

The Emergent Flocking Dynamics of Boids

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1 Introduction

This [simulation](#) of boids models how groups such as birds or fish move together in tightly coordinated groups using only simple local rules. We present a visual treatment of Craig Reynolds's boids model by simulating its three core behavioral rules: separation, alignment, and cohesion.

By integrating these forces numerically, each boid adjusts its velocity based only on looking at its immediate local surroundings. However, the superposition of all these decisions made locally produces the fluid-like motion of a flock, and is an example of an emergent behavior arising from simple components.

2 Computational Methods

At every time step, each boid obeys the following process:

1. Identify its neighborhood $\mathcal{N}_i = \{j : \|\mathbf{p}_j - \mathbf{p}_i\| \leq R\}$.
2. Compute three steering vectors based on alignment, cohesion, and separation.
3. Weight and sum these vectors to obtain the net acceleration \mathbf{a}_i .
4. Update velocity and position:

$$\mathbf{v}_i \leftarrow \mathbf{v}_i + \Delta t \mathbf{a}_i, \quad \mathbf{p}_i \leftarrow \mathbf{p}_i + \Delta t \mathbf{v}_i.$$

With this in mind, we now look at each steering component in greater detail:

Alignment

Each boid adjusts its velocity (both speed & direction) to match the average heading and speed of its neighbors. Within a perception radius R , define the neighbor set

$$\mathcal{N}_i = \{j : \|\mathbf{p}_j - \mathbf{p}_i\| \leq R\}, \quad N_i = |\mathcal{N}_i|.$$

Compute the average velocity of each boid's neighbors

$$\mathbf{v}_{\text{avg}} = \frac{1}{N_i} \sum_{j \in \mathcal{N}_i} \mathbf{v}_j.$$

The raw alignment steering force is then

$$\mathbf{f}_{\text{align}}^{\text{raw}} = \begin{cases} w_{\text{aln}} \left(v_{\text{max}} \frac{\mathbf{v}_{\text{avg}}}{\|\mathbf{v}_{\text{avg}}\|} - \mathbf{v}_i \right), & N_i > 0, \\ 0, & N_i = 0, \end{cases}$$

where w_{aln} scales the magnitude of the influence of alignment (set with sliders within the simulation) and v_{max} puts on a defined cruising speed while also capping boids to prevent them from accelerating without bound. This also drives each boid toward the averaged heading of its local flock.

Cohesion

Each boid applies an attractive force toward the center of mass of its neighbors, preventing the boids from drifting too far apart from their flock. Using the same neighborhood definition

$$\mathcal{N}_i = \{j : \|\mathbf{p}_j - \mathbf{p}_i\| \leq R\}, \quad N_i = |\mathcal{N}_i|,$$

compute the average position of the neighbors:

$$\mathbf{p}_{\text{avg}} = \frac{1}{N_i} \sum_{j \in \mathcal{N}_i} \mathbf{p}_j.$$

The raw cohesion steering force is then

$$\mathbf{f}_{\text{coh}}^{\text{raw}} = \begin{cases} w_{\text{coh}} \left(v_{\text{max}} \frac{\mathbf{p}_{\text{avg}} - \mathbf{p}_i}{\|\mathbf{p}_{\text{avg}} - \mathbf{p}_i\|} - \mathbf{v}_i \right), & N_i > 0, \\ \mathbf{0}, & N_i = 0, \end{cases}$$

where w_{coh} adjusts the strength of cohesion and v_{max} drives the boid's target cruising speed. This term draws each boid toward its local group's center of mass which keeps flock cohesion.

Separation

Each boid generates a repulsive force to avoid crowding/colliding and in that way steer away from neighbors that are too close. Again, same neighborhood definition with the same perception radius R

$$\mathcal{N}_i = \{j : \|\mathbf{p}_j - \mathbf{p}_i\| \leq R\}, \quad N_i = |\mathcal{N}_i|.$$

Now compute the unnormalized separation vector

$$\mathbf{s} = \sum_{j \in \mathcal{N}_i} \frac{\mathbf{p}_i - \mathbf{p}_j}{\|\mathbf{p}_i - \mathbf{p}_j\|}.$$

The raw separation steering force is then

$$\mathbf{f}_{\text{sep}}^{\text{raw}} = \begin{cases} w_{\text{sep}} \left(v_{\text{max}} \frac{\mathbf{s}/N_i}{\|\mathbf{s}/N_i\|} - \mathbf{v}_i \right), & N_i > 0, \\ \mathbf{0}, & N_i = 0, \end{cases}$$

where w_{sep} controls the repulsion strength and v_{max} again defines a consistent cruising speed. This force pushes boids away from close neighbors which is what helps them to prevent collisions and maintain space.

3 Results

3.1 Qualitative Observations

Figure 1 shows snapshots of the boid flock (on default setting $w = 1$) at three time points ($t = 0, 50$, and 100). We observe the eventual becoming of a cohesive travelling flock, transient state behavior¹, and cluster fragmentation under different weights of the steering parameters (alignment, cohesion, separation).

¹<https://pmc.ncbi.nlm.nih.gov/articles/PMC5686407/>

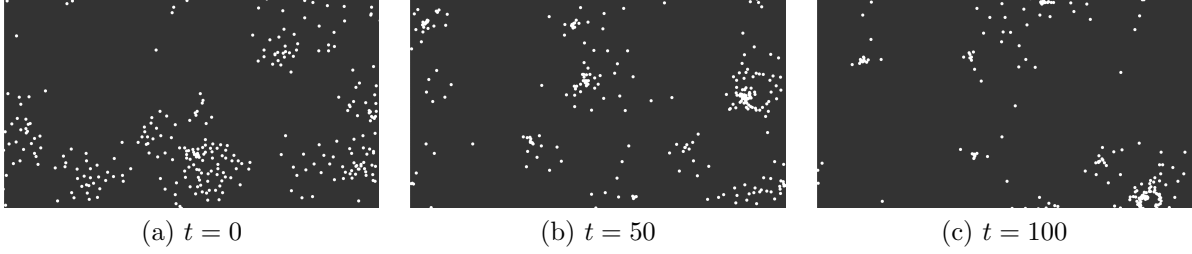


Figure 1: Simulation snapshots at different time steps with initial random positions (a), intermediate mixing (b), and eventual flocking (c).

3.2 Quantitative Metrics & Comparison

In order to compare how the three steering weights/parameters w_{aln} , w_{coh} , w_{sep} affect emergent flock behavior, we run simulations with combinations of representative parameters for each scenario. Table 1 indicates the observations made at $t = 100$ for each parameter setting.

Case	w_{aln}	w_{coh}	w_{sep}	Observed Behavior
A: Cohesive flock	1.5	1.5	1.0	Single, unified flock traveling coherently
B: Transient mixing	2.0	1.0	0.5	Vortex-like behavior, individual groups rejoining main flock
C: Fragmentation	0.5	0.5	1.5	Moving lattice, crystalline structure
D: Disorder	0.5	0.5	0.5	No coherent structure, random Brownian-like motion
E: Tight flock	2.0	2.0	0.5	Very compacted flocks, minimal internal spacing

Table 1: Behaviors at $t = 100$ under different steering parameters.

3.3 Discussion

The parameter sweep in Table 1 shows clear trends in how the (relative) behaviors or strengths of alignment, cohesion, and separation drive emergent flock behavior:

Case A: Cohesive flock

With relatively balanced parameters ($w_{\text{aln}} = w_{\text{coh}} = 1.5$, $w_{\text{sep}} = 1.0$), boids form a single, "normal" traveling flock that most closely replicates what people would usually see in real life.

Case B: Transient mixing

A high weight for the alignment parameter ($w_{\text{aln}} = 2.0$) with moderate cohesion and weak separation leads to vortex-like mixing before they eventually (slowly) reconverge into the main flock; the over-emphasis on matching neighbors' velocities causes the boids to overshoot.

Case C: Fragmentation

However, if we lower both alignment and cohesion while increasing separation ($w_{\text{sep}} = 1.5$), we find that this leads to a vibrating lattice structure. The repulsion is the bigger factor over attraction, so boids space apart at the cost of unified coherence.

Case D: Disorder

When all weights are weak ($w_{\text{aln}} = w_{\text{coh}} = w_{\text{sep}} = 0.5$), there is no single behavior that dominates. The result is random motion with minimal emergent structure. Hence, there needs to be sufficiently strong local rules in order to overcome the initial randomness.

Case E: Tight flock

Increasing both alignment and cohesion significantly ($w_{\text{aln}} = w_{\text{coh}} = 2.0$) with low separation produces a compact flock with boids crowding closely.

4 Conclusions

In this simulation, the interactions of three simple, local steering rules led to a variety of complex, lifelike emergent group behaviors. The boids are effectively decentralized: no single one of them has a "global" view, but still coherent patterns arise naturally through local feedback within each community of boids.