Stable Localized Patterns in Cross Diffusion and Chemotaxis Systems

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Department of Mathematics Chinese University of Hong Kong Joint work with T. Kolokolnikov, M. Ward Localized behavior in reaction-diffusion systems July 25, 2011, BIRS

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Outline of My Talk

 Part I: Localized Solutions in Cross-Diffusion Systems (joint work with T. Kolokolnikov) (SIAM J. Appl. Math. 2011 online)

 Part II: Localized Solutions in Chemotaxis System Modeling LA Crimes (joint work with T. Kolokolnikov and M. Ward)

Part I: Cross-Diffusion Systems

We first discuss pattern formations in a cross-diffusion system

Standard Diffusion: $\nabla(J), J = \nabla u$

Self Diffusion: $J = a(x, u) \nabla u$

Cross-diffusion: $J = a(x, u, v) \nabla u$

Q: Can cross-diffusion create stable patterns?

A model of cross-diffusion

We consider cross-diffusion model of Shigesada, Kawasaki and Teramoto (1979)

$$\begin{cases} u_t = \Delta \left[(d_1 + \rho_{12}v) \, u \right] + u(a_1 - b_1u - c_1v) \\ v_t = \Delta \left[(d_2 + \rho_{21}u) \, v \right] + v(a_2 - b_1u - c_1v) \\ \text{Neumann B.C. on } [a, b] \end{cases}$$
(1)

The kinetics are just the classic Lotka-Volterra competition model; d_1, d_2 represent self-diffusion Cross-diffusion (ρ_{12}, ρ_{21}) represent inter-species avoidance: abundance of v will cause u to diffuse faster and vice-versa.

Without cross-diffusion, only constant solution is stable [Kishimoto, 1981].

A well-studied toy model [Lou, Ni, Yotsutani, Wu, Xu] is [after scaling]:

$$\begin{cases} u_t = (\rho v u)_{xx} + u(a_1 - b_1 u - c_1 v) \\ v_t = dv_{xx} + v(a_2 - b_1 u - c_1 v) \end{cases}$$
(2)

with the following assumptions:

 $d \ll 1; \quad \rho \gg 1;$ all other parameters are positive and of O(1). (3) Biologically, when ρ is large, v acts as an inhibitor on u, so that udiffuses quickly in the regions of high concentration of v. This effect is believed to be responsible for the segregation of the two species.

Construction of steady state in 1D

- Lou, Ni, Yotsutani, 2004: Constructed a steady state in the form of a spike for u, and in the form of an inverted spike for v.
- More explicit computations [spike height] by Wu-Xu, 2010.
- Define

 $\tau = uv$

so that

$$0 = dv_{xx} + a_2v - b_2\tau - c_2v^2; \quad 0 = \rho\tau_{xx} + \tau \left(\frac{a_1}{v} - b_1\frac{\tau}{v^2} - c_1\right);$$
(4)

• In the limit $\rho \to \infty$ the shadow system is:

$$0 = dv_{xx} + a_2v - b_2\tau + c_2v^2;$$
 (5)

$$Lc_{1} = \int_{0}^{L} \left(\frac{a_{1}}{v} - b_{1}\frac{\tau}{v^{2}}\right).$$
 (6)

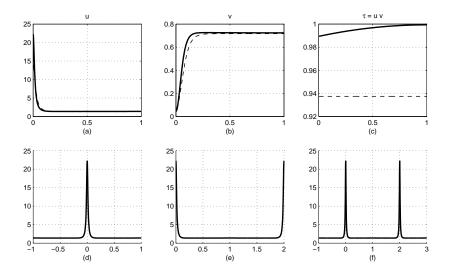
(Keener, 1981, Nishiura)

- ► the solution to dv_{xx} + a₂v b₂τ + c₂v² = 0 can be written as v = C₁ + C₂ tanh²(C₃x). Matching the integral condition gives
- asymptotic behavior

$$v(x) \sim \frac{a_2}{2c_2} \left[\frac{3}{2} \tanh^2 \left(\frac{x}{2\varepsilon} \right) + \delta \left(2 - 3 \tanh^2 \left(\frac{x}{2\varepsilon} \right) \right) \right];$$
$$u \sim \frac{\tau_0}{v(x)}$$

where

$$\begin{split} \varepsilon &:= \sqrt{\frac{2d}{a_2}} \qquad \text{[spike width]} \\ \delta &:= (\varepsilon/L)^{2/3} \frac{3}{4} \left(\frac{b_1}{b_2} \frac{\pi}{2}\right)^{2/3} \left(4\frac{a_1}{a_2} - \frac{b_1}{b_2} - 3\frac{c_1}{c_2}\right)^{-2/3} \quad \text{[spike height]} \\ \tau_0 &:= \frac{3}{16} \frac{a_2^2}{b_2 c_2}; \end{split}$$



v has an inverted spike

$$v(x) \sim \frac{a_2}{2c_2} \left[w(0) - w\left(\frac{x}{2\varepsilon}\right) + \delta\left(2 - 3\tanh^2\left(\frac{x}{2\varepsilon}\right)\right) \right]$$

$$w_{yy} - w + w^2 = 0; \quad w \to 0 \text{ as } |y| \to \infty, \quad w'(0) = 0.$$

▶ Note that $v(0) \sim \frac{a_2}{c_2} \delta = O(\varepsilon^{2/3}); \quad u(0) \sim O(\varepsilon^{-2/3}).$

This construction works as long as

$$\left(4\frac{a_1}{a_2} - \frac{b_1}{b_2} - 3\frac{c_1}{c_2}\right) > 0.$$

Question: is the solution stable ?

Linearized problem

Linearized equations are

$$\lambda \phi = d\phi_{xx} + a_2\phi - b_2\psi - c_2 2v\phi;$$

$$\lambda \left(\frac{1}{v}\psi - \frac{\tau}{v^2}\phi\right) = \rho\psi_{xx} + \left(\frac{a_1}{v} - b_1 2\frac{\tau}{v^2} - c_1\right)\psi + \left(-\frac{a_1\tau}{v^2} + 2b_1\frac{\tau^2}{v^3}\right)\phi.$$

Two kinds of eigenvalues

- ▶ large eigenvalues: $\lambda = O(1)$
- small eigenvalues: $\lambda = o(1)$

Principal stability result

Define

$$\rho_{K,\text{small}} := d^{-1/3} L^{8/3} \frac{c_2}{2} \left(\frac{b_1}{b_2} \frac{\pi}{2} \right)^{-2/3} \frac{a_2^{1/3}}{2^{1/3}} \left(4 \frac{a_1}{a_2} - \frac{b_1}{b_2} - 3 \frac{c_1}{c_2} \right)^{5/3};$$

$$\rho_b := 0.747 \rho_{K,\text{small}};$$

$$\rho_{K,\text{large}} := \rho_{K,\text{small}} \frac{2 \times 0.747}{1 - \cos\left[\pi \left(1 - 1/K\right)\right]}.$$
(9)

Then:

- A single boundary spike is stable for all ρ (not exponentially large in ε).
- A double-boundary steady state is stable if ρ < ρ_b and is unstable otherwise. The instability is due to a large eigenvalue.
- A K-interior spike steady state with $K \ge 2$ is stable if $\rho < \min(\rho_{K,\text{small}}, \rho_{K,\text{large}})$ and is unstable otherwise. When K = 1, it is stable provided that ρ is not exponentially large in ε .
- The critical scaling is

$$\rho = O(d^{-1/3}) = O(\varepsilon^{-2/3}) \gg 1.$$

Stability: small vs. large eigenvalues

- ▶ K spikes are always stable whenever $1 \ll \rho \ll d^{-1/3}$ and unstable when $K \ge 2$ and $\rho \gg d^{-1/3}$.
- ▶ Recall that $\rho_{K,\text{large}} := \rho_{K,\text{small}} \frac{2 \times 0.747}{1 \cos[\pi(1 1/K)]}$ and

$$\frac{2 \times 0.747}{1 - \cos\left[\pi \left(1 - 1/K\right)\right]} = \begin{cases} 1.494 > 1, & K = 2\\ 0.996 < 1, & K = 3\\ 0.875 < 1, & K = 4 \end{cases}$$

▶ ρ_{K,large} > ρ_{K,small} if K = 2 but ρ_{K,large} < ρ_{K,small} if K ≥ 3. It follows that the primary instability is due to small eigenvalues if K = 2but is due to large eigenvalues if K ≥ 3. This is in agreement with numerical simulations.

Boundary Conditions

Possible boundary conditions (as in van der Ploeg-Doelman, Indiana Univ.Math. J. 2005):

Config type	Boundary conditions for ϕ
1 interior spike on $[-L, L]$	$\phi'(0) = 0 = \phi'(L)$
even eigenvalue	
1 interior spike on $[-L, L]$	$\phi(0) = 0 = \phi'(L)$
odd eigenvalue	
2 1/2-spikes at $[0, L]$	$\phi'(0) = 0 = \phi(L)$
K spikes on $[-L, (2K-1)L],$	$\phi(L) = z\phi(-L), \qquad \phi'(L) = z\phi'(-L)$
Periodic BC	$z = \exp\left(2\pi i k/K\right), k = 0 \dots K -$
K spikes on $[-L, (2K-1)L],$	$\phi(L) = z\phi(-L), \qquad \phi'(L) = z\phi'(-L)$
Neumann BC	$z = \exp(\pi i k/K), k = 0 \dots K -$

(same BC for ψ)

Reduced problem, large eigenvalues

 Using asymptotic matching, eventually we get a new point-weight eigenvalue problem (PWEP):

$$\begin{cases} \lambda \Phi = \Phi_{yy} - \Phi + 2w\Phi - \chi \Phi(0) \\ \Phi \text{ is even and is bounded as } |y| \to \infty \end{cases}$$
 (PWEP)

where $w(y) = \frac{3}{2} \operatorname{sech}^2\left(\frac{y}{2}\right)$ satisfies

$$w_{yy} - w + w^2 = 0; \quad w \to 0 \text{ as } |y| \to \infty, \ w'(0) = 0.$$

For double-boundary spike,

$$\chi = \chi_b := \frac{\varepsilon^{-2/3}}{4\rho} \left(4\frac{a_1}{a_2} - \frac{b_1}{b_2} - 3\frac{c_1}{c_2} \right)^{5/3} c_2 \left(\frac{b_1}{b_2} \frac{\pi}{2} \right)^{-2/3} L^{8/3}.$$

For K spikes, Neumann BC, there are K choices for χ , namely

$$\chi = \frac{2}{1 - \cos \frac{\pi k}{K}} \chi_b, \quad k = 0 \dots K - 1 \text{ and } \chi = \text{very large positive}.$$

Analysis of $PWEP \ \lambda \Phi = \Phi_{yy} - \Phi + 2w\Phi - \chi \Phi(0)$

- $\lambda = 0, \ \Phi = w_y$ is a solution [corresponds to translation invariance]
- If χ = 0 then there is an unstable eigenvalue λ₁ > 0 and another eigenvalue λ₃ < 0.</p>
- Decompose:

$$\Phi(y) = \Phi^{\star} + \Phi_0(y); \quad \text{where} \quad \Phi^{\star} = \lim_{y \to \pm \infty} \Phi(y).$$

Then

$$\lambda \Phi^{\star} = -\Phi^{\star} - \chi \left(\Phi_0(0) + \Phi^{\star} \right)$$

and Φ_0 satisfies

$$\lambda \Phi_0 = \Phi_{0yy} - \Phi_0 + 2w\Phi_0 + 2w\Phi^*$$

so the PWEP becomes

$$\lambda \Phi_0 = \Phi_{0yy} - \Phi_0 + 2w\Phi_0 - \frac{2\chi}{\chi + \lambda + 1} \Phi_0(0)w$$
 (10)

$$\lambda \Phi_0 = \Phi_{0yy} - \Phi_0 + 2w\Phi_0 - \frac{2\chi}{\chi + \lambda + 1} \Phi_0(0)w$$

- Anzatz: if $\Phi_0 = w, \lambda = 0$ then $\chi = \frac{1}{2}$.
- \blacktriangleright Rigorous result: there is an unstable eigenvalue $\lambda>0$ for all $\chi<\frac{1}{2}$
- The above two facts seem to suggest: stability when $\chi > \frac{1}{2}$???)
- ▶ In the limit $\chi \to \infty$, the limiting problem is

$$\lambda \Phi_0 = \Phi_{0,yy} - \Phi_0 + 2w\Phi_0 - 2\Phi_0(0)w \tag{11}$$

Hypergeometric reduction

Theorem: the eigenvalues of $\lambda \Phi = \Phi_{yy} - \Phi + 2w\Phi - \chi \Phi(0)$ are given implicitly by:

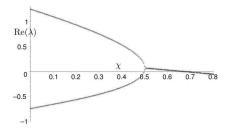
$$\lambda = -1 - \chi + 2\chi \Phi_0(0)$$

where

$$\Phi_0(0) = \frac{6\pi\lambda(\lambda+1)}{\sin(\pi\alpha)(4\lambda-5)(4\lambda+3)} - \frac{3}{2}\frac{1}{\lambda}{}_3F_2\left(\begin{array}{c}1,3,-1/2\\2+\alpha,2-\alpha\end{array};1\right)$$
$$\alpha = \sqrt{1+\lambda}$$

Similar idea has been used in

Doelman-Gardner-Kaper, Mem. AMS 2002, Indiana Univ. Math. J. 2001 Wei-Winter, MAA 2002, SIAM J.Math.Anal. 2003 Numerical result+winding argument of Ward-Wei EJAM 2003: all λ < 0 whenever χ > 0.669; stabilization is via a hopf bifurcation.



 desperate need for an analytical study of the limiting eigenvalue problem

$$\lambda \Phi_0 = \Phi_{0,yy} - \Phi_0 + 2w\Phi_0 - 2\Phi_0(0)w$$

Small eigenvalues

- Construct asymmetric spike steady states
- These bifurcate from the symmetric branch
- The instability thresholds for the small eigenvalues correspond precisely to this bifurcation point! Iron-Ward-Wei Phys D 2001 van der Ploeg-Doelman, Indiana Univ.Math. J. 2005 Proof is Needed !!!
- Main result: For 2 spikes, small eigenvalues is the dominant instability. For 3 or more, large eigenvalues dominate.

Radial equilibrium in two dimensions

Consider $\Omega \in \mathbb{R}^2$. Let w be the ground state in 2D:

$$\Delta w - w + w^2 = 0; \quad w \to 0 \text{ as } |y| \to \infty, \quad \max w = w(0)$$

and define

$$m := \max w(y) = w(0) \approx 2.39195.$$

Suppose that

$$\frac{a_1}{a_2} \left(2m-1\right) - \left(m-1\right) \frac{b_1}{b_2} - m\frac{c_1}{c_2} > 0 \tag{12}$$

and consider the asymptotic limit

$$d \ll 1; \quad \rho \gg 1. \tag{13}$$

If Ω is radially symmetric, there is a steady state at x = 0, in the form of an inverted spike for v. More precisely, we have

$$v(x) \sim \frac{1}{2m-1} \frac{a_2}{c_2} \left(1 - 2\delta\right) \left(\frac{w(0) - w\left(\frac{1-\delta}{\varepsilon}x\right)}{\varepsilon} + (2m-1)\delta\right);$$
$$u \sim \frac{\tau_0}{v(x)}$$

$$\Delta w - w + w^2 = 0; \quad w \to 0 \text{ as } |y| \to \infty, \quad \max w = w(0)$$

where

$$\varepsilon := \sqrt{\frac{(2m-1)d}{a_2}}; \quad \tau_0 := \frac{(m-1)m}{(2m-1)^2} \frac{a_2^2}{b_2 c_2}.$$
$$\delta \sim \frac{\varepsilon^2}{|\Omega|} \frac{4\pi b_1 m}{b_2 (2m-1)} \frac{1}{\left(\frac{a_1}{a_2} (2m-1) - (m-1)\frac{b_1}{b_2} - m\frac{c_1}{c_2}\right)};$$

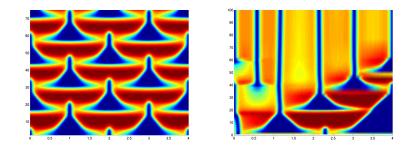
In particular,

$$v(0) \sim \frac{a_2}{c_2} \delta = O(d);$$
 $u(0) \sim \frac{(m-1)m}{(2m-1)^2} \frac{a_2}{b_2} \frac{1}{\delta} = O\left(\frac{1}{d}\right).$ (14)

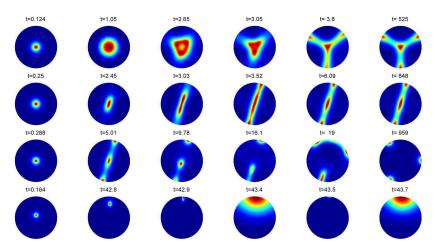
Stability:????

Interesting patterns: $\rho = O(1)$

Spike insertion, spatio-temporal chaos



Sensitivity to initial conditions. The left and right figure differ only in the initial conditions. On the left, symmetric initial conditions result in an intricate a time-periodic solution. On the right, the initial condition is the same as on the left, except for a shift of 0.1 units to the right. dynamics eventually settle to a 5-spike stable pattern.



 $\rho = 50, \ (a_1, b_1, c_1) = (5, 1, 1), \ (a_2, b_2, c_2) = (5, 1, 5)$

Row 1: $\rho = 2$. Spot splits into three spots. Row 2: $\rho = 4$. Initially, spot splits into two, final steady state consists of two boundary and one center spot. Row 3: $\rho = 6$. Row 4: $\rho = 500$. The interior

Part II: Localized Solutions in Crime Hotspot Model

UCLA Model of hot-spots in crime

- Recently proposed by Short Brantingham, Bertozzi et.al [PNAS, 2008].
- ▶ Very hot math: e.g. The New York Times, Dec 2010
- Crime is ubiquious but not uniformly distributed
 - some neigbourhoods are worse than others, leading to crime "hot spots"
 - Crime hotspots can persist for long time.

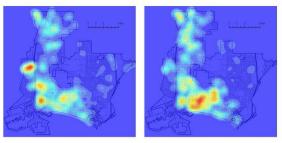


Fig. 1. Dynamic changes in residential burglary hotspots for two consecutive three-month periods beginning June 2001 in Long Beach, CA. These density maps were created using ArcGIS.

. . .

- Crime is temporally correlated:
 - Criminals often return to the spot of previous crime
 - If a home was broken into in the past, the likelyhood of subsequent breaking increases
 - Example: graffitti "tagging"
 - the motion of criminals towards higher attractiveness areas can be modeled by chemotaxis

Two-component model

$$A_t = \varepsilon^2 A_{xx} - A + \rho A + A_0$$

$$\tau \rho_t = D \left(\rho_x - 2 \frac{\rho}{A} A_x \right)_x - \rho A + \bar{A} - A_0.$$

- $\rho(x,t) \equiv$ density of criminals;
- $A(x,t) \equiv$ "attractiveness" of area to crime
- $A_0 = O(1) \equiv$ "baseline attractiveness "
- ▶ $D(-2\frac{\rho}{A}A_x)_x$ models the motion of criminals towards higher attractiveness areas
- $\bar{A} A_0 > 0$ is the baseline criminal feed rate
- We assume here:

$$\varepsilon^2 \ll 1, \quad D \gg 1.$$

Hot-spot steady state

$$0 = \varepsilon^2 A_{xx} - A + \rho A + A_0; \quad 0 = D \left(\rho_x - 2\frac{\rho}{A} A_x \right)_x - \rho A + \bar{A} - A_0$$

Key trick: ρ_x − 2^ρ/_AA_x = A² (ρA⁻²)_x. This suggests the change of variables:

$$v = \frac{\rho}{A^2};$$

so that

 $0 = \varepsilon^2 A_{xx} - A + vA^3 + A_0; \qquad 0 = D \left(A^2 v_x \right)_x - vA^3 + \bar{A} - A_0.$

• "Shadow limit" Large $D: v(x) \sim v_0;$

$$\varepsilon^2 A_{xx} - A + vA^3 + A_0 = 0;$$
 $v_0 \int_0^L A^3 dx = (\bar{A} - A_0) L.$

▶ Anzatz: $v_0 \ll 1$, $A \sim v_0^{-1/2} w(y)$, $y = x/\varepsilon$ where w is the ground state,

$$w_{yy} - w + w^3 = 0, \ w'(0) = 0, \ w \to 0 \text{ as } |y| \to \infty;$$

then

$$v_0 \sim \frac{\left(\int_{-\infty}^{\infty} w^3 dy\right)^2}{4L^2 \left(\bar{A} - A_0\right)^2} \varepsilon^2;$$
$$A(x) \sim \begin{cases} \frac{2L(\bar{A} - A_0)}{\varepsilon \int w^3} w(x/\varepsilon), & x = O\left(\varepsilon\right)\\ A_0, & x \gg O(\varepsilon). \end{cases}$$

Critical Scaling

Based on previous computations, we now set

$$\Omega = (-1, 1),$$

$$A = A_0 + \frac{1}{\epsilon}\hat{A}, \quad v = \epsilon^2 \hat{v}$$

$$D = \frac{\hat{D}}{\epsilon}$$

Then the steady-state problem becomes

$$0 = \varepsilon^2 \hat{A}_{xx} - \hat{A} + \hat{v} (\epsilon A_0 + \hat{A})^3$$
(15)

$$0 = \hat{D} \left((A_0 + \frac{1}{\epsilon} \hat{A})^2 \hat{v}_x \right)_x - \frac{1}{\epsilon} \hat{v} (\epsilon A_0 + \hat{A})^3 + \bar{A} - A_0.$$
(16)

Relation with A Schnakenberg Model

 The steady state problem in 1D is very close to the so-called Schnakenberg model

$$0 = \varepsilon^2 u_{xx} - u + vu^p$$
$$0 = Dv_{xx} + 1 - \frac{1}{\epsilon}vu^p$$

with p = 3

$$0 = \varepsilon^{2} \hat{A}_{xx} - \hat{A} + \hat{v} \hat{A}^{3}$$
(17)
$$0 = \hat{D} \left(A_{0}^{2} \hat{v}_{x} \right)_{x} - \frac{1}{\epsilon} \hat{v} \hat{A}^{3} + \bar{A} - A_{0}.$$
(18)

▶ To see this, we consider the following problem

$$\hat{D}(a(x)v_x)_x = f(x), v'(0) = 0$$
 (19)

we have

$$v(x) - v(0) = \frac{1}{\hat{D}} \int_0^x K_a(x, s) f(s) ds$$
 (20)

where

$$K_a(x,s) = \int_s^x \frac{1}{a(x)} dx$$

• Let us now consider $a(x) = (A_0 + \frac{\gamma}{\epsilon}w(\frac{x}{\epsilon}))^2$, where w > 0 and $w \sim e^{-|y|}$. Then

$$K_{a}(x,s) = K_{A_{0}^{2}}(x,s) + O(\epsilon|s-x|) + O(|[s,x] \cap (0,2\epsilon \ln \frac{1}{\epsilon})|)$$
(21)

Main stability result (1D)

► Main result: Consider K spikes on the domain of size 2KL. Then small eigenvalues become unstable if D > D_{c,small}; large eigenvalues become unstable if D > D_{c,small} where

$$\begin{split} D_{c,\text{small}} &\sim \frac{L^4}{\varepsilon^2} \frac{\left(\bar{A} - A_0\right)^3}{A_0^2 \pi^2} \\ D_{c,\text{large}} &\sim D_{c,\text{small}} \left(\frac{2}{1 - \cos \frac{\pi}{K}}\right) > D_{c,\text{small}} \end{split}$$

 Small eigenvalues become unstable before the large eigenvalues. • Example: Take $L = 1, \overline{A} = 2, A_0 = 1, K = 2, \varepsilon = 0.07$. Then $D_{c,\text{small}} = 20.67, D_{c,\text{large}} = 41.33.$

- if $D = 15 \implies$ two spikes are stable
- if $D = 30 \implies$ two spikes have very slow developing instability
- if $D = 50 \implies$ two spikes have very fast developing instability

very similar behavior to Schnakenberg model

$$0 = \varepsilon^2 u_{xx} - u + vu^3$$

$$0 = Dv_{xx} + 1 - \frac{1}{\epsilon}vu^3.$$

Iron-Wei-Winter, J.Math.Biol. 2003

Stability: large eigenvalues

Step 1: Reduces to the nonlocal eigenvalue problem (NLEP):

$$\lambda \phi = \phi'' - \phi + 3w^2 \phi - \chi \left(\int w^2 \phi \right) w^3 \qquad \text{where } w'' - w + w^3 = 0$$
(22)

with

$$\chi \sim \frac{3}{\int_{-\infty}^{\infty} w^3 dy} \left(1 + \varepsilon^2 D (1 - \cos \frac{\pi k}{K}) \frac{A_0^2 \pi^2}{4L^4 \left(\bar{A} - A_0\right)^3} \right)^{-1}$$

 This is an oversimplified problem but captures the main characteristics

Step 2: Key identity:
$$L_0w^2 = 3w^2$$
, where $L_0\phi := \phi'' - \phi + 3w^2\phi$. Multiply

$$\lambda \phi = \phi'' - \phi + 3w^2 \phi - \chi \left(\int w^2 \phi \right) w^3$$

by $w^2 \ {\rm and} \ {\rm integrate}$ to get

$$\lambda = 3 - \chi \int w^5 = 3 - \chi \frac{3}{2} \int w^3$$

Conclusion: (22) is stable iff $\chi > \frac{2}{\int w^3} \iff D > D_{c,\text{large}}$. This NLEP in 1D can be fully solved!!

Stability: small eigenvalues

- Compute asymmetric spikes
- They bifurcate from symmetric branch
- The bifurcation point is precisely when $D = D_{c,small}$.
- This is "cheating"... but it gets the correct threshold!!
- similar computations to Iron-Ward-Wei 2001 (for Gierer-Meinhardt system) Iron-Wei-Winter 2003 (for Schnakenberg model)

Two dimensions

$$\begin{aligned} A_t &= \varepsilon^2 \Delta A - A + \hat{v} A^3 + A_0 \\ \tau(A\hat{v})_t &= D \nabla \cdot \left(A^2 \nabla \hat{v} \right) - \hat{v} A^3 + \bar{A} - A_0 \\ Neumann \ BC \end{aligned} , \quad x \in \Omega \end{aligned}$$

- Steady-state: construction is similar to 1D, but no reduction to Schnakenberg model
- ▶ Stability: of *K* hot-spots:
 - If K = 1, then a single hot-spot is stable with respect to large eigenvalues, as long as D is not exponentially large in 1/ε.
 - $\blacktriangleright\,$ If $K\geq 2,$ then the steady state is stable with respect to large eigenvalues if

$$D < \frac{1}{\varepsilon^4} \ln \frac{1}{\varepsilon} \frac{\left(\bar{A} - A_0\right)^3 |\Omega|^3 A_0^{-2}}{4\pi K^3 \left(\int_{\mathbb{R}^2} w^3 dy\right)^2};$$
 (23)

and it is unstable otherwise.

▶ Instability thresholds occur when $D = O\left(\frac{\ln \varepsilon^{-1}}{\varepsilon^4 K^3}\right) \gg 1.$

Concluding Summary

- In both models, the instability thresholds occur close to the "shadow limit", i.e. the cross-diffusion term is very large.
- Steady-state computation is essentially a shadow system, but stability computations require more.
- Cross-diffusion (directed movement) can create stable multi-spike solutions even in the absence of spatial heterogenuity.
- Chemotaxis system (crime models) can also produce multiple stable patterns
- Stability analysis leads to novel, interesting and new eigenvalue problems