

(ii) chemotaxis

Deriving PDE models
continued

↳ movement in response to chemical signal

model organism: Dictyostelium discoideum
(slime mold)

Single-cellular amoeba, eats bacteria

low food $\rightarrow$ produce cAMP $\rightarrow$ aggregate

goal: model the process of secreting cAMP

& aggregating in response to cAMP gradients

[again, could use cell-level agent-based model, likely to be v. complicated]
How many variables?

two densities $\begin{cases} u(x,t) & \text{amoebae concentration} \\ c(x,t) & \text{concentration of cAMP} \end{cases}$

PDE model

note: run & tumble behavior

amoebae: diffusion + advection by external field

$v = \nabla c = \text{gradient of cAMP}$

(why not birth & death? always consider relevant timescales)

cAMP: diffusion + creation + degradation

$$\begin{cases} u_t = D_u \Delta u - \underbrace{\chi \nabla \cdot (u \nabla c)}_{\substack{\text{advection according} \\ \text{to cAMP gradient}}} \\ c_t = D_c \Delta c + \underbrace{\alpha u}_{\substack{\text{prod. by amoebae}}} - \underbrace{\beta c}_{\text{degradation}} \end{cases}$$

strength of response to gradient

Note: family of spatially homogeneous constant solns

say we set $c(x,0) = c_0$ ~~$c(x,0) = c_0$~~ ^{$u(x)$} ~~$c(x,0) = c_0$~~

$$u(x,0) = u_0$$

$$c(x,0) = c_0 \quad \forall x$$

$$\text{then } u_t = \cancel{D_u \Delta u}^0 - \chi \nabla \cdot (\cancel{u \nabla c})^0$$

$$c_t = \cancel{D_c \Delta c}^0 + \alpha u_0 - \beta c_0$$

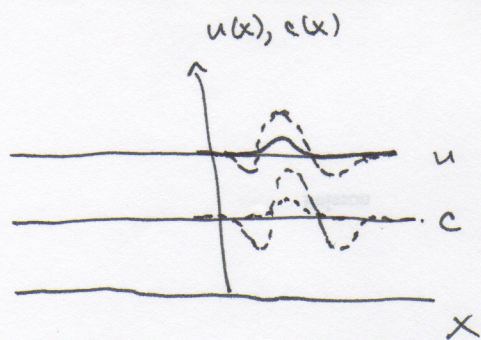
$$\Rightarrow \cancel{u_0} = \frac{\beta c_0}{\alpha} \text{ is steady soln for any } c_0$$

However, as w/ Turing bifurcation, spatially homogeneous steady state not nec. stable

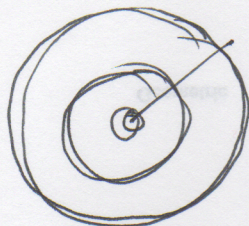
↳ recall in activator-inhibitor system key was to have inhibitor diffuse much faster than activator to get pattern forming instability

↳ here more complicated, but similar idea, ^{strength} (rate of cAMP prod. · rate of response) roughly speaking want $\alpha \chi$ to dominate

βD_u (rate of degradation · diffusion of amoebae)



↳ imagine radial



Only do if less than
(very heuristic, $\frac{1}{2}$ time gone)
like Bjorn's ex. of Turing?

(If response to CAMP $\neq$ incr.
CAMP prod. dominate
diffusion $\neq$ CAMP degradation
can see this kind of pattern
forming behavior)

Spiral Fundamentally 2D - can't look at quite
this way

(iii) Calcium waves:

Nature review ~10 yr. ago:

"Almost everything we do is controlled by Ca^{2+} - how we move, how our hearts beat, and how our brains process information and store memories."

→ believe information in signals may have to do w/ wave frequency rather than amplitude

- intracellular oscillations

↳ *Xenopus* oocytes: spatial waves that propagate through cell

- full purpose not known, but involved in avoidance of polyspermy (multiple fertilization) & other key developmental processes



like  1D front

but also see spiral waves
targets 



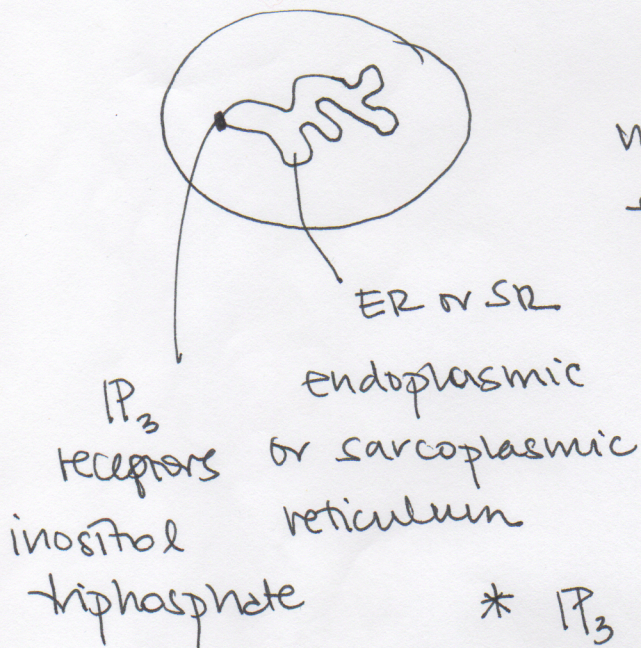
- intercellular oscillations

↳ airway epithelial cells: calcium waves generated by mechanical stimuli; purpose likely stimulation of cilia that move mucus: Ca^{2+} waves can coordinate movement of cilia (synchronized beat rhythm)

↳ contraction of heart muscle: [see visualizations on Scholarpedia]

- under normal conditions action potentials propagate along network of Purkinje fibers
(Ca^{2+} participates in creation of AP's)
- damaged tissue or changes in cellular structures can lead to abnormally long AP duration & shortened refractory times
 - ↳ can produce spiral waves that lead to tachycardia (heart rates 100-200 bpm)
 - or ventricular fibrillation (spiral breakup → often fatal)

=



Ca^{2+} concentration in cytosol much lower than surrounding extracellular matrix → actively pumped out (sequestered to ER)

* IP₃ → Ca^{2+} released from ER to cytoplasm

* Ca^{2+} exhibits positive feedback on short timescales (incr. activity of IP₃ receptors)

* Ca^{2+} inhibits negative feedback on longer timescales

Model for *Xenopus* oocytes

(can use stochastic Markov-type models for individual receptors — well just use density spatio-temporal (PDE) model)

$$P_t = D_P \Delta P - \overbrace{k_P P}^{\text{degradation}}$$

could include production
 ↳ assume longer timescale/
 not relevant here

$$C_t = D_C \Delta C + k_f \underbrace{\left(\mu_0 + \frac{\mu_1 P}{\mu_2 + P} \right)}_{\text{IP}_3 \text{ activation}} \underbrace{\left(b + \frac{(1-b)c}{k_1 + c} \right)}_{\text{calcium activation}} h - \underbrace{\frac{\gamma c}{k_2 + c}}_{\text{calcium inactivation}} + \underbrace{\beta}_{\text{pump out}} - \underbrace{\gamma}_{\text{leak in}}$$

rate const. / scaling factor

[note what happens when $p \rightarrow 0, \infty$; $c \rightarrow 0, \infty$]

$$\tau h_t = \frac{k_2}{k_2 + c} - h$$

~~say h starts at some const, $c \rightarrow 0$ what happens~~

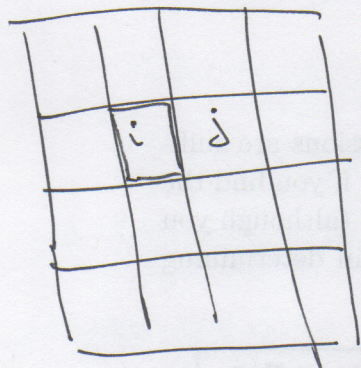
as $c \uparrow$, $\frac{k_2}{k_2 + c} \downarrow$ so $\sim$ exponential decay

as h gets small, $\frac{k_2}{k_2 + c} \rightarrow 0$ influx in c also gets small $\rightarrow 0$

but note time scales not same

↳ this is how we capture initial pos. feedback followed by later neg. feedback

To model intercellular waves, one approach is to use square lattice of cells w/ same internal dynamics as above, w/ Ca^{2+} across cells but IP_3 flux w/ gap junctions.



In particular use flux

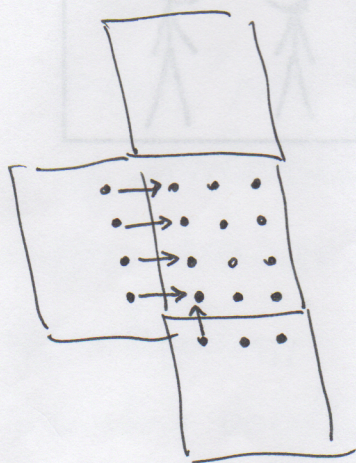
$$-k(p_i - p_j)$$

to couple cells

Solve numerically

using some # of grid points

inside cell and some # cells



Just scratching the surface - lots more in papers posted on canvas if interested
also Sneyd & Keener Math. Physiol.