

Stability of Stripe Solutions for the Gierer-Meinhardt Reaction-diffusion Model

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JohnHomenuke-ThesisPresentation.pdf](http://laplace.physics.ubc.ca/People/jhomenuk/JohnHomenuke-ThesisPresentation.pdf)

Outline

Introduction Activator-inhibitor systems and stability analysis of stripe solutions for the Gierer-Meinhardt (GM) Model.

Gierer-Meinhardt System

Numerics Crank-Nicholson and Multigrid schemes.

Stability Analysis Determining parameter spaces for which the stripe is unstable.

Compare Closed-form to Numerical Results

Remaining work / Questions

Introduction

Activator-Inhibitor Systems

- Two-component reaction-diffusion systems that form steady state solutions providing positional information.
- Applications to developmental biology.
- Activator undergoes autocatalysis to create local growths.
- Inhibitor suppresses the activator surrounding these centers.
- Both components diffuse. Inhibitor diffuses faster.
- Activator diffusion causes the steady-state.
- Play Animation
- Pattern provides positional information. Play Animation

Analyzing Stripe Solutions of the form $\text{sech}^2(x)$ for two kinds of instabilities: Breakup into spots and transverse evolution to zigzag patterns.

Gierer-Meinhardt System

Non-dimensionalized Form

$$\Omega = \{\mathbf{X} = (X_1, X_2) : -1 < X_1 < 1, 0 < X_2 < d_0\}, \quad t > 0$$

$$a = a(t, \mathbf{X}), \quad h = h(t, \mathbf{X})$$

$$\frac{\partial a}{\partial t} = \epsilon_0^2 \nabla^2 a - a + \frac{a^p}{h^q}, \quad \tau \frac{\partial h}{\partial t} = D \nabla^2 h - h + \frac{a^r}{\epsilon_0 h^s}$$

$$p > 1, q > 0, r > 1, s \geq 0, \quad \frac{qr}{p-1} - (s+1) > 0$$

$$\partial_n a = \partial_n h = 0, \quad \mathbf{X} \in \partial\Omega$$

Gierer-Meinhardt System

Non-dimensionalized Form

$$\Omega = \{\mathbf{X} = (X_1, X_2) : -1 < X_1 < 1, 0 < X_2 < d_0\}, \quad t > 0 \quad (1)$$

$$a = a(t, \mathbf{X}), \quad h = h(t, \mathbf{X}) \quad (2)$$

$$\frac{\partial a}{\partial t} = \epsilon_0^2 \nabla^2 a - a + \frac{a^p}{h^q}, \quad \tau \frac{\partial h}{\partial t} = D \nabla^2 h - h + \frac{a^r}{\epsilon_0 h^s} \quad (3)$$

$$p > 1, q > 0, r > 1, s \geq 0, \quad \frac{qr}{p-1} - (s+1) > 0 \quad (4)$$

$$\partial_n a = \partial_n h = 0, \quad \mathbf{X} \in \partial\Omega \quad (5)$$

Re-scaled Form

$$\Omega_l = \{\mathbf{x} = (x_1, x_2) : -l < x_1 < l, 0 < x_2 < d\}, \quad t > 0 \quad (6)$$

$$\frac{\partial a}{\partial t} = \epsilon^2 \nabla^2 a - a + \frac{a^p}{h^q}, \quad \tau \frac{\partial h}{\partial t} = \nabla^2 h - h + \frac{a^r}{\epsilon h^s} \quad (7)$$

$$l = 1/\sqrt{D}, \quad X = x/l, \quad d = d_0 l, \quad \epsilon = \epsilon_0 l \quad (8)$$

Numerics: Finite Differencing

Domain Discretize into an array of points.

$$\Omega \rightarrow \Omega_h = \{(x_i, y_j) : x_i = (i - 1)h, y_j = (j - 1)h\}$$

$$t = n\Delta t, \quad \Delta t = \lambda h, \quad 0 < \lambda < 1 \quad (\text{Courant number})$$

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Equations Differential operators \rightarrow Difference Operators via Taylor Series approximations.

$$\partial_t u \rightarrow \Delta_t(u^n) = \frac{u^{n+1} - u^n}{\Delta t} + O(\Delta t)$$

$$\partial_{xx} u \rightarrow \Delta_{xx}(u_i) = \frac{u_{i-1} - 2u_i + u_{i+1}}{h^2} + O(h^2)$$

Numerics: Finite Differencing

Domain Discretize into an array of points.

$$\Omega \rightarrow \Omega_h = \{(x_i, y_j) : x_i = (i - 1)h, y_j = (j - 1)h\} \quad (9)$$

$$t = n\Delta t, \quad \Delta t = \lambda h, \quad 0 < \lambda < 1 \quad (\text{Courant number}) \quad (10)$$

Equations Differential operators \rightarrow Difference Operators via Taylor Series approximations.

$$\partial_t u \rightarrow \Delta_t(u^n) = \frac{u^{n+1} - u^n}{\Delta t} + O(\Delta t) \quad (11)$$

$$\partial_{xx} u \rightarrow \Delta_{xx}(u_i) = \frac{u_{i-1} - 2u_i + u_{i+1}}{h^2} + O(h^2) \quad (12)$$

Crank-Nicholson Scheme Both sides centered at time index $n + 1/2$.

$$\partial_t u = \nabla^2 u + f(u, v) \quad (13)$$

$$\frac{1}{\Delta t} (u^{n+1} - u^n) = \frac{1}{2} (\nabla^2 u^{n+1} + \nabla^2 u^n + f(u^{n+1}, v^{n+1}) + f(u^n, v^n)) \quad (13)$$

$$u^{n+1} - \frac{\Delta t}{2} \left(\nabla^2 u^{n+1} + f(u^{n+1}, v^{n+1}) \right) = u^n + \frac{\Delta t}{2} \left(\nabla^2 u^n + f(u^n, v^n) \right) \quad (14)$$

Need to solve this *elliptic* equation at every time step.

Numerics: Multigrid

Solving Elliptic Equations Repeatedly apply a smoothing operator pointwise until convergence to true solution. Convergence rate depends on number of points.

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The Multigrid Advantage Convergence rate is independent of the number of points.

1. Transfer temporary solution to coarser grid.
2. Apply smoothing operator.
3. Transfer solution back to fine grid.

Use as many grids as necessary.

Numerics: Multigrid

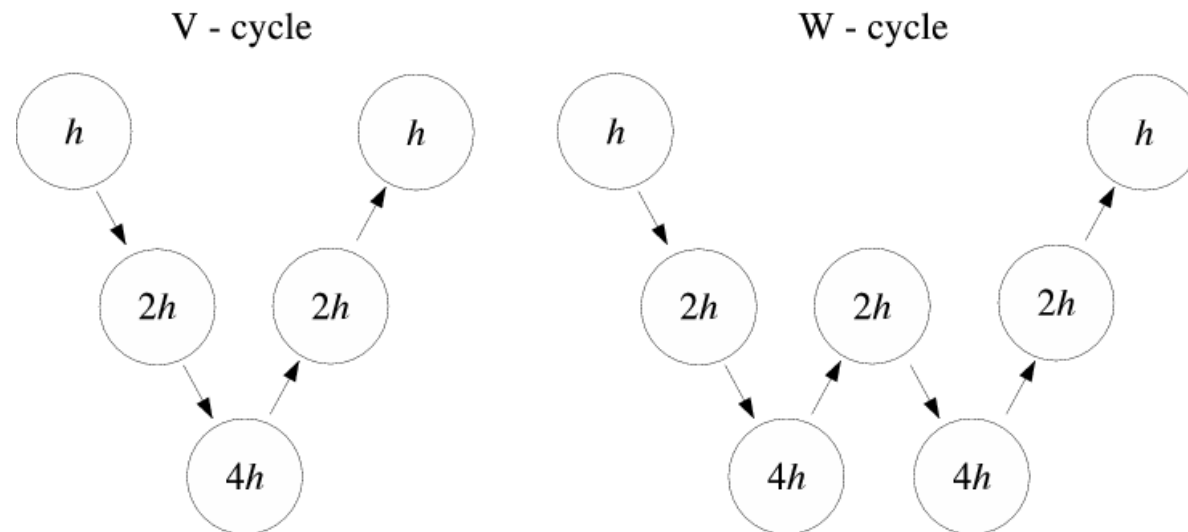
Solving Elliptic Equations Repeatedly apply a relaxation (smoothing) operator pointwise until convergence to true solution. Convergence rate depends on number of points.

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Use as many grids as necessary.

Schematic of Grid Transfers



Numerics: Multigrid

Restriction Operator \mathcal{R} Transfers the temporary solution to the next coarsest grid of mesh twice the current one using a weighted sum of the surrounding points.

$$\begin{bmatrix} \frac{1}{16} & \frac{1}{8} & \frac{1}{16} \\ \frac{1}{8} & \frac{1}{4} & \frac{1}{8} \\ \frac{1}{16} & \frac{1}{8} & \frac{1}{16} \end{bmatrix}$$

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Prolongation Operator \mathcal{P} Transfers the temporary solution to the next finest grid using a bilinear interpolation.

$$\begin{bmatrix} \frac{1}{4} & \frac{1}{2} & \frac{1}{4} \\ \frac{1}{2} & 1 & \frac{1}{2} \\ \frac{1}{4} & \frac{1}{2} & \frac{1}{4} \end{bmatrix}$$

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Relaxation Operator Newton Gauss-Seidel Method:

1. Nonlinear system $\mathbf{L}(\mathbf{u}) = \mathbf{f}$. (Our case: $\mathbf{L}(\mathbf{u}^{n+1}) = \mathbf{f}(\mathbf{u}^n)$).
2. Compute residual $\mathbf{r} = \mathbf{L}(\mathbf{u}^n) - \mathbf{f}(\mathbf{u}^n)$.
3. Compute Jacobian $J_{ij} = \partial L_{ij} / \partial u$ for $i = 1, \dots, D$ and $j = 1, \dots, D$. D is number of components.
4. Solve $\mathbf{J} \delta \mathbf{u} = \mathbf{r}$ to get adjustment to solution.
5. Repeat until local convergence is achieved.

Stability Analysis

Two Interaction Regimes

- Semi-strong: Diffusion coefficients $\epsilon_0^2 \ll 1$ and $D = O(1)$.
- Weak: Diffusion coefficients $\epsilon_0^2 \ll 1$ and $D = D_0\epsilon_0^2 \ll 1$.

Stability Analysis

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- Weak: Diffusion coefficients $\epsilon_0^2 \ll 1$ and $D = D_0\epsilon_0^2 \ll 1$.

Homoclinic Stripes Solution to steady state cross-section

$$a_{yy} - a + \frac{a^p}{h^q} = 0, D_0 \quad h_{yy} - h + \frac{a^r}{h^s} = 0$$

where $y = \epsilon_0 x_1$. Numerically determine $a(0)$ and $h(0)$. Then use an approximation in full solutions.

Stability Analysis

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Homoclinic Stripes Solution to steady state cross-section

$$a_{yy} - a + \frac{a^p}{h^q} = 0, \quad D_0 h_{yy} - h + \frac{a^r}{h^s} = 0 \quad (17)$$

where $y = \epsilon_0 x_1$. Numerically determine $a(0)$ and $h(0)$. Then use an approximation in full solutions.

Two Types of Instabilities for Homoclinic Stripes

- Varicose (breakup): Even perturbation eigenfunctions.
- Transverse (zigzag): Odd perturbation eigenfunctions.

Stability Analysis

Determining the Stability Use the ansatz

$$a = a_e(x_1/\epsilon_0) + \Phi(x_1/\epsilon_0)e^{\lambda t} \cos(mx_2), \quad h = h_e(x_1/\epsilon_0) + N(x_1/\epsilon_0)e^{\lambda t} \cos(mx_2) \quad (18)$$

in the original GM system. Then we get the eigenvalue problem

$$\begin{aligned} \Phi_{yy} - (1 + \mu)\Phi + \frac{pa_e^{p-1}}{h_e^q}\Phi - \frac{qa_e^p}{h_e^{q+1}}N &= \lambda\Phi \\ D_0N_{yy} - (1 + D_0\mu)N + \frac{ra_e^{r-1}}{h_e^s}\Phi - \frac{sa_e^r}{h_e^{s+1}}N &= \tau\lambda N \end{aligned}$$

which can be solved numerically to determine the range of $\mu = \epsilon_0^2 m^2$ for which $\Re(\lambda) > 0$.

Numerical Simulations

Initial Conditions

$$a(0, x_1, x_2) = a_e(0)\text{sech}^2(x_1/\epsilon_0), \quad h(0, x_1, x_2) = h_e(0)\text{sech}^2(x_1/\epsilon_0) \quad (19)$$

Solutions (Weak Interaction Regime)

1. $\epsilon_0 = 0.025$, $D_0 = 15.0$, $\tau = 0.01$, $(p, q, r, s) = (2, 1, 2, 0)$
Expected Behaviour: 8 spots followed by transverse migration.
Play Animation
2. $\epsilon_0 = 0.025$, $D_0 = 7.6$, $\tau = 0.01$, $(p, q, r, s) = (2, 1, 2, 0)$
Expected Behaviour: Zigzag pattern.
Play Animation
3. $\epsilon_0 = 0.025$, $D_0 = 6.8$, $\tau = 0.01$, $(p, q, r, s) = (2, 1, 2, 0)$
Expected Behaviour: No equilibrium stripe solution. Replication of stripe followed by breakup.
Play Animation

Remaining Work / Questions

Independent Residual Test Rewriting the program using a different scheme. If solutions are the same within truncation error, then they must be correct.

Comparison of Numerical to Theorized Results Behaviour on paper not reproduced precisely in numerical results.

- Different numbers of spots.
- No zigzag instabilities observed yet.

Semi-strong Regime Have yet to generate solutions in this regime.