

Simulation and Control of Multi-Agent Systems: A Minimal Evacuation Time Problem

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Motivation and Goals

- *Realistic Simulations* and *Optimal Control* of multi-agent systems require the reiterative solution of a large system of ODEs, which drives usually to **unaffordable computational costs**.
- We *approximate the problem with a kinetic PDE* and simulate its constrained evolution by means of *Binary Interaction algorithms*, which are able to reduce drastically the computational cost.
- Applications find place in several areas: *engineer, biology, robotics, computer graphics* . . .
- In [1] we are concerned with the **multi-scale modeling, control and simulation** of a **crowd leaving an unknown area**.



Problem setting

- We model human crowd leaving an *unknown environment*, Ω , under limited visibility. Due to their lack of information about the positions of exits, agents need to explore the environment first.
- Our aim is to *control the emergent dynamic* through the action of *few informed agents*, not recognized by the rest of the crowd.
- We consider N followers and M leaders whose interactions are described by

$$\begin{aligned} \dot{x}_i &= v_i, \\ \dot{v}_i &= S(x_i, v_i) + \sum_{j=1}^N F(x_i, v_i, x_j, v_j) + \sum_{l=1}^M F(x_i, v_i, y_l, w_l), \\ \dot{y}_k &= w_k = \sum_{j=1}^N K(y_k, x_j) + \sum_{l=1}^M K(y_k, y_l) + u_k \end{aligned} \quad (1)$$

where S represents a **self-propulsion term** and F, K the interaction kernels accounting the **social forces** of the dynamic: *alignment, repulsion and attraction, low visibility*.

- u_k is the **control** chosen as a solution in the set of *admissible controls* U_{adm} of

$$\min_{u(\cdot) \in U_{adm}} \{t > 0 \mid x_i(t) \notin \Omega, \forall i = 1, \dots, N\}, \quad \text{subject to (1).}$$

Simulation of the microscopic model

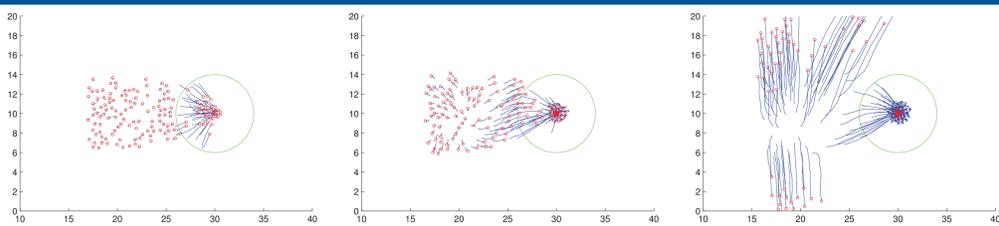


Figure : The exit is a point located at $E = (30, 10)$ which can be reached from any direction and it is visible inside the green circle. $N = 100$ Followers are initially randomly distributed in the domain $[17, 29] \times [6.5, 13.5]$ with velocity $(0, 0)$. **Without leaders** the total mass is not evacuated, farthest people split in several but cohesive groups with random direction and never reach the exit.

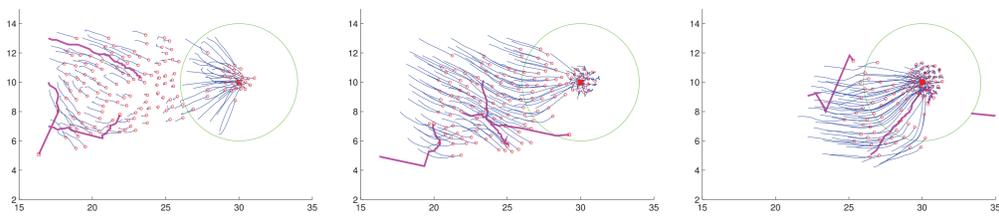


Figure : The optimal strategy u_k is obtained through a **modified compass search method**. **With $M = 3$ leaders** the optimal strategy u_k , prescribes that leaders *divert* some pedestrians from the right direction, so as not to steer the whole crowd to the exit at the same time. In this way **congestion is avoided** and pedestrian flow through the exit is increased.

Mesoscopic model

- When the **number of follower, N , is large**, a microscopic description of both populations is no more a viable option. Thus we consider the **evolution of the distribution of followers**, denoted by $f(x, v)$, together with the leaders, with empirical distribution $g(y, w)$.
- We assume that these densities satisfy the following coupled ODE-PDE system

$$\begin{aligned} \partial_t f + v \cdot \nabla_x f &= Q_F(f, f) + Q_L(f, g) \\ \dot{y}_k &= w_k = \int K(y_k, x) f(x, v) dx dv + \sum_{l=1}^M K(y_k, y_l) + u_k, \end{aligned} \quad (2)$$

where $Q_F, Q_L(f, g)$ represents two **Boltzmann-like operators** accounting the instantaneous rate of change of the particles' density.

Binary Interaction algorithms

- In order to simulate the dynamic of (2), we rely on **Binary interaction algorithms**.
- This approach is based on **Monte-Carlo procedure**, which allows to solve efficiently the interaction kernels, see [2, 3] for a detailed description.
- The main advantages with respect to standard methods are: the **linear cost** for the evaluation of the interaction kernels Q_F, Q_L , and a **full meshless** algorithm.

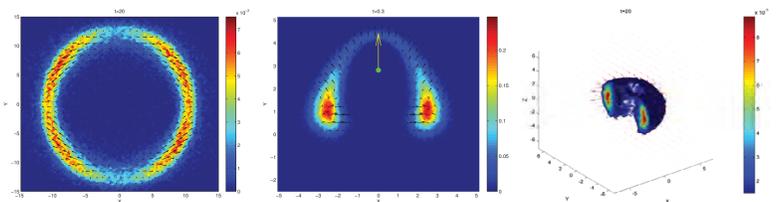


Figure : Simulations for the kinetic description of multi-agent systems through BI algorithms.

Simulations of the mesoscopic model

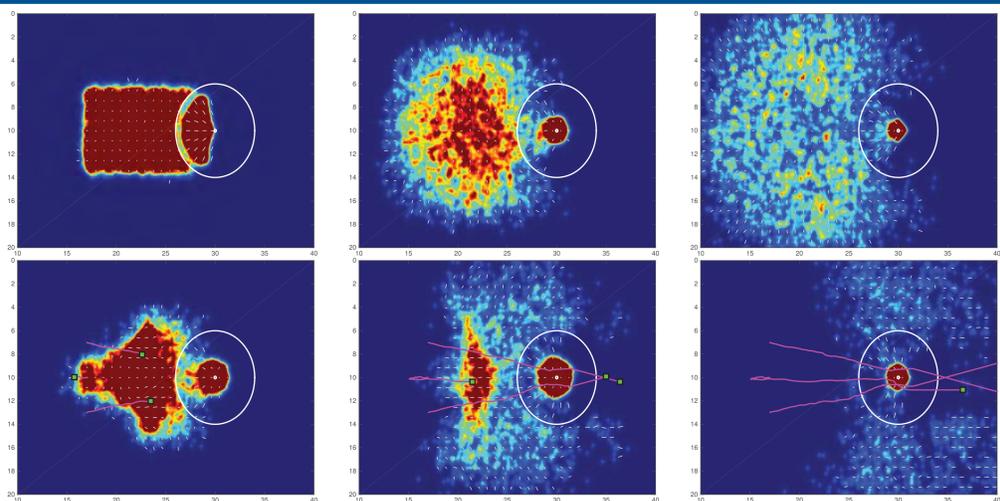


Figure : First row: no leaders. Second row: three leaders, optimal strategy u_k (computed via compass search methods).

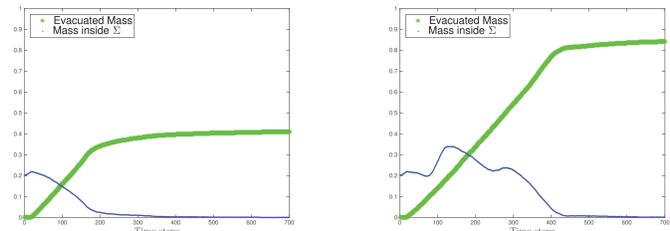


Figure : The **blue line** represents the **mass concentration** around the exit and **green line** the percentage of **total evacuated mass**, as functions of time. Case **without leaders** (left), total evacuated mass: 41.2%. Case **$M = 3$ leaders** (right) percentage of evacuated mass 85.2%. **Optimal strategies avoid congestions at exits.**

References

- [1] G. Albi and M. Bongini, E. Cristiani and D. Kalise, *Invisible Control of Self-Organizing Agents Leaving Unknown Environments*, Preprint No. IGDK-2015-09., (2015).
- [2] G. Albi and L. Pareschi, *Binary interaction algorithms for the simulation of flocking and swarming dynamics*, Multiscale Model. Simul., 11 (2013), pp. 1–29.
- [3] L. Pareschi and G. Toscani, *Interacting multi-agent systems. Kinetic equations & Monte Carlo methods*, Oxford University Press, USA, 2013.

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