

Three-dimensional mapping of intact ovaries reveals the aging dynamics of the ovarian reserve

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Supplementary Note: Modelling Oocyte Dynamics

S1 Problem Overview

We observe counts of oocytes in four developmental stages across multiple ovaries and ages (weeks):

Primordial (P), Primary (P2), Secondary (S), Antral (A).

For each ovary, we also have its organ volume. The goal is to model the dynamics of stage counts—including death and forward transitions—and to infer the underlying transition/death parameters together with standard errors.

S2 Data and Normalization

Let $y_{ik}(t)$ denote the observed count in stage $k \in \{0, 1, 2, 3\}$ for ovary i at week t , and let V_i be the ovary volume. To account for varying organ sizes, we analyze *densities* (counts per unit volume). This is implemented as:

$$\tilde{y}_{ik}(t) = \frac{y_{ik}(t)}{V_i/2}, \quad (1)$$

where the $1/2$ factor accounts for two ovaries per animal. After normalization, we stack all rows (ovaries $\times$ weeks) into a matrix $\mathbf{Y} \in \mathbb{R}^{n \times 4}$.

S3 Mechanistic Model: Compartments, Flows, and Death

We adopt a linear compartmental model with four states $\mathbf{N}(t) = (N_0(t), N_1(t), N_2(t), N_3(t))^\top$ representing the latent stage densities. The dynamics are governed by:

- **Forward transitions** from stage j to $j + 1$ with rate p_j .
- **Death** within each stage j with rate d_j .

The dynamics follow the linear system of ordinary differential equations (ODEs):

$$\frac{d}{dt}N_0 = -(d_1 + p_1)N_0, \quad (2)$$

$$\frac{d}{dt}N_1 = p_1N_0 - (d_2 + p_2)N_1, \quad (3)$$

$$\frac{d}{dt}N_2 = p_2N_1 - (d_3 + p_3)N_2, \quad (4)$$

$$\frac{d}{dt}N_3 = p_3N_2 - d_4N_3. \quad (5)$$

In matrix form, $\dot{\mathbf{N}}(t) = \mathbf{A} \mathbf{N}(t)$ with:

$$\mathbf{A} = \begin{bmatrix} -(d_1 + p_1) & 0 & 0 & 0 \\ p_1 & -(d_2 + p_2) & 0 & 0 \\ 0 & p_2 & -(d_3 + p_3) & 0 \\ 0 & 0 & p_3 & -d_4 \end{bmatrix}.$$

Throughout the main analysis, we fix $d_2 = d_3 = 0$ based on biological considerations and identifiability. The free parameters are $\boldsymbol{\theta} = (p_1, p_2, p_3, d_1, d_4)^\top$.

S4 Initial Conditions and Numerical Solution

For each ovary observed at week 5, we have a complete 4-vector of stage densities. We define the fitting initial condition as the mean across week-5 replicates:

$$\mathbf{N}_{\text{mean}}(5) = \frac{1}{|\mathcal{I}_5|} \sum_{r \in \mathcal{I}_5} \mathbf{Y}(r, \cdot).$$

Because the system is linear with constant coefficients, we compute $\mathbf{N}(t)$ using a robust ODE integrator (`solve_ivp`, RK45).

S5 Parameter Estimation via Nonlinear Least Squares

We treat the observations as noisy measurements: $\tilde{y}_{ik}(t_r) = N_k(t_r; \boldsymbol{\theta}) + \varepsilon_{ik}(t_r)$, where $\varepsilon_{ik}(t_r) \sim \mathcal{N}(0, \sigma_k^2)$. To improve numerical conditioning, we scale residuals by an empirical per-stage factor s_k :

$$r_{ik}(t_r; \boldsymbol{\theta}) = \frac{\tilde{y}_{ik}(t_r) - N_k(t_r; \boldsymbol{\theta})}{s_k}.$$

We estimate $\boldsymbol{\theta}$ by minimizing the sum of squared scaled residuals subject to bounds:

$$\hat{\boldsymbol{\theta}} = \arg \min_{\boldsymbol{\theta} \in [\text{LB}, \text{UB}]} \|\mathbf{r}(\boldsymbol{\theta})\|_2^2.$$

Standard errors are derived from the Jacobian $\mathbf{J}(\hat{\boldsymbol{\theta}})$ using the Gauss–Newton approximation: $\widehat{\text{Cov}}(\hat{\boldsymbol{\theta}}) \approx \hat{s}^2 (\mathbf{J}^\top \mathbf{J})^\dagger$, where $\hat{s}^2 = \text{RSS}/(m - p)$.

S6 Identifiability and Constraints

Even with all four series observed, some parameters are weakly identified. Trade-offs between death rates and preceding transition rates can yield similar flows. Fixing $d_2 = d_3 = 0$ alleviates this instability, particularly when secondary and antral counts are small and noisy.

S7 Results

The estimated parameters and associated uncertainties are provided in Figure 4.