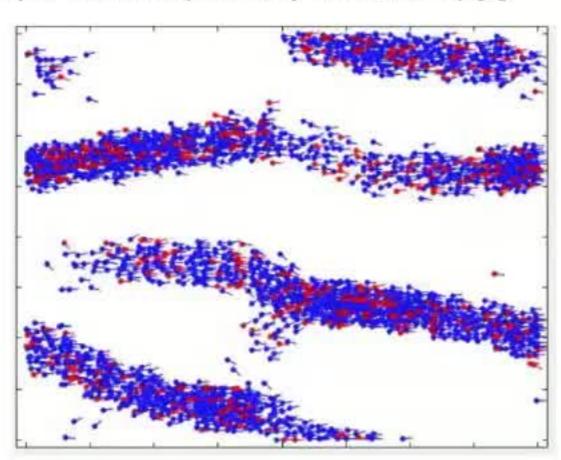
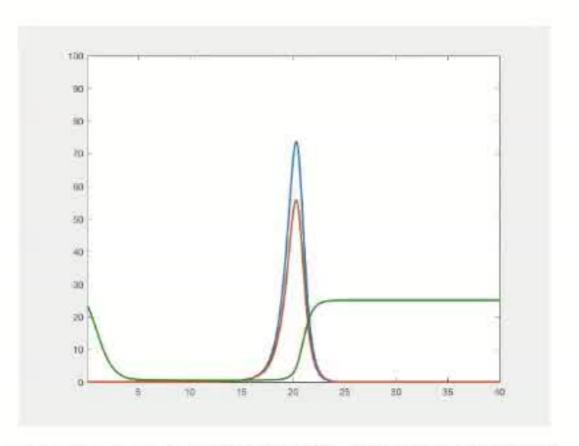
Agent-Based and Continuous Models of Locust Hopper Bands: Insights Gained Through the Lens of Dynamical Systems

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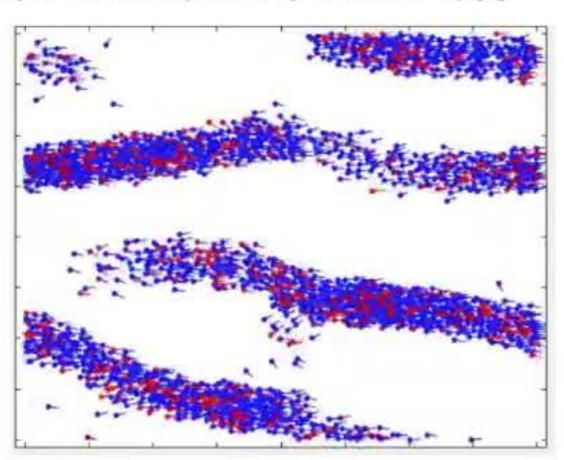




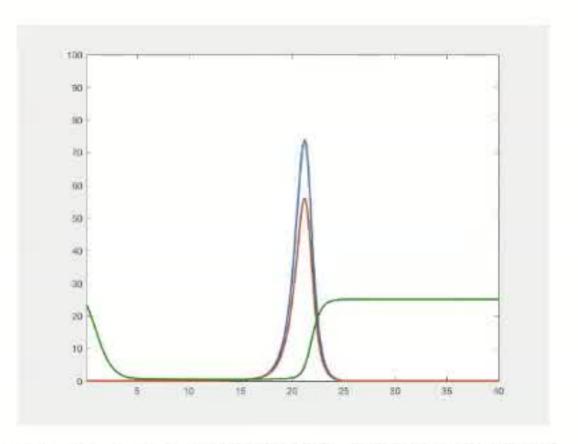
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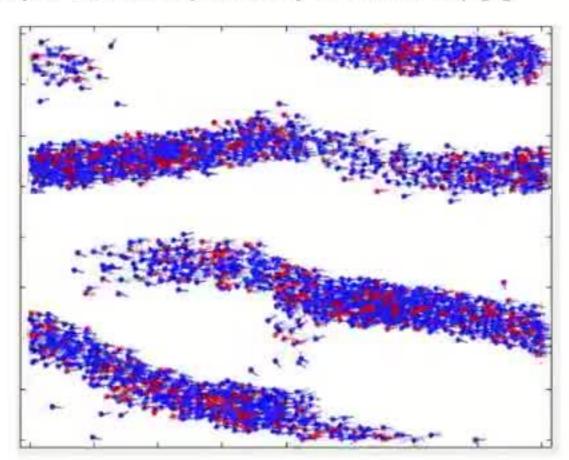




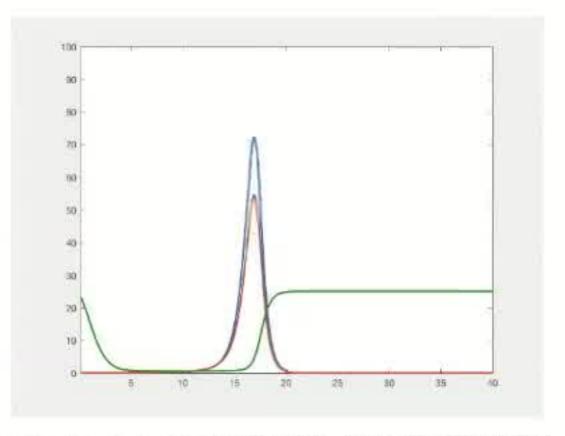
[2] Supported as part of an AMS MRC - NSF DMS 1641020

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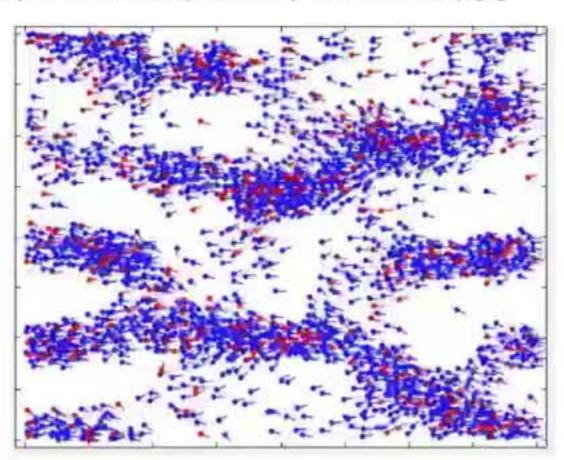




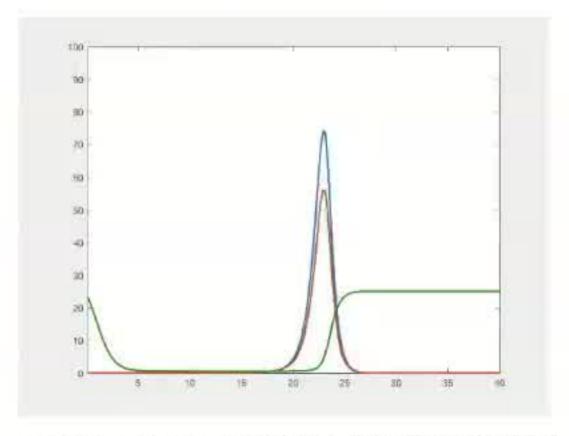


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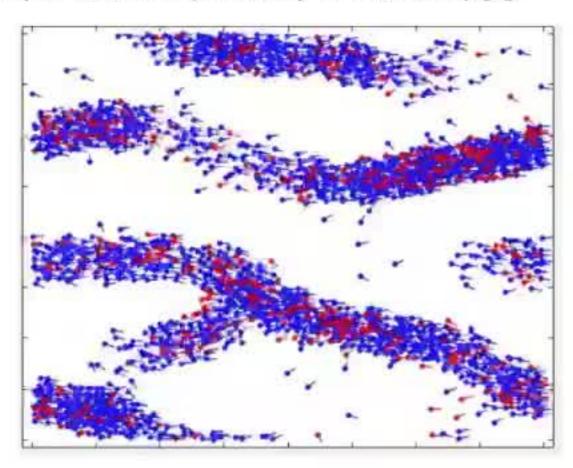




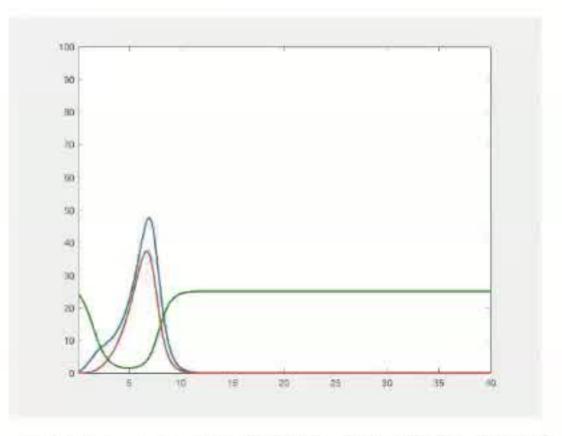


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Columnar Structure

Desert Locust Schistocera gregaria

Planar Front

Australian Plague Locust (Chortoicetes terminifera)

Model I: Exploring Columnar Structure





Model I: Exploring Columnar Structure





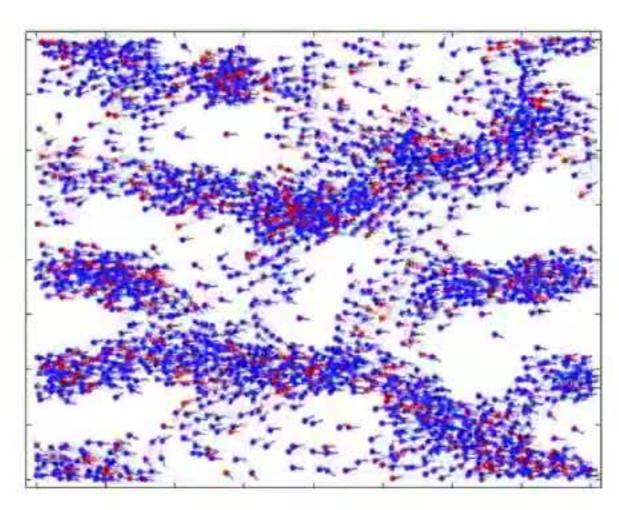
Agent- Based Model (ABM)



When Locusts Swarm

Kit Yates (Bath) & Jerome Buhl (Adelaide)

https://vimeo.com/144014400



Experiments/Observation



Agent-Based Model

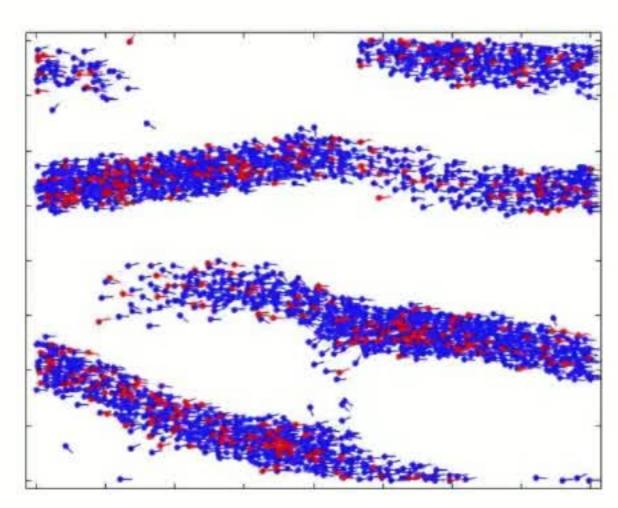
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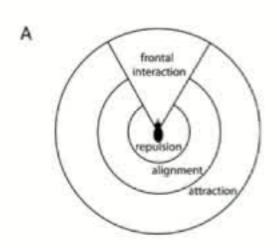


Experiments/Observation

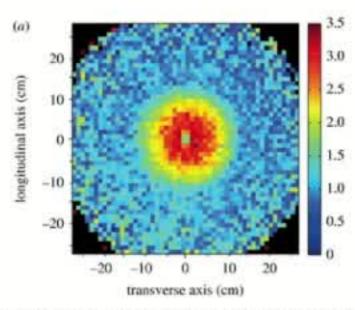


Agent-Based Model

Locust ABM - A Selective Review



Lukeman, Li & Edelstein-Keshet (2010)



Relative density of neighbors around a focal individual.

Buhl, Sword & Simpson (2012)

Three Zone Models:

- Lukeman, R., Li, Y.-X., & Edelstein-Keshet, L. (2010). Inferring individual rules from collective behavior. Proc. Natl. Acad. Sci. USA, 107(28), 12576–12580.
- Couzin, I., Krause, J., James, R., Ruxton, G., & Franks, N. (2002). Collective memory and spatial sorting in animal groups. J. Theor. Biol., 218, 1–11.
- J. M. Miller, A. Kolpas, J. P. J. Neto, and L. F. Rossi, A continuum three-zone model for swarms, Bulletin of mathematical biology, 74 (2012), pp. 536–61.
- Aoki, I. (1982). A simulation study on the schooling mechanism in fish. Bull. Jap. Soc. Sci. Fish. 48, 1081–1088.
- . Reynolds, C. W. (1987). Flocks, herds and schools: a distributed behavioral model. Comput. Graph. 21, 25-34.
- Huth, A. & Wissel, C. (1992). The simulation of the movement of fish schools. J. Theor. Biol. 156, 365–385.

Orientation Models:

- Vicsek, T., Czirók, A., Ben-Jacob, E., Cohen, I., & Shochet, O. (1995). Novel type of phase transition in a system of self-driven particles. Physical review letters, 75(6), 1226.
- Chaté, Hugues, et al. 'Modeling collective motion: variations on the Vicsek model.' The European Physical Journal B 64.3-4 (2008): 451-456.
- Grégoire, Guillaume, Hugues Chaté, and Yuhai Tu. "Moving and staying together without a leader." Physica D: Nonlinear Phenomena 181.3-4 (2003): 157-170.

Pause & Go Models:

- Ariel G, Ophir Y, Levi S, Ben-Jacob E and Ayali A (2014) Individuals' intermittent motion is instrumental to the formation and maintenance of swarms of marching locust nymphs. PLoS ONE 9(7): e101636.
- Nilsen, C., Paige, J., Warner, O., Mayhew, B., Sutley, R., Lam, M., Bernoff, A.J. and Topaz, C.M., 2013. Social aggregation in pea aphids: experiment and random walk modeling. PloS one, 8(12), p.e83343.

Locust ABM:

- Ariel, Gil, and Amir Ayali. "Locust collective motion and its modeling." PLoS computational biology 11.12 (2015): e1004522.
- Buhl, J., Gregory A. Sword, and Stephen J. Simpson. "Using field data to test locust migratory band collective movement models." Interface focus 2.6 (2012): 757-763.

ABM: State Variables

Consider N locusts indexed by the variable i = 1...N

Position: $\vec{x}_i = (x_i, y_i)$

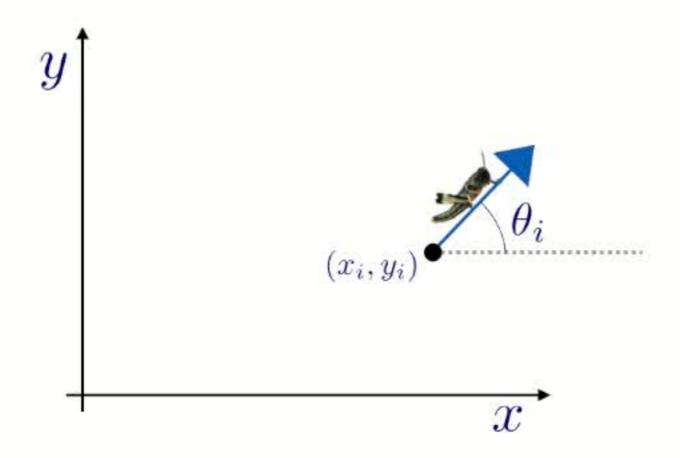
 $\vec{x}_i \in \mathbb{R}^2$

Orientation: θ_i

 $\theta_i \in \mathbb{S}^1$

State (stationary/moving): s_i

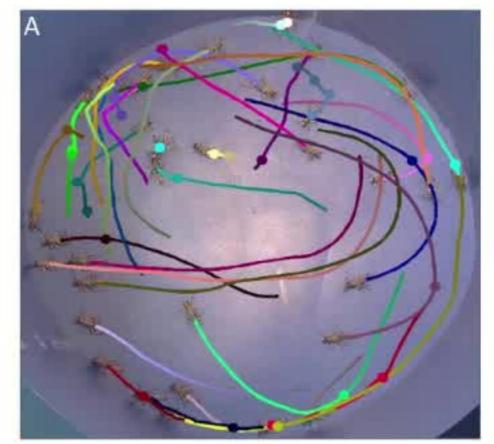
 $s_i \in \{0, 1\}$

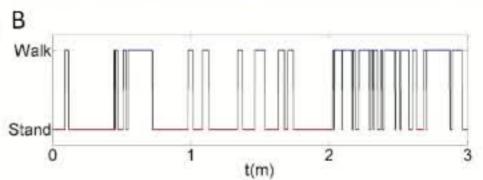


ABM: Pause/Go Motion

Motion tracking of locusts yields:

- Locusts exhibit pause/go motion.
- Pausing is more likely when a locust is blocked.
- Motion is more likely when a locust is nudged from behind.
- Locust speed is roughly constant.





Ariel, Ophir, Levi, Ben-Jacob & Ayali (2014) Individuals' intermittent motion is instrumental to the formation and maintenance of swarms of marching locust nymphs. PLoS ONE 9(7): e101636

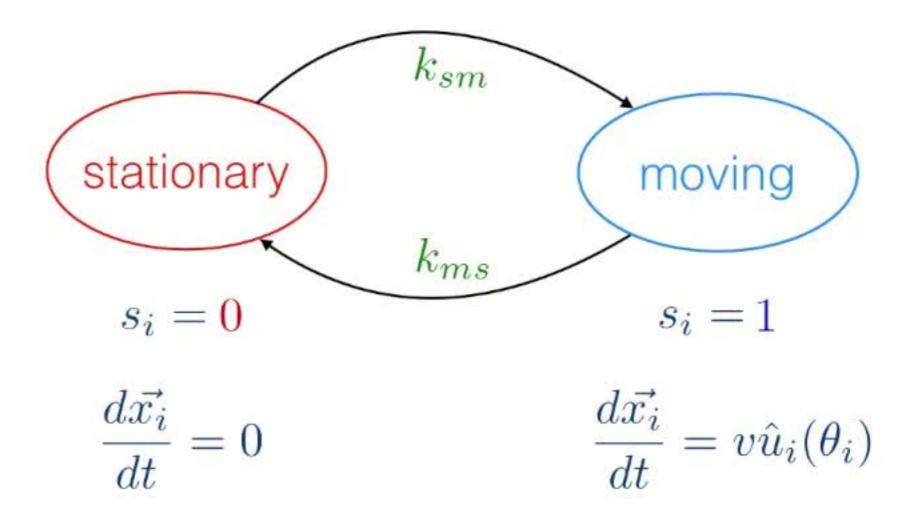
ABM: Pause/Go Motion

$$\frac{d\vec{x_i}}{dt} = s_i \mathbf{v} \hat{u}_i(\theta_i)$$

$$\hat{u}_i(\theta_i) = \langle \cos \theta_i, \sin \theta_i \rangle$$

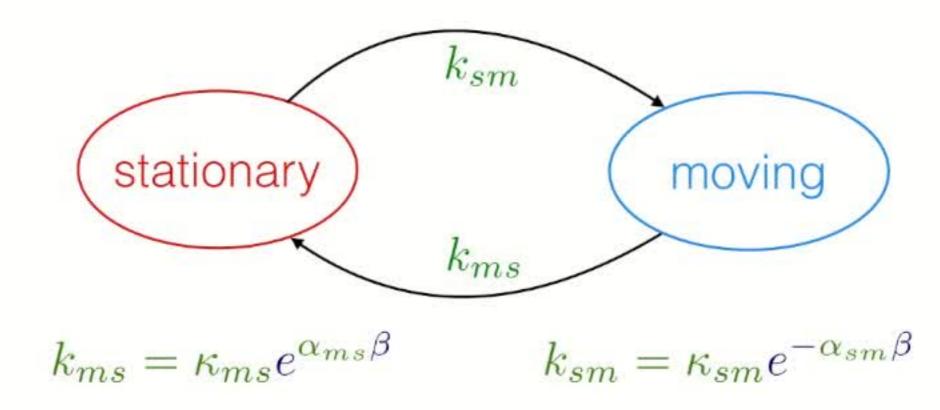
 $\mathbf{v} = \text{speed (constant)}$

Markov process for transition

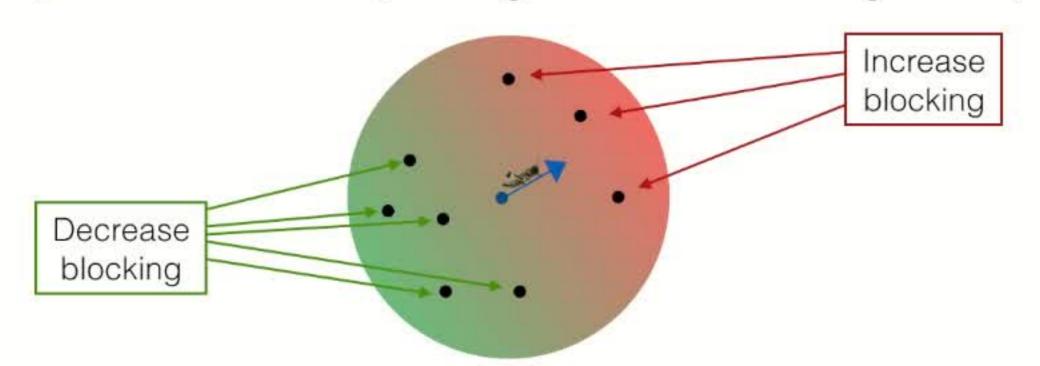


ABM: Pause/Go Motion (blocking)

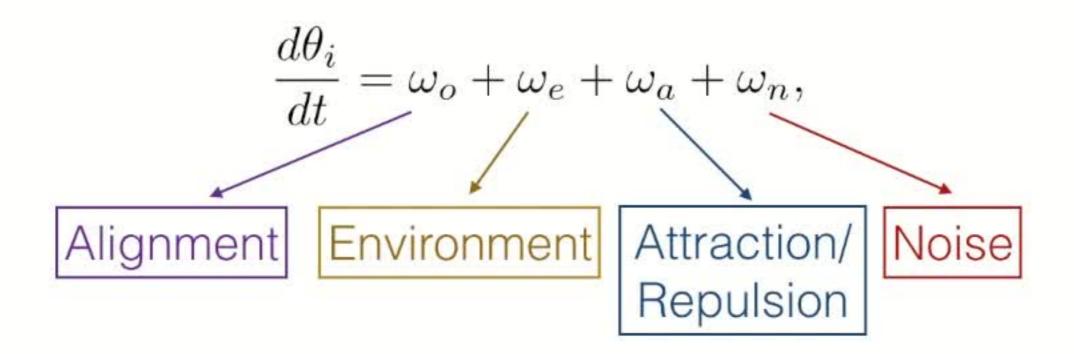
Markov process for transition

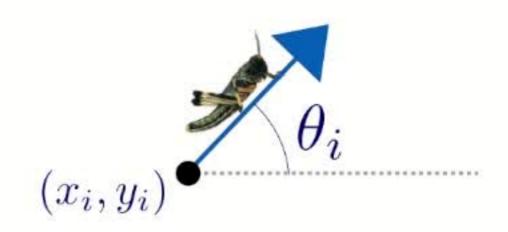


 β = blockscore (a weighted sum of neighbors)



ABM: Orientation





ABM: Orientation (Alignment)

$$\frac{d\theta_i}{dt} = \omega_o + \omega_e + \omega_a + \omega_n,$$

Alignment Environment Attraction/ Noise

Repulsion

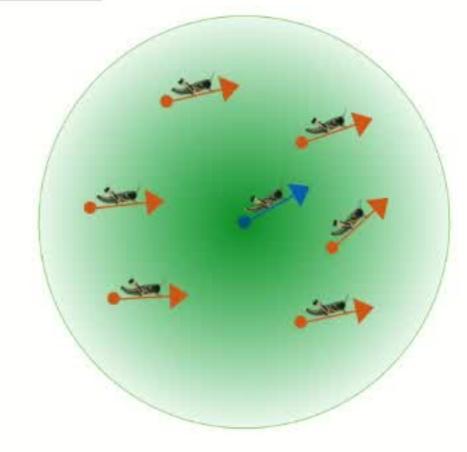
Kumamoto type coupling:

$$\omega_o = \sum_{\substack{j=1\\j\neq i}}^n c_o f_o(d_{ij}) \sin(\theta_j - \theta_i).$$

Weighted by distance, $d_{ij} = |\vec{x}_i - \vec{x}_j|$

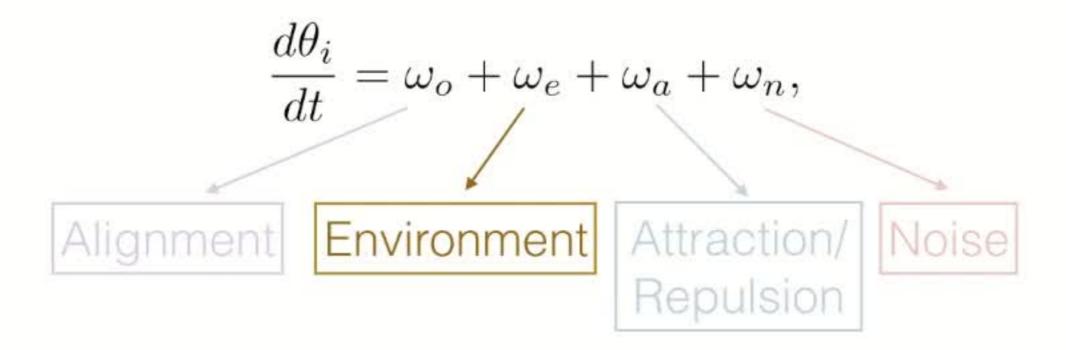
$$f_o(d) = \begin{cases} 1 - d/\ell_o & d < \ell_o \\ 0 & d \ge \ell_o \end{cases}$$

within a finite radius, ℓ_o .



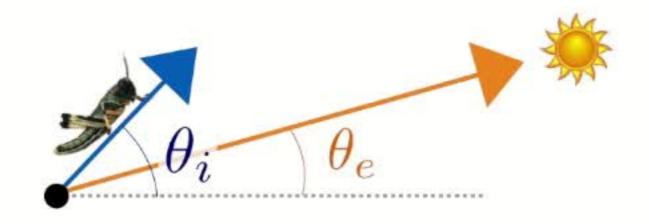


ABM: Orientation (Environment)



Locusts maintain a constant heading; usually downwind.

Environmental bias: $\omega_e = c_e \sin(\theta_e - \theta_i)$



ABM: Orientation (Attraction/Repulsion)

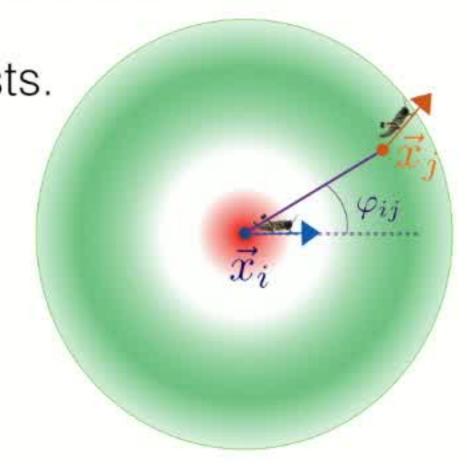
$$\frac{d\theta_i}{dt} = \omega_o + \omega_e + \omega_a + \omega_n,$$
 Alignment Environment Attraction/ Repulsion

Locust turns toward/away the weighted centroid of adjacent locusts.

$$\omega_a = \sum_{\substack{j=1\\j\neq i}}^n \left[\frac{c_r}{f_r} f_r(d_{ij}) - \frac{c_a}{f_a} f_a(d_{ij}) \right] \sin(\phi_{ij}),$$

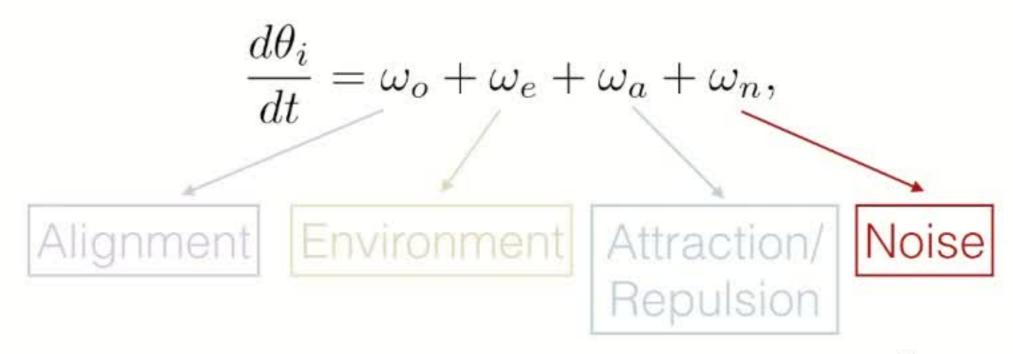
$$f_a(d) = \begin{cases} 1 - \frac{|d - l_w|}{|l_a - l_w|} & \text{for } |d - l_w| < |l_a - l_w| \\ 0 & \text{otherwise} \end{cases}$$

$$f_r(d) = \begin{cases} 1 - \frac{d}{l_r} & \text{for } d < l_r \\ 0 & \text{otherwise} \end{cases}$$

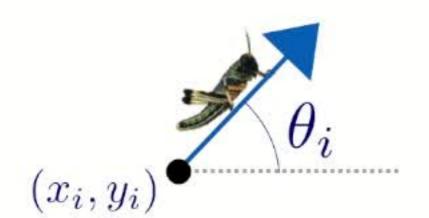


$$d_{ij} = |\vec{x}_i - \vec{x}_j|$$

ABM: Orientation (Noise)



White Noise:
$$\omega_n = \sqrt{c_n} \dot{W}$$



Change in $\theta_i(t)$ is normally distributed

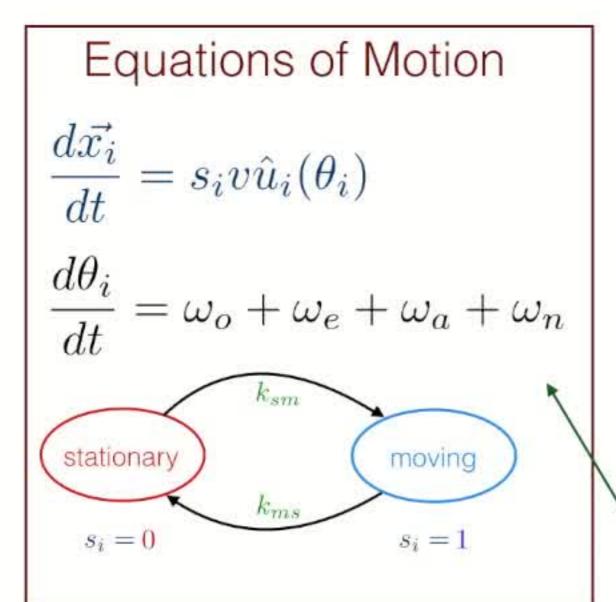
$$\theta_i(t + \Delta t) - \theta_i(t) \sim \mathcal{N}(0, c_n \Delta t)$$

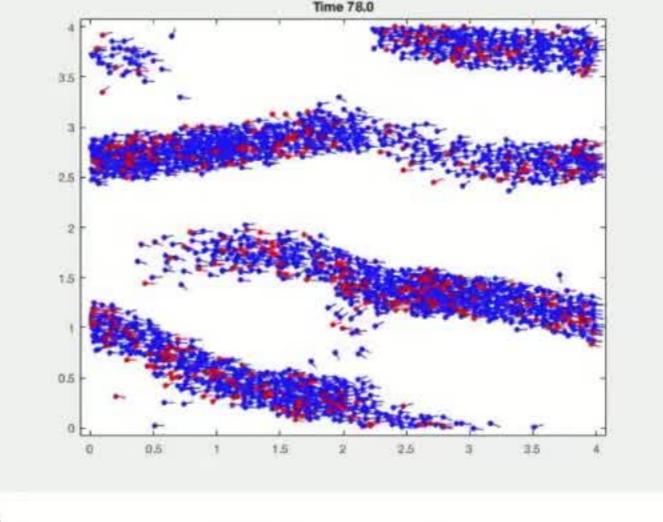
with variance $c_n \Delta t$.

ABM: Numerical Simulation

Consider N locusts indexed by the variable i = 1...N

Position: $\vec{x}_i \in \mathbb{R}^2$ Orientation: $\theta_i \in \mathbb{S}^1$ State: $s_i \in \{0, 1\}$





4000 agents

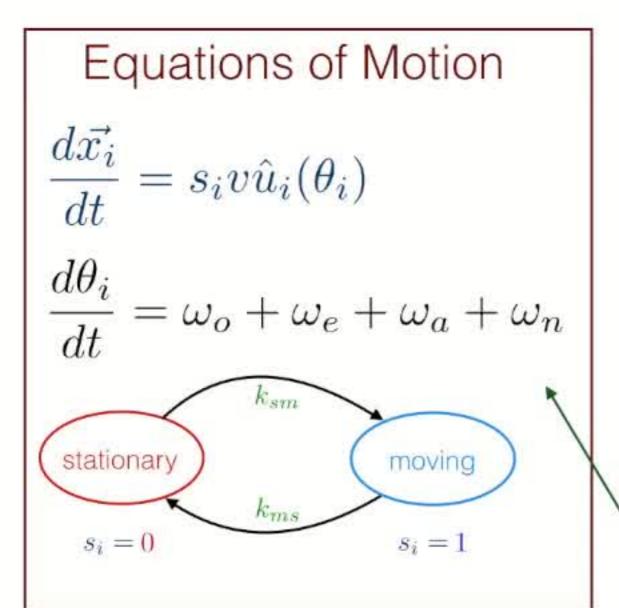
Integrate with a stochastic Modified Euler Method (Honeycutt, 1992)

14 Parameters (estimated from experimental data)

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4000 agents

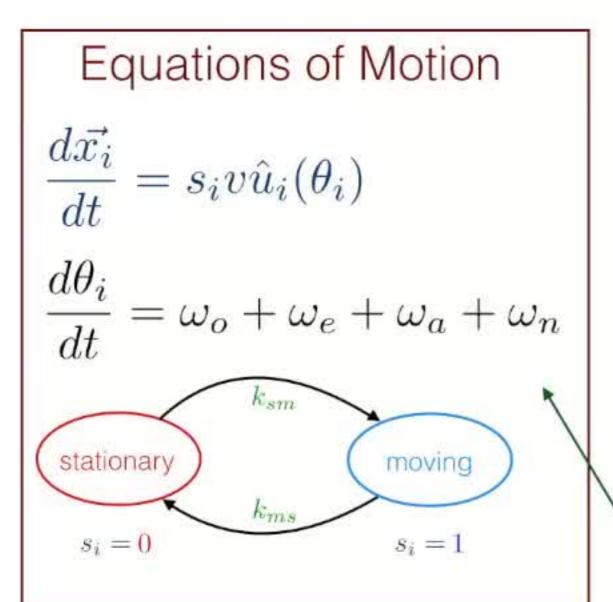
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Consider N locusts indexed by the variable i = 1...N

Position: $\vec{x}_i \in \mathbb{R}^2$ Orientation: $\theta_i \in \mathbb{S}^1$ State: $s_i \in \{0, 1\}$



Time 42.0

3.5

2.5

2

1.5

0 0.5

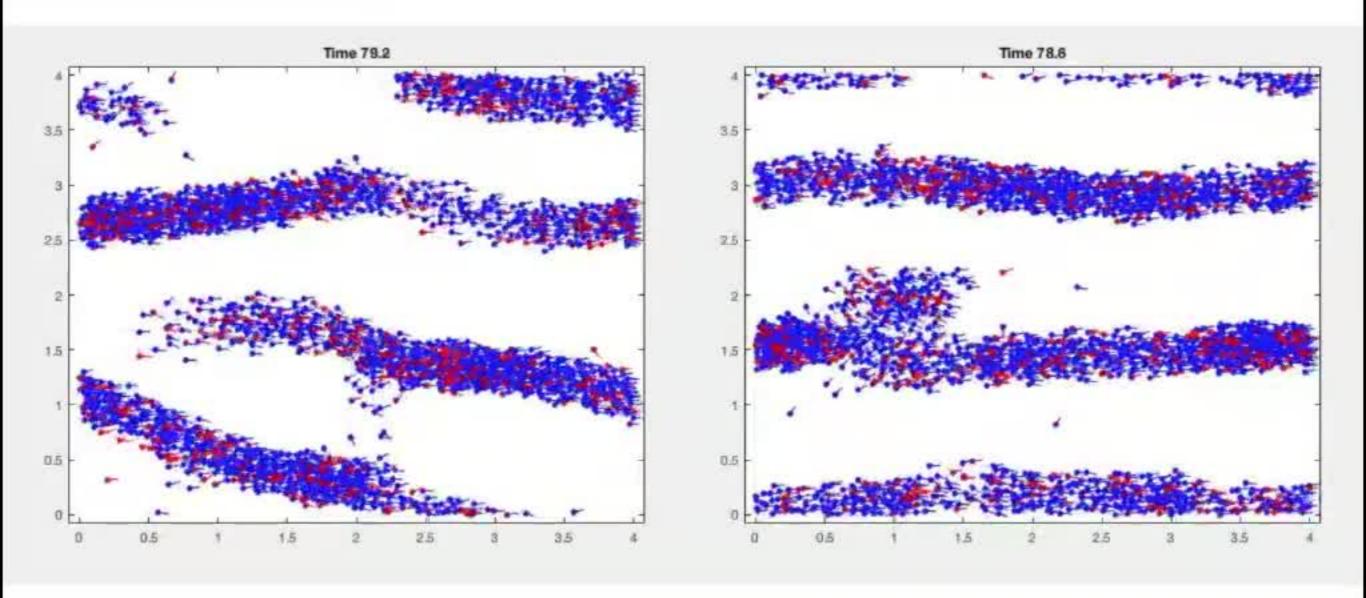
1 15 2 25 3 3.5 4

4000 agents

Integrate with a stochastic Modified Euler Method (Honeycutt, 1992)

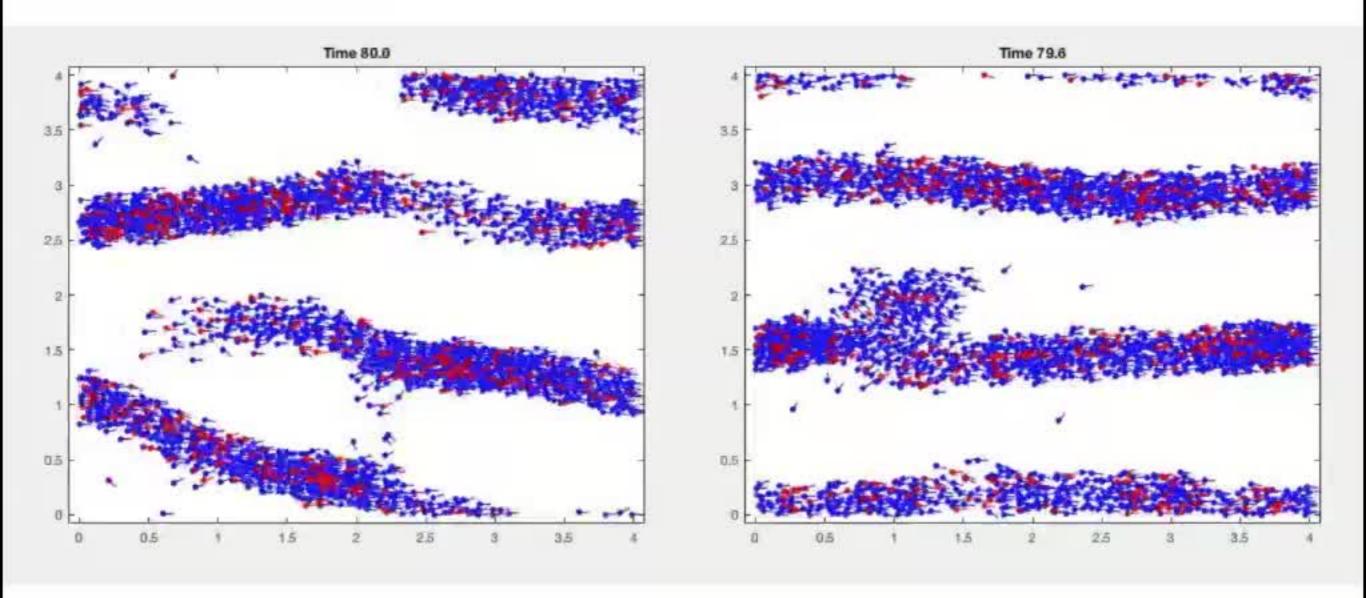
14 Parameters (estimated from experimental data)

Different Initial Conditions



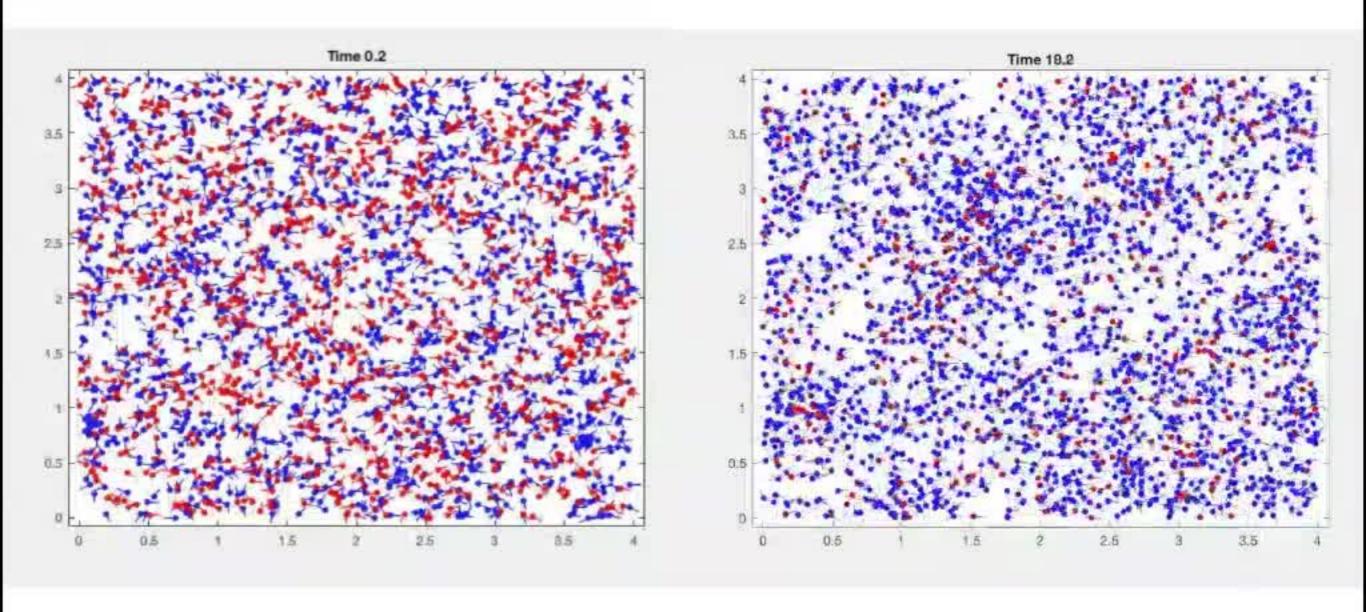
Red = Stationary Blue = Moving

Different Initial Conditions



Red = Stationary Blue = Moving

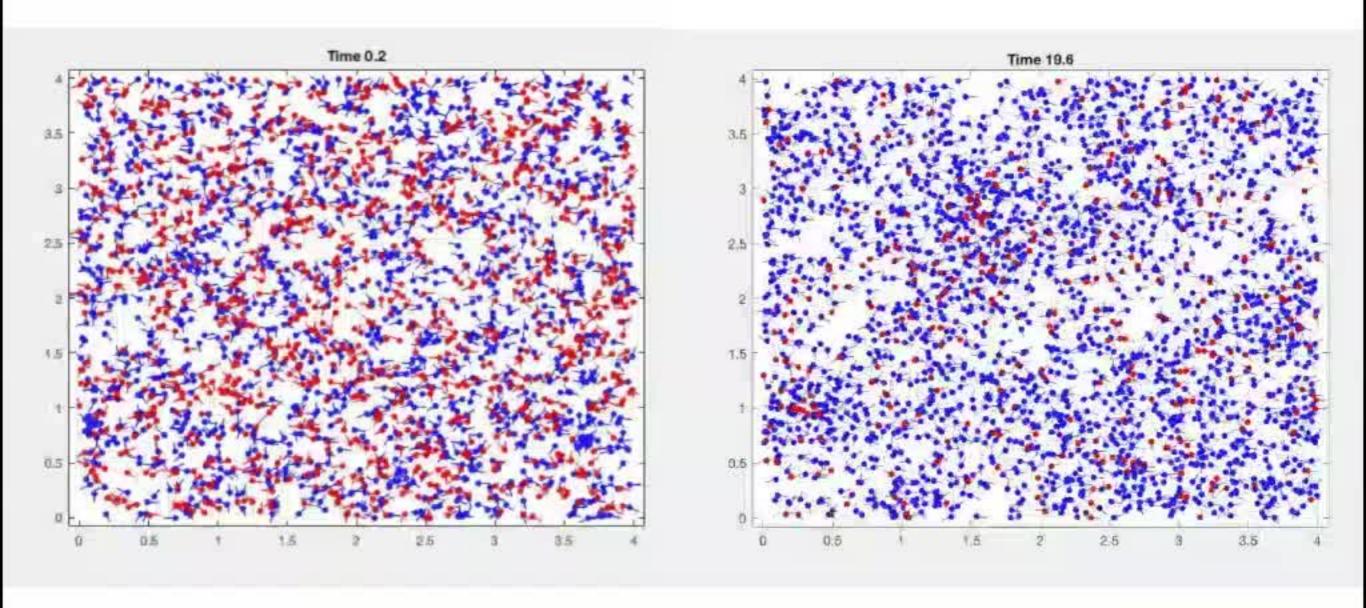
Alignment



Alignment

No Alignment

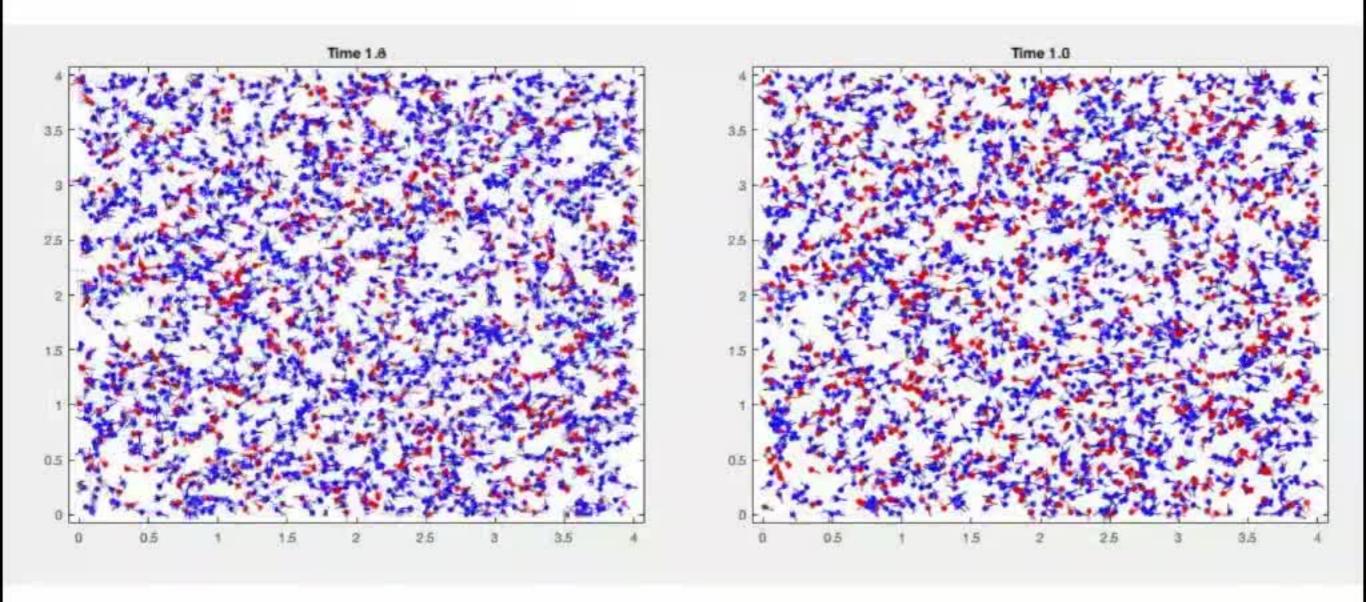
Alignment



Alignment

No Alignment

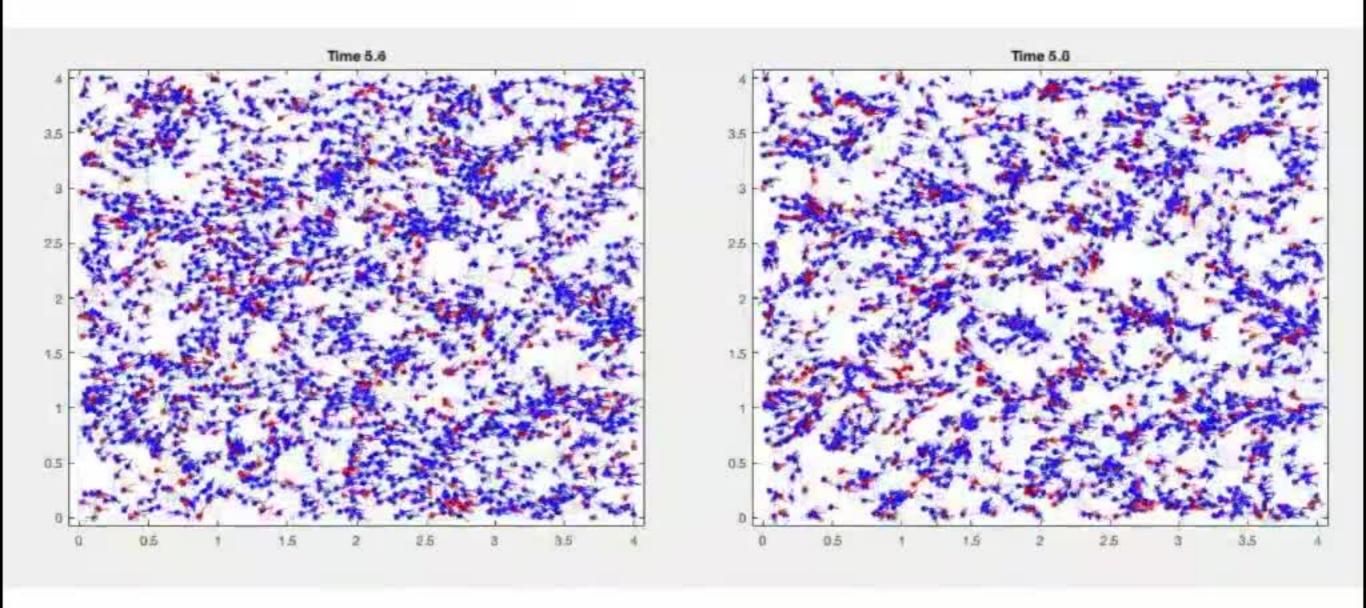
Repulsion



Repulsion

No Repulsion

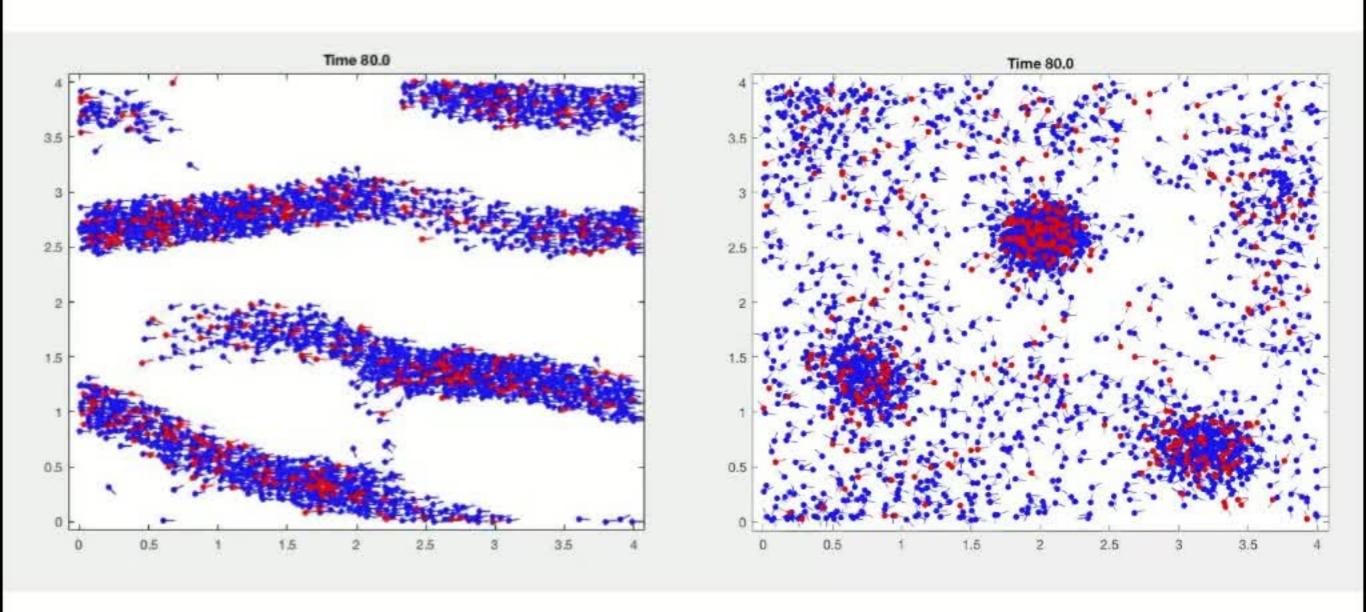
Repulsion



Repulsion

No Repulsion

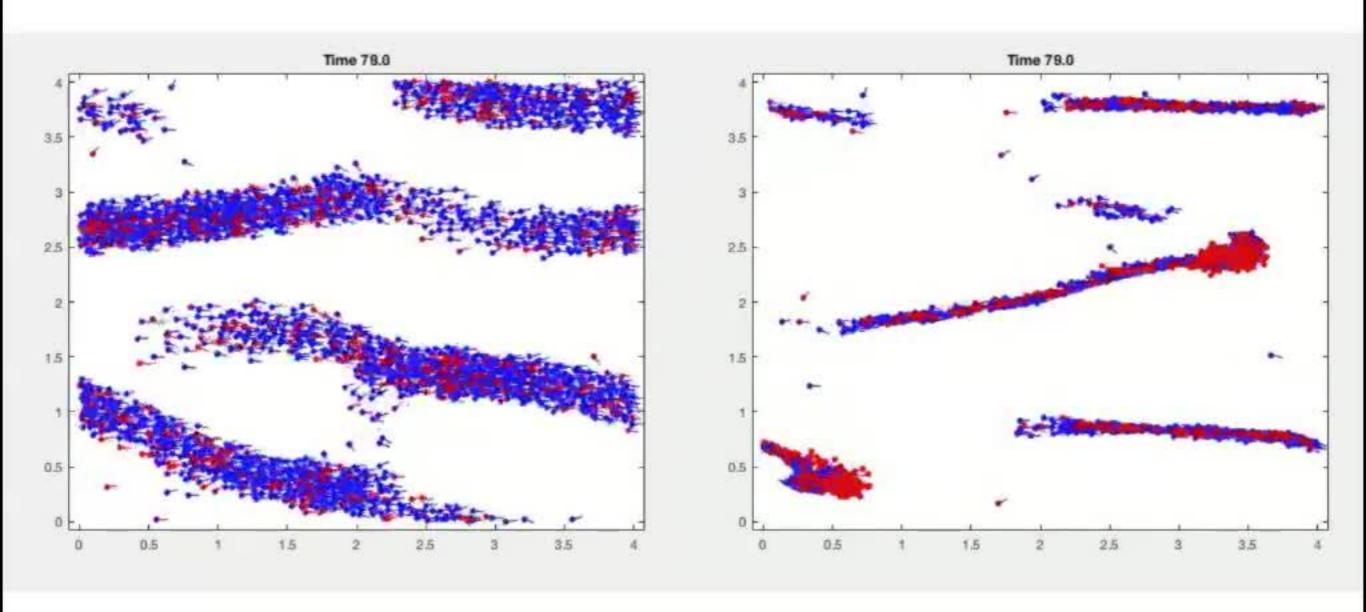
Alignment



Alignment

No Alignment

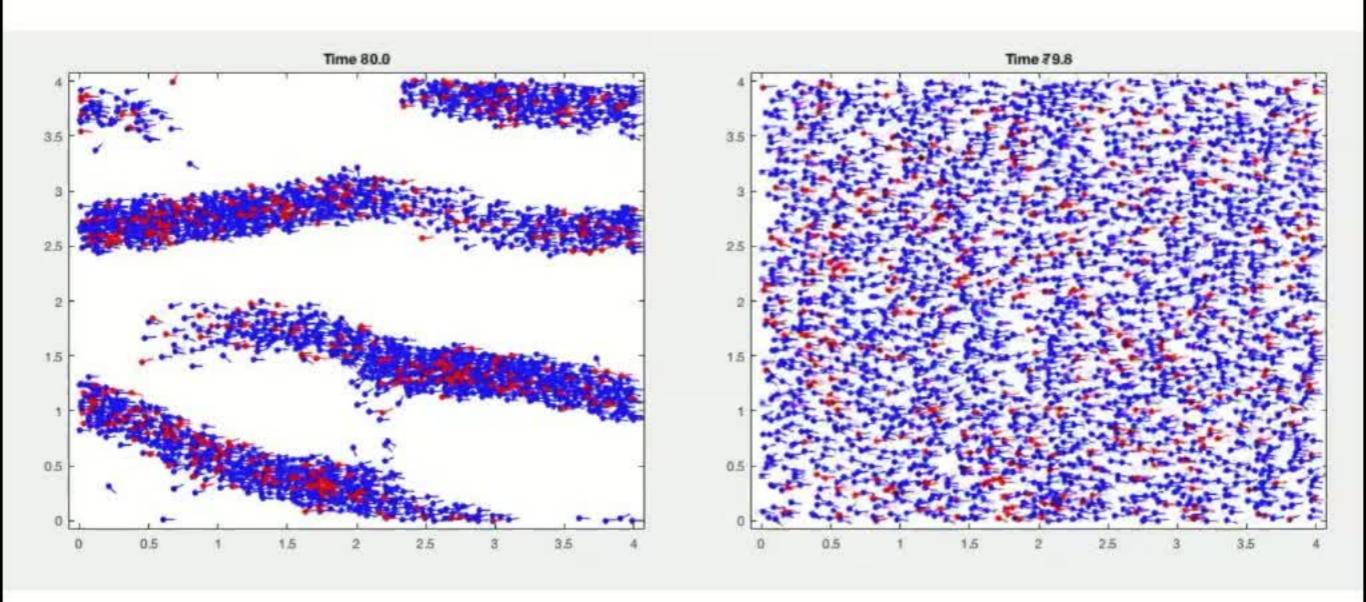
Repulsion



Repulsion

No Repulsion

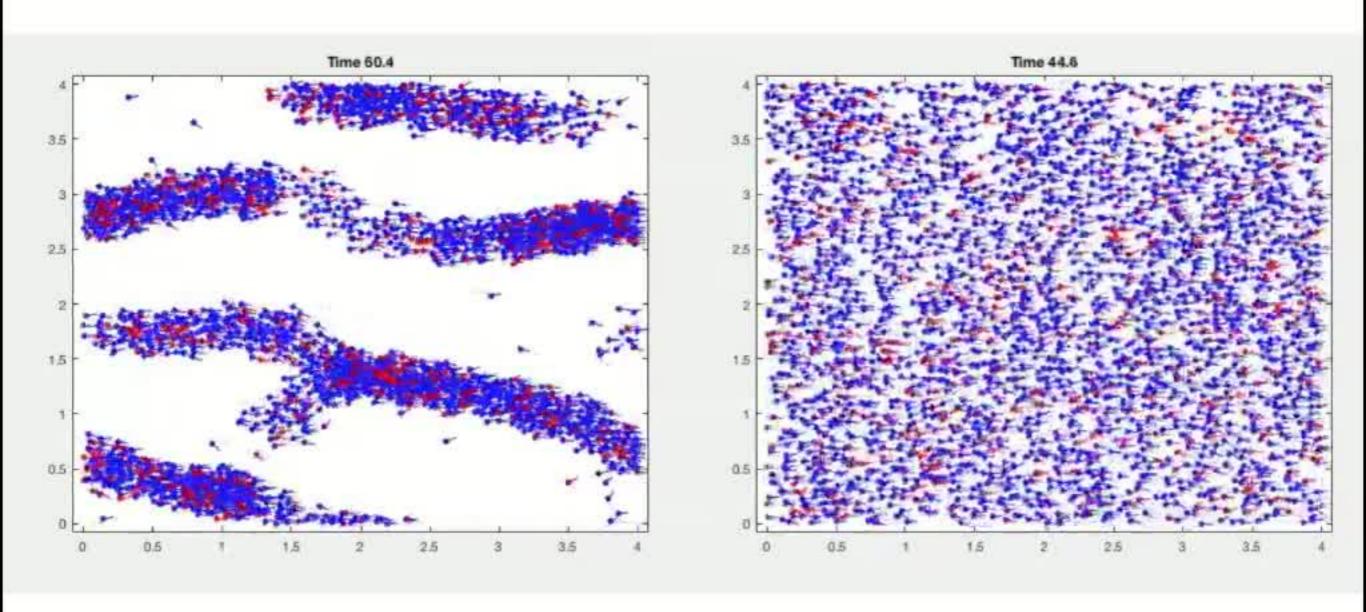
Attraction



Attraction

No Attraction

Attraction/Repulsion



Can we explain the role of attraction/repulsion in creating columnar structures?

Continuum Model

$$\frac{\partial S}{\partial t} = -k_{sm}S + k_{ms}M$$

$$\frac{\partial M}{\partial t} + \nabla \cdot (\vec{v}M) = k_{sm}S - k_{ms}M$$

$$\frac{d\theta_s}{dt} = \omega_o^s + \omega_e^s + \omega_a^s$$

$$\frac{d\theta_m}{dt} + \nabla \cdot (\vec{v}\theta_m) = \omega_o^m + \omega_e^m + \omega_a^m$$

S = Density of stationary locusts M = Density of moving locusts $\theta_s = \text{Angle of stationary locusts}$ $\theta_m = \text{Angle of moving locusts}$ $\rho = S + M$ (Total density)

$$k_{sm} = \kappa_{sm} e^{-\alpha_{sm}\beta_s} \quad \beta_s = \hat{u}(\theta_s) \cdot \int_{\Omega} \frac{\vec{y} - \vec{x}}{|\vec{y} - \vec{x}|} f_b(|\vec{y} - \vec{x}|) \rho(\vec{y}) \; d\vec{y}, \qquad \text{blocking}$$

$$\begin{aligned} \omega_o^m &= c_o \int_\Omega f_o(|\vec{y} - \vec{x}|) \left[S(\vec{y}) \sin \left[\theta_s(\vec{y}) - \theta_m(\vec{x}) \right] + M(\vec{y}) \sin \left[\theta_m(\vec{y}) - \theta_m(\vec{x}) \right] \right] \ d\vec{y}, \ \ \text{alignment} \\ \omega_e^m &= c_e \sin \left[\theta_e - \theta_m(\vec{x}) \right] \\ \omega_a^m &= -c_a \widehat{\vec{u}}_\perp(\theta_m) \cdot \int_\Omega \frac{\vec{y} - \vec{x}}{|\vec{y} - \vec{x}|} f_a(|\vec{y} - \vec{x}|) \rho(\vec{y}) \ d\vec{y}, \end{aligned} \qquad \text{attraction/repulsion}$$

Steady State

$$\frac{\partial S}{\partial t} = -k_{sm}S + k_{ms}M$$

$$\frac{\partial M}{\partial t} + \nabla \cdot (\vec{v}M) = k_{sm}S - k_{ms}M$$

$$\frac{d\theta_s}{dt} = \omega_o^s + \omega_e^s + \omega_a^s$$

$$\frac{d\theta_m}{dt} + \nabla \cdot (\vec{v}\theta_m) = \omega_o^m + \omega_e^m + \omega_a^m$$

S = Density of stationary locusts M = Density of moving locusts $\theta_s = \text{Angle of stationary locusts}$ $\theta_m = \text{Angle of moving locusts}$ $\rho = S + M$ (Total density)

$$S(\vec{x},t) = \bar{S}, \quad M(\vec{x},t) = \bar{M}, \quad \theta(\vec{x},t) = \theta_e$$

Orientation torques vanish by symmetry

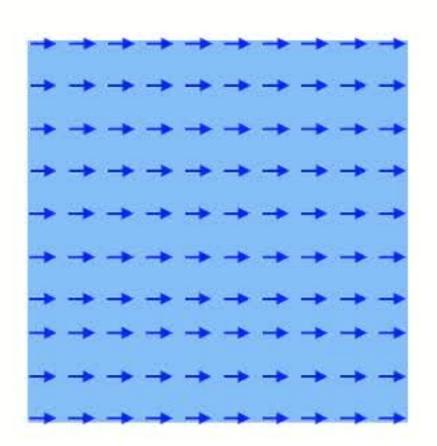
$$k_{sm} = \kappa_{sm} e^{-\alpha_{sm}\beta_s} \quad \beta_s = \hat{u}(\theta_s) \cdot \int_{\Omega} \frac{\vec{y} - \vec{x}}{|\vec{y} - \vec{x}|} f_b(|\vec{y} - \vec{x}|) \rho(\vec{y}) \; d\vec{y}, \; = 0 \quad \text{blocking}$$

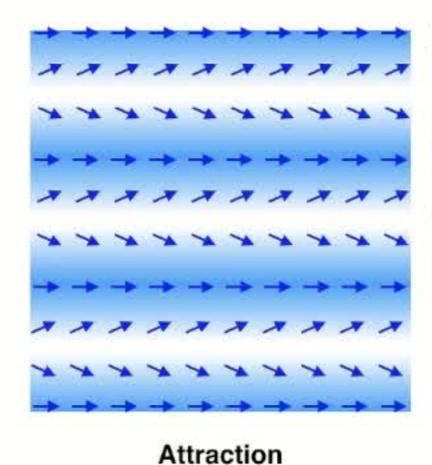
$$\begin{aligned} \omega_o^m &= c_o \int_\Omega f_o(|\vec{y} - \vec{x}|) \left[S(\vec{y}) \sin \left[\theta_s(\vec{y}) - \theta_m(\vec{x}) \right] + M(\vec{y}) \sin \left[\theta_m(\vec{y}) - \theta_m(\vec{x}) \right] \right] \ d\vec{y}, \ = \mathbf{0} \\ \omega_e^m &= c_e \sin \left[\theta_e - \theta_m(\vec{x}) \right] = \mathbf{0} \\ \omega_a^m &= -c_a \widehat{\vec{u}}_\perp(\theta_m) \cdot \int_\Omega \frac{\vec{y} - \vec{x}}{|\vec{y} - \vec{x}|} f_a(|\vec{y} - \vec{x}|) \rho(\vec{y}) \ d\vec{y}, \ = \mathbf{0} \end{aligned} \qquad \text{attraction/repulsion}$$

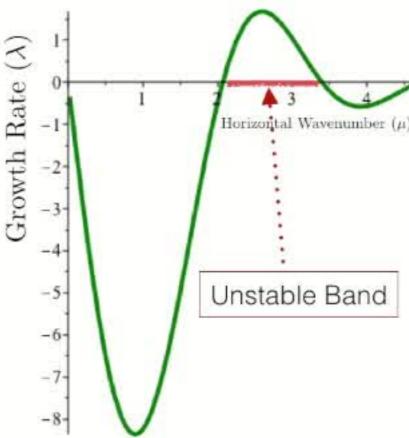
Transverse Stability Theory

$$\begin{bmatrix} S(x,y,t) \\ M(x,y,t) \\ \theta_s(x,y,t) \\ \theta_m(x,y,t) \end{bmatrix} = \begin{bmatrix} \overline{S} \\ \overline{M} \\ 0 \\ 0 \end{bmatrix} + \varepsilon \begin{bmatrix} \widehat{S} \\ \widehat{M} \\ \widehat{\theta}_s \\ \widehat{\theta}_m \end{bmatrix} e^{i\mu y + \lambda t},$$

S = Density of stationary locusts M = Density of moving locusts $\theta_s = \text{Angle of stationary locusts}$ $\theta_m = \text{Angle of moving locusts}$ $\rho = S + M$ (Total density)







No attraction

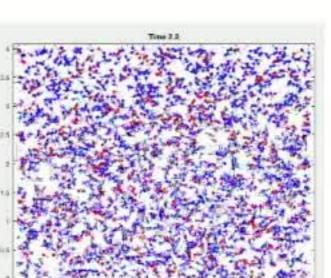
Horizontal modulation has unstable band of wavenumbers

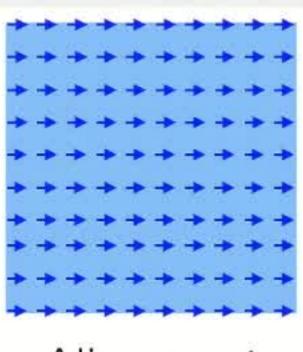
Horizontal modulation is neutrally stable

Growth Rate

Conclusions I

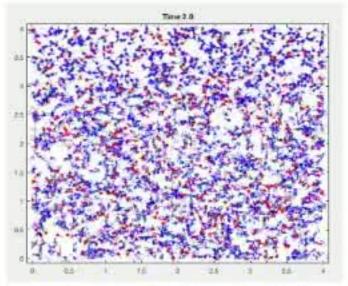
- Agent-Based Models, guided by biological observation, can capture hopper band formation.
- Homogenization yields a continuum model.
- Linear stability theory captures transition to band structured.
- Attraction and alignment are both necessary to obtain columns.

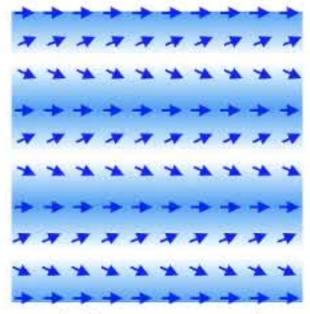












Alignment + Attraction

Model II: Resources, Feeding, and Planar Fronts

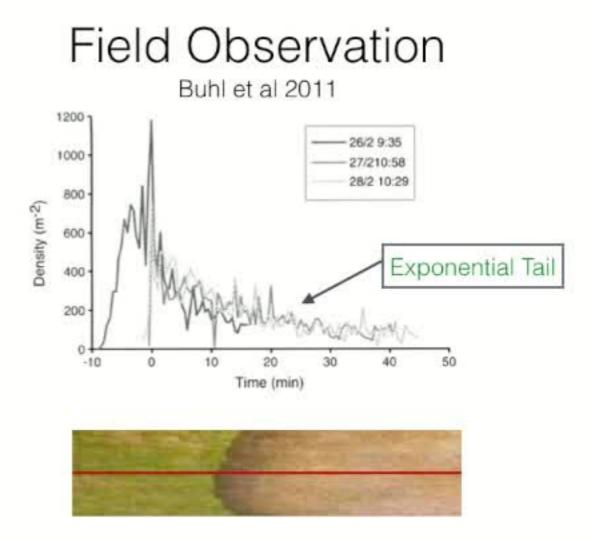




Planar fronts are observed for the Australian Plague Locust in resource rich environments

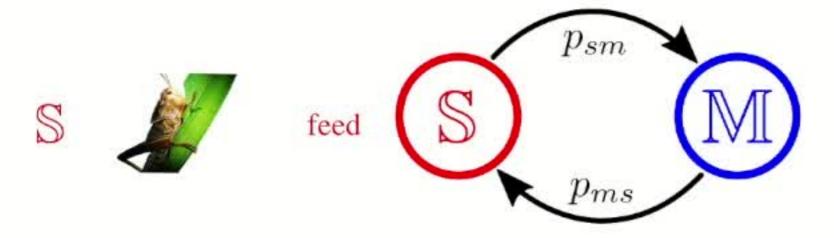
Locust Distribution in a Planar Front





Locust density is pulse-like with a steep rising front and an exponentially decaying tail.

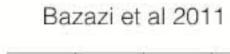
Agent- Based Model

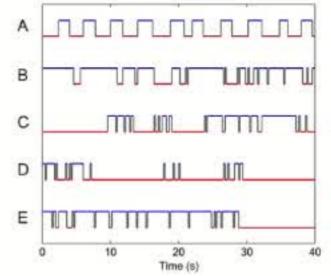


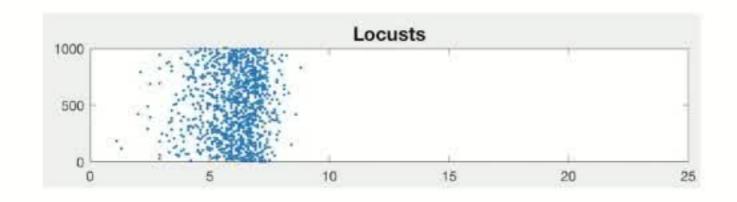


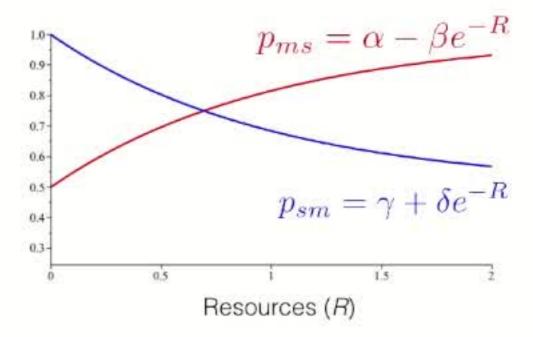
Lab Observation

- Model is one-dimensional.
- Locusts all move in the same direction.
- Locusts stop when eating
- Transition probabilities depend on resource availability.

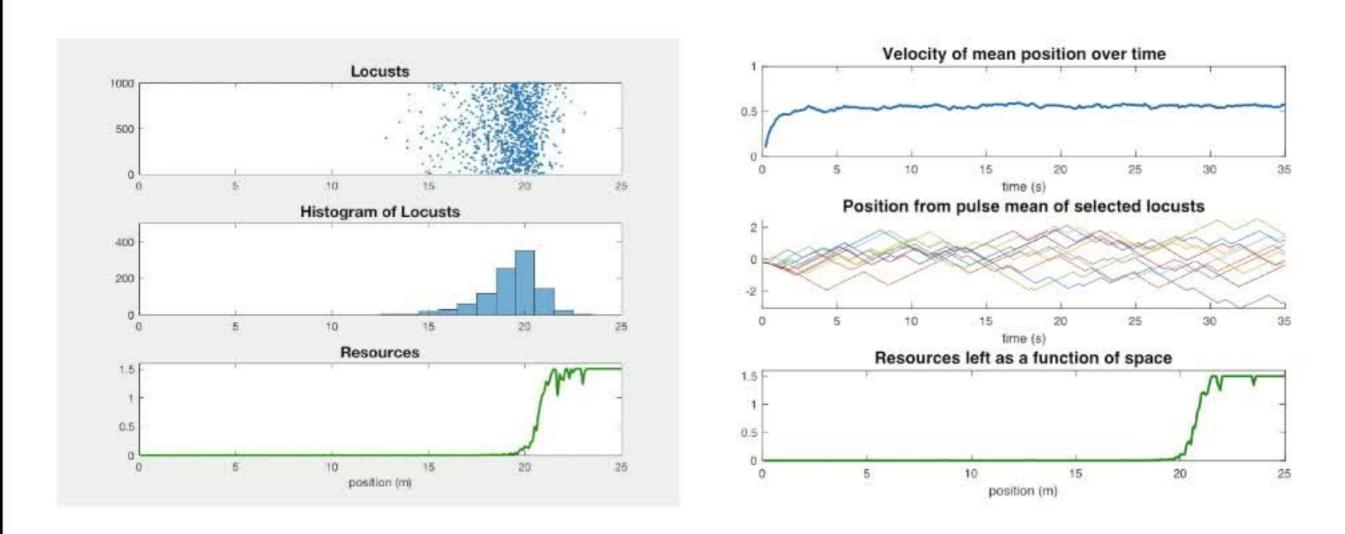








Numerics of ABM



Solution quickly relaxes to a traveling wave moving with roughly constant velocity.

PDE Model

Advection - Reaction Model

$$S_t = -k_{sm}(R) S + k_{ms}(R) M$$

$$M_t + vM_x = k_{sm}(R) S - k_{ms}(R) M$$

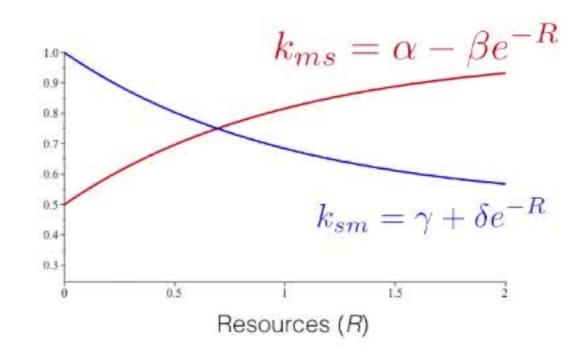
$$R_t = -\lambda SR$$

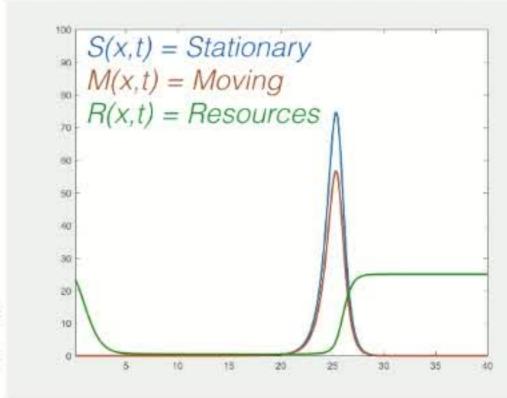
<u>Variables</u>

- S(x,t) = Stationary locust density
- M(x,t) = Moving locust density
- R(x,t) = Resource (food) density

<u>Parameters</u>

- v=Velocity of a moving locust
- $k_{sm}(R)$ = Stationary-to-moving transition rate
- $k_{ms}(R)$ = Moving-to-stationary transition rate
- λ = Resource consumption rate

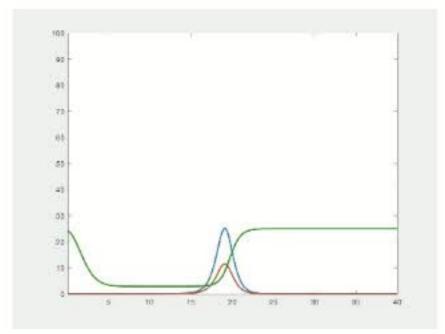


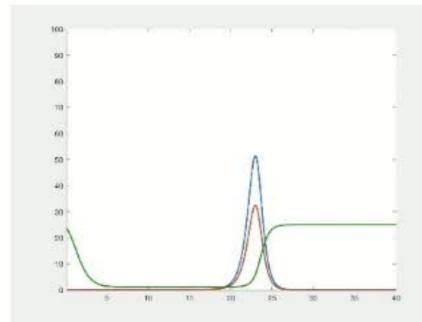


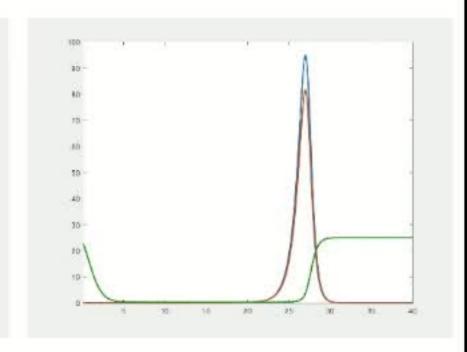
Traveling Pulse Solution

$$S(x,t) = Stationary$$

 $M(x,t) = Moving$
 $R(x,t) = Resources$







Mass = 100

Mass = 200

Mass = 400

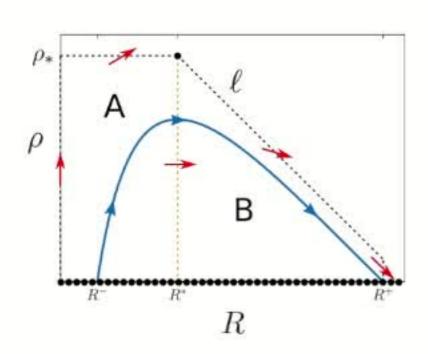
Traveling Wave Analysis

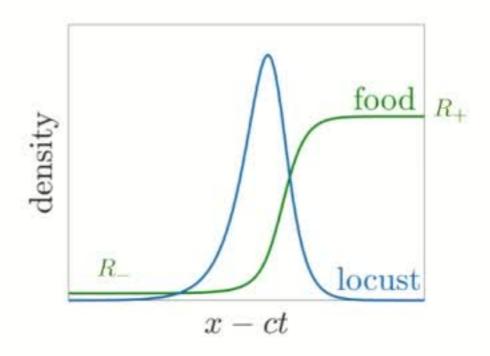
Moving reference frame $\xi = x - ct$ gives ordinary differential equation

$$R_{\xi} = \frac{1 - c}{c} \rho R$$

$$\rho_{\xi} = \left(\frac{k_{sm}}{c} - \frac{k_{ms}}{1 - c}\right) \rho$$

where $\rho = S + M$.



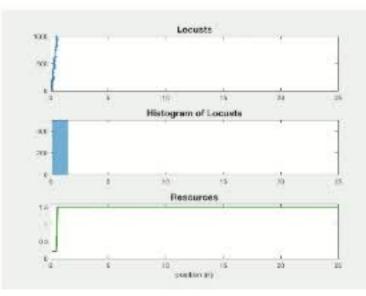


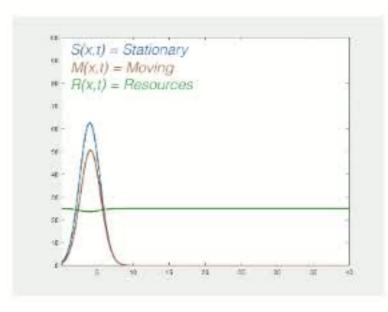
Total locust mass, M, and R_+ determine wave speed, c, and R_-

Conclusions II

- Agent-Based Models, guided by biological observation, can capture hopper band formation.
- Homogenization yields a continuum model.
- Phase plane analysis yields a traveling pulse solution.
- <u>Future research</u>: Construct a 2D model that combines social interactions and foraging.



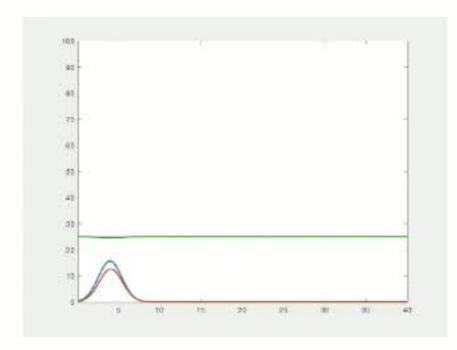


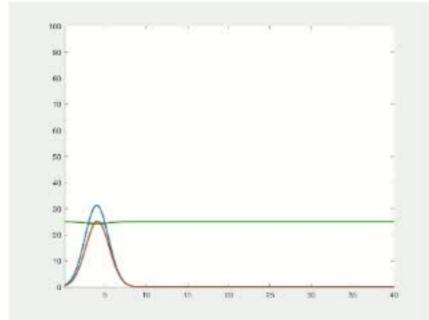


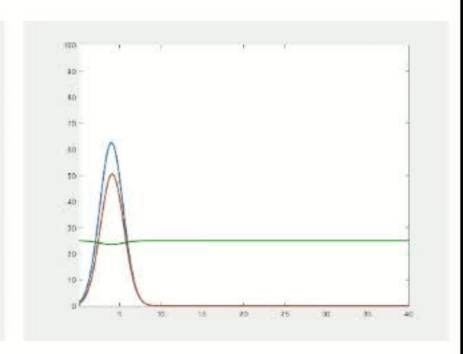
Traveling Pulse Solution

$$S(x,t) = Stationary$$

 $M(x,t) = Moving$
 $R(x,t) = Resources$







Mass = 100

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