

Embryonic semaphores Morphogens and contact-mediated signalling in developmental patterning

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Workshop on "Flags, Landscapes, Signaling"

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Positional information via morphogen gradients

- The concept of "positional information" relates to the idea that cells in developing embryos receive *external cues* to identify their positions within a coordinate system (with respect to boundaries), and then <u>interpret</u> this information to achieve distinct states.
- This can be achieved through the detection of the local concentration of an external gradient of signaling molecules (morphogens).
- A morphogen gradient arises from the secretion of molecules from a localized source, which diffuse and degrade over time.



Role of morphogen gradients

The graded distribution of morphogens allows the embryo to establish polarity along different axes.

The presence of genes that are expressed at high/low morphogen concentration aids in mesodermal patterning.



image source: H. L. Ashe & J. Briscoe, *Development* **133**: 385–394 (2006).





"Tuning" a flag



- * Is it possible to vary the size of cell-fate domains in a regulated manner? * Could inter-cellular communication play a role?
- * How do we prevent pathological outcomes that do not preserve the number and sequence of these domains?





The Notch signalling pathway





Three-gene interpretation module

- * We use a three-gene interpretation module* associated with the Sonic hedgehog (Shh) morphogen gradient in vertebrate neural tubes.
- Pax6 expression occurs even in the absence of the Shh morphogen.

* N. Balaskas et al., *Cell* **148**:273 (2012).

Model of Balaskas et al. (2012)

The time evolution of the expression levels of the three genes are described by:

- - α, β, γ : maximal growth rates
 - k_1, k_2, k_3 : decay rates
 - *K* : threshold of response function
 - h_1, h_2, h_3, h_4, h_5 : Hill coefficients
 - S_M : morphogen level

N. Balaskas et al., *Cell* **148**:273 (2012).

- whose concentration decays exponentially away from the source.
- a result of morphogen molecules binding to receptors on the cell membrane:

$$S_M(x) = S_N$$

* We consider a linear array of cells (i = 1, 2, ..., 30) responding to a morphogen (Shh)

This spatial profile is mirrored in the concentration of signaling molecules (S_M) released as

 $M_M(0) \exp(-x/\lambda_M)$

Behaviour of an uncoupled array of cells

We simulate this model over a linear array.

We track the expression levels *Y* of each of the three genes.

After an initial transient period. three cell-fate domains arise.

For simplicity, we assume that NICD either upregulates all of the genes or downregulates all of them.

Incorporating juxtacrine signaling

Incorporating juxtacrine signaling

If the concentrations of NICD and ligand are N^b and L, respectively, then:

 $\varphi_{1,2,3}$: strength of upregulation

 $\xi_{1,2,3}$: strength of downregulation

 K_N : threshold of response function

 $-k_3R$

N. Balaskas et al., *Cell* **148**:273 (2012).

В

For simplicity, we assume that the ligand is either activated by all of the genes or inhibited by each of them.

Ligand regulation

Modelling Notch signaling

The time evolution of the concentrations of ligand (L) and NICD (N^b) is given by:

$$\frac{dL}{dt} = \frac{\beta_L + \phi_4 \frac{B}{K} + \phi_5 \frac{W}{K} + \phi_6 \frac{R}{K}}{1 + \zeta_4 \frac{B}{K} + \zeta_5 \frac{W}{K} + \zeta_6 \frac{R}{K}} - \frac{L}{\tau_L}$$
$$\frac{dN^b}{dt} = \frac{\beta_{N^b} L^{trans}}{K + L^{trans}} - \frac{N^b}{\tau_{N^b}}$$

release of the receptors intracellular domain (N^b) to be released.

- - $\phi_{4.5.6}$: strength of promotion
 - $\zeta_{4.5.6}$: strength of repression
 - β_{L,N^b} : maximum growth rates
 - τ_{L,N^b} : mean lifetimes
- The trans-binding of Notch receptors to ligands (L^{trans}) of neighbouring cells leads to the
 - N. Balaskas et al., *Cell* **148**:273 (2012).

Classes of intercellular interactions

 θ_5

type III : $(\theta_i, 1, 0, \theta_j)$

j = 4,5,6.

For down $(\text{or } \phi = 0)$

 $\phi \geq 1$).

Using the same model, we can describe 4 classes of intercellular interaction, specified by the parameter set $(\varphi_i, \xi_i, \phi_j, \zeta_j)$, i = 1, 2, 3,

regulation,
$$\varphi = 0$$
.

For upregulation $\xi = 1$ (or $\zeta = 1$) while $\varphi \ge 1$ (or

type II : $(0, \theta_i, \theta_j, 1)$

type IV : $(\theta_i, 1, \theta_j, 1)$

Spatio-temporal evolution of the expression levels of the patterning genes for the four coupling types

Final expression levels of the patterning genes for the four coupling types

- * We consider 10^5 parameter sets $\Theta = \{\theta_1, ..., \theta_6\}$, where each θ_k is sampled from [1,10] for activation parameters and [0.1,1] for inhibition parameters.
- We quantify the variation in observed flags in terms of n_B, the number of *fate boundaries*, and d_H, the Hamming distance to the idealized flag (obtained in the absence of coupling).
- Repression of B, W, R almost always results in flags having two boundaries.
- * Coupling types I & II are much closer to the idealized flag in terms of d_H .

The Sobol' method

- * This is a variance-based sensitivity analysis technique that we can use to quantify the contribution of each of the parameters $\Theta = \{\theta_1, ..., \theta_6\}$ in determining the cell fates.
- ★ We consider the final state of each cell *i* to be represented by $F_i \in \{0,1,2\}$ corresponding to blue, white and red.
- * At each site *i* we compute the variance σ^2 in fate F_i across realizations.
- * We measure the first-order sensitivity indices *S*1 at each site *i* by computing the variance of $\langle F_i | \theta_j \rangle_{\theta_{k(\neq j)}}$ and normalising by the corresponding σ^2 .

Dependence of cell fate variation on location

- * The variance σ^2 is mostly negligible in types I & II.
- Only θ₂ and θ₃
 contribute significantly
 to all coupling types.
- The bulk of variation
 in *F_i* can be explained
 from *S*1 alone.

Sensitivity of segment lengths

- * For types I & II, the lengths of the white and red segments seen from the ensemble simulations are narrowly distributed around the values of the uncoupled case $(l_R^* = 10, l_W^* = 10).$
- * Pie charts display results of a sensitivity analysis in terms of l_R (outer), l_W (middle) and l_B (inner).

Dependence of boundary location on significant parameters

- * Increasing θ_2 expands both the R & B regions at the expense of W.
- * Increasing θ_3 expands both the R region with virtually no change to the B region.

Using a generalized architecture

- The modelling framework can be easily generalized to consider scenarios where each of the 6 interactions can correspond to either upregulation (+), downregulation (-) or no effect (\emptyset).
- * We can thus investigate $3^6 = 729$ different classes of intercellular coupling, and in each case we simulate the system dynamics with 10⁴ combinations of coupling strengths $\theta_1 \dots, \theta_6$.

- * For each of the 729 classes, we find the frequency (f_{FF}) that one observes "French flags", having exactly 3 distinct regions, occurring in the expected chromatic order.
- * The nature of regulation of ligand by patterning genes (governed by θ_{4-6}) plays an extremely minor role in regulating cell fate pattern.
- * We find that desired flags are almost never obtained if θ_{1-3} is in the "+" class, i.e. if NICD upregulates one of the genes.

θ_4	θ_{5}	θ_{6}
+	+	+
+	+	0
+	+	_
+	0	+
+	0	0
+	0	-
+	_	+
+	-	0
+	-	-
0	+	+
0	+	0
0	+	-
-	+	+
-	+	0
_	+	-
0	0	+
0	-	+
-	0	+
-	-	+
0	0	0
0	0	-
0	-	0
0	-	-
-	0	0
-	0	_
-	-	0

Conclusion (Part I)

- developing tissues.
- •
- This framework integrates two paradigms for investigating biological pattern formation:

 - A self-organised mechanism involving interactions between constituent species.

Publications:

https://github.com/boyonpointe/Notch-a-French-flag

Juxtacrine signaling can in principle play a key role adaptively regulating cellular differentiation in

This could potentially increase the robustness of the system in generating the desired flag by compensating for mutations affecting the production and / or interpretation of the morphogen.

A **boundary-organized** mechanism involving a prepattern, via a morphogen gradient

C. Kuyyamudi, S. N. Menon, and S. Sinha, *Phys. Rev. E* **103**, 062409 (2021).

C. Kuyyamudi, S. N. Menon, and S. Sinha, Indian J. Phys 96, 2657 (2022).

Development of cell-fate boundaries

As the embryo develops, sharp and robust boundaries are established between cells expressing different patterning genes. Their locations are invariant for a given species.

Development of a boundary between the expression domains of the genes goosecoid and brachyury in the Xenopus embryo.

image source: M. Artinger et al., Mech. Dev. 65:187-196 (1997).

Robustness and precision of cell-fate boundaries

The location of the boundary separating fate domains should ideally be <u>invariant</u> and <u>robust</u>.

If fate determination occurs on the basis of the (noisy) morphogen gradient alone, the boundary location will vary.

If we have two mutually inhibiting patterning genes expressed at low and high concentrations of morphogen, respectively, the fate of a cell at the interface of the two domains will be ambiguous.

Overview of conceptual framework

- We assume that the fates of each cell in a tissue subject to a morphogen gradient are governed by two patterning genes A and B.
- We model the stochastic dynamics of all variables X through a set of coupled equations of the form:

$$d\mathbf{X} = \mathscr{F}_X + \mathscr{G}_X dW$$

where \mathscr{F}_X are the deterministic components and the noise term $\mathscr{G}_X = \eta_X \mathbf{X}$.

* We consider a 1D array comprising N(=50) cells, with 6 species associated with each cell. Each species *X* a mean lifetime τ_X .

morphogen

gradient

* **morphogen** (*M*): Produced at a constant rate from a source located at the leftmost cell (site i = 1) and diffuses with strength D_M .

$$\mathcal{F}_M = \alpha_M \delta_{i,1} - D_M \nabla^2 M - I$$

* **downstream effector** (*S*): Produced as a result of *trans*-activation of receptors (*R*) that bind with ligands of neighbouring cells (L_{tr}).

$$\mathcal{F}_S = k_{tr} R L_{tr} - S/\tau_S$$

 M/τ_M

* **free ligands** (*L*): Produced at a constant rate in the absence of coupling, and production is *suppressed* when the concentration of downstream effector (*S*) increases sufficiently (viz. $S > K_5$).

Depleted when they bind to receptors of neighbouring cells (R_{tr}).

$$\mathscr{F}_{L} = \beta_{L_{0}} + \beta_{L} \frac{(K_{5})^{g}}{(K_{5})^{g} + S^{g}} - k_{tr} R_{tr} L - L/\tau$$

 free receptors (*R*): Produced at a constant rate in the absence of coupling, and production is *enhanced* when *A* & *B* are both expressed at sufficiently high levels (viz. *A*, *B* > *J*).

Depleted when they bind to ligands of neighbouring cells (L_{tr}).

$$\mathscr{F}_R = \beta_{R_0} + \beta_R \frac{A^g}{J^g + A^g} \frac{B^g}{J^g + B^g} - k_{tr} R L_{tr}$$

* **genes** (A & B): We assume that each cell has two mutually repressing genes that are expressed at different levels of morphogen concentration (for $A: M > K_1$, for $B: M > K_2$).

B is assumed to inhibit *A* <u>more strongly</u> than the reverse.

The downstream effector *S* released upon successful *trans*-activation can either upregulate or downregulate either *A* or *B*.

The deterministic components of the dynamics of *A* and *B* :

$$\mathcal{F}_{A} = \alpha_{A} \frac{M^{h}}{M^{h} + (K_{1})^{h}} \frac{(K_{3})^{h}}{B^{h} + (K_{3})^{h}} \Phi_{A} + \gamma_{A} \frac{S^{g}}{S^{g} + Q}$$
$$\mathcal{F}_{B} = \alpha_{B} \frac{M^{h}}{M^{h} + (K_{2})^{h}} \frac{(K_{4})^{h}}{A^{h} + (K_{4})^{h}} \Phi_{B} + \gamma_{B} \frac{S^{g}}{S^{g} + Q}$$

A
$$\Phi_A = Q^g/(Q^g + S^g)$$

T $\Phi_B = 1$
 $\gamma_A = 0$
S $\gamma_B = 0$

$$\begin{array}{c} \mathbf{A} & \Phi_A = 1 \\ \mathbf{\Phi}_B = 1 \\ \mathbf{\Phi}_B = 1 \\ \gamma_A > 0 \\ \mathbf{S} & \gamma_B = 0 \end{array}$$

$$\frac{1}{g} - A/\tau_A$$

 B/τ_B Qg

possible roles of the downstream effector:

B
$$\Phi_A = 1$$

T $\Phi_B = Q^g / (Q^g + S^g)$
 $\gamma_A = 0$
S $\gamma_B = 0$

Dynamics with and without intercellular interactions

expression levels.

eventual fates. domains.

- Large fluctuations in the gene
- Sustained ambiguity in cell fates at the interface of the domains.

- Cells rapidly converge to their
- Clear separation between fate

Increase in precision and robustness of the response

0.1

Steady-state distributions for the expression levels of *A* and *B* across realizations are extremely broad with a high degree of overlap near the boundary.

Sharply defined peaks in the distributions at high and low expression levels, with a clearly identifiable dominant gene at every cell, and a robust well-defined fate boundary.

Role of downstream effector on variability in boundary location

We consider the location of the <u>fate boundary</u> l_B for each of the four scenarios.

The variance of this quantity across realizations, $\sigma^2(l_R)$, is substantially reduced when $S \rightarrow A \text{ or } S \dashv B.$

Contours: Variance in the absence of coupling, $\sigma^2(l_B) \approx 1.38$.

Regions below dashed lines: Here $\langle l_B \rangle$ lies outside the range $10 \le \langle l_B \rangle \le 30$, where $\langle l_B \rangle = 20$ is seen in the uncoupled case.

Focusing on the boundary between cell-fates

We consider the dynamics of a pair of coupled cells, located at adjacent positions along the morphogen gradient.

The main observations of the full system can be reproduced in this simplified setting, allowing us to obtain analytical insights from this system.

Coupling leads to an increased separation between the levels of A & B.

Analytically capturing the noise reduction at the boundary

- We analytically capture the dynamics at the cell-fate boundary using the *linear noise approximation* (LNA).
- We formulate a set of effective single-step reactions describing the interactions between the species **X** of the two cells. We obtain the stoichiometric matrix **T** and propensities $\pi_a(\mathbf{X})$ of each reaction *a*.
- The LNA yields a matrix equation: $\mathbf{J}\Sigma + \Sigma \mathbf{J}^T = -\mathbf{B}\mathbf{B}^T$, • where $B_{ia} = T_{ia}\sqrt{\pi_a(\mathbf{X})}$ and **J** is the Jacobian matrix, calculated at the fixed point of the deterministic system.
- We solve this to obtain the covariance matrix Σ , which contains the variance in expression levels of each of the species along the diagonal. The resulting Fano factor $(\langle X^2 \rangle - \langle X \rangle^2) / \langle X \rangle$ is seen to capture the couplinginduced reduction in cell-fate variability at the boundary.

Intercellular interactions via notch signaling can enhance boundary precision even when the noise strength η is increased over two orders of magnitude.

Conclusion (Part II)

- Our model suggests that Notch signalling can enhance the **precision** and the **robustness** of cell-fate • boundaries by utilizing the *inherent asymmetry* in the response of patterning genes to an external morphogen gradient.
- An experimentally verifiable implication of our investigation is that notch signalling is more effective at reducing noise when it **upregulates** the patterning gene expressed at low morphogen concentration (or downregulates the gene that inhibits it).

C. Kuyyamudi, S. N. Menon, and S. Sinha, *Phys. Rev. E* **107**, 024407 (2023).

https://github.com/boyonpointe/Notch-enhanced-cell-fate-determination

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