

Analysing Robot Swarm Decision-making with Bio-PEPA

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joint work with

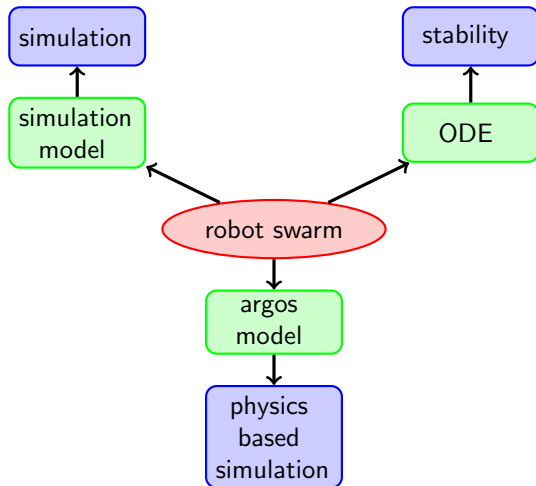
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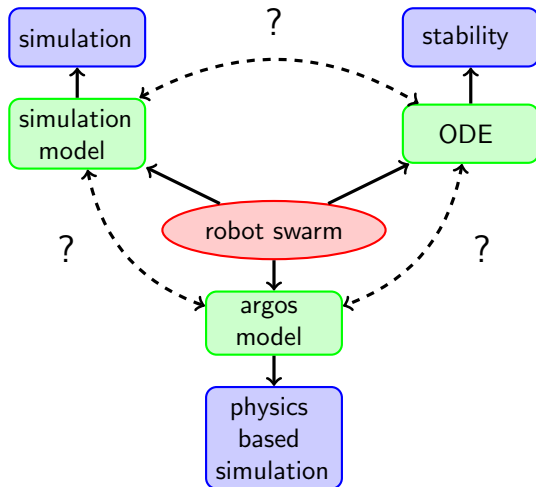
PRIN-CINA, February 4-6, Pisa

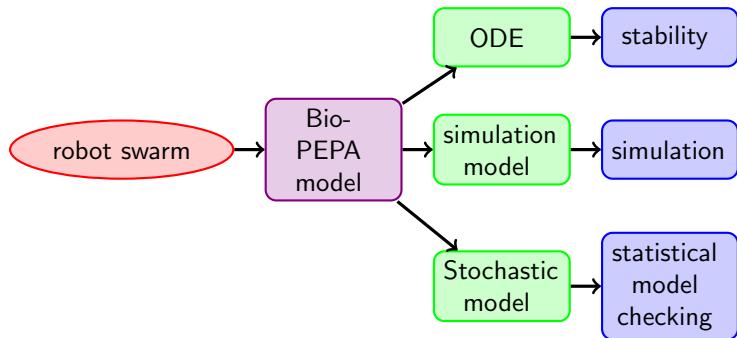
(ANTS 2012, Brussels, 12-14 September 2012)

Swarm Robotic Systems Analysis



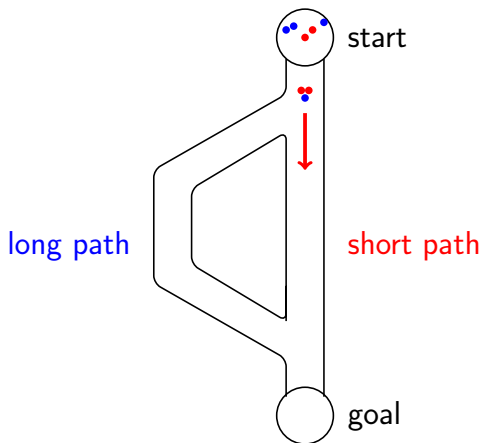
Swarm Robotic Systems Analysis



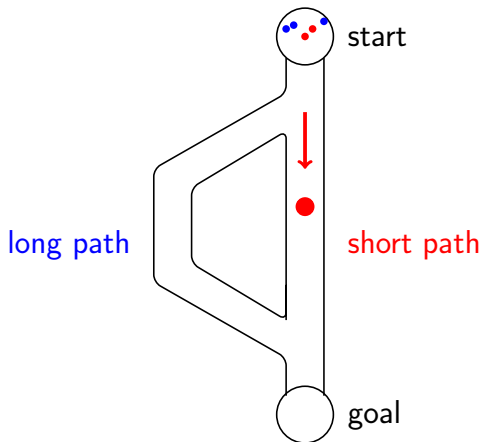


- Consistency: formal relation between models and analyses results
- One model using agent based perspective and composition
- Scalability: abstraction from agent identity
- Design phase analysis: select most promising solutions for further study

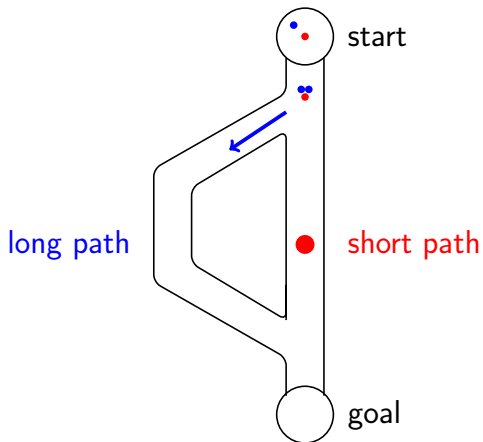
Variant of the well-known double bridge experiment:



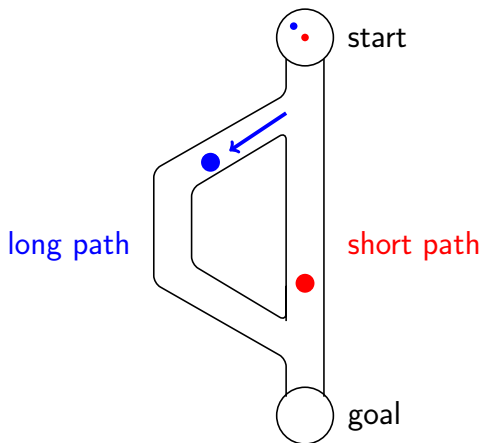
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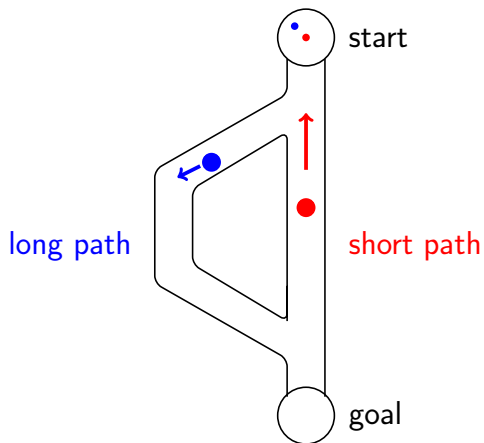
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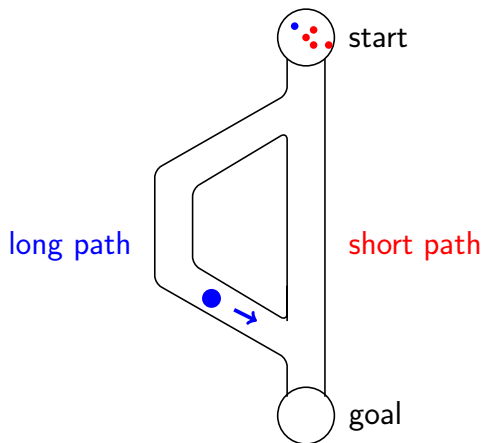
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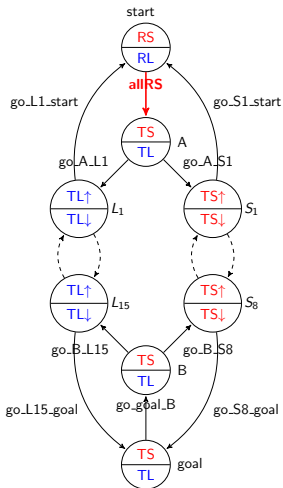
“Can **formal methods** contribute to model and analyse the **emergent** behaviour of **agent-based** Robot Swarms?”

- Can emergent behaviour in time and space be addressed?
- Is the method scalable?
- Can we deal with the formation of teams of robots?
- Can we handle action durations in a sufficiently accurate way?
- Can we deal with spatial aspects?

- What is the probability of convergence on each path?
- What are the average number of teams formed until convergence?
- How much time passes before convergence occurs?
- What is the average transient behaviour of the system?

Bio-PEPA model of Swarm Decision Making

[Bio-PEPA: Jane Hillston et al., 2009]



location top : size = 10000, type = compartment

location start in top : size = 1000, type = compartment

...

location L15 in top : size = 1000, type = compartment

location goal in top : size = 1000, type = compartment

$$\begin{aligned} \text{Robo_start_fromS} &= (\text{allRS}, 3) \downarrow \text{Robo_start_fromS@start} + \\ & (\text{RS2RL1}, 2) \downarrow \text{Robo_start_fromS@start} + \\ & (\text{RS1RL2}, 1) \downarrow \text{Robo_start_fromS@start} + \\ & (\text{go.S1_start}, 3) \uparrow \text{Robo_start_fromS@start}; \end{aligned}$$

$$\begin{aligned} \text{Teams_A_S} &= (\text{allRS}, 1) \uparrow \text{Teams_A_S@A} + \\ & (\text{RS2RL1}, 1) \uparrow \text{Teams_A_S@A} + \\ & \text{go.A.S1} \downarrow \text{Teams_A_S@A}; \end{aligned}$$

$$\begin{aligned} \text{Teams_A_L} &= (\text{allRL}, 1) \uparrow \text{Teams_A_L@A} + \\ & (\text{RS1RL2}, 1) \uparrow \text{Teams_A_L@A} + \\ & \text{go.A.L1} \downarrow \text{Teams_A_L@A}; \end{aligned}$$

$$\begin{aligned} \text{system} &::= (\text{Robo_start_fromS@start}(\text{SS}) < * > \text{Robo_start_fromL@start}(\text{SL}) \\ & < * > \dots < * > \text{Teams_B_fromL@B}(0)) \end{aligned}$$

RSS = Robo_start_fromS

RSL = Robo_start_fromL

kineticLawOf allRS : $\frac{(RS)}{(RS)+(RL)} * \frac{(RS-1)}{(RS-1)+(RL)} * \frac{(RS-2)}{(RS-2)+(RL)} * \text{move};$

pSSL = $\frac{(RS)}{(RS)+(RL)} * \frac{(RS-1)}{(RS-1)+(RL)} * \frac{(RL)}{(RS-2)+(RL)}$

kineticLawOf RS2RL1 : $(pSSL + pSLS + pLSS) * \text{move};$

kineticLawOf go.A.S1 : $\text{walk} * \text{Teams_A_S@A};$

SS = 16, SL = 16, move = 0.28, walk = 0.1, k = (32 minstart)/3

Validation of results

Three approaches:

- Stochastic simulation (Gillespie) via Bio-PEPA plugin tools
- Statistical model checking via Bio-PEPA tools + PRISM tool
- Fluid approximation (ODE) via Bio-PEPA tools

Download and info:

- <http://homepages.inf.ed.ac.uk/jeh/Bio-PEPA/biopepa.html>
- <http://www.prismmodelchecker.org>

Statistical Model Checking

e.g. [Peyronnet et al., 2004, Kwiatkowska et al., 2011]

Validation for values of k ranging from 1 to 10 (active teams) of the following quantitative properties [Montes de Oca et al., 2011]:

- Probability of convergence on short path
- Expected team formations until convergence
- Expected convergence time

There exist several variants of Statistical Model Checking.

We use the Confidence Interval (CI) method to approximate:

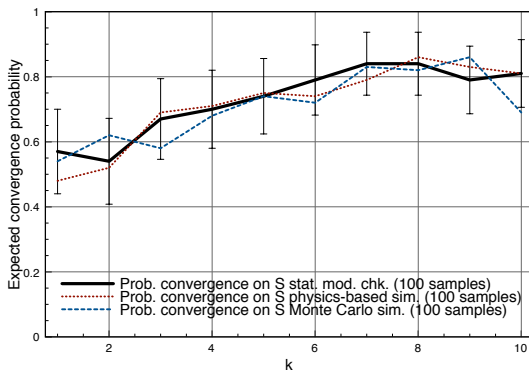
- Probability
- Rewards

SMC: Probability of convergence on short path

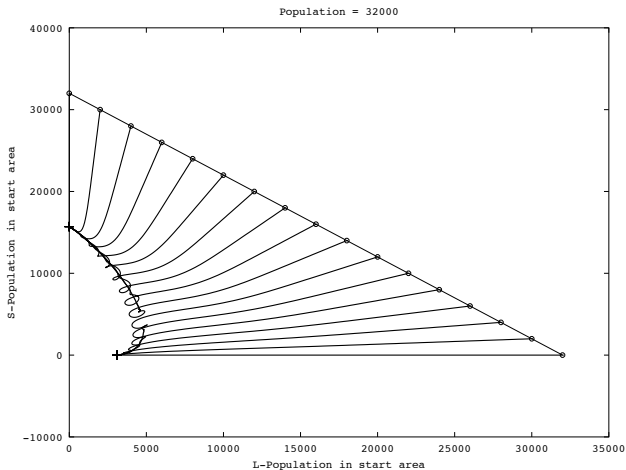
$$P = ? [! \text{“Convergence_on_L”} \cup \text{“Convergence_on_S”}]$$

What is the probability ($P = ?$) to see no (!) convergence on L until (U) convergence on S takes place?

“Convergence_on_S” = all the 32 robots are either in a team on the short path or in the S-population in the start area or at the goal area.



Fluid approximation (for $k=7$)



Phase-space diagram of S-population versus L-population in the start area for a population of 32,000 robots.

Conclusions:

- Validation shows promising correspondence to results in the literature
- Modelling of populations, teams, locations, distances, probabilities
- Various analysis methods have been applied on same model
- Efficient analysis times, also addressing emergent behaviour

Discussion:

- Revisiting the language: 'Swarm-PEPA'? (WP4, T4.2)
- Further analysis techniques to support design of emergent behaviour? (WP3, T3.2)
- Spatial aspects of SRS: beyond simple locations? (Self)-Adaptive systems?
- Modeling language for autonomic systems?
E.g. Service Component Ensemble Language (SCEL) language of EU ASCENS/QUANTICOL projects.