set heading <- rnd(36  
}  
reflex aggregate when  
  let candidates
```



- Any computer program
- Expert systems
- Finite state automata
- Task-based architectures
- Perception-decision-action architectures
- Planning architectures
- Neural networks
- Bayesian networks...

Agent-based models are generative: they represent behaviours at a «micro-level» and generate outcomes, with simulations, at a «macro-level».



Agent-based models are well adapted to situations where...

1. it is difficult to test hypotheses solely based on observations of the reference system
2. the actors of a reference system are **heterogeneous**
3. it is possible to identify intermediary **levels/organisations** that influence the dynamics of the reference system
4. the **level of analysis/observation** is not fixed
5. changes at the macro-level should be outcomes, and not inputs, of the model



As such, they represent an invaluable tool for building models of complex systems

Some remarks concerning Agent-Based Modelling

► Advantages

- Hypotheses expressed at the individual level
- Modelling the dynamics
- Models are experimental objects (simulation)

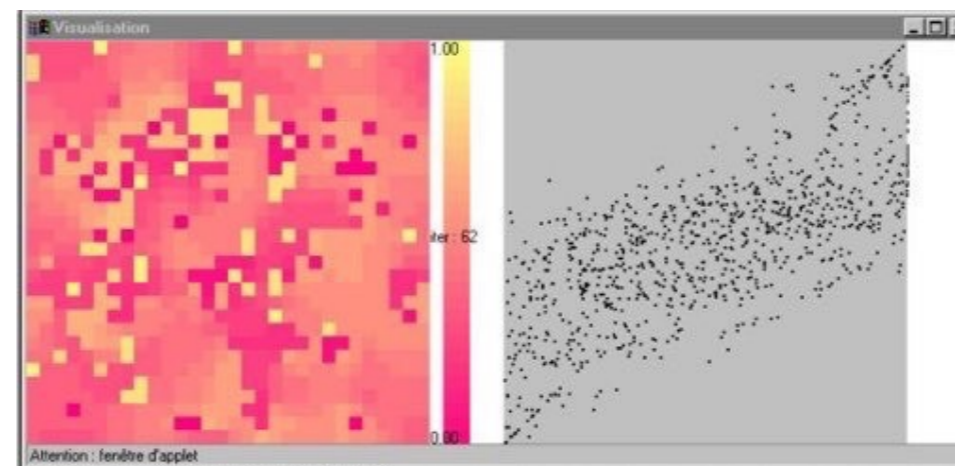
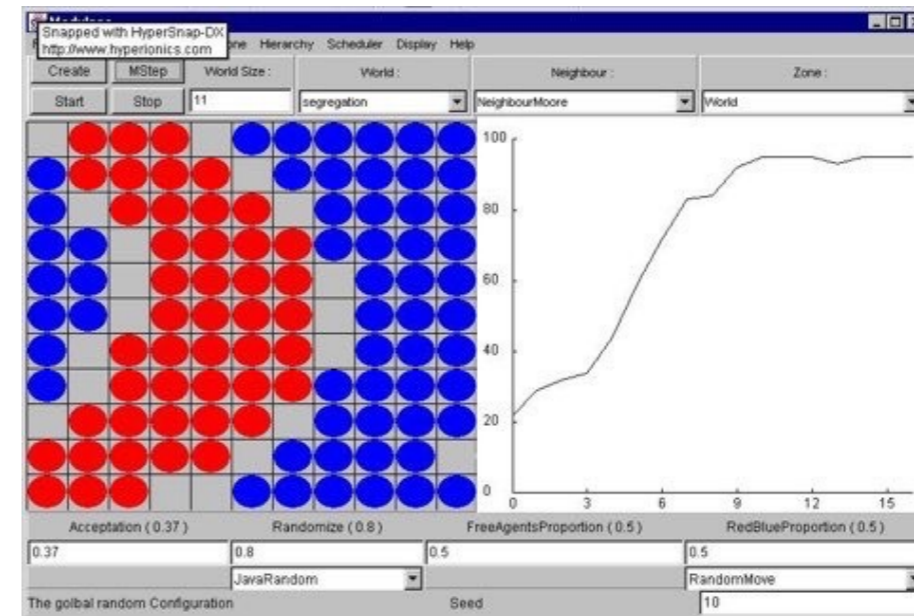
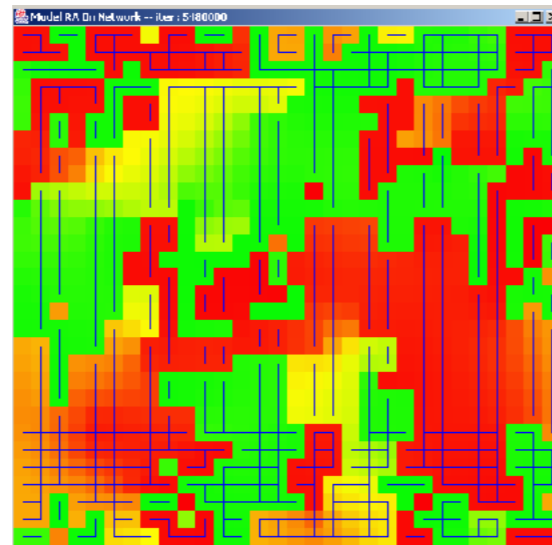
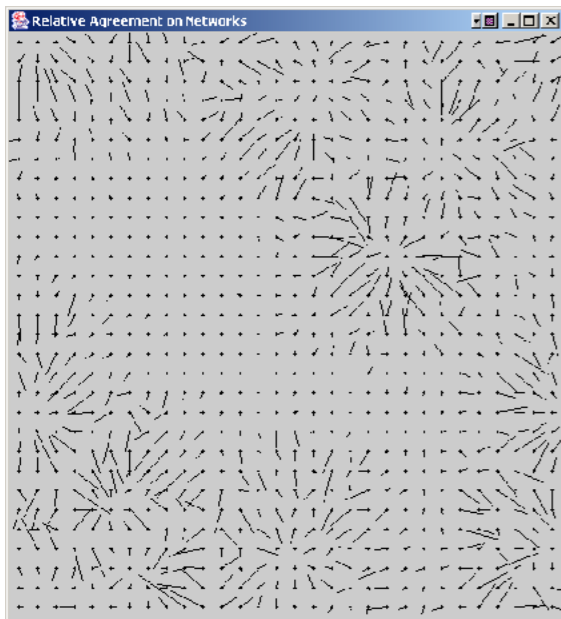
► Drawbacks

- Reproduction of the complexity of the real system (micro/macro link)
- Difficulty to understand how the results are produced
- Validation

Use of modelling

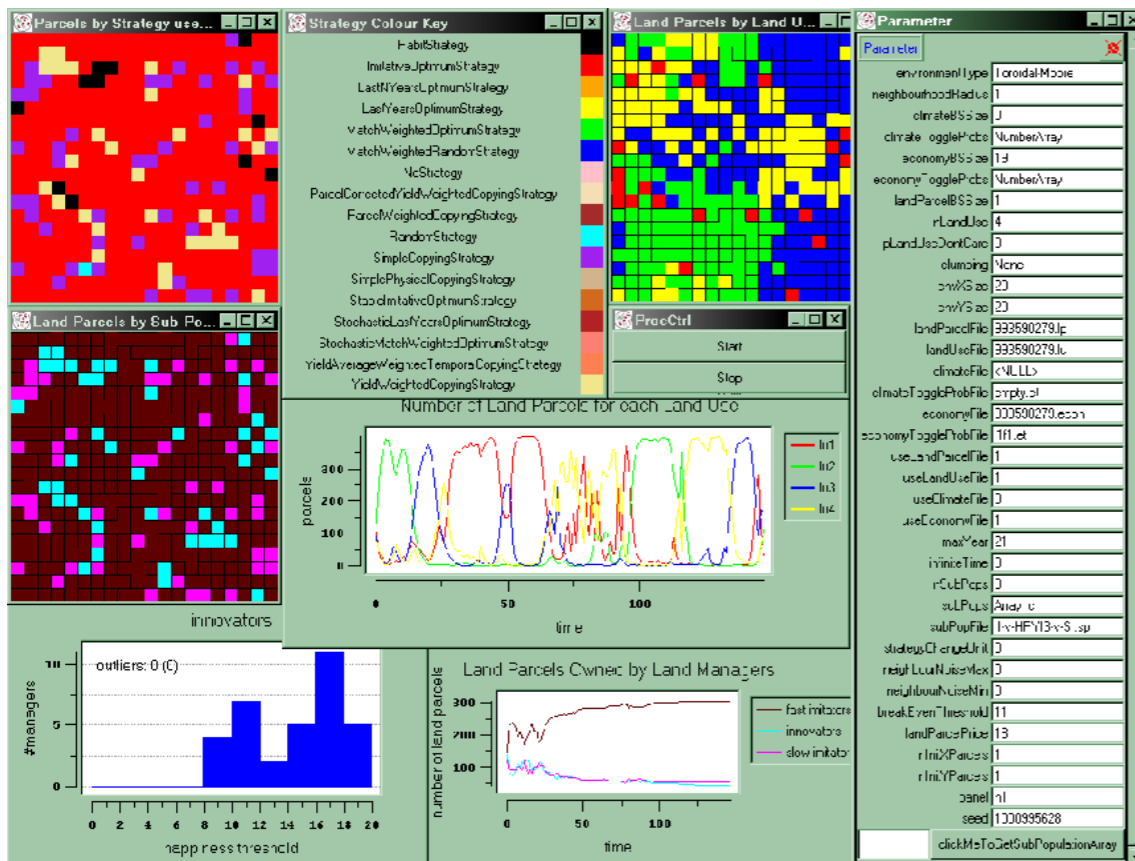
► To understand

- Test/elaboration of hypotheses, prospective simulation
- Formalization/verification of sociological theories



Use of modelling

- ▶ To understand
- ▶ To decide
 - Predictive simulation for decision-making
 - Test of scenarios through simulation
 - Artefacts for helping negotiation and/or coordinated management



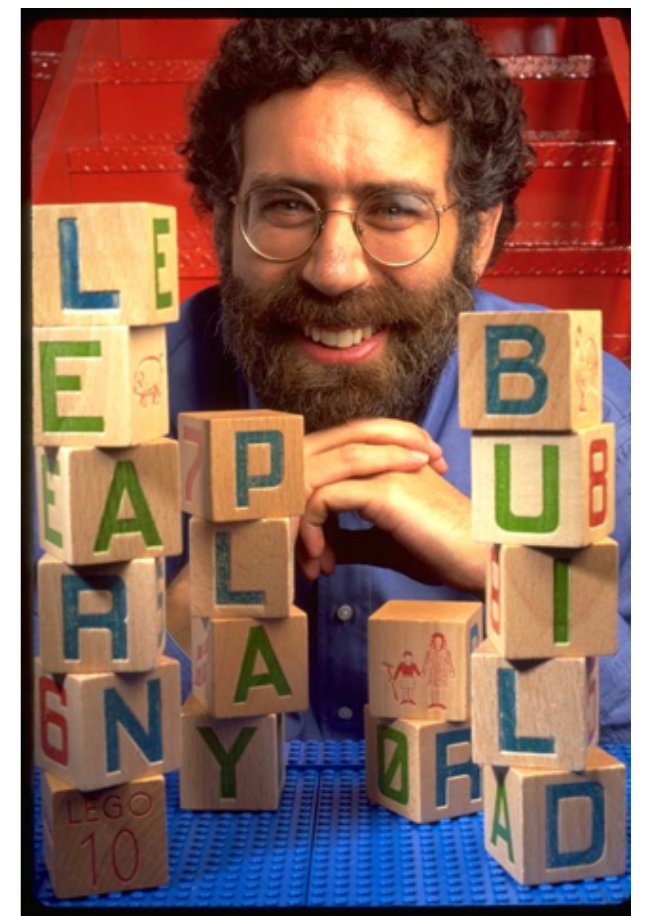
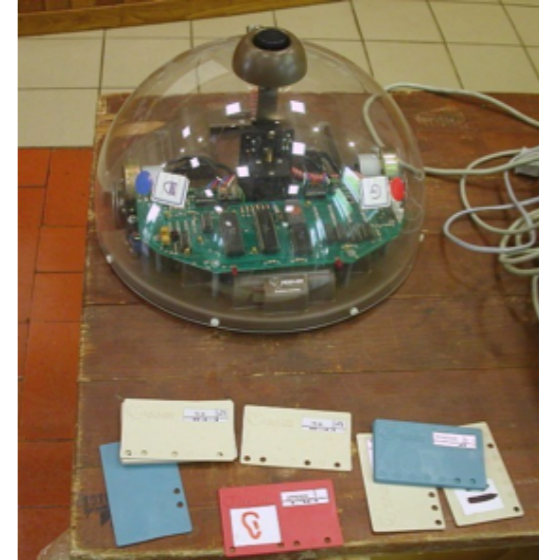
Use of modelling

- ▶ To understand
- ▶ To decide
- ▶ **Interactive simulation**
 - « Serious » Gaming (SimCity)
 - Training, teaching



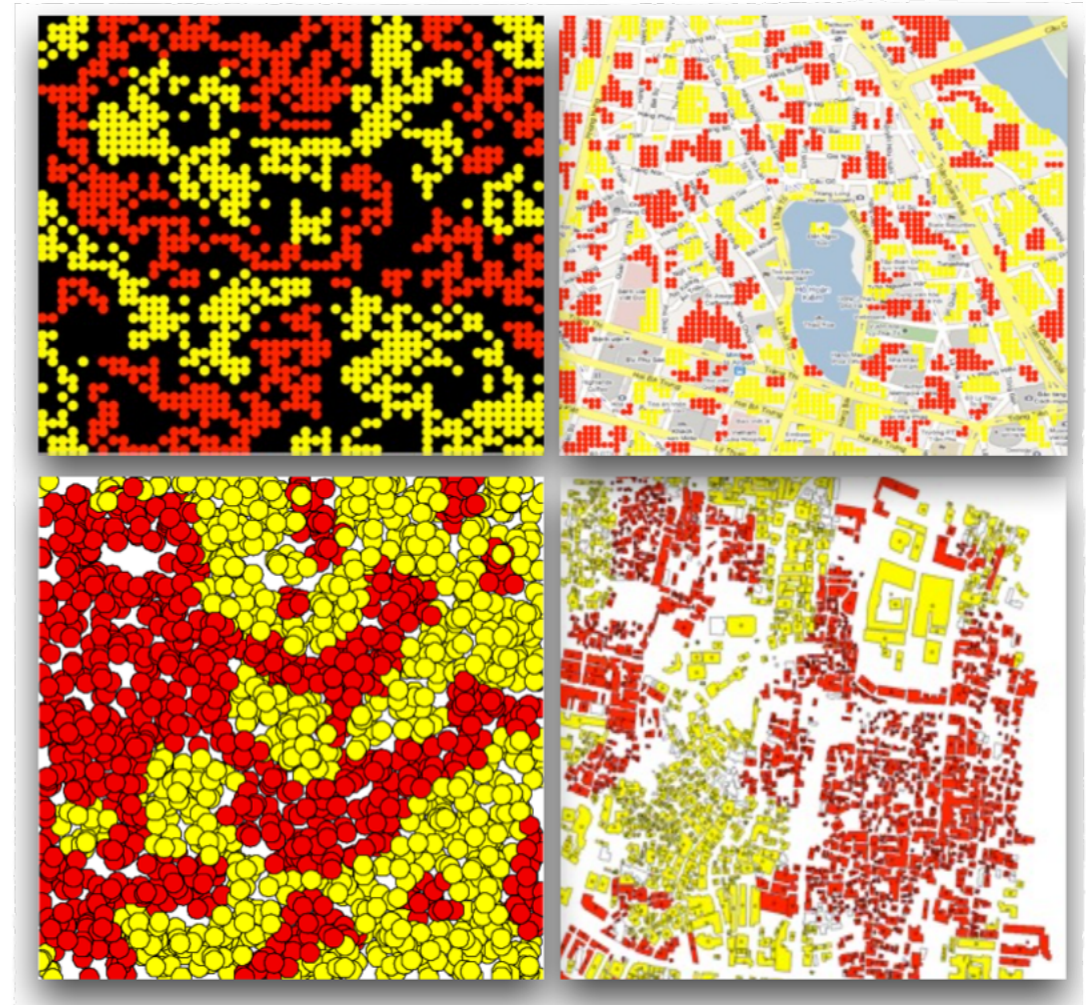
Rebuild phenomena to understand them

- ▶ Use Modelling not Models !!!
- ▶ Rebuilding phenomena enable to better understand them
 - Pappert (Logo), Resnick (NetLogo)
- ▶ What could be the individual behaviours producing a given collective phenomenon?
 - Experimental approach by trial and errors
 - ex: Regroupment of ants dead bodies outside of the anthill

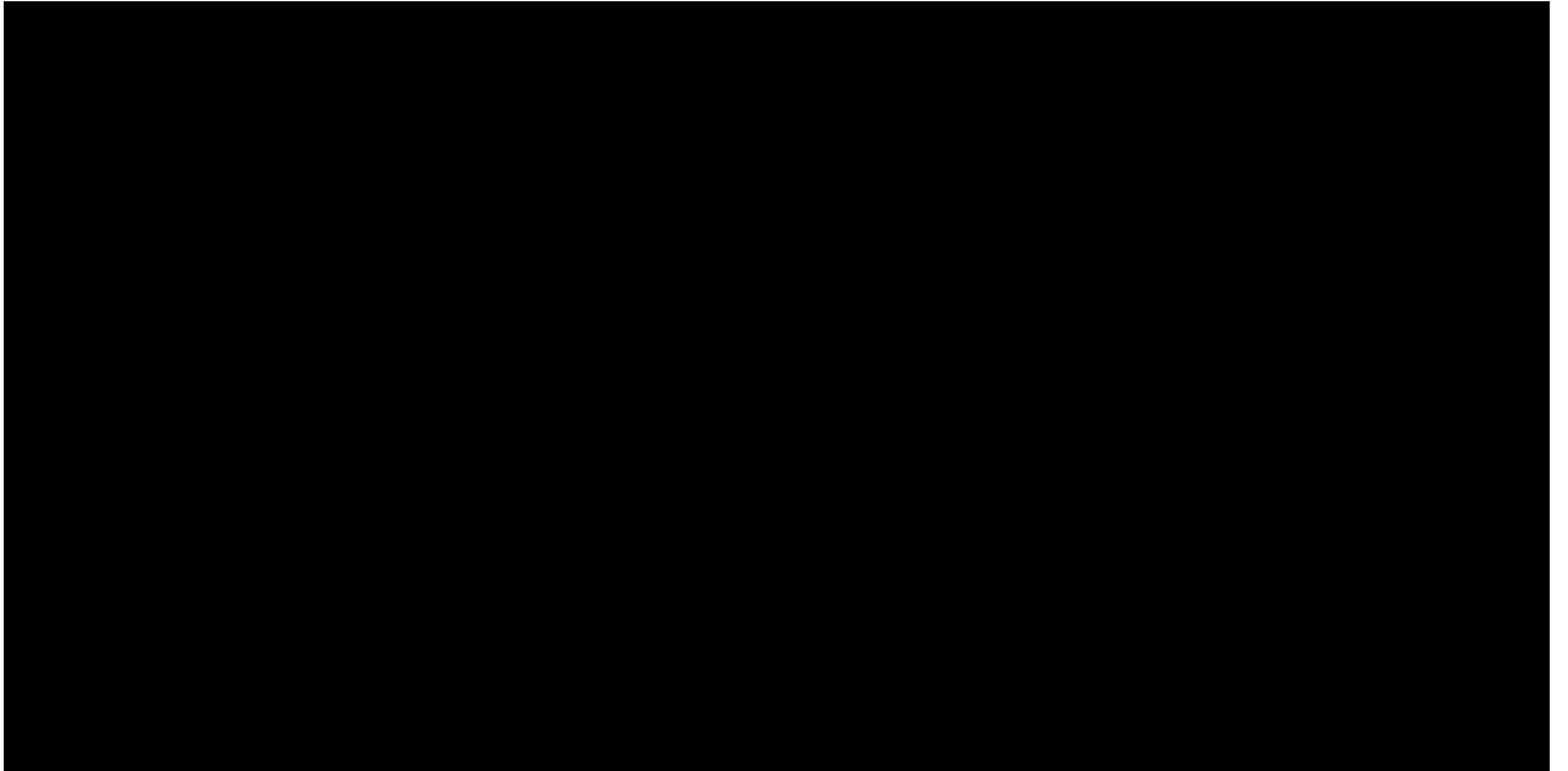


Agent-Based Modelling and Simulation

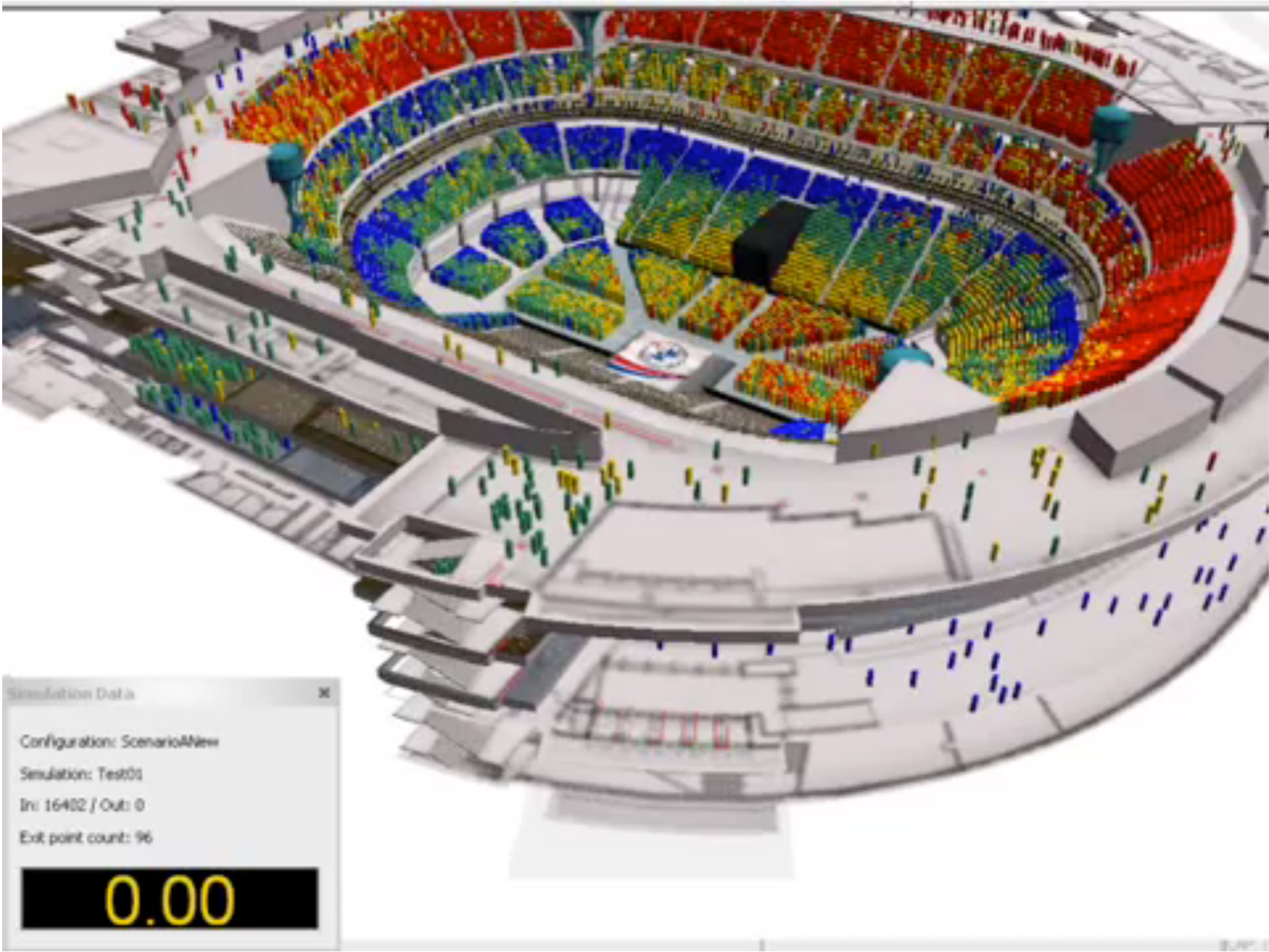
Some examples

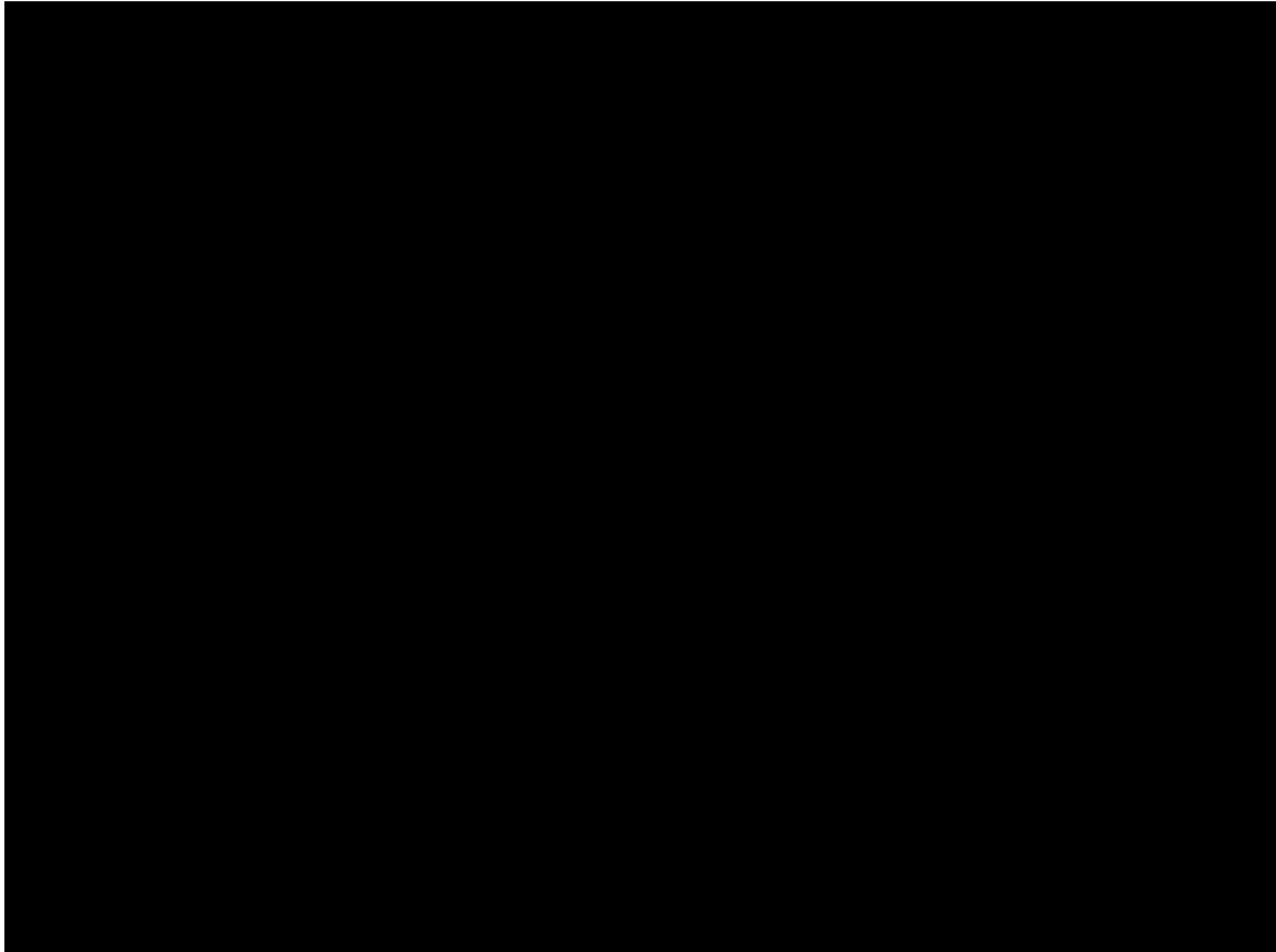


Crowd dynamics

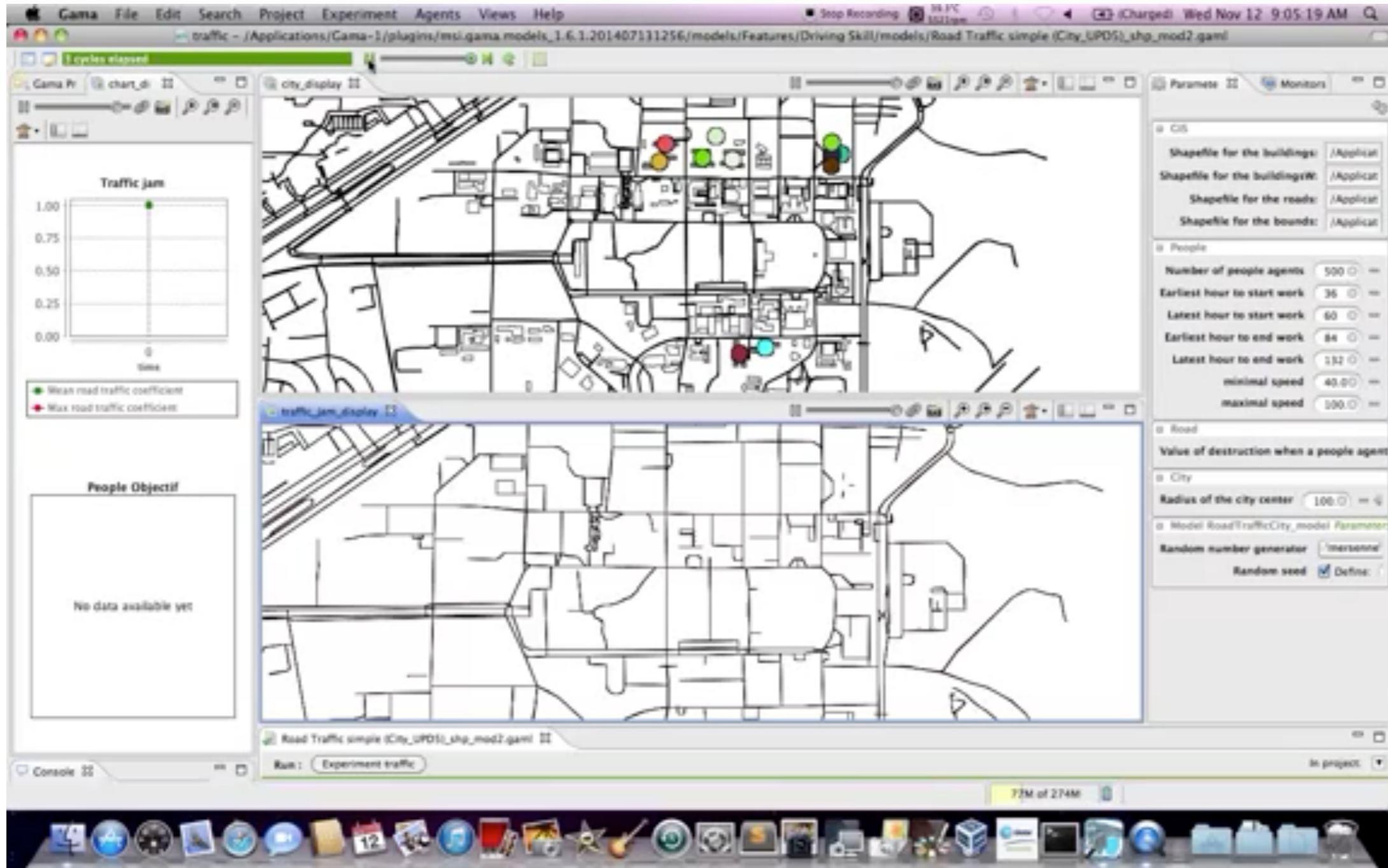


Evacuation





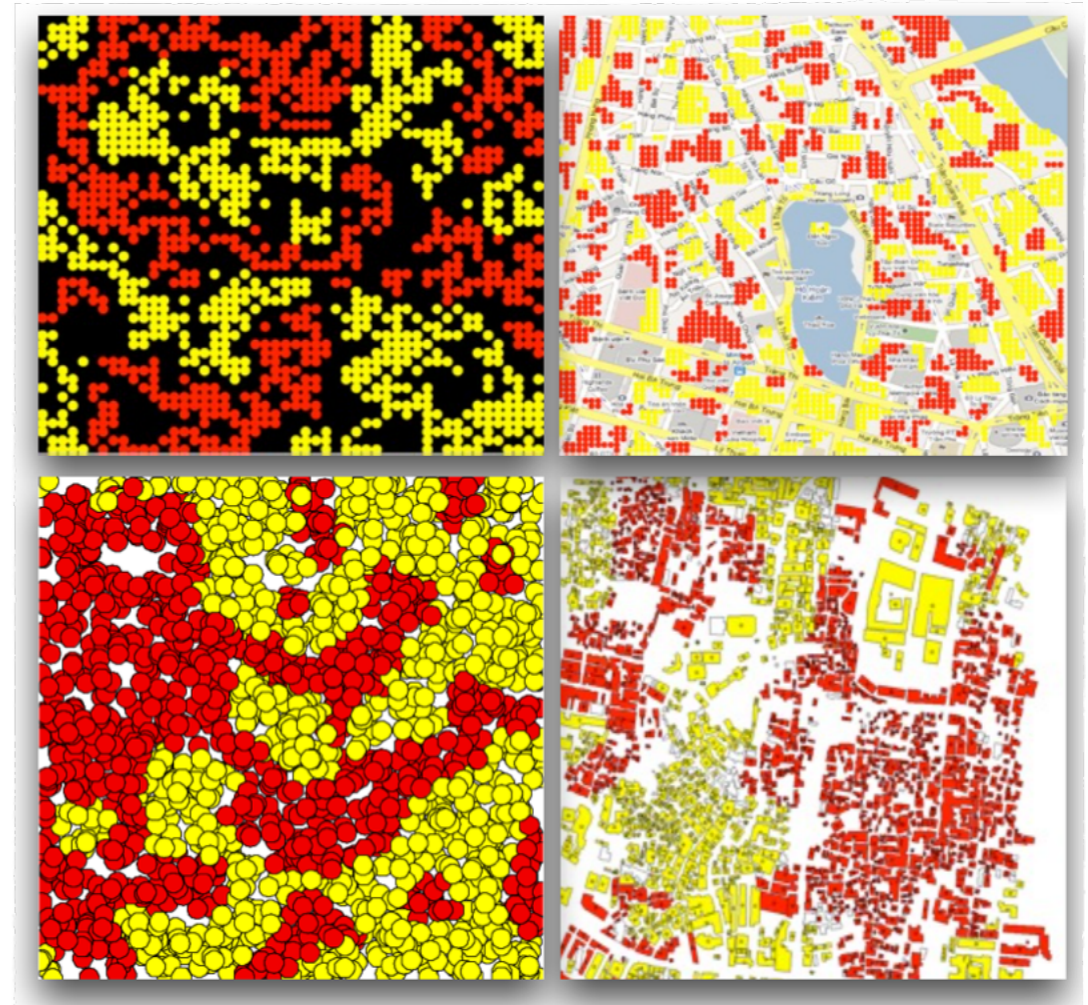
Traffic



Agent-Based Modelling and Simulation

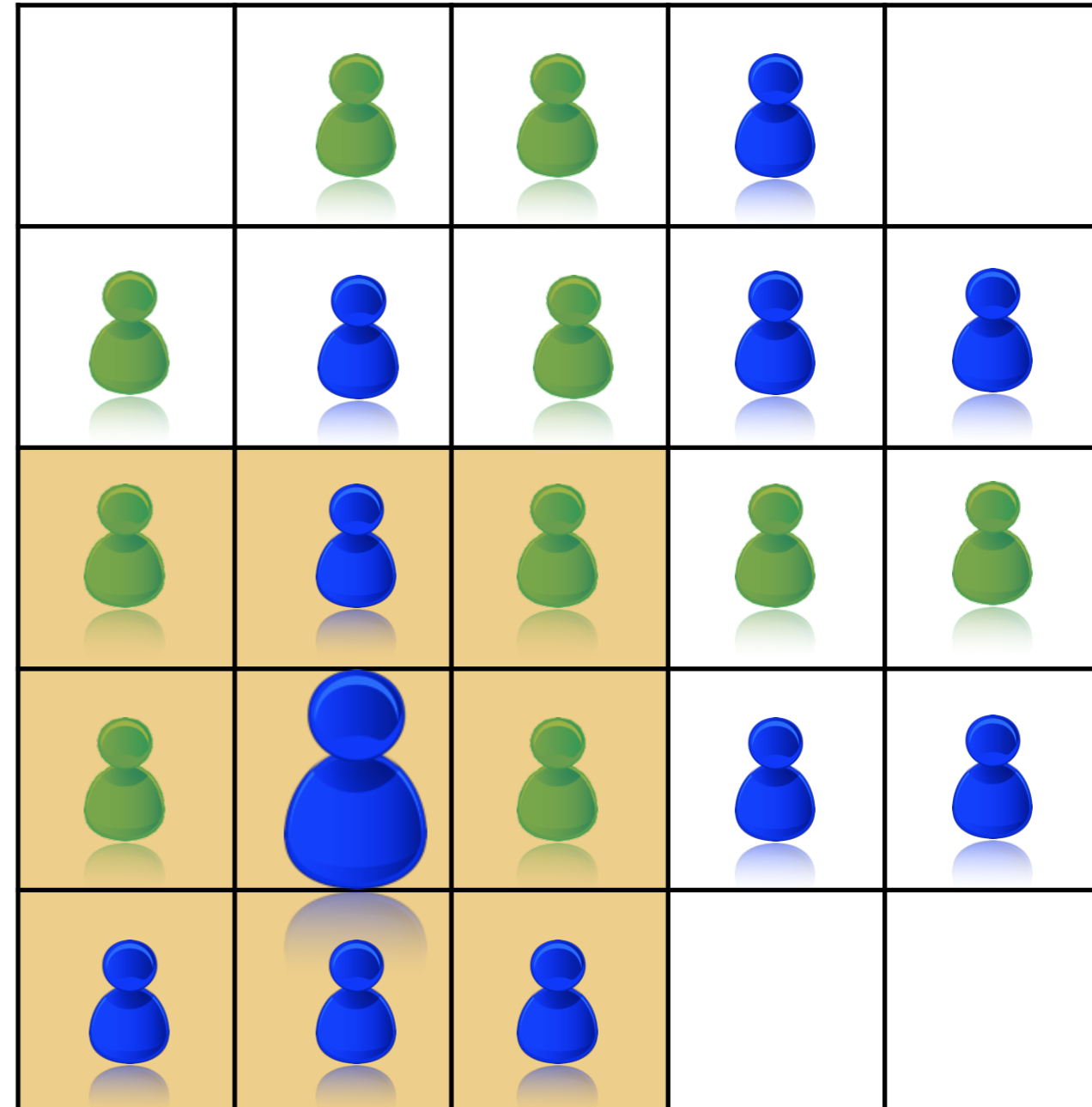
The Schelling model

(a.k.a. the HelloWorld of the social simulation)



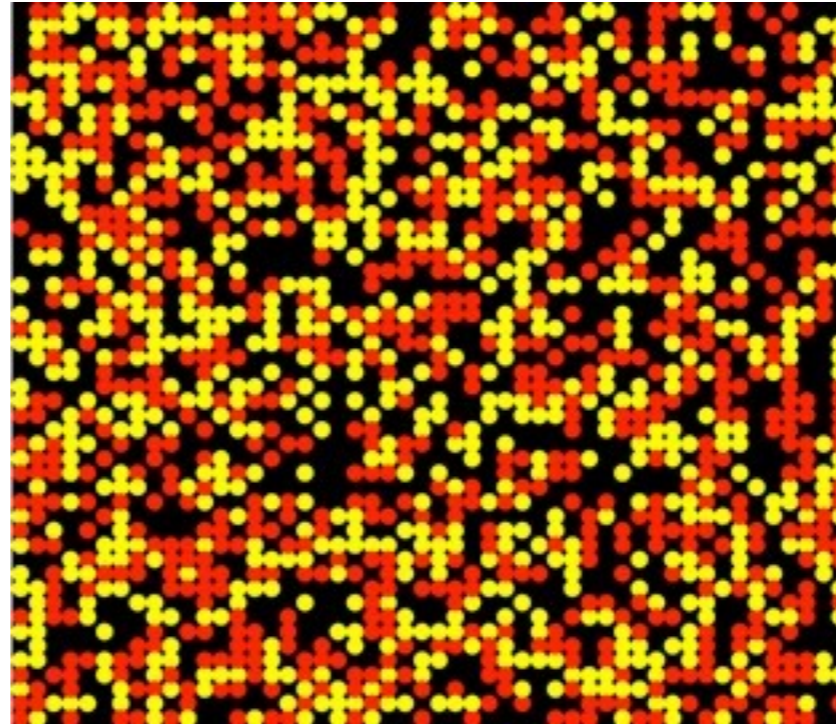
Description

- ▶ In a simple agent-based model, agents are located in a discrete environment (grid). Each agent has a color, a perception of its neighbours and a **preference**: the minimal rate of neighbours of the same color that the agent can accept.
- ▶ **One behaviour**: if the rate of neighbours with the same color as the agent is lower than the agent preference, then it moves to another random free cell.

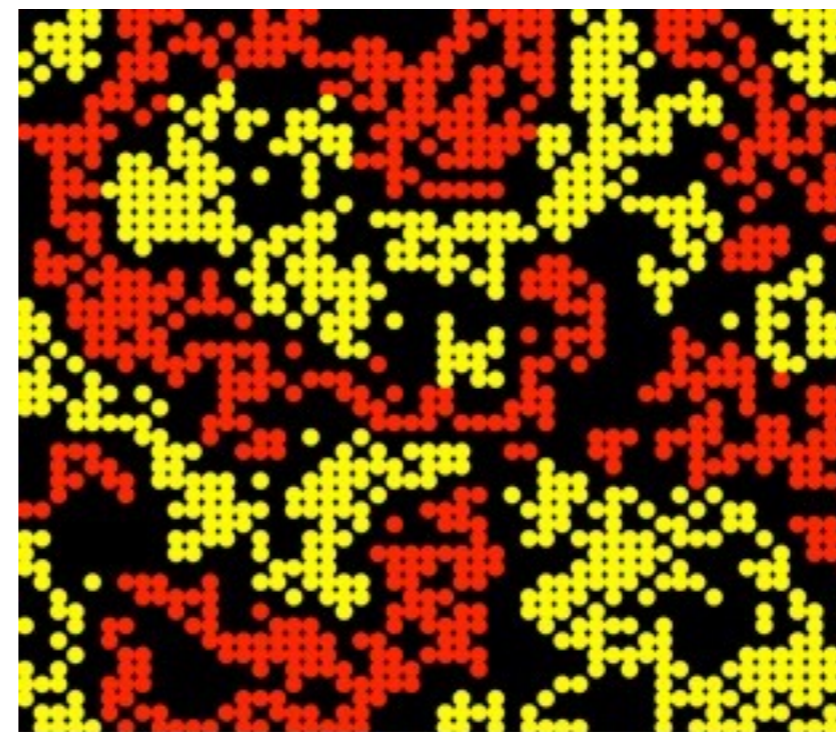


Results

- ▶ The main result shows that, even with a low individual preference (35%), 2 groups can be fully segregated.
- ▶ This phenomenon cannot be deduced analytically from the knowledge of individual preferences.



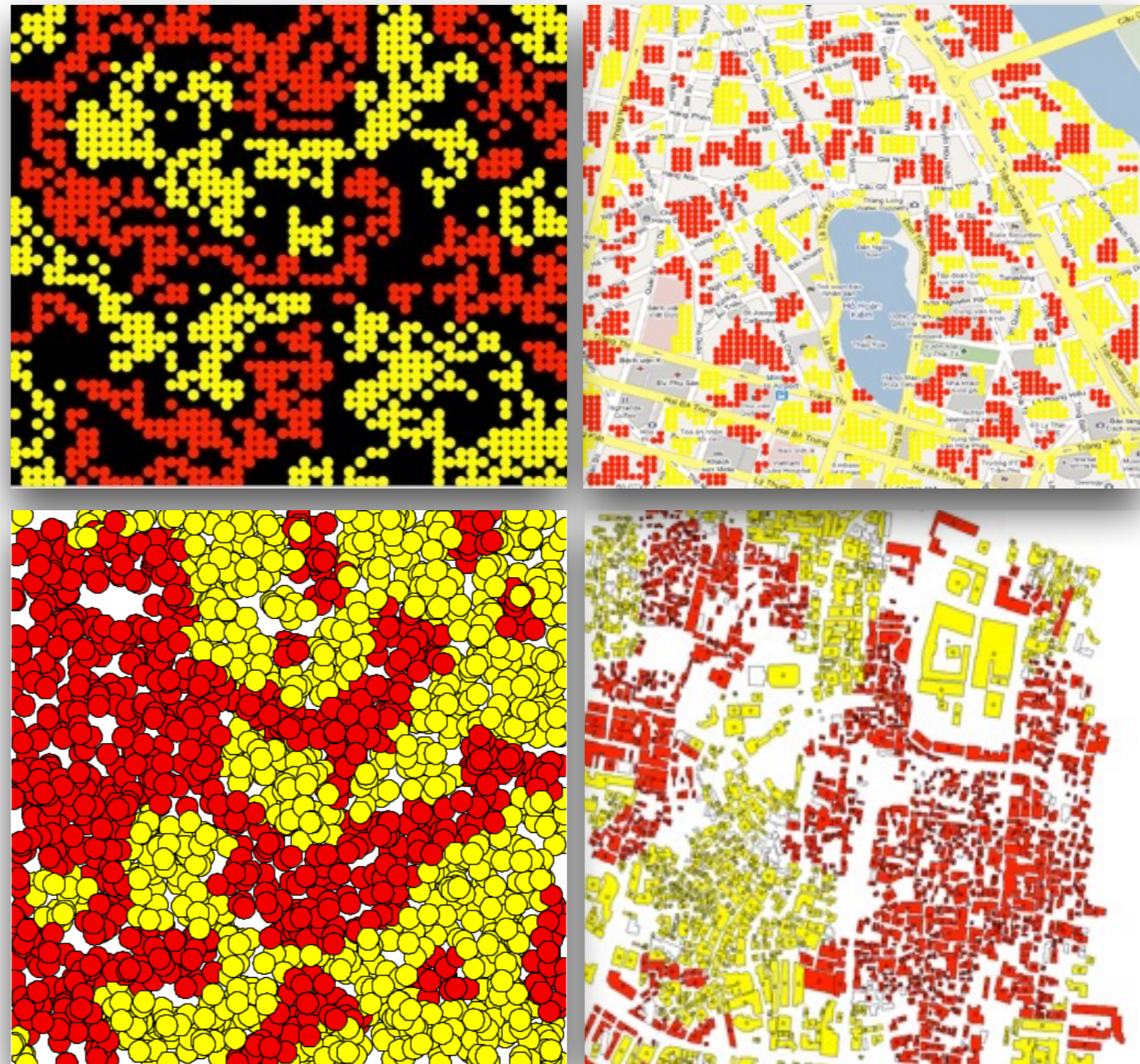
Initial situation
Grid : 50 x 50
1200 agents
Individual preference: 35%
Segregation index:
(\sum similar neighbours / \sum neighbours)
: 49,9 %



final situation (equilibrium)
Segregation index : 94 %

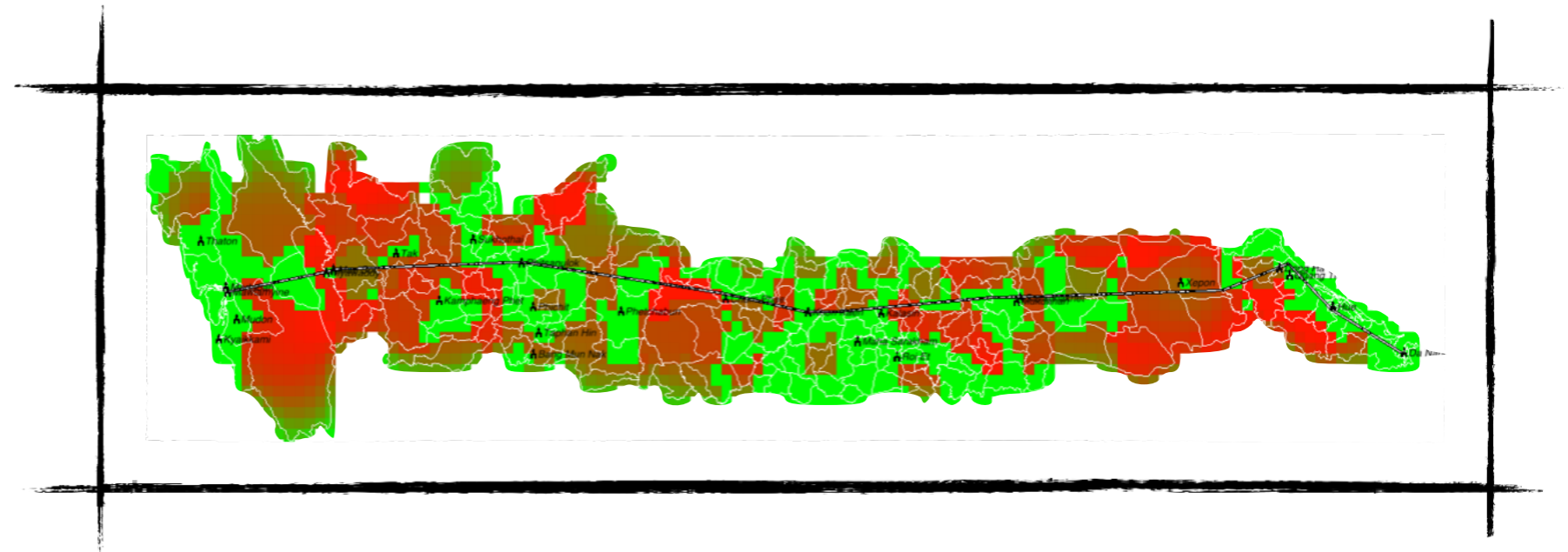
Spatially explicit segregation

- ▶ An interest of agent-based models is that they can be initially designed simple and becomes more and more complex, e.g. by using realistic data.
- ▶ The same segregation model can be used on continuous or discrete (grid) environment, on grid built using Google Maps or using GIS data.



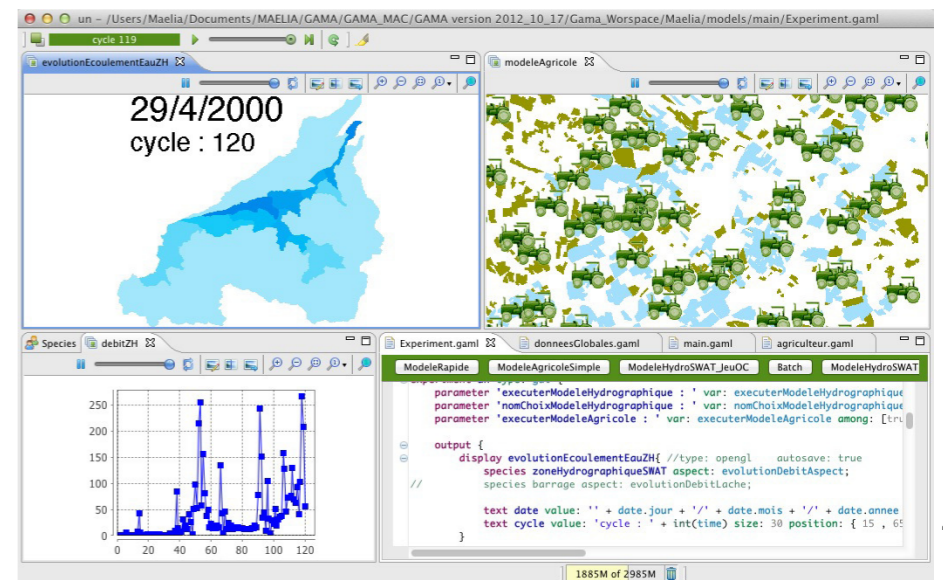
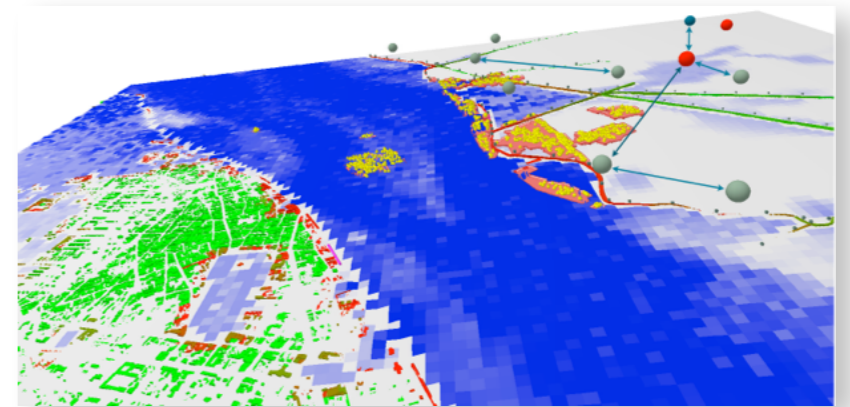
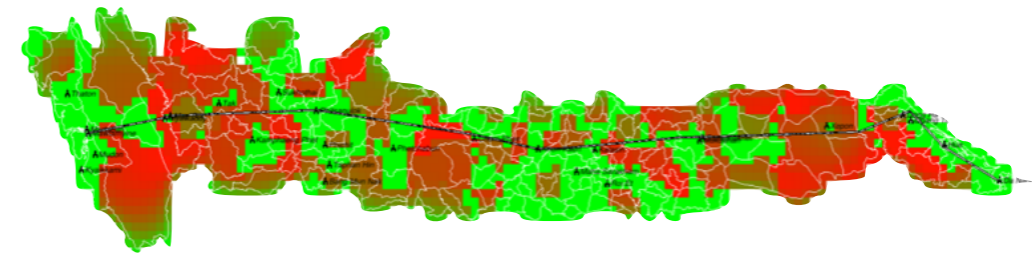
Voir aussi : Crooks, A. T. (2010), Constructing and Implementing an Agent-Based Model of Residential Segregation through Vector GIS, *International Journal of Geographical Information Science*, 24(5): 661-675.

Examples of modelling projects



Some examples...

- ▶ Understanding and exploring urban spatial dynamics
 - ▶ Case study: Can Tho (Vietnam)
- ▶ What incidence do the economic exchanges between ASEAN countries have on the dengue spread?
 - ▶ Case study : Vietnam-Myanmar corridor
- ▶ Reproducing and exploring past events using agent-based geo-historical models
 - ▶ Case study: floods of 1926, Hanoi (Vietnam)
- ▶ Assessment of the social, economic and environmental impacts of the various alternative of definition and management of (new) water Volume Available for Agriculture
 - ▶ Case study: Adour-Garonne basin (France)



Details about vocabulary

- ▶ Agent-based Modeling (ABM)
- ▶ Individual-based Modeling (IBM)
- ▶ Multi-Agent-Based Simulation (MABS)

Sources

- ▶ François Bousquet
- ▶ Alexis Drogoul
- ▶ Philippe Caillou