## Data Science Course Understading swarm behaviour

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### **Presentation Overview**

- Review past models (simple speed coupling, couzin model, vicsek model)
- Module 1: Effective leadership and decision-making in animal groups on the move (Couzin et al)
- Module 2: Self-Propelled Particles with Soft-Core Interactions: Patterns, Stability, and Collapse (D'Orsogna et al)

# Review past models (simple speed coupling, vicsek model, couzin model)

Simple speed coupling: weighted speed

Particle adapts a fraction of its nearest neighbour's speed.

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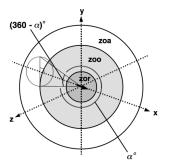
#### Vicsek model

- introduce interaction zones: repulsion and alignment
- add noise

## Review past models: simple speed coupling, vicsek model, couzin model

#### Couzin model

 different zones of neighbourhoods: repulsion, orientation, attraction



**Figure :** Representation of an individual in the model centred at the origin: zor = zone of repulsion, zoo = zone of orientation, zoa = zone of attraction.  $\alpha = field$  of perception

We introduce a bias: orientation of  $\geq 1$  particle (the "informed" one/ "scout") and merge orientation and attraction zone.

#### **THEORY**

**New parameters**: weight of bias, proportion of bias, group direction

**Goal:** Which parameter sets give a nice group movement, and how does the behaviour change when we change the parameters?

update of  $d_i$  in each zone:

$$\begin{split} \mathbf{d}_i(t+\Delta t) &= -\sum_{j\neq i} \frac{\mathbf{c}_j(t) - \mathbf{c}_i(t)}{|\mathbf{c}_j(t) - \mathbf{c}_i(t)|} \\ \mathbf{d}_i(t+\Delta t) &= \sum_{j\neq i} \frac{\mathbf{c}_j(t) - \mathbf{c}_i(t)}{|\mathbf{c}_j(t) - \mathbf{c}_i(t)||} + \sum_{j=1} \frac{\mathbf{v}_j(t)}{|\mathbf{v}_j(t)|} \\ \mathbf{d}_i{}^{'}(t+\Delta t) &= \frac{\hat{\mathbf{d}}_i(t+\Delta t) + \omega \mathbf{g}_i}{|\hat{\mathbf{d}}_i(t+\Delta t) + \omega \mathbf{g}_i|} \end{split}$$

#### **IMPLEMENTATION**

- periodic boundary conditions
- virtual interaction of agents
- explicit Euler method:  $v_{n+1} = v_n + \tau \cdot f(t_n, v_n)$

**SIMULATION** on specific parameter set

### **QUANTIFICATION**

- Accuracy
- Group elongation
- Two scouts

**SIMULATION** optimised

**Motivation**: Simulate swarming of multiagent systems in order to understand the operating principles of natural swarms, e.g. swarming behaviour of *M. xanthus* cells.



We choose a **kinetic theory** based approach.

The model traits are:

- self-propulsion
- friction
- interaction between particles: repulsion and attraction

Consider N interacting, self-propelled particles governed by the following equations of motion

$$\frac{d\vec{x_i}}{dt} = \vec{v_i} \tag{1}$$

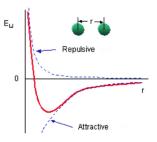
$$F = m \frac{d\vec{v_i}}{dt} = (\alpha v - \beta |\vec{v_i}|^2) \vec{v_i} - \vec{\nabla_i} U(\vec{x_i})$$
 (2)

where U is a pairwise interaction potential and  $\alpha, \beta > 0$  are values for propulsion and friction forces.

For *U* we choose the *Morse potential* 

$$U(\vec{x_i}) = \sum_{j \neq i} \left[ -C_a e^{-|\vec{x_i} - \vec{x_j}|/l_a} + \underbrace{C_r e^{-|\vec{x_i} - \vec{x_j}|/l_r}}_{\text{repulsion}} \right]$$
(3)

where  $C_a$ ,  $C_r$  denote attractive and repulsive strengths and  $I_a$ ,  $I_r$  their respective length scales.

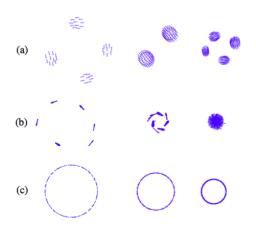


#### IMPLEMENTATION

- rigid boundary conditions
- no virtual interactions
- initial conditions of random distribution
- explicit Euler method to solve the ODEs

### **SIMULATION**

INSERT one example simulation



**Figure :** Catastrophic geometry.
(a) Clumps. (b) Ring Clumping. (c) Rings.

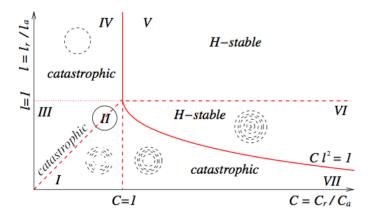
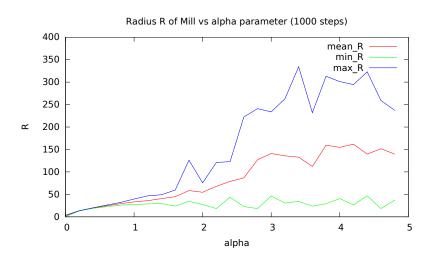


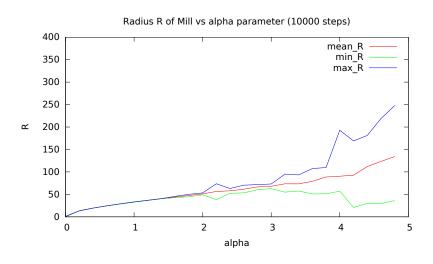
Figure: H-stability phase diagram of the Morse potential

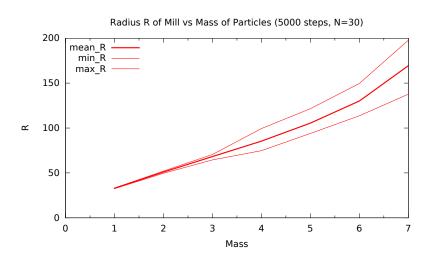
INSERT own simulations

### **QUANTIFICATION**

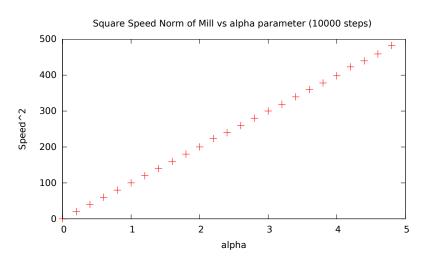
- parameter sets for different regions
- testing interesting trends, e.g.  $R \propto \frac{\alpha m}{\beta N}$  (from setting  $F_{centrifugal} = F_{centripetal}$ )

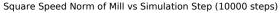


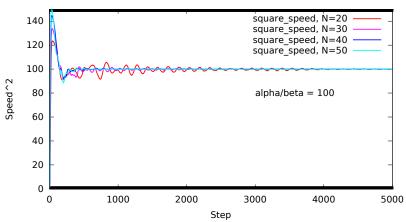




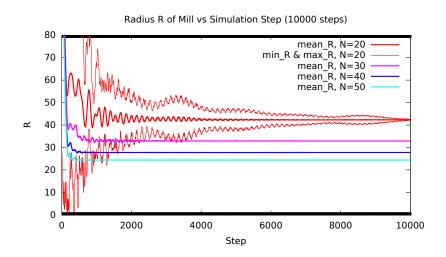
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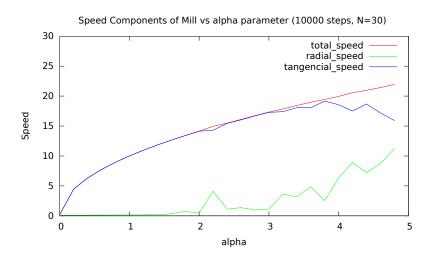


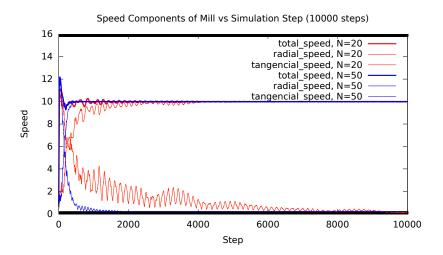




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- confirmation of ring formation
  - $R_{max}$ ,  $R_{mean}$ ,  $R_{min}$  converge to same value
  - $v_{radial} o 0$  and  $v_{tangential} o |v_i|$  for  $t o \infty$







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- confirmation of ring formation
  - $R_{max}$ ,  $R_{mean}$ ,  $R_{min}$  converge to same value
  - $v_{radial} \rightarrow 0$  and  $v_{tangential} \rightarrow |v_i|$  for  $t \rightarrow \infty$
- influence of increasing N on the shape



