Part V

Spatial interactions and pattern models

Turing Model I: mathematics and simulation Week 10, 2019

Patterns in Nature



The detailed gene regulatory networks for sea urchin development Endomesoderm Specification up to 30 Hours This model is frequently revised. It is based on the latest The current VfA incli Maternal Inputs laboratory data, some of which is not yet published. Smadar de Leon, Joel Sagar Damle, Andrew R published data, is bas Mat G-cadherin Isabelle Peter (endod Additional data sources for selected notes: 1: McClay lab; 2: Angerer lab; and Joel Smith (CP dom Mat cß Mat Wnt6 Mat Otx 3,4: McClay lab; 5: Rogers and Calestani, 2010; 6: Croce and McClay and expression d frizzled unkn mes/end rep KGSK-3 Su(H):NIC SU(H) Ubiq nβ-TCF Runx Nucl. nβ-TCF ECNS SoxB1 Hnf 6 Wnt16 βαOtx Notch Activin B Blimpl r11pm Wnt8 Hox11/13b Delta Hox11/13b PMC unkn repr Wnt8 unkn mes activ Ese Prox Gene X Ubiq Blimp1 Oral NSM βαOtx SoxC ES Su(H):N r11pm Delta γ(2) α TBr Alx1 HesC Otx Ubiq Notch Vegf3 Veg1 Endoderm Tel Erg Hex Tgif FoxN2/3 nß-TCF Aboral NSM Brn1/2/4 FoxO VEGFR Veg2 Endoderm Veg1 Ectoderm z 166 marily shows the endomesoderm network architecture ne addition of all PMC components starting at 6 hours, h signal from PMC to Veg2, the presence of Wnt8 in tures are no longer present by 21 hours. Consult the SuTx Decorin Msp130 Dpt FoxF Smo PMC Primary mesenchyme cells Ubiq -FvMo1,2,3 NSM non-skeletogenic mesoderm Sm30 G-cadherin Ficolin Small Mic/CP Abo NSM Diff. Oral NSM Diff. Copyright © 2001–2011 Hamid Bolouri and Dic Davidson

Ubiq=ubiquitous; Mat = maternal; activ = activator; rep = repressor, unkn = unknown; Nucl. = nuclearization; χ = β-catenin source; nβ-TCF = nuclearized b-β-catenin-Tcf1; ES = early signal; ECNS = early cytoplasmic nuclearization system; Zyg. N. = zygotic Notch

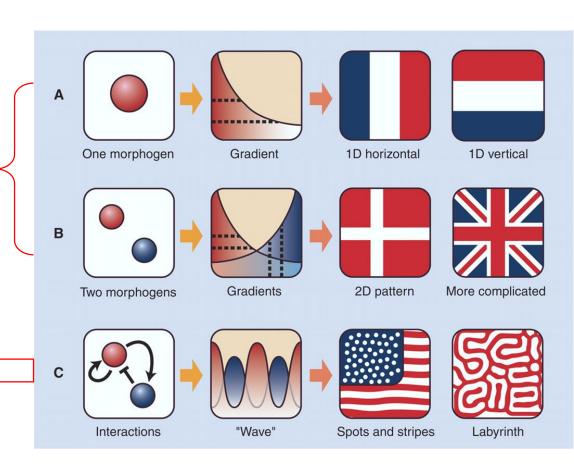
What do we mean by pattern?

 Stable behavior in time (most commonly, cell differentiation) that is regular in space (regular ="following rules")

Two main rules of pattern formation

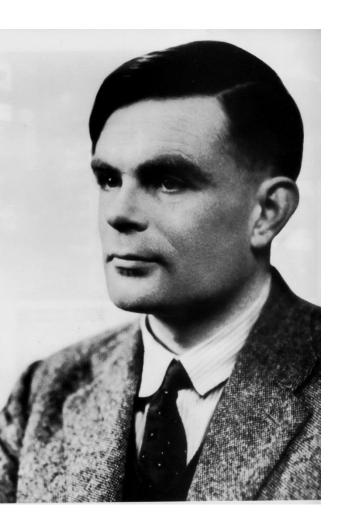
Morphogen gradient

- Positional information laid out externally
- Cells respond passively (gene expression and movement)
- Reaction-diffusion systems
 - Pattern formation autonomous
 - Typically involve mutual signaling and movement



Reaction-Diffusion Model as a Framework for Understanding Biological Pattern Formation, S Kondo and T Miura, Science 329, 1616 (2010) 5

Alan Turing (1912-1954)



- One of greatest scientists in 20th century
- Designer of Turing machine (a theoretical computer) in 1930's
- Breaking of U-boat Enigma, saving battle of the Atlantic
- Initiate nonlinear theory of biological growth



THE CHEMICAL BASIS OF MORPHOGENESIS

By A. M. TURING, F.R.S. University of Manchester

(Received 9 November 1951—Revised 15 March 1952)

It is suggested that a system of chemical substances, called morphogens, reacting together and diffusing through a tissue, is adequate to account for the main phenomena of morphogenesis. Such a system, although it may originally be quite homogeneous, may later develop a pattern or structure due to an instability of the homogeneous equilibrium, which is triggered off by random disturbances. Such reaction-diffusion systems are considered in some detail in the case of an isolated ring of cells, a mathematically convenient, though biologically unusual system. The investigation is chiefly concerned with the onset of instability. It is found that there are six essentially different forms which this may take. In the most interesting form stationary waves appear on the ring. It is suggested that this might account, for instance, for the tentacle patterns on *Hydra* and for whorled leaves. A system of reactions and diffusion on a sphere is also considered. Such a system appears to account for gastrulation. Another reaction system in two dimensions gives rise to patterns reminiscent of dappling. It is also suggested that stationary waves in two dimensions could account for the phenomena of phyllotaxis.

The purpose of this paper is to discuss a possible mechanism by which the genes of a zygote may determine the anatomical structure of the resulting organism. The theory does not make any new hypotheses; it merely suggests that certain well-known physical laws are sufficient to account for many of the facts. The full understanding of the paper requires a good knowledge of mathematics, some biology, and some elementary chemistry. Since readers cannot be expected to be experts in all of these subjects, a number of elementary facts are explained, which can be found in text-books, but whose omission would make the paper difficult reading.

What did he discover

$$\partial_t u_1 = f_1(u_1, u_2) + D_1 \partial_x^2 u_1,$$
 Generalized reaction-
 $\partial_t u_2 = f_2(u_1, u_2) + D_2 \partial_x^2 u_2,$ diffusion equations

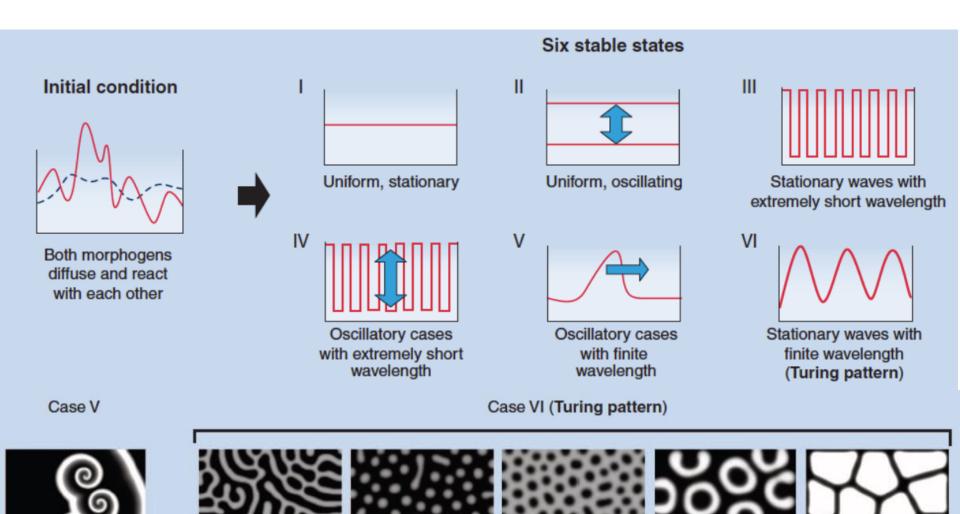
His insights:

- ≥2 interacting chemicals are needed for pattern formation to occur.
- 2. Diffusion in a reacting systems can be a destabilizing influence.
- 3. Instability can cause growth of structure at a particular wavelength
- 4. Diffusion coefficients of two reagents differ substantially.

The underlying reactions relates to gene regulations and signal transduction networks

Alan Turing - one of the first systems biologists?

Six different models and patterns



Stability analysis of Turing's model

Turing's surprising and important discovery was that there are conditions: the spatially uniform state is stable in the absence of diffusion but can be unstable to nonuniform perturbation precisely because of diffusion.

$$\partial_t u_1 = f_1(u_1, u_2) + D_1 \partial_x^2 u_1,$$

$$\partial_t u_2 = f_2(u_1, u_2) + D_2 \partial_x^2 u_2,$$

Stationary uniform base solution: $\mathbf{u}_{b} = (\mathbf{u}_{1b}, \mathbf{u}_{2b})$ so that:

$$f_1(u_{1b}, u_{2b}) = 0,$$

 $f_2(u_{1b}, u_{2b}) = 0.$

By linearizing about the base solution, perturbation $(\delta u1(t,x), \delta u2(t,x))$ follows the his equations:

$$\partial_t \delta u_1 = a_{11} \delta u_1 + a_{12} \delta u_2 + D_1 \partial_x^2 \delta u_1, \partial_t \delta u_2 = a_{21} \delta u_1 + a_{22} \delta u_2 + D_2 \partial_x^2 \delta u_2.$$

$$a_{ij} = \frac{\partial f_i}{\partial u_j}\Big|_{\mathbf{u}_b} \quad \mathbf{A} = \partial \mathbf{f}/\partial \mathbf{u}$$

It can be further expanded into:

$$\delta \mathbf{u} = \delta \mathbf{u}_q \, e^{\sigma_q t} \, e^{iqx} = \left(\begin{array}{c} \delta u_{1q} \\ \delta u_{2q} \end{array} \right) e^{\sigma_q t} e^{iqx}$$

with growth rate σ_{α} and wave number q.

Stability analysis of Turing's model

It can be converted to an eigenvalue problem:

$$\mathbf{A}_q \, \delta \mathbf{u}_q = \sigma_q \, \delta \mathbf{u}_q$$

where

$$\mathbf{A}_q = \mathbf{A} - \mathbf{D}q^2 = \begin{pmatrix} a_{11} - D_1 q^2 & a_{12} \\ a_{21} & a_{22} - D_2 q^2 \end{pmatrix}$$

For given eigenvalue σ_{iq} , the particular solution with wave number q:

$$\left(c_{1q}\,\delta\mathbf{u}_{1q}\,e^{\sigma_{1q}t}+c_{2q}\,\delta\mathbf{u}_{2q}\,e^{\sigma_{2q}t}\right)e^{iqx}$$

The uniform solution u_b is stable if both eigenvalues σ_{iq} have negative real parts for all wave number q.

$$0 = \det (\mathbf{A}_q - \sigma_q \mathbf{I}) = \sigma_q^2 - (\operatorname{tr} \mathbf{A}_q)\sigma_q + \det \mathbf{A}_q$$

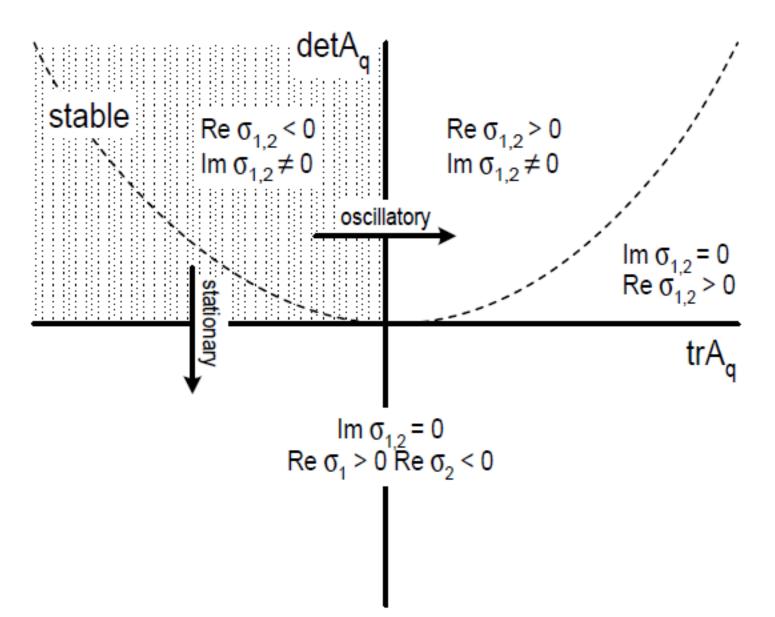
$$\sigma_q = \frac{1}{2} \text{tr} \mathbf{A}_q \pm \frac{1}{2} \sqrt{(\text{tr} \mathbf{A}_q)^2 - 4 \det \mathbf{A}_q}$$

The criteria for stability (both)

$$tr \mathbf{A}_q = a_{11} + a_{22} - (D_1 + D_2)q^2 < 0,$$

$$\det \mathbf{A}_q = (a_{11} - D_1 q^2)(a_{22} - D_2 q^2) - a_{12} a_{21} > 0.$$

Stability regions of Turing Model



Stability criteria

The criteria for stability (both)

$$tr \mathbf{A}_q = a_{11} + a_{22} - (D_1 + D_2)q^2 < 0,$$

$$\det \mathbf{A}_q = (a_{11} - D_1 q^2)(a_{22} - D_2 q^2) - a_{12} a_{21} > 0.$$

If uniform state is stable ($D_1 = D_2 = 0$)

$$a_{11} + a_{22} < 0,$$

 $a_{11}a_{22} - a_{12}a_{21} > 0.$

So $tr \mathbf{A}_q$ always negative, only way for diffusion to destabilize the uniform state is $\det \mathbf{A}_q < 0$. Taking derivative of $\det \mathbf{A}_q$ against q^2 , we can find most unstable q^2_m :

 $q_m^2 = \frac{D_1 a_{22} + D_2 a_{11}}{2D_1 D_2}.$

$$\det \mathbf{A}_{q_m} = a_{11}a_{22} - a_{12}a_{21} - \frac{(D_1a_{22} + D_2a_{11})^2}{4D_1D_2}.$$

If detAq<0, then $D_1a_{22} + D_2a_{11} > 2\sqrt{D_1D_2(a_{11}a_{22} - a_{12}a_{21})}$

Therefore a11 and a22 must have opposite sign, a12 and a21 also must have.

Stability criteria

Assuming a11>0, a22<0 let's define two diffusion lengths:

$$l_1 = \sqrt{\frac{D_1}{a_{11}}}$$
 and $l_2 = \sqrt{\frac{D_2}{-a_{22}}}$,

$$l_2 = \sqrt{\frac{D_2}{-a_{22}}},$$

Then
$$q_m^2 = \frac{1}{2} \left(\frac{1}{l_1^2} - \frac{1}{l_2^2} \right) > \sqrt{\frac{a_{11}a_{22} - a_{12}a_{21}}{D_1D_2}}$$
.

Unstable condition: $I_2>I_1$, As a11>0, "activator", a22<0, 'inhibitor". 12>11: local activation with long range inhibition.

This local activation with long range inhibition mechanism is not explicitly expressed by Alan Turing. 20 years later, it was rediscovered and explicitly stated by Grierer and Meinhardt:



The authors of the activator-inhibitor model: Hans Meinhardt (to the left) and Alfred Gierer.

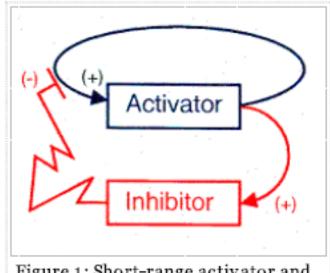


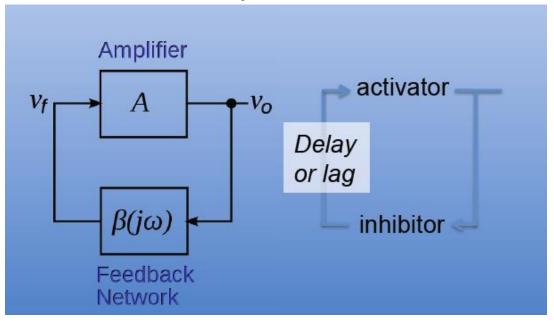
Figure 1: Short-range activator and long-range inhibitor in Gierer-Meinhardt model

$$rac{\partial a}{\partial t} =
ho \, rac{a^2}{h} - \mu_a a + D_a \, rac{\partial^2 a}{\partial x^2} +
ho_a$$

$$rac{\partial h}{\partial t} =
ho a^2 - \mu_h h + D_h \, rac{\partial^2 h}{\partial x^2} +
ho_{h5}$$

Intuitive understanding of Turing patterning

 As a spatial generalization of the delayedfeedback system oscillation in time



If feeding back an opposing signal at a different point in time can produce oscillation in time, the feeding back an opposing signal at a different point in space might be able to produce oscillation in space.

Harmonic oscillator (electronics)

Biological oscillator

The Inhibitor must act over a longer characteristic spatial scale than the activator

Intuitive understanding of Turing patterning

As a spatial generalization of bistability

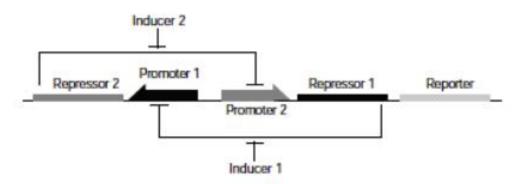


Figure 1 Toggle switch design. Repressor 1 inhibits transcription from Promoter 1 and is induced by Inducer 1. Repressor 2 inhibits transcription from Promoter 2 and is induced by Inducer 2.

Gardner, Cantor and Collins, 2000

Dynamical systems can be bistable (or multistable) if there is nonlinear positive feedback.

Autoactivation must be more than linear sensitive.

$$\frac{du}{dt} = \frac{\alpha_1}{1 + v^{\beta}} - u$$

$$\frac{dv}{dt} = \frac{\alpha_2}{1 + u^{\gamma}} - v$$
High State
Low state

Phenomena of Turing pattern

- Local rules, no long distance planning
- Exponentially amplification of small perturbation
- Select certain wavelengths for amplification

Matlab examples of 1D Grierer-Meinhardt model with uniform initial condition

- File: meinh1dflat.m
 - Parameters: nx, nt, Da, Dc, noise level
 - Results:
 - figure 1: spatial-temporal diagram of activator: x:time, y:space
 - figure 2: movie of activator and inhibitor
 - Run 4 times to show different results each time

Matlab examples of 2D Grierer-Meinhardt model with uniform initial condition

- File2: meinh2dflat.m
 - Parameters: Da, Dc
 - Results:
 - figure 1: movie of activator figure 2
- File3: meinh2dstripe.m
 - Adding saturation of activation of a to a
 - Parameters: sat (saturation of a to a)
 - Results:
 - figure 1: movie of activator figure 2
 - Run 2 times to compare no saturation vs saturation.

A real application of Turing pattern on Science

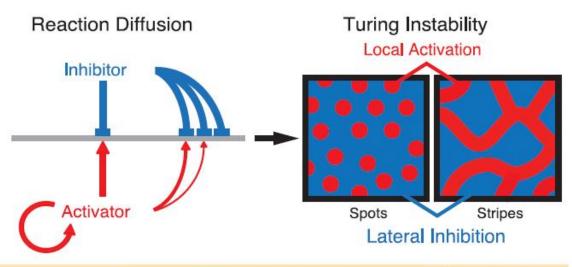
MEMBRANES

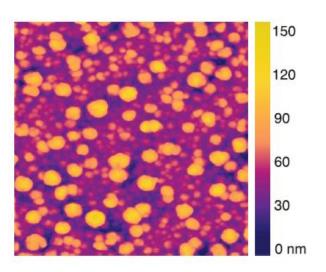
Polyamide membranes with nanoscale Turing structures for water purification

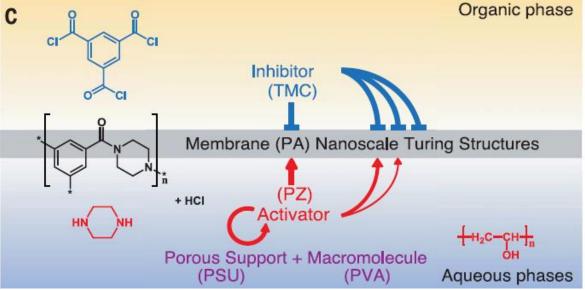
Zhe Tan, 1 Shengfu Chen, 1 Xinsheng Peng, 2 Lin Zhang, 1* Congjie Gao 1,3

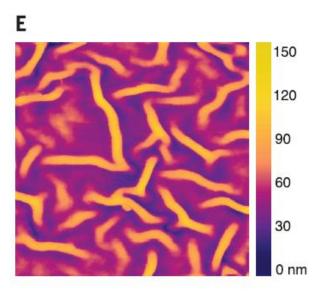
The emergence of Turing structures is of fundamental importance, and designing these structures and developing their applications have practical effects in chemistry and biology. We use a facile route based on interfacial polymerization to generate Turing-type polyamide membranes for water purification. Manipulation of shapes by control of reaction conditions enabled the creation of membranes with bubble or tube structures. These membranes exhibit excellent water-salt separation performance that surpasses the upper-bound line of traditional desalination membranes. Furthermore, we show the existence of high water permeability sites in the Turing structures, where water transport through the membranes is enhanced.

A real application of Turing pattern on Science









Occasionally the field of pattern formation is expanding, such as embryo development

• Simple implementation: time-dependent adjustment of dx, dy.

Matlab examples of 1D Grierer-Meinhardt model with growth domain

- File: meinh1dexpand.m
 - Parameters: nx, nt, Da, Dc
 - Results:
 - figure 1: spatial-temporal diagram of activator: x:time, y:space
 - figure 2: movie of activator and inhibitor in expanding space
 - Run 4 times to show different results each time

Matlab examples of 2D Grierer-Meinhardt model with growth domain

- File: meinh2dexpand.m
 - Parameters:
 - Results:
 - figure 1: movie of activator in expanding space
- File: meinh2dexpandstripe.m
 - Parameters:
 - Results:
 - figure 1: movie of activator in expanding space

This model can be used to predict the timing and location of new

hair follicle formation in mouse embryo

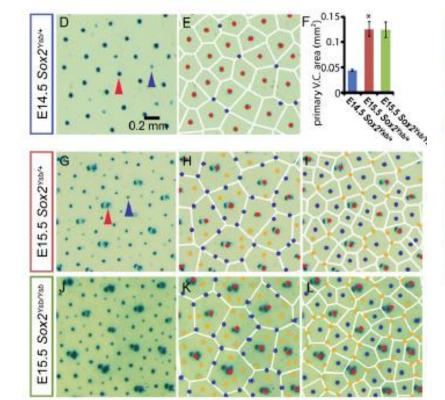
Predicting the spatiotemporal dynamics of hair follicle patterns in the developing mouse

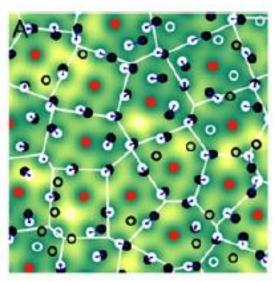
Chi Wa Cheng^{a,1}, Ben Niu^{a,1}, Mya Warren^{b,1}, Larysa Halyna Pevny^{c,2}, Robin Lovell-Badge^d, Terence Hwa^{b,3}, and Kathryn S. E. Cheah^{a,3}

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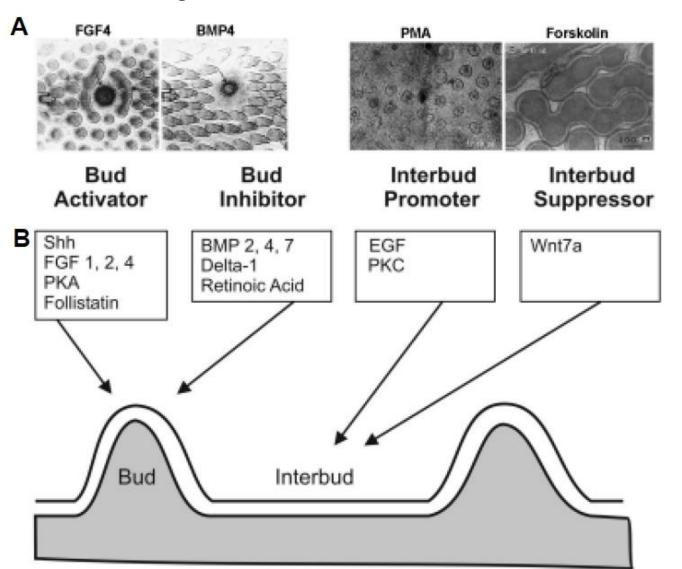






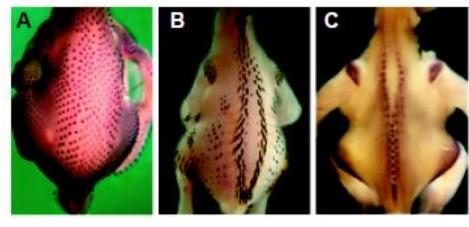
Chicken feather bud pattern and Turing modeling

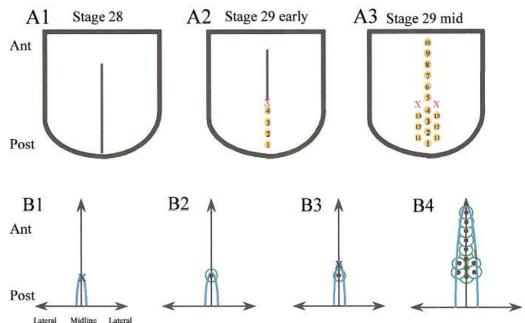
With belief in Turing model, biologists have extensively screening for activators and inhibitors.



Chicken feather bud has very different pattern

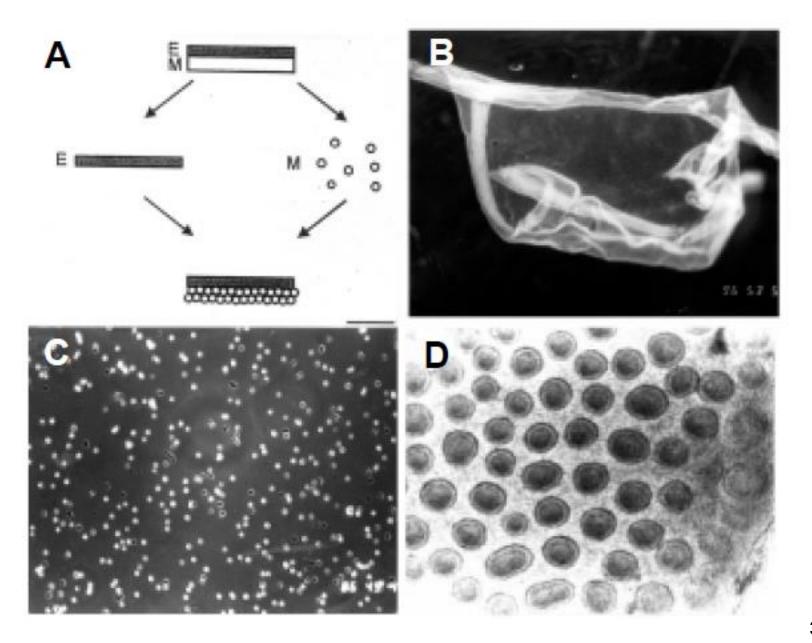
- Spatially regular (close to hexagonal)
- Temporally sequential
- Faster than expansion





Is it possible that regular chicken feather bud patterning follows Turing mechanism?

- Turing patterns are inherently spatial temporal heterogeneous.
- In vitro reconstitution experiments also demonstrated this.



Matlab examples of 1D Grierer-Meinhardt model with spatial localized initial condition

- File: meinh1dseed.m
 - Parameters: nx (250-350), noise to initial condition
 - Results:
 - figure 1: spatial-temporal diagram of activator: x:time, y:space
 - figure 2: movie of activator and inhibitor in expanding space

Without noise, to a degree, 1D model can generate sequential regular pattern

Matlab examples of 2D Grierer-Meinhardt model with spatial localized initial condition

- File: meinh2dseed.m
 - Parameters:
 - Results:
 - figure 1: spatial-temporal diagram of activator: x:time, y:space
 - figure 2: movie of activator and inhibitor in expanding space

Without noise, 2D model cannot generate sequential regular pattern

Possible mechanisms on top of Turing model for chicken feather bud

- "Competent wave":
- 1.Move from posterior to anterior, middle line to laterial
- Enable Turing pattern formation within area of competent
- 3. Mesenchymal cell competent (differentiation?)

Solid experimental proof of Turing model

 Next week, we will discuss a few critical experimental studies in the last two years.