

## Example, Turing diffusive instability in 1D

Recall  $\begin{cases} \frac{\partial c_1}{\partial t} = R_1(c_1, c_2) + D_1 \frac{\partial^2 c_1}{\partial x^2} \\ \frac{\partial c_2}{\partial t} = R_2(c_1, c_2) + D_2 \frac{\partial^2 c_2}{\partial x^2} \end{cases}$

With  $A = \begin{pmatrix} \frac{\partial R_1}{\partial c_1}(\bar{c}_1, \bar{c}_2) & \frac{\partial R_1}{\partial c_2}(\bar{c}_1, \bar{c}_2) \\ \frac{\partial R_2}{\partial c_1}(\bar{c}_1, \bar{c}_2) & \frac{\partial R_2}{\partial c_2}(\bar{c}_1, \bar{c}_2) \end{pmatrix}$ ,  $(\bar{c}_1, \bar{c}_2)$  a spatially uniform steady state ( $R_1(\bar{c}_1, \bar{c}_2) = R_2(\bar{c}_1, \bar{c}_2) = 0$ )

we have diffusive instability if

1.  $\text{Tr } A = a_{11} + a_{22} < 0$
2.  $\det A = a_{11}a_{22} - a_{12}a_{21} > 0$
3.  $a_{11}D_2 + a_{22}D_1 > 2\sqrt{D_1 D_2 \det A}$

Unstable perturbations,  $e^{\sigma t} \cos qx$  with  $\sigma > 0$ , then appear for  $q$  with

$$(q^2 - \frac{u}{2})^2 < \frac{u^2}{4} - w, \quad u = \frac{a_{11}D_2 + a_{22}D_1}{D_1 D_2}, \quad w = \frac{\det A}{D_1 D_2}$$

Ex (~problem 11.18 in EK)

A predator-prey model with spatial diffusion is given by

$$\begin{cases} U_t = U + \frac{1}{2}U^2 - UV + \frac{1}{2}U_{xx} & (D_1 = \frac{1}{2}) \\ V_t = UV - V^2 + DV_{xx} & (D_2 = D) \end{cases} \quad \text{For a comparison with Lotka-Volterra, see discussion in EK.}$$

$u(t, x)$  = prey population,  $v(t, x)$  = predators

Spatially uniform steady states:

$$\left\{ \bar{U}(1 + \frac{1}{2}\bar{U} - \bar{V}) = 0 \right. \quad (1)$$

$$\left. \bar{V}(\bar{U} - \bar{V}) = 0 \right. \Rightarrow \bar{V} = 0 \text{ or } \bar{V} = \bar{U}$$

$$\Rightarrow (\bar{U}_1, \bar{V}_1) = (0, 0) \text{ and } (\bar{U}_2, \bar{V}_2) = (2, 2)$$

$$A(u, v) = \begin{pmatrix} 1+u-v & -u \\ v & u-2v \end{pmatrix} \Rightarrow$$

$$A_1 = A(0, 0) = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}, \lambda_1 = 1 > 0, \lambda_2 = 0 \Rightarrow \text{unstable without diffusion} \\ \Rightarrow \text{diffusive instability not possible}$$

$$A_2 = A(2, 2) = \begin{pmatrix} 1 & -2 \\ 2 & -2 \end{pmatrix} \quad \begin{cases} 1. \text{Tr } A_2 = -1 < 0 \\ 2. \det A_2 = 2 > 0 \end{cases} \Rightarrow (2, 2) \text{ stable without diffusion}$$

$$3. \text{ becomes } 1 \cdot D - 2 \cdot \frac{1}{2} > 2\sqrt{\frac{1}{2} \cdot D \cdot 2} \Leftrightarrow D - 1 > 2\sqrt{D} \Rightarrow$$

$$(\sqrt{D} - 1)^2 > 2 \Rightarrow \sqrt{D} > \sqrt{2} + 1 \Rightarrow D > 3 + 2\sqrt{2} \approx 5.8$$

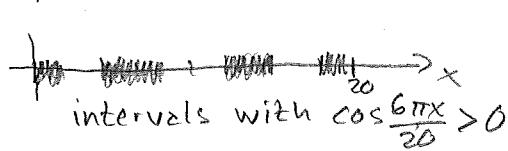
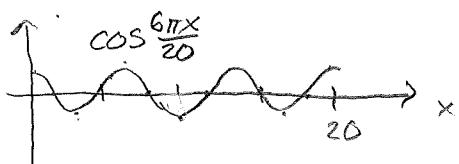
condition on  $D$  for diffusive instability

$$\text{Take } D=6 \Rightarrow u = \frac{6-1}{6 \cdot \frac{1}{2}} = \frac{5}{3} \text{ and } w = \frac{2}{3} \Rightarrow \frac{u^2}{4} - w = \frac{1}{36}$$

unstable perturbations for  $-\frac{1}{6} < q^2 - \frac{5}{6} < \frac{1}{6} \Leftrightarrow \frac{2}{3} < q^2 < 1$

With  $q = \frac{n\pi}{L}$  and  $L=20$  for example:

$$\frac{2}{3} < \frac{\pi^2}{20^2} n^2 < 1 \Leftrightarrow 27.0 < n^2 < 40.5 \Rightarrow n=6 \text{ only possibility}$$



This is the pattern that can appear due to diffusive instability with our choices of  $D$  and  $L$ .

Note: if  $D$  very near  $3+2\sqrt{2}$ ,  $q^2 = \frac{\pi^2 n^2}{L^2}$  must be very near  $\frac{5}{6}$  and maybe no integer  $n$  satisfies the inequality

This model shows that diffusion can lead to patterns in prey and predator populations. 2D models may be more interesting.

## Pattern formation in 2 space dimensions

24.3

First we analyze the aggregation of cellular slime molds.

$a(t, x, y)$  = slime mold amoeba

$c(t, x, y)$  = cAMP concentration

Use  $\nabla f = (f_x, f_y)$  gradient,  $\nabla \cdot (v_1, v_2) = \frac{\partial v_1}{\partial x} + \frac{\partial v_2}{\partial y}$  divergence of vector  
 $\Rightarrow \nabla \cdot \nabla f = f_{xx} + f_{yy} = \nabla^2 f$  Laplace of  $f$  (works also in 3D)

1D

$$\frac{\partial a}{\partial t} = -\frac{\partial}{\partial x} (\bar{J}_{rd} + \bar{J}_{ch})$$

$$\bar{J}_{rd} = -\mu \frac{\partial a}{\partial x}$$

$$\bar{J}_{ch} = \chi a \frac{\partial c}{\partial x}$$

$$\frac{\partial c}{\partial t} = D \frac{\partial^2 c}{\partial x^2} + fa - kc$$

2D (& 3D)

$$\frac{\partial a}{\partial t} = -\nabla \cdot (\bar{J}_{rd} + \bar{J}_{ch})$$

$$\bar{J}_{rd} = -\mu \nabla a \quad \left. \begin{array}{l} \\ \end{array} \right\} \text{vectors}$$

$$\bar{J}_{ch} = \chi a \nabla c$$

$$\frac{\partial c}{\partial t} = D \nabla^2 c + fa - kc$$

For ≥ 2D we have

$$\left\{ \begin{array}{l} a_t = \mu \nabla^2 a - \chi \nabla \cdot (\underbrace{a \nabla c}_{\text{non-lin. product rule}}) = \mu \nabla^2 a - \chi \nabla a \cdot \nabla c - \chi a \nabla^2 c \\ c_t = D \nabla^2 c + fa - kc \end{array} \right. \quad \xrightarrow{\text{2D}}$$

$$\left\{ \begin{array}{l} a_t = \mu(a_{xx} + a_{yy}) - \chi(a_x c_x + a_y c_y) - \chi a(c_{xx} + c_{yy}) \\ c_t = D(c_{xx} + c_{yy}) + fa - kc \end{array} \right.$$

Homogeneous steady state  $\bar{a}, \bar{c}$  if all derivatives 0  $\Rightarrow f\bar{a} = k\bar{c}$  as in 1D

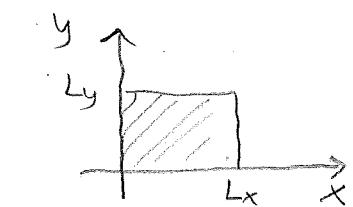
$$\text{Perturbations } \left\{ \begin{array}{l} a(t, x, y) = \bar{a} + \tilde{a}(t, x, y) \\ c(t, x, y) = \bar{c} + \tilde{c}(t, x, y) \end{array} \right. \quad \begin{array}{l} \text{constant} \\ \text{small} \end{array}$$

Neglecting 2<sup>nd</sup>-order terms, we find the linearized system

$$\left\{ \begin{array}{l} \tilde{a}_t = \mu(\tilde{a}_{xx} + \tilde{a}_{yy}) - \chi \bar{a}(\tilde{c}_{xx} + \tilde{c}_{yy}) \\ \tilde{c}_t = D(\tilde{c}_{xx} + \tilde{c}_{yy}) + f\bar{a} - k\bar{c} \end{array} \right. , \text{ just as in 1D}$$

With a domain  $0 < x < L_x, 0 < y < L_y$ ,  
 and no-flux Neumann boundary conditions,  
 look for solutions

$$\left\{ \begin{array}{l} \tilde{a}(t, x, y) = A e^{\sigma t} \cos q_1 x \cos q_2 y \\ \tilde{c}(t, x, y) = C e^{\sigma t} \cos q_1 x \cos q_2 y \end{array} \right. \Rightarrow$$



(we could also study, e.g., circular domains but this requires more new studies of solution expressions)

$$\begin{cases} \sigma \tilde{\alpha} = \mu(-q_1^2 \tilde{\alpha} - q_2^2 \tilde{\alpha}) - \chi \tilde{\alpha} (-q_1^2 \tilde{c} - q_2^2 \tilde{c}) \\ \sigma \tilde{c} = D(-q_1^2 \tilde{c} - q_2^2 \tilde{c}) + f \tilde{\alpha} - k \tilde{c} \end{cases}$$

Let  $Q^2 = q_1^2 + q_2^2$  and divide by  $e^{\sigma t} \cos q_1 x \cos q_2 y \Rightarrow$

$$\begin{cases} \sigma A = -\mu Q^2 A + \chi \tilde{\alpha} Q^2 C \\ \sigma C = -D Q^2 C + f A - k C \end{cases} \quad \text{exactly the same as in 1D but } q \text{ replaced by } Q$$

$\Rightarrow$  solutions with  $A \neq 0, C \neq 0$  and  $\sigma > 0$  ( $\sigma_2 < 0$  always) if

$$\underline{\mu(DQ^2+k) < f\chi\tilde{\alpha}} \quad \text{condition for aggregation}$$

$$q_1 = \frac{m\pi}{L_x}, q_2 = \frac{n\pi}{L_y}, m, n = 0, 1, 2, \dots \Rightarrow Q^2 = \left(\frac{m^2}{L_x^2} + \frac{n^2}{L_y^2}\right)\pi^2$$

If  $L_x > L_y$  ( $\frac{1}{L_x^2} < \frac{1}{L_y^2}$ ), the first pattern to appear is  $m=1, n=0$ , i.e.

$\cos \frac{\pi x}{L_x}$  ( $m=n=0$  not a "pattern"). The second is either  $m=0, n=1$ ,

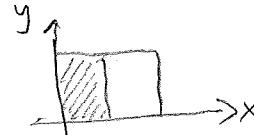
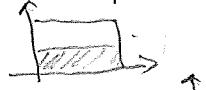
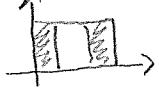
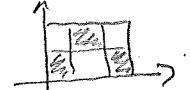
$\cos \frac{m y}{L_y}$  if  $\frac{1}{L_y^2} < \frac{4}{L_x^2}$  ( $L_x < 2L_y$ ), or  $m=2, n=0$ ,  $\cos \frac{2\pi x}{L_x}$  if  $L_x > 2L_y$ .

Patterns of amoeba concentrations that can appear satisfy

$$\underline{\mu(D\pi^2(\frac{m^2}{L_x^2} + \frac{n^2}{L_y^2}) + k) < f\chi\tilde{\alpha}}$$

See pictures, page 524 in EK

Ex  $L_x = \sqrt{2}$ ,  $L_y = 1$ ,  $Q^2 = (\frac{m^2}{2} + n^2)\pi^2$ , order of appearing:

<u><math>m, n</math></u>	<u><math>\frac{m^2}{2} + n^2</math></u>	<u>function</u>	
1, 0	1/2	$\cos \frac{\pi x}{\sqrt{2}}$	
0, 1	1	$\cos \pi y$	
1, 1	3/2	$\cos \frac{\pi x}{\sqrt{2}} \cos \pi y$	
2, 0	2	$\cos \frac{2\pi x}{\sqrt{2}}$	
2, 1	3	$\cos \frac{2\pi x}{\sqrt{2}} \cos \pi y$	
0, 2	4	$\cos 2\pi y$	
3, 0	9/2	$\cos \frac{3\pi x}{\sqrt{2}}$	
3, 1	11/2		
2, 2	6		
:	:		:

shows where function positive  
(where slime molds move lumped together)

Linear combinations of patterns satisfying the aggregation condition give more general patterns for  $\tilde{\alpha}$  (and  $\tilde{c}$ )