Agent Based Modelling and Simulation in Systems Biology: Mobile Aspects

Emanuela Merelli Università di Camerino

Reykjavik University - 18 September 2007

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Plan

- Computational Systems Biology
- Strength and weakness of the formal approach
- Multi-level modeling in systems biology
- Agent-based modeling and simulation
- Case Study: from carbohydrate oxidation process
- UNICAM CoSy Research Group
- LITBIO project
- Erasmus Mundus Master Course in Formal Methods for Systems Engineering (FOMSE)
- Conclusion

Strength and weakness of the formal approaches Multi-level modelling for systems biology Case Study: Oxidation of carbohydrate The challenge of the 21st century Insights the Cell during the Carbohydrate Oxidation

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The challenge of the 21st century

The challenge of the 21st century will be to understand **how** biological molecules in living systems **integrate** to complex systems, how these systems **function** and their **evolution**.

We are scaling up from computational molecular biology (i.e. genomes, transcriptomes, proteomes, metabolomes) to **computational systems biology**

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Systems Biology

The science of **building models** of biological systems, from cells to organisms, by:

- systematically perturbing the system (biologically, genetically, or chemically); and

- monitoring gene, protein and other pathways.

Models are used in **drug research** for understanding their effects and side-effects and for building personalized medicine

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. Systems Biology is typically require tools to

- integrate the huge amount of biological data in order to
- understand the behaviour of biological systems, and to
- study the relationships and interactions between the various parts of a biological system, such as organelles, cells, physiological systems, organisms etc.

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From local interactions to global behaviour

Most of the biological systems are complex in nature

- have many of its components coupled in a non-linear fashion
- exhibit the emergence property from simpler interactions rules, i.e. the formation of complex patterns

▶ ...

Thus the proposal of a **multi-level approach** in modelling biological systems would allow zooming-in and zooming-out into through models

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From molecules to cell compartments, organisms, ... ecosystems



the global behaviour of the system can be determined by defining the lower-level interaction rules among components $\langle \Box \rangle \langle \Box \rangle$

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Glycolysis metabolic pathway

 $\textit{glucose} + 2\textit{NAD}^+ + 2\textit{ADP} + 2\textit{Pi} \rightarrow 2\textit{pyruvate} + 2\textit{NAD} + 2\textit{H}^+ + 2\textit{ATP} + 2\textit{H}_2\textit{O}$



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The Carbohydrate Oxidation is the energy production process performed by two active components of the Cell: Cytoplasm and Mitochondrion



Metabolic pathways

Cytoplasm (without oxygene)

- Glycolysis (convert 1 glucose into 2 pyruvate + ...)
- Lactic Fermentation (reduce pyruvate to lactate)
- Alcoholic Fermentation (reduce pyruvate to ethanol)

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- Mithocondrion (with oxygene)
 - Mithocondrial Inner Membrane
 - Transportation (pyrutave frpm cytop. to MM)
 - Respiration Chain ()
 - Mithocondrial Matrix
 - Partial oxidation of pyruvate
 - Kreb's cycle

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Cytoplasm: a lower-level bio-complexity

To model a portion $(10^{-15} \text{ liters})$ of a cytoplasm, we must simulate **the movement** and the **interactions** of about **27 millions of autonomous entities**.

Each interaction has associated 2 kinetic constants;

in many cases those values are still undiscovered; each day one of them could be discovered by one of the hundreds of research groups distributed all over the world



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Multi-level modelling

A model (M) for a system (S) and experiment (E) is anything to which E can be applied in order to answer questions about S M. Minsky, Models, minds, machines. 1965

This definition implies the co-existence of several models for any system.

A model can be distinguished w.r.t their organizational level:

- Is the system dscribed at macro or at meso ot at micro level?
- Are concentration changes the focus of interest, or the behaviour of individuals?

ODE Stochastic π -calculus A promising approach

It has been observed that the language used by biologists for verbal communications is **vague** and avoid clear prediction.

Also, the use of colourful diagrams, so much beloved from biologists, is often usually useless for computer reasoning and quantitative analysis.

Formal techniques are instead likely to give the chance to inherit the wide range of existing methods and tools provided by computer science (such as property design and verification, and automated reasoning).

ODE Stochastic π -calculus A promising approach

ODE limitations and Multi-Scaling power

The Ordinary Differential Equation (ODE) is the classical approach arising from the biochemical point of view.

A network of interactions (chemical reactions) between molecules (metabolites, proteins, genes) is established and ODE are used to numerically describe the continuous variations in the concentration of substances.

ODE beset with the combinatorial explosion

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ODE Stochastic π -calculus A promising approach

Process abstraction

Using the process abstraction opens up new possibilities for understanding molecular systems.

Computers and biomolecular systems both start from a small set of elementary components from which, layer by layer, more complex entities are constructed with evermore sophisticated functions.

The framework should also allow the system to be **zoomed-in** and **zoomed-out** at different levels of abstraction.

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ODE Stochastic π -calculus A promising approach

Stochastic π -calculus

In recent years, there has been increasing interest in computational models of biological systems based on various calculi of communicating processes such as the **stochastic** π -**calculus**.

Regev, A., Silverman, W., Shapiro, E.: Representation and simulation of biochemical processes using the pi-calculus process algebra. In: Pacific Symposium of Biocomputing 6 (PSB 2001)

Regev, A.: Representation and simulation of molecular pathways in the stochastic pi-calculus. In: 2nd workshop on Computation of Biochemical Pathways and Genetic Networks, Heidelberg, Germany (2001)

In 2003, Aviv Regev and Edu Shapiro proposed "the $\pi\text{-}\mathbf{calculus}$ as an abstraction for biomolecular modelling"

ODE Stochastic π -calculus A promising approach

Albeit π -calculus allows to reflect the inherent concurrency, communication and stochasticity of cellular systems, **each new case reveals drawbacks** in modelling biological systems

The minimality of the calculus makes it heavy to manage models with **complex coordination** and dependencies. It forces the modeler to encode high-level interaction patterns into low-level means offered by the calculus.

The encodings makes models **harder to write and understand**, and obscure the original high-level biological model with a plethora of irrelevant computational details.

ODE Stochastic π -calculus A promising approach

A criticism to π -calculus

Perhaps it depends on \ldots the asymmetry of the send/receive communication model

Whenever two molecules A and B bind to form a complex AB, they each create a private channel. The protocol for exchanging these new private channels is itself complicated by the directionality of communication

[D. Duchier, C. Klutter *Biomolecular agents as multi-behavioural concurrent objects* ENTCS, 150, 2006]

ODE Stochastic π -calculus A promising approach

A criticism to π -calculus

Perhaps it depends on \ldots the asymmetry of the send/receive communication model

not very suitable to model the simultaneous exchange!

Whenever two molecules A and B bind to form a complex AB, they each create a private channel. The protocol for exchanging these new private channels is itself complicated by the directionality of communication

[D. Duchier, C. Klutter *Biomolecular agents as multi-behavioural concurrent objects* ENTCS, 150, 2006]

ODE Stochastic π -calculus A promising approach

Coordination, in CS, refers to languages and models for coordinating, configuring, and, generally, glueing together components. e.g., LINDA

Computational Systems Biology Strength and weakness of the formal approaches

Multi-level modelling for systems biology Case Study: Oxidation of carbohydrate ODE Stochastic π -calculus A promising approach



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ODE Stochastic π -calculus A promising approach

Shall **agent-based paradigm** help to overcome the limitation of the understandability of the π -calculus by offering a **multi-level modelling environment** and ...

... exploit the Grid computational resources the simulation of biological systems?

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What is an agent? Autonomous agent Situated agent Flexible agent Reactive Agent Proactive Agent and Mobile Agent

Milner's agent point of view

In 1989 Milner wrote ...

- "each of the several part of the system has its own identity, which persist through time; we shall call these parts agents"
- … Communication and concurrency notions, both essential in understanding complex dynamic systems. On the one hand such a system has diversity, being composed of several parts each acting concurrently with, and independently of, other parts; on the other hand a complex system has unity (or else we shall not call it a system) achieved through communication among its parts.
- ... underling both these notions is the assumption that each of the several parts of such a system has its own identity, which persists through time; we shall call these parts agents
- … This lead to use the term agent very broadly, to mean any system whose behaviour consists of discrete actions; an which for one purpose we take to be atomic may, for other purpose, be decomposed into sub-agents acting concurrently and interacting.

What is an agent? Autonomous agent Situated agent Flexible agent Reactive Agent Proactive Agent and Mobile Agent

An **agent** is an "encapsulated" computer system **situated** in some environment and capable of **flexible**, **autonomous** action in that environment in order to meet its design objectives

> [M. Wooldridge, Agent-based software engineering. In IEE Proceedings of Software Engineering, 1997]

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What is an agent? Autonomous agent Situated agent Flexible agent Reactive Agent Proactive Agent and Mobile Agent

An autonomous agent is a computer system capable of flexible, **autonomous** (problem-solving) action, situated in dynamic, open, unpredictable and typically multi-agent domains.

The agent has the control over its internal state and over its own behaviour

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What is an agent? Autonomous agent Situated agent Flexible agent Reactive Agent Proactive Agent and Mobile Agent

A situated agent is a computer system capable of flexible, autonomous (problem-solving) action, **situated** in dynamic, open, unpredictable and typically multi-agent domains.

The agent perceives the environment through sensors and acts through effectors

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What is an agent? Autonomous agent Situated agent Flexible agent Reactive Agent Proactive Agent and Mobile Agent

A computer system capable of **flexible**, autonomous (problem-solving) action, situated in dynamic, eventually open, unpredictable and typically multi-agent domains.

reactive: the agent responds in timely fashion to environmental change

proactive: the agent acts in anticipation of future goals

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What is an agent? Autonomous agent Situated agent Flexible agent Reactive Agent Proactive Agent and Mobile Agent

Reactive agents can be defined by a 6-tuple < E, P, A, see, do, action >

where

- *E* is the set of states for the environment
- P is a partition of E (representing a relevant abstraction of the environment from the agent's point of view)
- A is a set of actions
- see : $E \rightarrow P$
- action : $P \rightarrow A$
- $do: A \times E \to E$

These agents observe the environment (see), find the appropriate action (action), and act (do)

What is an agent? Autonomous agent Situated agent Flexible agent Reactive Agent **Proactive Agent** and Mobile Agent

 $\begin{array}{l} \mbox{Proactive agents can be defined by a 9-tuple} \\ < D, E, P, A, d_0, see, KB, do, action > \end{array}$

where

- D is a set of predicate calculus databases
- ► *E*, *P*, *A*, see and *do* are the same as for reactive agents
- ▶ d₀ is the initial KB
- action : $D \times P \rightarrow A$
- $\blacktriangleright KB: D \times P \to D$

These agents observe the environment (*see*), choose their next action according to some kind of reasoning on their current knowledge (*action*), act (*do*), and update their knowledge base

What is an agent? Autonomous agent Situated agent Flexible agent Reactive Agent Proactive Agent and Mobile Agent

Mobile agents are agents that move across distributed environments (execution platforms)

Mobile agents for Global computing and **mobile code** for Grid computing are new paradigms for distributed computing that complement such technologies as distributed objects, remote compute servers, and distributed programming systems such as PVM and MPI

What is an agent? Autonomous agent Situated agent Flexible agent Reactive Agent Proactive Agent and Mobile Agent

Multi-agent system

The description of a system consists of a collection of **interactive agents** and a set of coordination rules.

An agent has a name and a set of behaviours. He can adopts **different behaviours** at differents times depending on the environmental conditions.

At **meso level** of modelling biological systems, the elementary interactions consist of the binding or unbinding of two agents, the modification of the state of a interface (site), and the deletion or creation of an agent.

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Carbohydrate oxidation: the PASSI agent identification UML diagram



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Roles in Lactic Fermentation function



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UML Class Enzymes Enzima +initialize(+reagisci(Enzima Enzima Fermentazione Glicolisi Alcolica TSIdentificator : Identificator



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UML Sequence diagram of the preliminary phase of glycolysis



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An agent-based "calculator" for glycolysis

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But ... Zooming-in

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But ... molecules move



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... and then interact by affinity



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... and then interact by affinity



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 reaction 1: GLC + ATP -> F6P + ADP:
 reaction_2: F6P + ATP -> FBP + ADP;
 reaction_3: FBP => 2 * GAP;
 reaction 4: GAP + NAD -> DPG + NADH:
 reaction_5: DPG + ADP => PEP + ATP;
 reaction 6: PEP + ADP -> PYR + ATP:

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reaction_7: PYR -> ACA;

reaction_8: ACA + NADH => EtOH + NAD;

reaction 9: AMP + ATP => 2 * ADP:

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            Case Study: Oxidation of carbohydrate
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Glycolysis

$glucose + 2ATP + 4ADP + 2Pi + 2NADox \rightarrow$ $2pyruvate + 2ADP + 4ATP + 2NADrid + 2H_2O$

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Step Reactions Substrates Products	Enzyme Catalyser	$\Delta G^{o'}$ in Kcal mol ⁻¹
1 glucose, ATP \longrightarrow glucose-6-phosphate, ADP, H ⁺	hexokinase	-4.0
2 glucose-6-phosphate	phos.glu.isomerase	+0.4
3 fructose-6-phosphate, ATP \longrightarrow fructose1,6-bisphosphate, ADP, H ⁺	phosphofructokinase	-3.4
4 fructose1,6-bisphosphate glyceraldehyde 3-phosphate	aldolase	+3.7
5 dihydroxyacetone phosphate glyceraldehyde 3-phosphate	trio . phos . isomerase	+1.8
6 glyceraldehyde 3-phosphate, 1,3-bisphosphoglycerate, Pi, NAD ⁺ NADH, H ⁺	glyceraldehyde 3phos.dehydro.se	+1.5
7 1,3-bisphosphoglycerate, ADP - 3-phosphoglycerate, ATP	phosphoglyceratekinase	-4.5
8 3-phosphoglycerate - 2-phosphoglycerate	phosphoglyceromutase	+1.1
9 2-phosphoglycerate \longrightarrow phosphoenolpyruvate, H ₂ O	enolase	+0.4
10 phosphoenolpyruvate, ADP, $H^+\ \rightleftarrows\ enolpyruvate,\ ATP$	pyruvatekinase	-7.5
11 enolpyruvate \longrightarrow pyruvate		

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LITBIO Project



OGSA

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Erasmus Mundus Master Course in *Formal Methods for Systems Engineering* (FOMSE)

The master curricula are specified by three tracks. Every student chooses a track, which defines the mobility plan (between two sites, at the end of the first year) of the student. The choice is driven by the specialization/thesis/research interest specified in the name of the track. Every track has a minimum number of ECTS (30) covering the following Basic Topics:

- Process Algebra
- Distributed and Concurrent Computation
- Theory of Computation

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Erasmus Mundus Partnership



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Master Tracks

- Systems Biology: Year 1 Reykjavik (Basic Topics + Courses in Bioinformatics and Systems Biology + free choiches). Year 2 Camerino (Specialization + Research Credits + Thesis)
- Software Systems Engineering: Year 1 Camerino (Basic Topics + Service Oriented Architectures + free choices). Year 2 Eindhoven (Specialization + Research Credits + Thesis)
- Intelligent Systems Year 1 Eindhoven (Basic Topics + Algorithms for Intelligent Systems + free choices). Year 2 Reykjavik (Specialization + Research Credits + Thesis)

Researchers

Diletta Cacciagrano Nicola Cannata Rosario Culmone Maria Rita Di Berardini Emanuela Merelli Luca Tesei

Phd Students

Ezio Bartocci Francesco De Angelis Francesca Piersigilli Barbara Re Oliviero Riganelli Leonardo Vito *Roberta Alfieri*

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