

# COS598H Final Presentation: Stability Analysis in Swarms

Zachary Stier

May 3, 2019

# Outline

- \* Motivation: simulating flocks
- \* Gazi and Passino's model
- \* Global convergence
- \* Agentwise convergence

# Background: modeling flocks

It would be nice to have a realistic model of how organisms flock

# Background: modeling flocks

It would be nice to have a realistic model of how organisms flock  
It would also be nice to understand mathematics of such models

# Background: modeling flocks

It would be nice to have a realistic model of how organisms flock  
It would also be nice to understand mathematics of such models  
Unfortunately...

# Background: modeling flocks (boids)

That simulation is based on Craig Reynolds' 1986 **boids** model

# Background: modeling flocks (boids)

That simulation is based on Craig Reynolds' 1986 **boids** model  
(*boi*d is short for *bird-oid object*)

# Background: modeling flocks (boids)

That simulation is based on Craig Reynolds' 1986 **boids** model

(*boi*d is short for *bird-oid object*)

Three conditions enforced on individuals:



# Background: modeling flocks (boids)

That simulation is based on Craig Reynolds' 1986 **boids** model

(*boi*d is short for *bird-oid object*)

Three conditions enforced on individuals:

(a) **alignment**\* – tendency to align parallel to avg. of proximal boids

# Background: modeling flocks (boids)

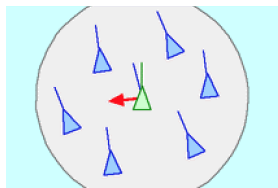
That simulation is based on Craig Reynolds' 1986 **boids** model

(*boid* is short for *bird-oid object*)

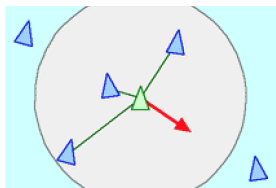
Three conditions enforced on individuals:

- (a) **alignment\*** – tendency to align parallel to avg. of proximal boids
- (b) **separation** – tendency to avoid collision
- (c) **cohesion** – tendency to avoid divergence

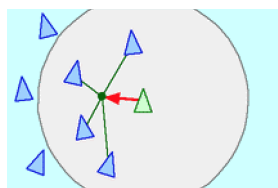
# Background: modeling flocks (boids)



(a) alignment\*



(b) separation



(c) cohesion

IMAGE SOURCE: C. REYNOLDS

# A model for global (b) & (c)

We'll see here a model that implements separation and cohesion

We'll then prove some results about it

Model due to Gazi & Passino, "Stability Analysis in Swarms," 2003

# A model for global (b) & (c)

$M$  agents, indexed by  $i \in [M]$ , with positions  $\mathbf{x}^i \in \mathbf{R}^n$   
 $\mathbf{x}^i$  is implicitly  $\mathbf{x}^i(t)$ ; the time-parameter is suppressed for convenience  
All agents always know all others' positions, velocities with no delay  
Agents are governed by the first-derivative relation

$$\dot{\mathbf{x}}^i = \sum_{j=1, \neq i}^M g(\mathbf{x}^i - \mathbf{x}^j)$$

where  $g : \mathbf{R}^n \rightarrow \mathbf{R}^n$  implements pairwise separation/cohesion

# The artificial social potential function $g$

We will consider the **artificial social potential function**

$$g(\mathbf{x}) := - \left( a - b \exp \left( - \frac{\|\mathbf{x}\|^2}{c} \right) \right) \mathbf{x}$$

# Our first result about this model

Define the **center** in the obvious way:

$$\bar{\mathbf{x}} := \frac{1}{M} \sum_{i=1}^M \mathbf{x}^i$$

and the  $i$ th **offset** as

$$\mathbf{e}^i := \mathbf{x}^i - \bar{\mathbf{x}}$$

## Lemma 1

$\bar{\mathbf{x}}$  is stationary.

# Our first result about this model

## Lemma 1

$\bar{\mathbf{x}}$  is stationary.

Proof.

$$\dot{\bar{\mathbf{x}}} = \frac{1}{M} \sum_{i=1}^M \dot{\mathbf{x}}^i = -\frac{1}{M} \sum_{i=1}^M \sum_{j=1, \neq i}^M \mathbf{v}_{ij} g_1(\mathbf{v}_{ij}) = 0$$

where  $\mathbf{v}_{ij} := \mathbf{x}^i - \mathbf{x}^j$  and  $g_1(\mathbf{v}) := \frac{\langle g(\mathbf{v}), \mathbf{v} \rangle}{\|\mathbf{v}\|^2}$ . ■



# Lyapunov function candidates

Given

$$\dot{x} = f(x) \tag{1}$$

for  $f : \mathbf{R}^n \rightarrow \mathbf{R}^n$ , consider  $V \in \mathcal{C}^1(\mathbf{R}^n \rightarrow \mathbf{R})$  locally PSD on  $U \subset \mathbf{R}^n$

# Lyapunov function candidates

Given

$$\dot{x} = f(x) \tag{1}$$

for  $f : \mathbf{R}^n \rightarrow \mathbf{R}^n$ , consider  $V \in \mathcal{C}^1(\mathbf{R}^n \rightarrow \mathbf{R})$  locally PSD on  $U \subset \mathbf{R}^n$   
If  $\nabla V \cdot g$  is NSD then the sol'n to (1) is asymptotically stable in  $U$

# Lyapunov function candidates

Given

$$\dot{x} = f(x) \tag{1}$$

for  $f : \mathbf{R}^n \rightarrow \mathbf{R}^n$ , consider  $V \in \mathcal{C}^1(\mathbf{R}^n \rightarrow \mathbf{R})$  locally PSD on  $U \subset \mathbf{R}^n$   
If  $\nabla V \cdot g$  is NSD then the sol'n to (1) is asymptotically stable in  $U$   
 $V$  is a **Lyapunov function candidate** (LFC)

Attraction and repulsion balance at

$$\delta := \sqrt{c \ln \frac{b}{a}}$$

# Free agents

Attraction and repulsion balance at

$$\delta := \sqrt{c \ln \frac{b}{a}}$$

Definition (free agent)

A **free agent at time**  $t$  is an agent at distance more than  $\delta$  from every other agent.

# Free agents

Attraction and repulsion balance at

$$\delta := \sqrt{c \ln \frac{b}{a}}$$

## Definition (free agent)

A **free agent at time  $t$**  is an agent at distance more than  $\delta$  from every other agent.

## Lemma 2

Free agent  $i$  at time  $t$  that is also farther than  $\delta$  from  $\bar{\mathbf{x}}$  has instantaneous motion towards  $\bar{\mathbf{x}}$ .

## Lemma 2

Free agent  $i$  at time  $t$  that is also farther than  $\delta$  from  $\bar{\mathbf{x}}$  has instantaneous motion towards  $\bar{\mathbf{x}}$ .

## Proof.

Manipulation of def. of  $\bar{\mathbf{x}}$  gives

$$\dot{\mathbf{x}}^i = -aM\mathbf{e}^i + b \sum_{j=1, \neq i}^M \exp\left(-\frac{\|\mathbf{v}_{ij}\|^2}{c}\right) \mathbf{v}_{ij}$$

Using LFC  $V_i := \frac{1}{2}\|\mathbf{e}^i\|^2$ , bound

$$\dot{V}_i \leq -aM\|\mathbf{e}^i\|^2 + b\|\mathbf{e}^i\| \sum_{j=1, \neq i}^M \exp\left(-\frac{\|\mathbf{v}_{ij}\|^2}{c}\right) \|\mathbf{v}_{ij}\|$$

*cont. on next slide*

## Lemma 2

Free agent  $i$  at time  $t$  that is also farther than  $\delta$  from  $\bar{\mathbf{x}}$  has instantaneous motion towards  $\bar{\mathbf{x}}$ .

Proof.

Using  $\|\mathbf{v}_{ij}\| > \delta$  to bound the exponential term, we have

$$\dot{V}_i \leq -a\|\mathbf{e}^i\|^2 - (M-1) \left( a\|\mathbf{e}^i\| - b\delta \exp\left(-\frac{\delta^2}{c}\right) \right) \|\mathbf{e}^i\|$$

where the second term is NSD by choice of  $\delta$  and assumption that  $\|\mathbf{e}^i\| \geq \delta$ ; the inequality simplifies to

$$\dot{V}_i \leq -a\|\mathbf{e}^i\|^2 = -2aV_i.$$

i.e. we see that  $\dot{\mathbf{x}}^i$  is in a direction such that  $\|\mathbf{e}^i\|$  decreases. ■



# Convergence?

What can we say about convergence of agents in the swarm?

# Convergence?

What can we say about convergence of agents in the swarm?

We *cannot* say that each agent converges to  $\bar{\mathbf{x}}$

# Convergence?

What can we say about convergence of agents in the swarm?

We *cannot* say that each agent converges to  $\bar{\mathbf{x}}$

Instead, define

$$\varepsilon := \frac{b}{a} \sqrt{\frac{c}{2e}}$$

and

$$\hat{t} := -\frac{\ln \varepsilon}{a} + \frac{1}{a} \ln \left( 2 \max_{i \in [M]} V_i(0) \right)$$

# Convergence?

What can we say about convergence of agents in the swarm?

We *cannot* say that each agent converges to  $\bar{\mathbf{x}}$

Instead, define

$$\varepsilon := \frac{b}{a} \sqrt{\frac{c}{2e}}$$

and

$$\hat{t} := -\frac{\ln \varepsilon}{a} + \frac{1}{a} \ln \left( 2 \max_{i \in [M]} V_i(0) \right)$$

## Theorem 1

Each agent in the swarm will lie inside  $B_\varepsilon(\bar{\mathbf{x}})$  by time  $\hat{t}$ .

# Convergence?

## Theorem 1

Each agent in the swarm will lie inside  $B_\varepsilon(\bar{\mathbf{x}})$  by time  $\hat{t}$ .

## Proof.

Fix any  $i \in [M]$ . Recall:

$$\dot{V}_i \leq -aM\|\mathbf{e}^i\|^2 + b\|\mathbf{e}^i\| \sum_{j=1, \neq i}^M \exp\left(-\frac{\|\mathbf{v}_{ij}\|^2}{c}\right) \|\mathbf{v}_{ij}\|$$

The linear-exponential term obtains its maximum  $\sqrt{\frac{c}{2e}}$  at  $\|\mathbf{v}_{ij}\| = \sqrt{\frac{c}{2}}$  so  $\dot{V}_i < 0$  if

$$\|\mathbf{e}^i\| > \frac{b(M-1)}{aM} \sqrt{\frac{c}{2e}} =: \varepsilon'$$

*cont. on next slide*

# Convergence?

## Theorem 1

Each agent in the swarm will lie inside  $B_\varepsilon(\bar{\mathbf{x}})$  by time  $\hat{t}$ .

## Proof.

Note that  $\frac{M}{M-1}\varepsilon' = \varepsilon$ , i.e.  $\varepsilon' < \varepsilon$ . Therefore  $\mathbf{x}^i$  will not converge to anywhere outside  $B_{\varepsilon'}(\bar{\mathbf{x}}) \subset B_\varepsilon(\bar{\mathbf{x}})$ .

To compute finite-time convergence, recall:

$$\dot{V}_i \leq -2aV_i \implies V_i(t) \leq V_i(0)e^{-2at}$$

so  $\|\mathbf{e}^i\| = \varepsilon$  by time

$$t_i \leq -\frac{1}{2a} \ln \left( \frac{\varepsilon^2}{2V_i(0)} \right)$$

which simplifies to the expression for  $\hat{t}$ . ■

# Convergence!

Despite issues with repulsion and mutual convergence, we can still characterize the asymptotic convergence of the swarm

# Convergence!

Despite issues with repulsion and mutual convergence, we can still characterize the asymptotic convergence of the swarm

Let  $\Omega$  be the invariant set of equilibrium points, i.e. states  $\vec{x}$  with  $\dot{\vec{x}} = 0$



# Convergence!

Despite issues with repulsion and mutual convergence, we can still characterize the asymptotic convergence of the swarm

Let  $\Omega$  be the invariant set of equilibrium points, i.e. states  $\vec{x}$  with  $\dot{\vec{x}} = 0$

## Theorem 2

The swarm converges, as  $t \rightarrow \infty$ , to a value in  $\Omega$ .

i.e. the swarm converges to a constant state

# Convergence!

## Theorem 2

The swarm converges, as  $t \rightarrow \infty$ , to a value in  $\Omega$ .

## Proof.

Consider the LFC (also an artificial potential function)

$$J(\vec{\mathbf{x}}) = \frac{1}{2} \sum_{i=1}^{M-1} \sum_{j=i+1}^M \left( a \|\mathbf{v}_{ij}\|^2 + bc \exp\left(-\frac{\|\mathbf{v}_{ij}\|^2}{c}\right) \right)$$

Since  $\nabla_{\mathbf{x}^i} J(\vec{\mathbf{x}}) = -\dot{\mathbf{x}}^i$ , at all times  $t$ :

$$\dot{J}(\vec{\mathbf{x}}) = - \sum_{i=1}^M \|\dot{\mathbf{x}}^i\|^2 \leq 0$$

By **LaSalle's invariance principle**,  $\vec{\mathbf{x}} \rightarrow \{\vec{\mathbf{x}} : \dot{J}(\vec{\mathbf{x}}) = 0\}$ . ■

- \* V. Gazi and K. M. Passino. “Stability Analysis of Swarms.” In *IEEE Transactions on Automatic Control* 48.4 (2003).
- \* C. Reynolds. “Flocks, Herds, and Schools: A Distributed Behavioral Model.” In *Computer Graphics* 21.4 (1987).