Adaptive Automata Community Detection and Clustering - A generic methodology -

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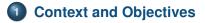
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- 2 Self-Organization Modeling
- Spatial and Behavioral Modeling Based on Community Detection
- Application: Geographical Information Systems and Agent-Based Modeling mixing
- 5 Conclusion and Perspective





Conclusion

Context and Objectives

- Complexity theory cover wide area of Systems in Science, making relevant links between social, biological and physical systems;
- In these complex systems, spatial structures emerge from interacting entities crossed by energetic fluxes;
- These emergent spatial structure are self-organizations, controled by some global objectives;
- The communities computed in the following respect these characteristics.





Conclusion

Complexity Concept Approach

Complexity Analysis is based on conceptual functions:

- Complexity is based on multi-description
 - Multi-scale, Multi-actor and Multi-disciplinary
 - Micro/Macro interaction in multi-scale description
- Emergent self-organizations and associated morphologies
 - Dynamic of the organizations: their evolution and their adaptation
 - Hierarchical structure of organization
 - Organization feed-back on the entities (country laws feed-back on the cities management)





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Conclusion

Self-Organization Modeling

Emergent Computing Classification

- Cellular Automata (CA) can model urban land-use or regional dynamics.
 - From Schelling's model, we can study by CA simulation, the segregation-like phenomena

 Agent-based modelling extends the basic diffusion rules of CA to more sophisticated behavioral processes.
 We developped in the following our specific agent behavior modeling using automata with multiplicities.





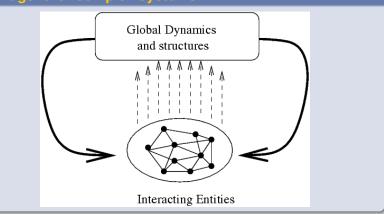
Community Detect.

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Conclusion

Agent-Based Modeling

Multi-agent for complex systems





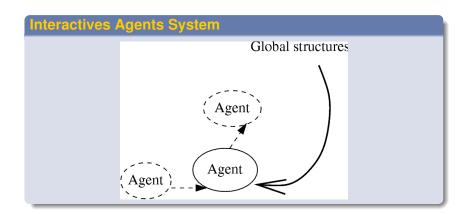
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Context

Agent-Based Modeling







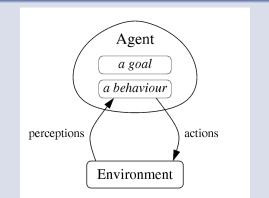
Context

GIS & ABM C

Conclusion

Agent-Based Modeling

Agents modeling





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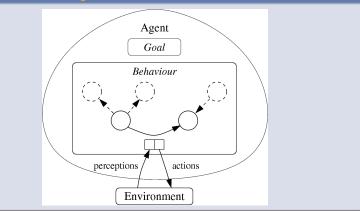
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Conclusion

Agent-Based Modeling

Automata-based Agent Behavior





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Automata-based Agent Behavior Modeling

Automata with multiplicities

Agent behavior is modelled by automata with multiplicities which is defined by

- A set of perception represented by an alphabet
- A set of actions represented by a semi-ring K
- A set of states with a subset of initial states and a subset of final states
- A set of transitions between states which is generate by a perception in input and which generate an action in output





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Automata-based Agent Behavior Modeling

Automata with multiplicities

Because the set of actions *K* is a **semi-ring**,

- we can represented the automata using a linear representation (vectors and matrices),
- we can defined many kinds of operators on these automata and so improve automatic processes on agent management





- Strategy modeling using *probabilistic* automata for game theory
- Automata based model for player behavior with adversory
 2 behavioral states: Cooperate (s1: C) or Defect (s2:D)





Example

- Strategy modeling using *probabilistic* automata for game theory
- Automata based model for player behavior with adversory
 - 2 behavioral states: Cooperate (s1: C) or Defect (s2:D)
 - Probabilistic transition from one state to another according to what make the adversory at the previous step

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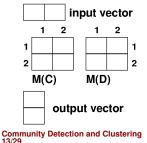




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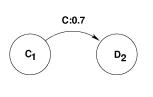




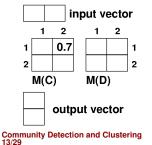
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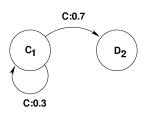
LINEAR REPRESENTATION



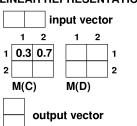
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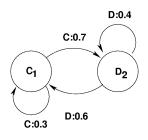
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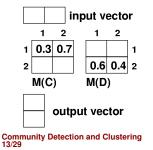
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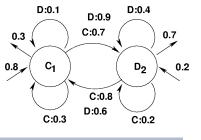




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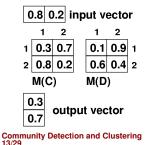


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LINEAR REPRESENTATION





- Context and Objectives
- 2 Self-Organization Modeling

Spatial and Behavioral Modeling Based on Community Detection

- Application: Geographical Information Systems and Agent-Based Modeling mixing
- 6 Conclusion and Perspective



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Context

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Spatial and Behavioral Modeling Based on Community Detection

Spatial agent and associated distances

- A spatial agent is defined by
 - Spatial coordinates
 - A behavior modeled by an automaton with multiplicities
- A **spatial distance** between 2 agents, can be computed according to their spatial coordinates
- A **behavioral distance** between 2 agents can be compute by the distance between the vectors which stores all the coefficients of the linear representation of the agent behavior automata.

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Spatial and Behavioral Modeling Based on Community Detection

Community Definition

• A **community** is a system or an organization which is characterized by a *spatial* property, a *behavior* property and the interaction between the both.

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Conclusion

Spatial and Behavioral Modeling Based on Community Detection

Genetic operators on automata

 Using the matrix representation of the automata with multiplicities, we can defined the classical genetic operators: duplication, crossing-over and mutation, using a chromosome composed of each lines of the matrices of the linear representation.





Example ... following

Genetic on strategy modeling for player behavior

- Genetic operators deal with **population** of **individuals** (here player behavior modeled by automata).
- **Individual** is described by a **chromosome** which is a sequence of **alleles** (elementary information).
- Here, the chromosomes are coding the transition matrices of the automata linear representation.

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• Here, an allele is a matrix line ..





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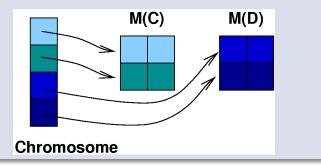


Conclusion

Example ... following

Genetic on strategy modeling for player behavior

 ... and the chromosome is the set of the matrix lines of all the transition matrices



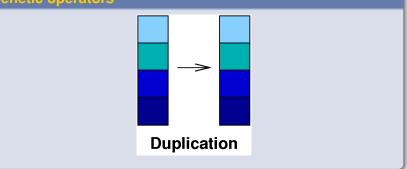


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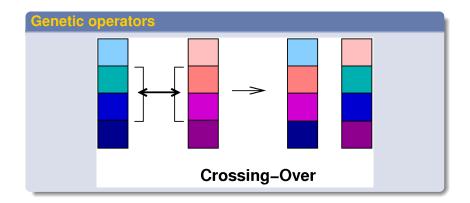














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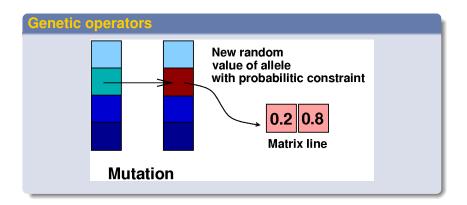


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Conclusion

Example ... following





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Context

Conclusion

Spatial and Behavioral Modeling Based on Community Detection

Community Detection associated to fitness function

- We can defined the fitness of each agent as following:
 - We compute his neighbourhood, using the *spatial distance*
 - We sum the **behavioral distance** of the agent itself with all the others agents included in the neighbourhood
 - We define the fitness as the inverse of the average of the previous sum.
- Self-organization communities emerge from the use of this fitness inside a genetic algorithm.





Spatial and Behavioral Modeling Based on Community Detection

Community Detection associated to fitness function

Let V_x a neighbourhood of the agent *x*, relatively to its spatial location. We define f(x) the agent fitness of the agent *x* as :

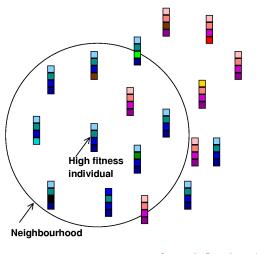
$$f(x) = \begin{cases} \frac{card(\mathcal{V}_x)}{\sum\limits_{y_i \in \mathcal{V}_x} d(x, y_i)^2} & \text{if } \sum\limits_{y_i \in \mathcal{V}_x} d(x, y_i)^2 \neq 0\\ \infty & \text{otherwise} \end{cases}$$

where d(x, y) is the behavioral semi-distance between the two agents x and y.

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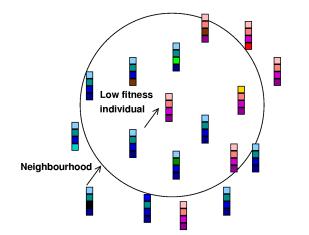




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Conclusion

Application :GIS and Agent-Based Modeling Mixing

Geographical Information Systems - GIS

Nowadays, the geographical information is a very wide knowledge database

- GIS allow to store, manage and compute all this information
- Wide-world communication improve the interaction networks dealing with Geo-Politic



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Conclusion

Application :GIS and Agent-Based Modeling Mixing

Geographical Information Systems - GIS

 \rightarrow The future improvement of GIS with automatic self-organization processes (like the communities detection proposed) can be one of the major aspect of the increasing of the world complexity to be controled as a whole, with the tools from the complexity concepts





Conclusion

Application :GIS and Agent-Based Modeling Mixing

Agent-based mixing

- Our goal is to include the community detection as a agent-based self-organization processus inside GIS
- We use Repast and its extension proposed by ESRI: Agent Analyst







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Conclusion

Conclusion and Perspectives

The future of the methodology and its applications

- The methodology proposed here, is based on community clustering (spatial and behavioral control) emerging from complex evolutive agents systems described by automata
- Concrete applications can be developped using GIS mixed with adaptive/genetic agent modeling
- Practical applications can be various and are under development now, specificaly on urban dynamics (economical, environmental, social or cultural development)

