

# Sparse Control of Alignement models

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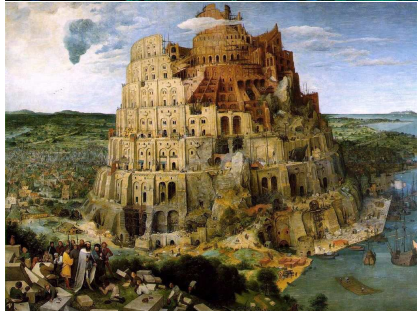
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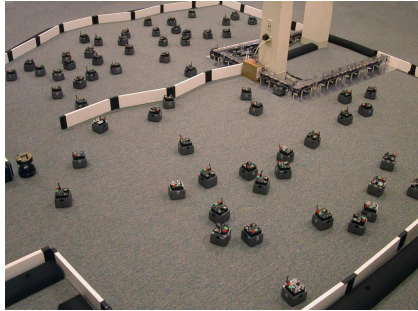


# Emergent behaviors in multi-agent systems

- Birds Flocking; Fish Shoaling; Locusts Swarming;
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Applications:

- Forcing emergent behaviors in swarms of robots;
- Avoid financial crisis or “Black Swans”.



# The Cucker–Smale model

## Cucker–Smale (2007)

$$\begin{cases} \dot{x}_i = v_i \\ \dot{v}_i = \frac{1}{N} \sum_{j=1}^N a(\|x_i - x_j\|)(v_j - v_i) \end{cases} \quad (CS)$$

where  $x_1, \dots, x_N, v_1, \dots, v_N \in \mathbb{R}^d$  and  $a$  is positive and nonincreasing.

- $x_i \in \mathbb{R}^d$  is the *main state* (e.g. position);
- $v_i \in \mathbb{R}^d$  is the *consensus parameter* (e.g. velocity);
- $a(\cdot)$  represents *the communication rate*.

In Cucker–Smale (2007) :  $a(x) = \frac{1}{(1 + x^2)^\beta}$ ,  $\beta > 0$ .

# The Cucker–Smale model

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where  $x_1, \dots, x_N, v_1, \dots, v_N \in \mathbb{R}^d$  and  $a$  is positive and nonincreasing.

In matrix notation:

$$\begin{cases} \dot{x} = v \\ \dot{v} = -L_x v, \end{cases}$$

where  $L_x$  is the Laplacian. <sup>1</sup>

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<sup>1</sup>Given a nonnegative symmetric  $N \times N$  matrix  $A = (a_{ij})_{i,j}$ , the *Laplacian*  $L$  of  $A$  is defined by  $L = D - A$ , with  $D = \text{diag}(d_1, \dots, d_N)$  and  $d_k = \sum_{j=1}^N a_{kj}$ .

# Consensus

- The mean velocity  $\bar{v} = \frac{1}{N} \sum_{i=1}^N v_i$  is conserved.

## Definition (Consensus point)

A steady configuration of System (CS)  $(x, v) \in (\mathbb{R}^d)^N \times (\mathbb{R}^d)^N$  where

$$v_1 = \dots = v_N$$

is called a *consensus point*.

- The dynamics originating from a consensus point  $(x, v)$  is given by rigid translation  $x(t) = x + t\bar{v}$ .





# Unconditional consensus emergence

- the bilinear form  $B(u, v) = \frac{1}{2N^2} \sum_{i,j=1}^N \langle u_i - u_j, v_i - v_j \rangle = \frac{1}{N} \sum_{i=1}^N \langle u_i, v_i \rangle - \langle \bar{u}, \bar{v} \rangle$ ,
- the *dispersion*  $X(t) := B(x(t), x(t)) = \frac{1}{2N^2} \sum_{i,j=1}^N \|x_i(t) - x_j(t)\|^2$ ,
- the *disagreement*  $V(t) := B(v(t), v(t)) = \frac{1}{2N^2} \sum_{i,j=1}^N \|v_i(t) - v_j(t)\|^2$ .

# Unconditional consensus emergence

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Theorem (Cucker–Smale '07; Ha–Tadmor '08; Carrillo et al. '08)

If

$$a(x) = \frac{1}{(1+x^2)^\beta} \quad \text{and} \quad \beta \leq \frac{1}{2},$$

every solution of (CS) tend to consensus.

- Moreover if  $\beta < 1/2$  there exists  $\alpha > 0$  such that for every  $t > 0$ ,

$$X(t) \leq \bar{X} \quad \text{and} \quad V(t) \leq V(0)e^{-t\alpha}$$

# Conditional consensus

## Proposition (Ha–Ha–Kim, 2010)

Let  $(x_0, v_0) \in (\mathbb{R}^d)^N \times (\mathbb{R}^d)^N$  be such that  $X_0 = B(x_0, x_0)$  and  $V_0 = B(v_0, v_0)$  satisfy

$$\sqrt{V_0} \leq \int_{\sqrt{X_0}}^{\infty} a(\sqrt{2Nr}) dr.$$

Then the solution of (CS) with initial data  $(x_0, v_0)$  tends to consensus.

# Non-Consensus

Two agents moving on  $\mathbb{R}$  with  $a = 2/(1 + x^2)$ .

- $x(t) = x_1(t) - x_2(t)$  relative main state
- $v(t) = v_1(t) - v_2(t)$  relative consensus parameter

Then

$$\begin{cases} \dot{x} = v \\ \dot{v} = -\frac{v}{1+x^2} \end{cases}$$

with  $x(0) = x_0$  and  $v(0) = v_0 > 0$ . The solution of this system is

$$v(t) - v_0 = -\arctan x(t) + \arctan x_0.$$

If  $\arctan x_0 + v_0 > \pi/2$  ( $\implies \arctan x_0 + v_0 \geq \pi/2 + \varepsilon$  for some  $\varepsilon > 0$ ) then

$$v(t) \geq -\arctan x(t) + \pi/2 + \varepsilon > \varepsilon,$$

for every  $t > 0$ . In other words, the system does **not** tend to consensus.

# Self-organization

Admissible controls, accounting for the external field, are measurable functions  $u = (u_1, \dots, u_N) : [0, +\infty) \rightarrow (\mathbb{R}^d)^N$  satisfying the  $\ell_1^N - \ell_2^d$ -norm constraint

$$\sum_{i=1}^N \|u_i(t)\| \leq M,$$

for every  $t > 0$ , for a given  $M > 0$ .

## Cucker–Smale System

$$\begin{cases} \dot{x}_i(t) = v_i(t), \\ \dot{v}_i(t) = \frac{1}{N} \sum_{j=1}^N a(\|x_j(t) - x_i(t)\|)(v_j(t) - v_i(t)) \end{cases} \quad (CCS)$$

for  $i = 1, \dots, N$ , and  $x_i \in \mathbb{R}^d$ ,  $v_i \in \mathbb{R}^d$ .

# Self-organization Vs Organization via intervention

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## Controlled Cucker–Smale System

$$\begin{cases} \dot{x}_i(t) = v_i(t), \\ \dot{v}_i(t) = \frac{1}{N} \sum_{j=1}^N a(\|x_j(t) - x_i(t)\|)(v_j(t) - v_i(t)) + u_i(t), \end{cases} \quad (CCS)$$

for  $i = 1, \dots, N$ , and  $x_i \in \mathbb{R}^d$ ,  $v_i \in \mathbb{R}^d$ .

# Totally distributed control

## Proposition

*For every initial condition  $(x_0, v_0) \in (\mathbb{R}^d)^N \times (\mathbb{R}^d)^N$  and  $M > 0$  there exists  $T > 0$  and  $u : [0, T] \rightarrow (\mathbb{R}^d)^N$  with  $\sum_{i=1}^N \|u_i(t)\| \leq M$  for every  $t \in [0, T]$  such that the associated solution tends to **consensus***

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Consider the solution with initial data  $(x_0, v_0)$  associated with the feedback

$$u(t) = -\alpha(v(t) - \bar{v}(t)) \quad \text{with } 0 < \alpha \leq \frac{M}{N\sqrt{B(v_0, v_0)}}.$$

Then,

$$\begin{aligned} \frac{d}{dt}V(t) &= \frac{d}{dt}B(v(t), v(t)) \\ &= -2B(L_x v(t), v(t)) + 2B(u(t), v(t)) \\ &\leq 2B(u(t), v(t)) = -2\alpha B(v - \bar{v}, v - \bar{v}) = -2\alpha V(t). \end{aligned}$$

Therefore  $V(t) \leq e^{-2\alpha t}V(0)$  and  $V(t) \rightarrow 0$  exponentially fast as  $t \rightarrow \infty$ .



# Greedy control: the variational principle

For every  $(x, v) \in (\mathbb{R}^d)^N \times (\mathbb{R}^d)^N$  and  $M > 0$ , let  $U(x, v)$  be defined as the set of solutions of the variational problem

$$\min B(u, v)$$

$$\text{subject to } \sum_{i=1}^N \|u_i\| \leq M,$$

# Greedy control: the variational principle

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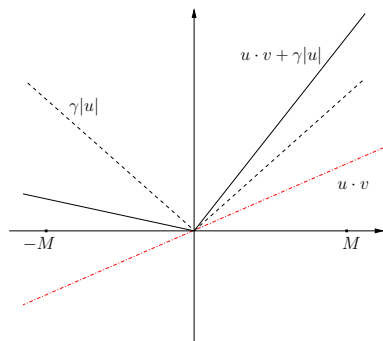
$$\begin{aligned} \min \left( B(u, v) + \frac{\gamma(x)}{N} \sum_{i=1}^N \|u_i\| \right) \\ \text{subject to } \sum_{i=1}^N \|u_i\| \leq M, \end{aligned}$$

where

$$\gamma(x) = \int_{\sqrt{B(x,x)}}^{\infty} a(\sqrt{2Nr}) dr.$$

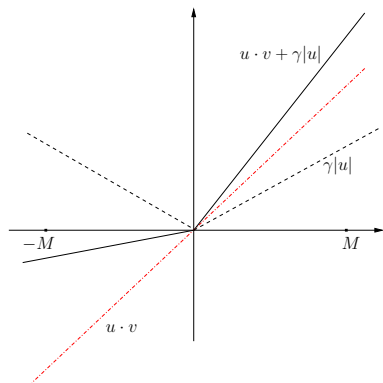
# Geometric interpretation (scalar case)

Case  $|v| < \gamma$



$u = 0$  unique solution in  $[-M, M]$

Case  $|v| > \gamma$



$|u| = M$  unique solution in  $[-M, M]$

# Greedy controls are stabilizing

## Theorem

*For every initial pair  $(x_0, v_0) \in (\mathbb{R}^d)^N \times (\mathbb{R}^d)^N$ , the differential inclusion*

$$(\dot{x}, \dot{v}) \in \{(v, -L_x v + u) \mid u \in U(x, v)\}$$

*with initial condition  $(x(0), v(0)) = (x_0, v_0)$  is well-posed and its solutions converge to consensus as  $t$  tends to  $+\infty$ .*

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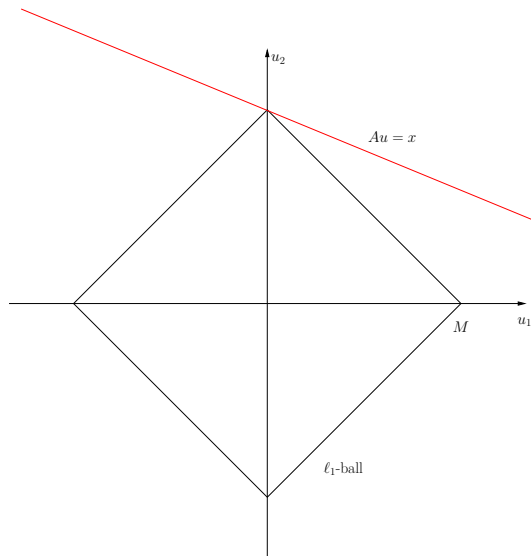
with initial condition  $(x(0), v(0)) = (x_0, v_0)$  is well-posed and its solutions converge to consensus as  $t$  tends to  $+\infty$ .

For almost every  $(x, v) \in (\mathbb{R}^d)^N \times (\mathbb{R}^d)^N$  the solution of

$$\min \left( B(u, v) + \frac{\gamma(x)}{N} \sum_{i=1}^N \|u_i\| \right) \text{ subject to } \sum_{i=1}^N \|u_i\| \leq M,$$

is **sparse**, in particular it has at most **one nonzero component**.

# Sparsity: the geometric interpretation



Consider the case  $d = 1$ ,  
 $N = 2$ . We want to find

$$\max\{Au = x\},$$

subject to

$$|u_1| + |u_2| \leq M.$$

# Sparse control: explicit construction

Let  $v_{\perp}(t) = v(t) - \bar{v}(t)$ .

- If  $\|v_{\perp_i}\| \leq \gamma(x)$  for every  $i = 1, \dots, N \implies$  consensus region reached  
 $\implies u_1 = \dots = u_N = 0$ .

# Sparse control: explicit construction

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## The sparse control

- If  $\|v_{\perp_i}\| \leq \gamma(x)$  for every  $i = 1, \dots, N \implies$  consensus region reached  
 $\implies u_1 = \dots = u_N = 0$ .
- Otherwise let  $j \in \{1, \dots, N\}$  be the smallest index such that

$$\|v_{\perp_j}\| = \max_{1 \leq i \leq N} \|v_{\perp_i}\|$$

then

$$u_j = -M \frac{v_{\perp_j}}{\|v_{\perp_j}\|}, \quad \text{and} \quad u_i = 0 \quad \text{for every } i \neq j.$$



# Sparse control: explicit construction

Let  $v_{\perp}(t) = v(t) - \bar{v}(t)$ .

## The sparse control or “Shepherd dog” strategy

- If  $\|v_{\perp_i}\| \leq \gamma(x)$  for every  $i = 1, \dots, N \implies$  consensus region reached  $\implies u_1 = \dots = u_N = 0$ .
- Otherwise let  $j \in \{1, \dots, N\}$  be the smallest index such that

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then

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(Loading 4 agents)

(Loading 5 agents)

(Loading 6 agents)

(Loading 20 agents)

(Loading 50 agents)

# Optimality of the greedy strategy

All the controls in  $U(x, v)$  are of the form

$$u_i = \begin{cases} 0 & \text{if } v_{\perp i} = 0 \\ -\alpha_i \frac{v_{\perp i}}{\|v_{\perp i}\|} & \text{if } v_{\perp i} \neq 0 \end{cases}$$

for  $\alpha_i \geq 0$ ,  $\sum_{i=1}^N \alpha_i \leq M$ .

## Proposition

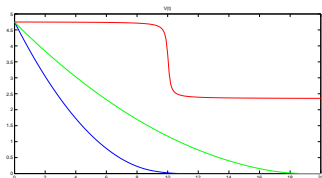
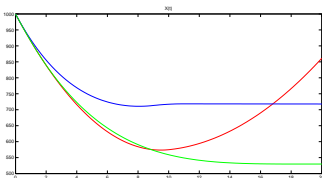
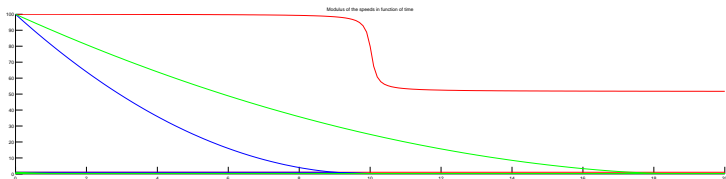
*The sparse feedback control  $u$ , associated with the solution  $((x(t), v(t)))$  is a minimizer of*

$$\mathcal{R}(t, u) = \frac{d}{dt} V(t),$$

*over all possible feedback controls in  $U(x(t), v(t))$ . In other words, the sparse feedback control  $u$  is the **best choice** in terms of the rate of convergence to consensus.*

# Numerical simulations I

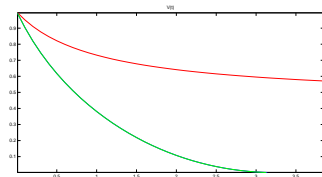
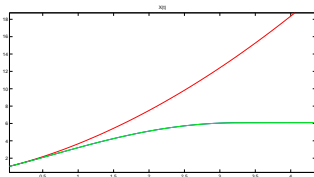
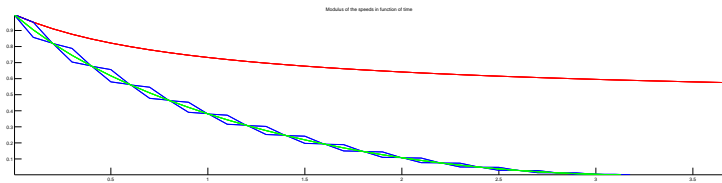
- $N = 20$ ;
- $x_1(0) = (-100, 0)$  and  $v_1(0) = (-1, 0)$ ;
- $x_2(0) = (100, 0)$  and  $v_2(0) = (-10, 0)$ ;
- $x_3(0) = \dots = x_{20}(0) = (0, 0) = v_3(0) = \dots = v_{20}(0)$ ;





# Numerical simulations II: Fully symmetric case

- $N = 4$
- $x_1(0) = (-1, 0)$  and  $v_1(0) = (-1, 0)$ ;
- $x_2(0) = (0, 1)$  and  $v_2(0) = (0, 1)$
- $x_3(0) = (1, 0)$  and  $v_3(0) = (1, 0)$
- $x_4(0) = (0, -1)$  and  $v_4(0) = (0, -1)$



# Sample and hold

## Definition (Sampling solution)

Let  $U \subset \mathbb{R}^m$ ,  $f : \mathbb{R}^n \times U \rightarrow \mathbb{R}^n$  be continuous and locally Lipschitz in  $x$  uniformly on compact subset of  $\mathbb{R}^n \times U$ . Given a feedback  $u : \mathbb{R}^n \rightarrow U$ ,  $\tau > 0$ , and  $x_0 \in \mathbb{R}^n$  we define the *sampling solution* of the differential system

$$\dot{x} = f(x, u(x)), \quad x(0) = x_0,$$

as the continuous (actually piecewise  $C^1$ ) function  $x : [0, T] \rightarrow \mathbb{R}^n$  solving recursively for  $k \geq 0$

$$\dot{x}(t) = f(x(t), u(x(k\tau))), \quad t \in [k\tau, (k+1)\tau]$$

using as initial value  $x(k\tau)$ , the endpoint of the solution on the preceding interval, and starting with  $x(0) = x_0$ . We call  $\tau$  the *sampling time*.

## Theorem

Fix  $M > 0$  and consider the sparse feedback control  $u$ . Then for every initial condition  $(x_0, v_0) \in (\mathbb{R}^d)^N \times (\mathbb{R}^d)^N$  there exists  $\tau_0 > 0$  small enough, such that for all  $\tau \in (0, \tau_0]$  the sampling solution of (CCS) associated with the control  $u$ , the sampling time  $\tau$ , and initial pair  $(x_0, v_0)$  reaches the consensus region in **finite time**.

In particular the systems reaches the consensus region within time

$$T_0 = \frac{2N}{M}(\sqrt{V(0)} - \gamma(\bar{X}))$$

where  $\bar{X} = 2B(x_0, x_0) + \frac{2N^4}{M^2}B(v_0, v_0)^2$

# Local Controllability

Let  $\mathcal{V}_f = \{(v_1, \dots, v_N) \in (\mathbb{R}^d)^N \mid v_1 = \dots = v_N\}$ .

## Theorem

- For every  $(\tilde{x}_0, \tilde{v}_0) \in (\mathbb{R}^d)^N \times \mathcal{V}_f$ ,
- for almost every  $(\tilde{x}_1, \tilde{v}_1) \in (\mathbb{R}^d)^N \times \mathcal{V}_f$ ,
- for every  $\delta > 0$ , and
- for every  $i = 1, \dots, N$

there exist  $T > 0$  and a control  $u : [0, T] \rightarrow [0, \delta]^d$  steering the system from  $(\bar{x}, \bar{v})$  to  $(\tilde{x}, \tilde{v})$ , with the property  $u_j(t) = 0$  for every  $j \neq i$  and every  $t \in [0, T]$ .

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$$\begin{cases} \dot{x}^k = v^k \\ \dot{v}^k = -L_{\bar{x}}v^k + Bu, \end{cases}$$

for every  $k = 1, \dots, d$  where  $B = (1, 0, \dots, 0)^t$

Therefore we reduce the investigation of the Kalman condition for a linear system on  $\mathbb{R}^N$  of the form

$$\dot{v} = (-L_{\bar{x}})v + Bu.$$

## Corollary

*For every  $M > 0$ , for every initial condition  $(x_0, v_0) \in (\mathbb{R}^d)^N \times (\mathbb{R}^d)^N$ , for almost every  $(x_1, v_1) \in (\mathbb{R}^d)^N \times \mathcal{V}_f$ , there exist  $T > 0$  and a componentwise and time sparse control  $u : [0, T] \rightarrow (\mathbb{R}^d)^N$ , satisfying  $\sum_{i=1}^N \|u_i(t)\| \leq M$ , such that the corresponding solution starting at  $(x_0, v_0)$  arrives at the consensus point  $(x_1, v_1)$  within time  $T$ .*

# Optimal Control

We consider the optimal control problem

$$\begin{cases} \dot{x}_i(t) = v_i(t), \\ \dot{v}_i(t) = \frac{1}{N} \sum_{j=1}^N a(\|x_j(t) - x_i(t)\|)(v_j(t) - v_i(t)) + u_i(t), \end{cases} \quad (\text{CCS})$$

with running cost, for  $\gamma > 0$ ,

$$\int_0^T \left( \sum_{i=1}^N \left( v_i(t) - \frac{1}{N} \sum_{j=1}^N v_j(t) \right)^2 + \gamma \sum_{i=1}^N \|u_i(t)\| \right) dt.$$

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- High dimensional of the state space  $\implies$  difficulties in applying theoretical results (PMP);
- Codimension of the non-sparse manifold in the space of  $(x, v, p_x, p_v)$ ;



(Loading 6 agents - optimal control)

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# Reference

- Marco Caponigro (CNAM), Massimo Fornasier (TU München), Benedetto Piccoli (Rutgers), Emmanuel Trélat (Paris 6),  
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**Thank you!**

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