



Simulation



Introduction



Outline

- Introduction
- Multiagent-based Simulation approach
- Multiagent simulation platform
- Turtlekit



Simulation

Introduction

■ Simulation

- “*The process of designing a model of a real system and conducting experiments with this model for the purpose either of understanding the behaviour of the system or of evaluating various strategies (within the limits imposed by a criterion or a set of criteria) for the operation of the system.*” (Shannon 1976)
- “*Simulation is the discipline of designing a model of an actual or theoretical physical system, executing the model on a digital computer, and analyzing the execution output*” (Fishwick 1994)

■ Model

- “To an observer B, an object A is a model of an object A to the extent that B can use A to answer questions that interest him about A. “ [Minsky, 1965]

R. E. Shannon. Simulation modeling and methodology. In Proc. of the 76 Bicentennial Conference on Winter Simulation, pages 9-15, 1976.

P. A. Fishwick. Computer simulation : Growth through extension. In Society for Computer Simulation, pages 3-20, 1994

M. L. Minsky. Matter, mind, and models. In Proc. of the Intern. Federation of Information Processing Congress, vol. 1, pages 45-49, 1965.

Simulation

Introduction

■ Simulation supports

- Understanding, Exploration, Clarification
 - to understand the behavior of the reference system thanks to a model that is considered as a miniature reproduction of the reference system.
- *Validation, Assessment, Verification*
 - to test an hypothesis of the reference system, to validate or to certify the underlying theory.
- Control, action, control
 - to support a decision process or a control that will influence the state of the real reference system.
- Forecast, Prediction, Anticipation
 - to predict the possible evolutions of the reference system following evolutions or disturbances.
- *Communication, Formation, Visualization*
 - to show and share the model of the dynamic of the reference system.

Simulation

Methodologies

■ Fishwick

- The *model design* associates the real system with a representation of this system (*the model*).
 - This model is built from real observations (objective) or knowledge (subjective).
 - Data are usually formalized using formal semantics or mathematical logic to reduce ambiguities as much as possible.
 - It is then converted to algorithms,
- The *model execution* phase is the processing of the algorithm to produce numerical outputs.
- The *execution analysis phase*, deals with the analysis and confrontation of the results of the program with the behaviors observed in the model.

Fishwick, P.: Computer simulation: growth through extension. IEEE Potential February/ March (1996) 24 to 27

Drogoul, A., Vanbergue, D., & Meurisse, T. (2003). Multi-agent based simulation: Where are the agents?. In *Multi-agent-based simulation II* (pp. 1-15). Springer Berlin Heidelberg.

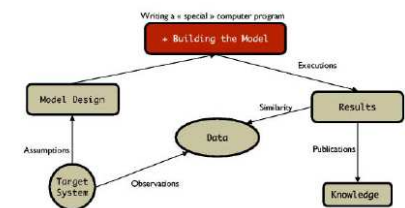
Multiagent-based Simulation approach

Simulation

Methodologies

■ Gilbert and Troitzsch

- Refine the Fishwick proposal with the addition of the model building phase
- The initial model is written into a computer program: the operational model
 - Adaptation to the simulator
 - There are both operational models and Simulators.
 - The differences between models introduces bias.



Drogoul, A., Vanbergue, D., & Meurisse, T. (2003). Multi-agent based simulation: Where are the agents?. In *Multi-agent-based simulation II* (pp. 1-15). Springer Berlin Heidelberg.

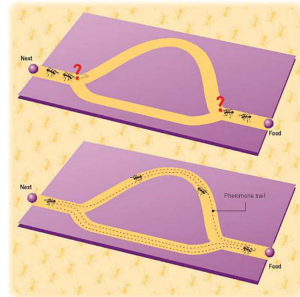
Gilbert, N., Troitzsch, K.G.: Simulation for the Social Scientist. Open University Press (1999)

Multiagent-based Simulation approach bottom-up modeling approach

- **Microscopic level: simulation of the behavior of the components of the real system.**
 - The components: the agents
 - Their relation: interaction and organization at a micro level.
- **Macroscopic level: Observation, Analyze of properties of the multiagent system.**

■ Example: Ants

- Micro level: ants are agents which put pheromones in the environment
- Macro level: the shortest path



Multiagent-based Simulation approach overview

- **A Multiagent-based Simulation (MABS) is a microscopic simulation model**
 - *A Multi-agent system*: the multiagent model of an actual or theoretical physical system
 - *Simulation*: controls of the evolution of the model in time.
- **Advantages**
 - MABS supports
 - Multi-level modeling:
 - Different models of "individuals": from simple entities to more complex ones.
 - Different levels of representation: "individuals" and "groups" within an unified conceptual framework.
 - The simulation of complex systems:
 - Structure preserving modeling of the simulated reality,
 - simulation of proactive behaviors,
 - Parallel computations,
 - Dynamic simulation scenarios
- **Limits**
 - Computation costs

Multiagent-based Simulation approach Contributors

- **The Thematician (expert of the domain)**
 - *Role*: Defines the intention of the simulation process.
 - *Result*: the *domain model* which describes the multiagent model of the reality. The agents are informally associated to the components of the system and their relations are identified (interaction, organization).
- **The modeler**
 - *Role*: He translates the knowledge of the thematician.
 - *Result*: the *design model* where the agents are a refinement of the agents in the *domain model*. Their properties are expressed using concepts taken from multiagent domain (behavioral model, communications, ...)
- **The Computer Scientist**
 - *Role*: He designs the operational model and writes the computer program.
 - *Result*: the computational system where agents are computational agents.

Multiagent simulation platform

Multiagent, simulation platforms

Typology

- **The operational, simulated model can be executed on a**
 - Generic multiagent platforms
 - *Advantage*: the computer scientist knows his environment, i.e. the platform and the related multiagent model.
 - *Limit*: The platform must be adapted (or not) to support the simulation,
 - *Example*: JASON, JADE (Tapas, PlaSMA), MASH, MADKIT (Turtlekit)
 - Generic simulation platforms
 - *Advantage*: the computer scientist can use the same environment for different design models.
 - *Limit*: a new operational model has to be built for each new simulations.
 - *Example*: MASON, SWARM, GAMA, CORMAS, TURTLEKIT, REPASt, NETLOGO, ...
 - specialized simulation platforms
 - *Advantage*: some parts of the operational model can be already available.
 - *Limit*: adaptation to a new platform.
 - *Example* (traffic simulation platform): Archisim ,MATSim, MITSIMlab, ...

Scheduler

Temporal models

- **Discrete time model**
 - Time advances in discrete step, which are integer multiples of some basic period such as 1 second, 1 day or ...
 - If the state at time t is q and the input time t is x , then the state at time $t+1$ will be $\delta(q,x)$ and the output y at time t will be $\lambda(q,x)$
 - δ is called the state transition function
 - λ is called the output function
- **Discrete Time Simulation**

```

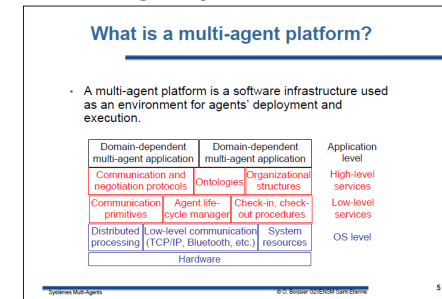
Ti = ti, Tf = tf
x(0) = v0, ..., x(9) = v9
q(0) = q0
t = Ti
while (t <= Tf) {
    y(t) = λ(q(t),x(t))
    q(t+1) = δ(q(t),x(t))
    t = t+1
}

```

Multiagent, simulation platforms

Components

■ Components of a multiagent platform



■ A scheduler

- A temporal model: discrete, continuous, event
- A scheduling policy
 - Synchronization of agent evolution
 - Simulation of the simultaneity

Scheduler

Temporal models

- **Discrete event models**
 - appropriate for those systems for which changes in system state occur only at discrete points in time.
 - A discrete points in time is called an event.
- **Discrete Event Simulation**
 1. Initialize the state variables
 2. Initialize the 'collection of pending events'
 3. Initialize the simulation clock
 4. while (there are pending events to be handled){
 - Remove the pending event (E) with the smallest timestamp (t)
 - Set simulation clock to that time t
 - Execute the event handler for event E

Scheduler

Multiagent scheduler

scheduling algorithm based on a continuous temporal model

```

T duration of the simulation
time = System.time();
T = time + T
Agents= {agents of the simulation};
For (a: agent)
    activate(a)
while (time < T)
    time = System.time();
    
```

scheduling algorithm based on a discrete temporal model

```

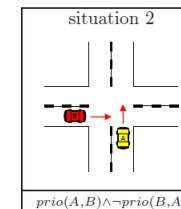
T duration of the simulation
time = 0;
Agents= {agents of the simulation};
while (time < T){
    For (a: Agents){
        \activate(a)
        a.ContextComputation()
        a.DecisionProcess()
        a.actionProcessing()
    }
    time++
}
    
```

Scheduler

Simultaneity problem

■ Discrete simulation

- Let t be the simulation time value and $a_i(t, q(t))$ the action of the i^{th} agent following the current state of the simulated system $q(t)$
- How to ensure that $q(t)$ will be the same for a_i and a_{i+1} since $a_i(t, q(t))$ modifies the current state



Dealing with Multi-Agent Coordination by Anticipation: Application to the Traffic Simulation at Junctions.
A Doniec, S Espié, R Mandiau, S Piechowiak - EUMAS, 2005

Simultaneity problem

Solution

■ No solution

- The most current solution,
- The action of an agent should not change the world in an important way. The micro coordination problems resulting of the scheduling process are not taken into account.
- The consequences of this choice have to be taken into account

■ The scheduling policy

- The activation order of the agents is randomized
- If the number of agents and simulation steps are important then no agent should be advantaged.
- If the simulation must be replayed, the random process has to be taken into account by the simulation model.

■ A dedicated mechanism

- The agents are activated in the same simulation state and the antagonism between their action is resolved by a decision process.
- Influence / reaction model: The agents do not directly act in the simulation but emit influences that are validated by the decision process.

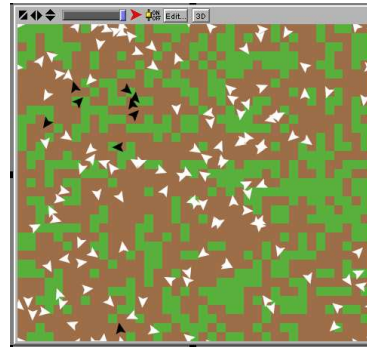
Ferber, J., & Müller, J. P. (1996, December). Influences and reaction: a model of situated multiagent systems. In *Proceedings of Second International Conference on Multi-Agent Systems (ICMAS-96)* (pp. 72-79).

Turtlekit

Illustrative Example

■ Prey and Predators

- A multiagent model
 - Environment: a grid
 - Prey: reactive agents who avoid the predators
 - Predator: communicative agents who coordinate to catch the preys
 - When three predators are around a prey, this last one die
- Simulation
 - A scheduling process
 - Temporal model
 - Activation process



Turtlekit

overview

■ Supports the simulation of heterogeneous multiagent model

- The superclass **AbstractAgent** contains the methods for the
 - Management of the life cycle:
 - activate(); end(); launchAgent(...); killAgent(...)
 - Communication management:
 - broadcastMessage(...); sendMessage(...); nextMessage(); isMessageBoxEmpty(); receiveMessage(Message m);
 - Organization management
 - createGroup(); leaveGroup(); requestRole(); getRoles(); isGroup(...); getAgentsWithRole()

■ Example

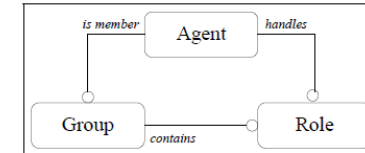
```
public void setup(){
    playRole("predator");
    ACLMessage m = new ACLMessage("INFORM","I'm a new predator");
    broadcastMessage("Turtlekit","HUNT","predator",m);
}
```

Turtlekit

overview

■ Plugin of the Madkit platform dedicated to the simulation

- Supports the Madkit organizational model: the AGR model



- Interaction are regulated by the organizational model
 - The communications are regulated following the organizational model
 - The perception can be implemented following the organizational model

Turtlekit

overview

■ Class Agent

- Inherits of the superclass AbstractAgent
- Implements the Runnable Interface
 - Methods “to control” the its thread
 - exitImmediatelyOnKill(); live(); pause(int t); run().
 - Additional methods for communications
 - waitNextMessage(); waitNextMessage(long timeout)

■ Example

```
public void live() {
    while (true) {
        Message m = waitNextMessage();
        if (m instanceof ACLMessage)
            handleMessage((ACLMessage)m);
    }
}
```

Turtlekit

overview

■ Class Turtle

- Inherits of the superclass AbstractAgent
- Do not implements the Runnable interface
- Additional methods
 - Related to the simulation process
 - setup, activate, end
 - A turtle is a situated agent, he has methods to
 - To be located in the environment
 - setX, xcor, dx, distance, towards, getHeading,
 - To move in the environment
 - moveto, fd, home, turnLeft, turnRight
 - To perceive the environment
 - countTurtlesAt, countTurtleHere, turtlesAt, turtlesHere

■ Example

```
public void setup(){
    playRole("predator");
    randomHeading();
    setColor(Color.red);
    if (countTurtlesHere(>0))
        fd(1);
}
```

```
private boolean caught(){
    int cpt=0;
    for(int i=1;i<=1;i++){
        for(int j=-1;j<=1;j++){
            if (! (i==0 && j==0) ){
                Turtle[] tur = turtlesAt(i,j);
                if (tur!= null && tur.length>0 &&
                    tur[0].isPlayingRole("predator"))
                    cpt++;
            }
        }
    }
    if (cpt>3) return true;
    return false;
}
```

Turtlekit

Simulation

■ Pseudo activation algorithm

time = 0; Turtle = {turtles of the simulation}; T duration of the simulation
While (time < T)

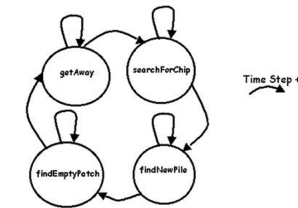
For (t: Turtle)

currentAction = scheduler.getCurrentActionTurtle(t)

nextAction = activate(t,currentAction)

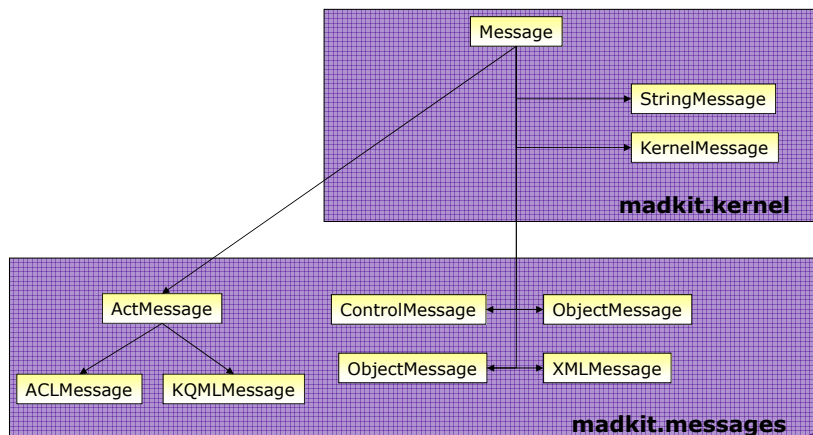
scheduler.setCurrentActionTurtle(t,nextAction)

time++



Madkit

Message



Madkit

Message

■ Message

- getCreationDate; getReceiver; getSender

■ ActMessage

- getAction; getContent; getFieldValue;
getInReplyTo; getObject; setContent; setField;
setInReplyTo; setObject

■ ACLMessage

- getAct, getPerformative; setPerformative;
getReceivers;removeReceiver;clearAllReceiver.