

# Comparing Models of Delta-Notch Signalling

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# Introduction

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# The Signalling Pathway

- ▶ Ligands bind to a Notch receptor.
- ▶ This releases NICD in the cell.
- ▶ NICD promotes Notch, inhibits Delta, and promotes Serrate.
- ▶ *Cis-inhibition* occurs between molecules from the same cell.

## Previous Models

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- ▶ Is a deterministic ODE model.
- ▶ Not accurate to the biochemistry.
- ▶ Uses computational simulations.
- ▶ Looks at multiple 1D/2D domains.

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We investigate how the results of these models change under the **Stochastic Differential Equation** (SDE) and **Agent-Based** formalisms.

## Our Assumptions

- ▶ All ligands are Delta molecules (so we ignore Serrate).
- ▶ Delta molecules only bind to Notch receptors from neighbouring cells.
- ▶ Reaction events follow a Poisson point process.
- ▶ Notch and Delta production are hill functions of NICD.



## Our Chemical Equations

Consider a cell  $A$  with neighbour  $B$ . The Notch, Delta, and NICD concentrations of cell  $A$  are given by  $N_A$ ,  $D_A$ , and  $I_A$  respectively (and similarly for cell  $B$ ).

1.  $N_A + D_B \rightarrow I_A$ : Binding event.
2.  $N_A \rightarrow \emptyset$ : Notch decay.
3.  $D_A \rightarrow \emptyset$ : Delta decay.
4.  $I_A \rightarrow \emptyset$ : NICD decay.
5.  $\emptyset \rightarrow N_A$ : Notch production.
6.  $\emptyset \rightarrow D_A$ : Delta production.

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In the **SDE** model, we reintroduce randomness by adding an independent *Wiener Process* to each equation in the ODE model. This recreates some the stochasticity of the agent-based model.

# Derivation

## Preliminaries and Assumptions

Reaction	Rate	Description
$N_A + D_B \rightarrow I_A$	$k_T N_A D_B$	A Notch receptor from cell $A$ binds to a delta ligand from cell $B$ , triggering the release of NICD in cell $A$ .
$N_A \rightarrow \emptyset$	$\gamma N_A$	A Notch receptor from cell $A$ decays.
$D_A \rightarrow \emptyset$	$\gamma D_A$	A Delta receptor from cell $A$ decays.
$I_A \rightarrow \emptyset$	$\gamma_I I_A$	A NICD molecule from cell $A$ decays.
$\emptyset \rightarrow N_A$	$H^+(I_A)$	A Notch receptor in cell $A$ is produced
$\emptyset \rightarrow D_A$	$H^-(I_A)$	A Delta ligand in cell $A$ is produced

Table 1: A list of possible reactions and their rates in a cell  $A$  with neighbour  $B$ .

## Preliminaries and Assumptions

Since there are 6 possible reactions per cell, a system with  $k$  cells will have  $6k$  possible reactions to keep track of.

In the following derivations, we focus on tracking a single cell (cell  $A$ ) for simplicity.

# Derivation of Kolmogorov Forward Equation

Let us define

- ▶  $P(n, d, i, t)$ : the probability of having  $n$  Notch receptors,  $d$  Delta ligands, and  $i$  NICD molecules at time  $t$ .
- ▶  $D_{\text{ext}}$ : the average Delta concentration over all neighbours of cell  $A$ .

The KFE describes how  $P(n, d, i, t)$  changes over a small time step  $\Delta t$ :

$$\begin{aligned} P(n, d, i, t + \Delta t) = & P(n, d, i, t) \\ & + P(n - 1, d, i, t) \cdot H^+(i) \Delta t \quad (\text{Notch production}) \\ & + P(n, d - 1, i, t) \cdot H^-(i) \Delta t \quad (\text{Delta production}) \\ & + P(n + 1, d, i, t) \cdot \gamma(n + 1) \Delta t \quad (\text{Notch decay}) \\ & + P(n, d + 1, i, t) \cdot \gamma(d + 1) \Delta t \quad (\text{Delta decay}) \\ & + P(n, d, i + 1, t) \cdot \gamma_I(i + 1) \Delta t \quad (\text{NICD decay}) \\ & + P(n + 1, d, i - 1, t) \cdot k_T(n + 1) D_{\text{ext}} \Delta t \quad (\text{Trans-activation}) \\ & - P(n, d, i, t) \cdot [H^+(i) + H^-(i) + \gamma n + \gamma d + \gamma_I i + k_T n D_{\text{ext}}] \Delta t \end{aligned}$$



## Derivation of Kolmogorov Forward Equation

Rearranging to find the time derivative and taking the limit as  $\Delta t \rightarrow 0$ :

$$\begin{aligned}\frac{\partial P(n, d, i, t)}{\partial t} = & P(n-1, d, i, t) \cdot H^+(i) \\ & + P(n, d-1, i, t) \cdot H^-(i) \\ & + P(n+1, d, i, t) \cdot \gamma(n+1) \\ & + P(n, d+1, i, t) \cdot \gamma(d+1) \\ & + P(n, d, i+1, t) \cdot \gamma_I(i+1) \\ & + P(n+1, d, i-1, t) \cdot k_T(n+1)D_{\text{ext}} \\ & - P(n, d, i, t) \cdot [H^+(i) + H^-(i) + \gamma n + \gamma d + \gamma_I i + k_T n D_{\text{ext}}]\end{aligned}$$

## Matrix Formulation

The Kolmogorov forward equation can be expressed in matrix form as:

$$\frac{d\vec{P}(t)}{dt} = \mathbf{A}\vec{P}(t)$$

where  $\vec{P}(t)$  is a column vector containing the probabilities for all possible states, and  $\mathbf{A}$  is the transition rate matrix.

Each entry  $A_{i,j}$  represents the transition rate from state  $i$  to state  $j$ . The diagonal elements  $A_{i,i}$  contain the negative sum of all outgoing rates from state  $i$ .

## Matrix Formulation

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We flatten the 3D state space into a 1D list.

- ▶ We encode  $(n,d,i)$  into a single index — just like how multidimensional data (e.g., images, simulation grids) are stored in computer memory.

## Matrix Formulation

To construct the matrix  $\mathbf{A}$ , we first establish a one-to-one mapping between the three-dimensional state space  $(n, d, i)$  and a one-dimensional index  $k$ .

We set  $n_{max}$ ,  $d_{max}$ , and  $i_{max}$  as large enough bounds for Notch, Delta, and NICD in cell  $A$ . These define the size of our 3D “box” of possible states.

*Remark:* These are not strict limits but chosen large enough to technically capture all relevant dynamics.

## Matrix Formulation

For a system with “maximum” values  $n_{max}$ ,  $d_{max}$ , and  $i_{max}$ , we define:

$$k(n, d, i) = n \cdot (d_{max} + 1) \cdot (i_{max} + 1) + d \cdot (i_{max} + 1) + i$$

for  $0 \leq n \leq n_{max}$ ,  $0 \leq d \leq d_{max}$ ,  $0 \leq i \leq i_{max}$ .

The inverse mapping gives:

$$\begin{aligned} n(k) &= \left\lfloor \frac{k}{(d_{max} + 1) \cdot (i_{max} + 1)} \right\rfloor \\ d(k) &= \left\lfloor \frac{k \bmod ((d_{max} + 1) \cdot (i_{max} + 1))}{i_{max} + 1} \right\rfloor \\ i(k) &= k \bmod (i_{max} + 1) \end{aligned}$$

## Matrix Formulation

With this mapping, we can now define the elements of matrix  $\mathbf{A}$ . Let  $k$  and  $k'$  be the indices corresponding to states  $(n, d, i)$  and  $(n', d', i')$  respectively:

$$A_{k,k'} = \begin{cases} H^+(i) & \text{if } (n', d', i') = (n + 1, d, i) \quad (\text{Notch production}) \\ H^-(i) & \text{if } (n', d', i') = (n, d + 1, i) \quad (\text{Delta production}) \\ \gamma(n) & \text{if } (n', d', i') = (n - 1, d, i) \quad (\text{Notch decay}) \\ \gamma(d) & \text{if } (n', d', i') = (n, d - 1, i) \quad (\text{Delta decay}) \\ \gamma_I(i) & \text{if } (n', d', i') = (n, d, i - 1) \quad (\text{NICD decay}) \\ k_T(n)D_{\text{ext}} & \text{if } (n', d', i') = (n - 1, d, i + 1) \quad (\text{Trans-activation}) \\ -v & \text{if } k' = k \quad (\text{Diagonal terms}) \end{cases}$$

# Matrix Formulation

Note that  $A_{k,k'}$  represents the transition rate from state  $k$  to state  $k'$ . The diagonal elements  $A_{k,k}$  are the negative sum of all outgoing rates from state  $k$ :

$$A_{k,k} = -v = -[H^+(i) + H^-(i) + \gamma n + \gamma d + \gamma_I i + k_T n D_{\text{ext}} + k_T d N_{\text{ext}}]$$

where  $(n, d, i)$  is the state corresponding to index  $k$ .

The resulting matrix  $\mathbf{A}$  is a sparse matrix with the following properties:

- ▶ Dimension:  $(n_{\text{max}} + 1) \cdot (d_{\text{max}} + 1) \cdot (i_{\text{max}} + 1) \times (n_{\text{max}} + 1) \cdot (d_{\text{max}} + 1) \cdot (i_{\text{max}} + 1)$
- ▶ Number of non-zero elements:  $\approx 7 \cdot (n_{\text{max}} + 1) \cdot (d_{\text{max}} + 1) \cdot (i_{\text{max}} + 1)$



## Matrix Formulation

The corresponding transition matrix **A** would be:

$$\mathbf{A} = \begin{pmatrix} -v_{000} & \alpha_{000,001} & \alpha_{000,010} & \alpha_{000,011} & \alpha_{000,100} & \alpha_{000,101} & \alpha_{000,110} & \alpha_{000,111} \\ \alpha_{001,000} & -v_{001} & \alpha_{001,010} & \alpha_{001,011} & \alpha_{001,100} & \alpha_{001,101} & \alpha_{001,110} & \alpha_{001,111} \\ \alpha_{010,000} & \alpha_{010,001} & -v_{010} & \alpha_{010,011} & \alpha_{010,100} & \alpha_{010,101} & \alpha_{010,110} & \alpha_{010,111} \\ \alpha_{011,000} & \alpha_{011,001} & \alpha_{011,010} & -v_{011} & \alpha_{011,100} & \alpha_{011,101} & \alpha_{011,110} & \alpha_{011,111} \\ \alpha_{100,000} & \alpha_{100,001} & \alpha_{100,010} & \alpha_{100,011} & -v_{100} & \alpha_{100,101} & \alpha_{100,110} & \alpha_{100,111} \\ \alpha_{101,000} & \alpha_{101,001} & \alpha_{101,010} & \alpha_{101,011} & \alpha_{101,100} & -v_{101} & \alpha_{101,110} & \alpha_{101,111} \\ \alpha_{110,000} & 0 & H^+(0) & 0 & H^-(1) & 0 & -v_{110} & \gamma_I \\ \alpha_{111,000} & 0 & 0 & H^+(0) & 0 & H^-(1) & H^+(1) & -v_{111} \end{pmatrix}$$

where:

$$\alpha_{ndi,n'd'i'} = H^+(i) + H^-(i) + \gamma n + \gamma d + \gamma_I i + k_T n D_{\text{ext}} + k_T d N_{\text{ext}}$$

## Remarks

Let us consider

- ▶ A one-cell system:  $P(n, d, i, t + \Delta t)$   
A  $k$ -cell system:  $P(n_1, d_1, i_1, n_2, d_2, i_2, \dots, n_k, d_k, i_k, t + \Delta t)$
- ▶  $D_{\text{ext}}$  (the average Delta concentration over all neighbours of cell  $A$ ).

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A  $k$ -cell system:  $P(n_1, d_1, i_1, n_2, d_2, i_2, \dots, n_k, d_k, i_k, t + \Delta t)$
- ▶  $D_{\text{ext}}$  (the average Delta concentration over all neighbours of cell  $A$ ).

The complexity of a single cell highlights the need for numerical methods to model interactions between cells.

# Results

# Comparing Two-Cell Models

## Noise in the SDE Model

## Stability Analysis

We tested model stability by perturbing 5 parameters:  $N_m$ ,  $D_m$ ,  $K_T$ ,  $\gamma$ , and  $\gamma_I$ .

- ▶  $N_m$ ,  $D_m$  are the maximum rates of Notch/Delta production.
- ▶  $K_T$  is the binding rate.
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Due to computational limitations, we investigated the 2-dimensional subspaces of our 5-dimensional parameter space using the following algorithm:

- ▶ Run the two-cell model on a  $25 \times 25$  grid of points spanning 2 OOMs.
- ▶ Take the convex hull of the points at which the cells differentiated.



# Stability Analysis

# Simulations on Linear Domains

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# Patterns on Linear Domains

# Simulations on Hexagonal Domains

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