

MoK: Stigmergy Meets Chemistry to Exploit Social Actions for Coordination Purposes

Stefano Mariani, Andrea Omicini
{s.mariani, andrea.omicini}@unibo.it

Dipartimento di Informatica: Scienza e Ingegneria (DISI)
ALMA MATER STUDIORUM—Università di Bologna

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- 1 Introduction
- 2 Stigmergy in Natural and Artificial Systems
- 3 The Molecules of Knowledge Model
 - MoK Overview
 - MoK Self-Organisation
 - The MoK Model as a BIC Model
- 4 Toward Self-Organising, Social Workspaces
- 5 Conclusion



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The Challenge

Socio-technical systems are becoming increasingly complex and thus difficult to design, mostly due to the unpredictability of human interactions [Omicini, 2012].

KIE

Furthermore, such systems often represent Knowledge Intensive Environments (KIE) [Bhatt, 2001], that is, they are meant to manage a huge amount of (possibly heterogeneous) information [Ossowski and Omicini, 2002].



A Path To Follow I

Coordination models have been developed to cope with software systems' increasing complexity, whose main source is the interaction space such systems have to manage [Gelernter and Carriero, 1992, Papadopoulos and Arbab, 1998, Omicini and Viroli, 2011].

Non-determinism

Such interactions have been for long recognised as an undesirable source of (uncontrollable) non-determinism, harnessing correctness, reliability, and predictability of systems.



A Path To Follow II

Then, **probability** entered the picture as a means to model, govern, and predict non-determinism, making it become a source of solutions rather than problems.

Nature-inspired Systems

Natural systems from chemistry, biology, physics, sociology, and the like are widely recognised for their capability to “reach order out of chaos” through **adaptiveness** and **self-organisation**, hence they have been proficiently used as metaphors for the engineering of coordination models & systems [Omicini, 2013].



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(Cognitive) Stigmergy & BIC I

In complex social systems, such as human organisations, animal societies, and artificial multi-agent systems as well, interactions between individuals are *mediated by the environment*, which “records” all the *traces* left by their actions [Weyns et al., 2007].

Stigmergy

Trace-based communication is related to the notion of *stigmergy*, firstly introduced in the biological study of social insects [Grassé, 1959].



(Cognitive) Stigmergy & BIC II

Furthermore:

- modifications to the environment are often amenable of a *symbolic interpretation*
- interacting agents feature *cognitive abilities* that can be proficiently exploited in stigmergy-based coordination

Cognitive Stigmergy

When traces becomes signs, stigmergy becomes *cognitive stigmergy* [Omicini, 2012].



(Cognitive) Stigmergy & BIC III

Another step beyond (cognitive) stigmergy is taken by Behavioral Implicit Communication (BIC), in which coordination among agents can be based on the observation and interpretation of *actions as wholes*, rather than solely of their effects on the environment.

BIC

In BIC, actions themselves become the “message”, often intentionally sent through the environment in order to obtain collaboration [Castelfranchi et al., 2010].



Computational requirements I

Moving from stigmergy to BIC, a list of *desiderata* emerge to be supported by the coordination abstraction:

- reification of agent's (inter)actions
- recording of their *contextual properties*
- recording of their “traces”
- capability of reaction to traces emission
- topology-related aspects management
- ontology management for symbolic interpretation



Computational requirements II

Among the many sorts of coordination models [Gelernter and Carriero, 1992], tuple-based ones [Gelernter, 1985] have been already taken as a reference for stigmergic coordination—including cognitive and BIC [Omicini, 2012].

Tuple-based Coordination

(Logic) Tuple-based infrastructures such as TuCSoN [Omicini and Zambonelli, 1999], feature space-time situatedness, probabilistic primitives and coordination programmability^a, providing us with all the necessary tools to fully support stigmergy, cognitive stigmergy, and BIC.

^aThanks to the ReSpecT language [Omicini, 2007].



(Bio)Chemistry & Biochemical Tuple Spaces

Biochemical Tuple Spaces

Biochemical tuple spaces are a stochastic extension of the LINDA framework [Gelernter, 1985] inspired both by chemistry and biology proposed in [Viroli and Casadei, 2009].

The idea is to attach to each tuple a “concentration”, which can be seen as a measure of the *pertinency/activity* of the tuple: the higher it is, the more likely and frequently the tuple will influence system coordination.

Concentration of tuples evolves at a certain rate due to stochastic chemical rules installed into the tuple space, which acts as a chemical simulator [Gillespie, 1977].



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MoK Overview

The MoK model was introduced in [Mariani and Omicini, 2013] as a framework to conceive, design, and describe knowledge-oriented, self-organising coordination systems.

Main Ideas

- information should autonomously link together
- users should see information spontaneously manifest to and diffuse toward them



Formal MoK I

Atoms

Produced by a source and conveying an atomic piece of information, atoms should also store ontological metadata to ease automatic processing:

$$\text{atom}(\text{src}, \text{val}, \text{attr})^c$$

Molecules

MoK heaps for information aggregation, molecules cluster together semantically related atoms:

$$\text{molecule}(\text{Atoms})^c$$


Formal MoK II

Enzymes

Enzymes represent the reification of (epistemic) knowledge-oriented (inter-)actions, and are meant to participate biochemical reactions to properly increase molecules' concentration^a:

$$\text{enzyme}(\text{Molecule})^c$$

^aThe term “molecule” will be used also for “atom” in the following.

MoK Function

As a knowledge-oriented model, MoK must have a way to determine the semantic correlation between information. Therefore, the MoK function should be defined, taking two molecules and returning a value $m \in [0, 1]$:

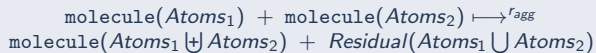
$$\mathcal{F}_{mok}: \text{Molecule} \times \text{Molecule} \mapsto [0, 1]$$

Formal MoK III

Biochemical Reactions

The behaviour of a MoK system is determined by biochemical reactions, which stochastically drive molecules aggregation, as well as reinforcement, decay, and diffusion:

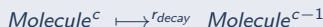
Aggregation — Bounds together semantically related molecules



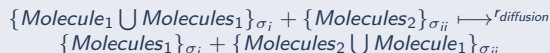
Reinforcement — Consumes an enzyme to reinforce the related molecule



Decay — Enforcing time situatedness, molecules should fade away as time passes



Diffusion — Analogously, space situatedness is inspired by biology and therefore based upon *diffusion*



Formal MoK IV

Other aspects like topology, information production and consumption are addressed by:

Compartments — the conceptual *loci* for all other MoK abstractions, providing the notions of **locality** and **neighbourhood**

Sources — each one associated to a compartment, MoK sources are the origins of atoms, which are continuously injected at a given rate

Catalysts — the abstraction for knowledge **prosumers**, who emit enzymes whenever they interact with their compartment



Outline

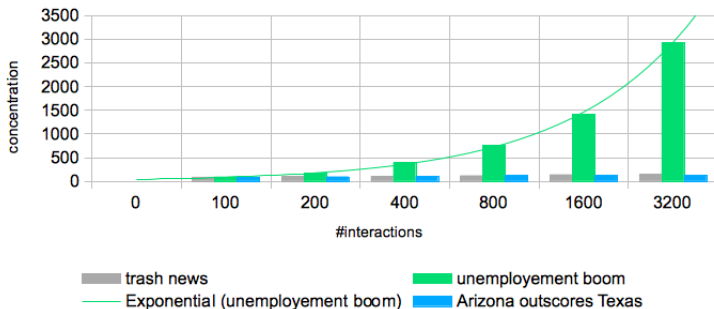
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“Smart migration” I

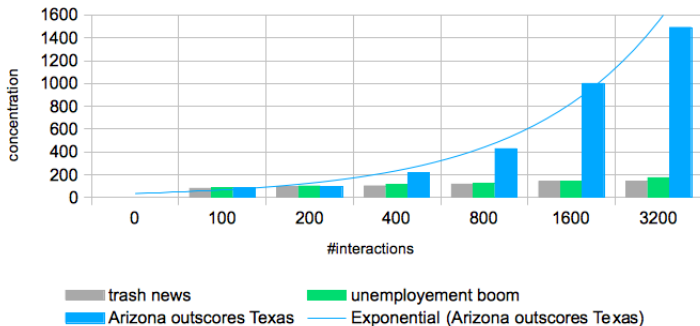
A “producer” compartment diffuses a collection of different atoms to the neighbouring compartments “economics” and “sports”. We expect the system to reach an “equilibrium” in which the two topic-oriented compartments are mainly populated by topic-compliant news molecules.

“Economics” Compartment



"Smart migration" II

"Sports" Compartment



The stochastic equilibrium between diffusion, reinforcement and decay laws, makes a "smart migration" pattern appear by emergence.



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MoK Limitations

MoK currently lacks three of the computational requirements needed to support BIC-based coordination:

- making traces of agents' interactions available for observation to other agents
- making agents' interactions themselves available for other agents inspection
- explicitly record contextual information about such actions

In fact:

- enzymes are **consumed** and **cannot diffuse**, hence their observation is restricted to the environment
- contextual information is **only implicitly** conveyed by enzymes, being them produced at a given time in users' own compartment



MoK Extension I

σ Descriptor

We can keep track of contextual information regarding agents interactions by associating each enzyme to a *descriptor* σ of the compartment they were released into

$$\text{enzyme}(\sigma, \text{Molecule})^c$$

Such a descriptor could store any meta-information about the compartment useful to better understand the action: its current time, place, and so on.



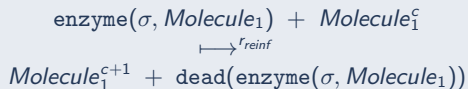
MoK Extension II

Then, we should:

- allow enzymes to participate in MoK diffusion reactions, so that both other agents and compartments can observe and use them
- produce a “dead enzyme” whenever enzymes are involved in a reinforcement reaction, subject to decay and diffusion

“Trace” Reinforcement

MoK reinforcement reaction should then be rewritten as follows:



Furthermore, decay and diffusion reactions should apply respectively to dead enzymes solely (decay) and both enzymes (diffusion).



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Current Prototype

A first prototype of the MoK model is actually implemented upon the TuCSoN coordination infrastructure [Omicini and Zambonelli, 1999] featuring ReSpecT *tuple centres* [Omicini, 2007].

TuCSoN Mapping

- logic tuples are MoK atoms, molecules, enzymes and biochemical reactions declarative specification
- ReSpecT reactions installed in tuple centres implement Gillespie's chemical simulation algorithm [Gillespie, 1977]
- ReSpecT tuple centres represent MoK compartments
- TuCSoN agents reify with MoK sources and users (catalysts)



Next Steps I

Such prototype implementation paves the way toward a much more complex and general idea of self-organising workspaces [Omicini, 2011] a full-fledged MOK system would support.

What Is Done

In this context, MOK already support knowledge aggregation and organisation:

- the former by the molecule abstraction, actually reifying semantic relationships among different information chunks
- the latter by the combined contribution of MOK diffusion, reinforcement, and decay, in which enzymes play a critical role



Next Steps II

Socio-technical system for knowledge-oriented coordination should properly handle pervasive computing scenarios where knowledge is pervasively distributed and is to be accessible ubiquitously.

What Still To Do

The *data in the cloud* paradigm could inspire a novel knowledge in the cloud paradigm, aiming at transforming data clouds into semantic clouds, spontaneously and continuously *self-(re)organising* as a consequence of semantic correlations between information chunks and social actions carried out by knowledge workers.



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Conclusion

- We discussed principles borrowed from cognitive and behavioural (social) sciences. . .
- . . . then showed how to exploit them in computational systems. . .
- . . . thanks to the biochemically-inspired MoK model, aimed to deal with the issue of coordination in knowledge-intensive environments

Future

By sharing some of our ideas regarding the future of socio-technical systems, we hope new collaborations between the research fields of sociology, cognitive sciences, and coordination models and languages will arise.



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