



Exponential Flocking under Communication Failures for some Cucker-Smale models

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(in collaboration with É. Flayac)

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Outline of the talk

Multi-agent systems and pattern formation

A quick overview of the Cucker-Smale flocking

Cucker-Smale flocking under communication failures

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Multi-agent systems – *Some illustrations*

A multi-agent system is a large ensemble of interacting things



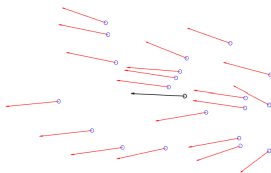
Introduction – *Pattern formation*

Central observation (pattern formation)

Interacting multi-agent systems may form **interesting global structures** starting from **elementary** interaction rules.

Examples of classical patterns

- ◇ **Consensus** (everybody tends to agree on something) :
↪ *aggregation models in biology, opinion models, etc...*
- ◇ **Flocking** (everybody goes in the same direction) :
↪ *flocks of birds, herds analysis, opinion formation, etc...*



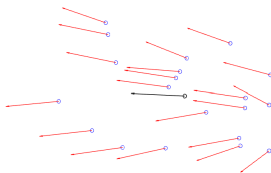
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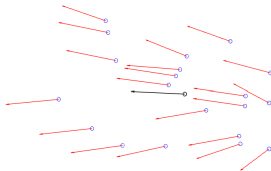
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Multi-agent systems – *Weighted Cucker-Smale model*

Let $N \geq 1$ and $(\mathbf{x}, \mathbf{v}) := (x_1, \dots, x_N, v_1, \dots, v_N) \in (\mathbb{R}^{2d})^N$ be s.t.

$$(CS') \quad \begin{cases} \dot{x}_i(t) = v_i(t), \\ \dot{v}_i(t) = \frac{1}{N} \sum_{j=1}^N \xi_{ij}(t) \phi(|x_i(t) - x_j(t)|) (v_j(t) - v_i(t)), \end{cases}$$

where

- ◇ $\phi(\cdot)$ is **radial** \rightsquigarrow interactions depend on **relative distance**,
- ◇ $\xi_{ij}(\cdot) \in L^\infty(\mathbb{R}_+, [0, 1])$ are **symmetric communication rates**

\rightsquigarrow **Intuitive idea** : each agent tries to **align** its velocity with that of the other agents, with a **weight** depending on their relative distance.

Main question for today

Under which hypotheses will (CS') converge towards alignment ?

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Alignment models (Cucker & Smale '07)

Consider the **full-communication** case where $\xi_{ij}(\cdot) \equiv 1$, i.e.

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where $\phi \in \text{Lip}(\mathbb{R}_+, \mathbb{R}_+^*)$ is a **positive** and **non-increasing** kernel.

Definition (Asymptotic flocking for (CS))

A solution $(x(\cdot), v(\cdot))$ of (CS) **converges to flocking** if

$$\sup_{t \geq 0} |x_i(t) - \bar{x}(t)| < +\infty \quad \text{and} \quad \lim_{t \rightarrow +\infty} |v_i(t) - \bar{v}| = 0.$$

where $\bar{x}(\cdot)$ and \bar{v} are the position-velocity barycenters.

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Cucker-Smale systems – *Convergence analysis framework*

First question

How do we characterise flocking formation ?

↪ **Good idea** : find a suitable **dissipative Lyapunov** structure

Definition (Variance and standard deviation)

Consider the **variance bilinear form**

$$B : (\mathbf{x}, \mathbf{y}) \in (\mathbb{R}^d)^N \times (\mathbb{R}^d)^N \mapsto \frac{1}{N} \sum_{i=1}^N \langle \mathbf{x}_i, \mathbf{y}_i \rangle - \langle \bar{\mathbf{x}}, \bar{\mathbf{y}} \rangle.$$

Characterisation of flocking

A solution $(\mathbf{x}(\cdot), \mathbf{v}(\cdot))$ of (CS) converges to flocking **if and only if**

$$\sup_{t \geq 0} X(t) < +\infty \quad \text{and} \quad \lim_{t \rightarrow +\infty} V(t) = 0,$$

where $X(t) := \sqrt{B(\mathbf{x}(t), \mathbf{x}(t))}$ and $V(t) := \sqrt{B(\mathbf{v}(t), \mathbf{v}(t))}$.

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Generalised flocking – *Underlying dissipative structure*

Definition (Graph Laplacians)

Consider the operator $\mathbf{L} : (\mathbb{R}^d)^N \rightarrow \mathcal{L}((\mathbb{R}^d)^N)$ defined by

$$(\mathbf{L}(\mathbf{x})\mathbf{v})_i = \frac{1}{N} \sum_{j=1}^N \phi(|\mathbf{x}_i - \mathbf{x}_j|)(\mathbf{v}_i - \mathbf{v}_j),$$

for all $i \in \{1, \dots, N\}$.

Proposition (Interesting things about graph Laplacians)

- ◇ Allows to rewrite (CS) as the **semilinear** system

$$\dot{\mathbf{x}}(t) = \mathbf{v}(t), \quad \dot{\mathbf{v}}(t) = -\mathbf{L}(\mathbf{x}(t))\mathbf{v}(t).$$

- ◇ It holds that $B(\mathbf{L}(\mathbf{x})\mathbf{v}, \mathbf{v}) \geq 0$ for all $\mathbf{x}, \mathbf{v} \in (\mathbb{R}^d)^N$.
- ◇ The strength of the interactions in the system is quantified by the **eigenvalues** of $\mathbf{L}(\mathbf{x}(t))$.

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Cucker-Smale systems – *Known convergence results*

Observation : $\dot{X}(t) \leq V(t)$ and $\dot{V}(t) \leq -\phi(2\sqrt{N}X(t))V(t)$

Theorem (Flocking formation) [Cucker & Smale '07, Ha & Liu '09]

Suppose that $\phi(\cdot) \in \text{Lip}(\mathbb{R}_+, \mathbb{R}_+^*)$ is **positive** and **non-increasing**.

- ◇ If $\phi(\cdot)$ is **lower-bounded**, then flocking **always occurs**
 $\rightsquigarrow V(\cdot)$ uniformly exponentially converges towards 0,
- ◇ If $\phi(\cdot)$ **vanishes at infinity** but $\phi \notin L^1(\mathbb{R}_+, \mathbb{R}_+)$, then flocking **always occurs** as well!
 \rightsquigarrow one has $\phi \notin L^1 \implies X(t) \leq X_M$ for some $X_M > 0$,
- ◇ If $\phi(\cdot)$ **vanishes at infinity** and $\phi \in L^1(\mathbb{R}_+, \mathbb{R}_+)$, then flocking **may fail to occur...** \rightsquigarrow One needs small $(X(0), V(0))$.

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Theorem (Flocking formation) [Cucker & Smale '07, Ha & Liu '09]

Suppose that $\phi(\cdot) \in \text{Lip}(\mathbb{R}_+, \mathbb{R}_+^*)$ is **positive** and **non-increasing**.

- ◇ If $\phi(\cdot)$ is **lower-bounded**, then flocking **always occurs**
 $\rightsquigarrow V(\cdot)$ uniformly exponentially converges towards 0,
- ◇ If $\phi(\cdot)$ **vanishes at infinity** but $\phi \notin L^1(\mathbb{R}_+, \mathbb{R}_+)$, then flocking **always occurs** as well!
 \rightsquigarrow one has $\phi \notin L^1 \implies X(t) \leq X_M$ for some $X_M > 0$,
- ◇ If $\phi(\cdot)$ **vanishes at infinity** and $\phi \in L^1(\mathbb{R}_+, \mathbb{R}_+)$, then flocking **may fail to occur...** \rightsquigarrow One needs small $(X(0), V(0))$.

\hookrightarrow We will henceforth assume that $\phi \notin L^1(\mathbb{R}_+, \mathbb{R}_+)$.

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Outline of the talk

Multi-agent systems and pattern formation

A quick overview of the Cucker-Smale flocking

Cucker-Smale flocking under communication failures

Generalised flocking – *Modelling communication failures*

Cucker-Smale system (Back to communication failures)

The weighted Cucker-Smale model can be rewritten as

$$\dot{\mathbf{x}}(t) = \mathbf{v}(t), \quad \dot{\mathbf{v}}(t) = -\mathbf{L}(t, \mathbf{x}(t))\mathbf{v}(t),$$

where we introduce the **time-dependent** graph-Laplacian

$$(\mathbf{L}(t, \mathbf{x})\mathbf{v})_i = \frac{1}{N} \sum_{j=1}^N \xi_{ij}(t) \phi(|x_i - x_j|) (v_i - v_j)$$

Problem (Communication weights)

↪ The $\xi_{ij}(\cdot)$ may vanish or be small on possibly long time-intervals.

Main question

Under what kind of assumption on $\xi_{ij}(\cdot)$ do we recover flocking ?

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First idea : Agents i and j directly communicate if $\xi_{ij}(t) > 0$, but they can still **indirectly communicate** when $\xi_{ij}(t) = 0$.



Figure: In both situation $\xi_{34}(t) = 0$ but agents 3 and 4 communicate

Second idea : All the agents do not **need** to interact at all times, \hookrightarrow we only need a lower-bound on the **average interactions**.

\rightsquigarrow **Persistence conditions** on the $\xi_{ij}(\cdot)$ using a graph-Laplacian

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Generalised flocking – *Persistence and graph Laplacian*

Definition (Graph Laplacian of the weights)

Consider the operator $\mathbf{L}_\xi : \mathbb{R}_+ \rightarrow \mathcal{L}((\mathbb{R}^d)^N)$ defined by

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Definition (Persistence condition)

The weights $\xi_{ij}(\cdot)$ satisfy the **persistence condition** (PE) if

$$B\left(\left(\frac{1}{\tau} \int_t^{t+\tau} \mathbf{L}_\xi(s) ds\right) \mathbf{v}, \mathbf{v}\right) \geq \mu B(\mathbf{v}, \mathbf{v}), \quad (\text{PE}_{\tau, \mu})$$

for all $(t, \mathbf{v}) \in \mathbb{R}_+ \times (\mathbb{R}^d)^N$, where $(\tau, \mu) \in \mathbb{R}_+^* \times (0, 1]$.

Heuristic meaning

On every time window of length τ , the total average of interactions (both **direct** and **indirect**) between pairs of agents is **at least** μ .

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Generalised flocking – *Main convergence result*

Theorem (Flocking formation)[B. & Flayac '21]

Let $(\mathbf{x}^0, \mathbf{v}^0) \in (\mathbb{R}^d)^N \times (\mathbb{R}^d)^N$ and assume the following holds.

- (i) The weights $\xi_{ij}(\cdot)$ satisfy (PE) for some $(\tau, \mu) \in \mathbb{R}_+ \times (0, 1]$.
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$$\phi(r) \geq \frac{K}{(\sigma + r)^\beta}.$$

Then, the solution $(\mathbf{x}(\cdot), \mathbf{v}(\cdot))$ of (CS') **converges to flocking**.

Remarks (sharpness of the assumptions)

- ◊ Hypothesis (ii) is probably not sharp... we expect $\beta \in (0, 1)$
- ◊ The persistence assumption (PE) **seems quite sharp** !
 - Average connectivity is necessary, even for consensus
 - Quantifying dissipation estimates are needed for flocking

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Generalised flocking – *Elements of proof*

Problem : Lack of dissipation for $V(\cdot) \rightsquigarrow$ strict Lyapunov design !

- 1) Define the **time-varying** family of operators

$$\psi_\tau(t) = (1 + c^2)\tau \text{Id} - \frac{1}{\tau} \int_t^{t+\tau} \int_t^s \mathbf{L}(\sigma, \mathbf{x}(\sigma)) d\sigma ds$$

and consider the candidate **Lyapunov functional**

$$\mathcal{V}_\tau(t) = \lambda(t)V(t) + \sqrt{B(\psi_\tau(t)\mathbf{v}(t), \mathbf{v}(t))}$$

where $\lambda(\cdot)$ is a tuning parameter (e.g. Mazenc & Malisoff '09).

- 2) After some (heavy) Lyapunov using (PE) and (ii), we have

$$V(T) \lesssim V(0) \exp \left(- C_{\tau, \mu} T_\lambda^{\frac{1-2\beta}{2(1-\beta)}} \right)$$

where T_λ is the **existence horizon** of $\lambda(\cdot) \rightsquigarrow T \rightarrow +\infty$!

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where $\lambda(\cdot)$ is a tuning parameter (e.g. Mazenc & Malisoff '09).

- 2) After some (heavy) Lyapunov using (PE) and (ii), we have

$$V(T) \lesssim V(0) \exp \left(- C_{\tau, \mu} T_\lambda^{\frac{1-2\beta}{2(1-\beta)}} \right)$$

where T_λ is the **existence horizon** of $\lambda(\cdot) \rightsquigarrow T \rightarrow +\infty$!

Generalised flocking – *Elements of proof*

Problem : Lack of dissipation for $V(\cdot) \rightsquigarrow$ strict Lyapunov design !

- 1) Define the **time-varying** family of operators

$$\psi_\tau(t) = (1 + c^2)\tau \text{Id} - \frac{1}{\tau} \int_t^{t+\tau} \int_t^s \mathbf{L}(\sigma, \mathbf{x}(\sigma)) d\sigma ds$$

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Generalised flocking – *Conclusion and open perspective*

Conclusive remarks

We have identified a synthetic and fairly minimalist persistence condition for multi-agent flocking via Lyapunov methods \rightsquigarrow **Nice !**

Open questions and coming work

- 1) Recover the **sharp** exponent range $\beta \in (0, 1)$
 \hookrightarrow Try a **better Lyapunov design!**
- 2) Carry out the analysis using L^∞ -Lyapunov functionals
 \hookrightarrow Necessary to obtain **uniform estimates** w.r.t. N and study **graph-limits** as $N \rightarrow +\infty$!
- 3) Adapt our methodology to **random communication failures**
 \hookrightarrow Case where $(t, \omega) \mapsto \xi_{ij}(t, \omega)$ are **stochastic processes**

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All in all...

Thank you for your attention !

