

# Spatial Transcriptomics Simulation Model

We developed a computational framework for generating large-scale, biologically realistic spatial transcriptomics datasets. The model synthesizes gene expression counts for  $G$  genes across a  $W \times H$  two-dimensional grid of spatial locations (spots), providing explicit control over spatial expression patterns, sparsity, and count variability.

## Spatial Covariance Modeling

For genes with spatially localized expression, we modeled the spatial correlation using the Matérn covariance kernel. To ensure clear parameterization, we define the Matérn function [PMID: 40399936; PMID: 39865128] as a correlation function  $\kappa_\nu(r)$ , which is a normalized form of the covariance kernel with unit variance ( $\sigma^2 = 1$ ). For two spots separated by Euclidean distance  $r = \|\mathbf{s}_i - \mathbf{s}_j\|$ , the correlation is:

$$\kappa_\nu(r) = \frac{2^{1-\nu}}{\Gamma(\nu)} \left( \frac{\sqrt{2\nu}r}{\ell} \right)^\nu K_\nu \left( \frac{\sqrt{2\nu}r}{\ell} \right)$$

where  $\nu > 0$  is the smoothness parameter,  $\ell > 0$  is the characteristic length scale,  $K_\nu(\cdot)$  is the modified Bessel function of the second kind, and  $\Gamma(\cdot)$  is the gamma function. The length scale  $\ell$  controls the range of spatial correlation, while the smoothness parameter  $\nu$  controls the differentiability of the spatial field.

To ensure numerical stability, we evaluated the kernel using its integral representation, which is well-suited to Gauss-Laguerre quadrature. Using  $n = 64$  quadrature nodes  $\{x_j, w_j\}$ , we approximated the kernel as:

$$\kappa_\nu(r) \approx \frac{1}{2^{\nu-1}\Gamma(\nu)} \sum_{j=1}^{64} w_j x_j^{\nu-1} \exp\left(-\frac{\nu r^2}{4\ell^2 x_j}\right)$$

For each  $(\ell, \nu)$  pair, kernel values were precomputed radially and truncated when falling below a user-specified cutoff  $c_{\text{cut}}$ , defining an adaptive spatial support radius.

## Generation of Spatially Variable Gene Expression

The latent mean expression  $\mu_g(\mathbf{s})$  for gene  $g$  at spot  $\mathbf{s}$  is modeled as a sum of a baseline expression level and contributions from a set of expression hotspots  $H_g$ :

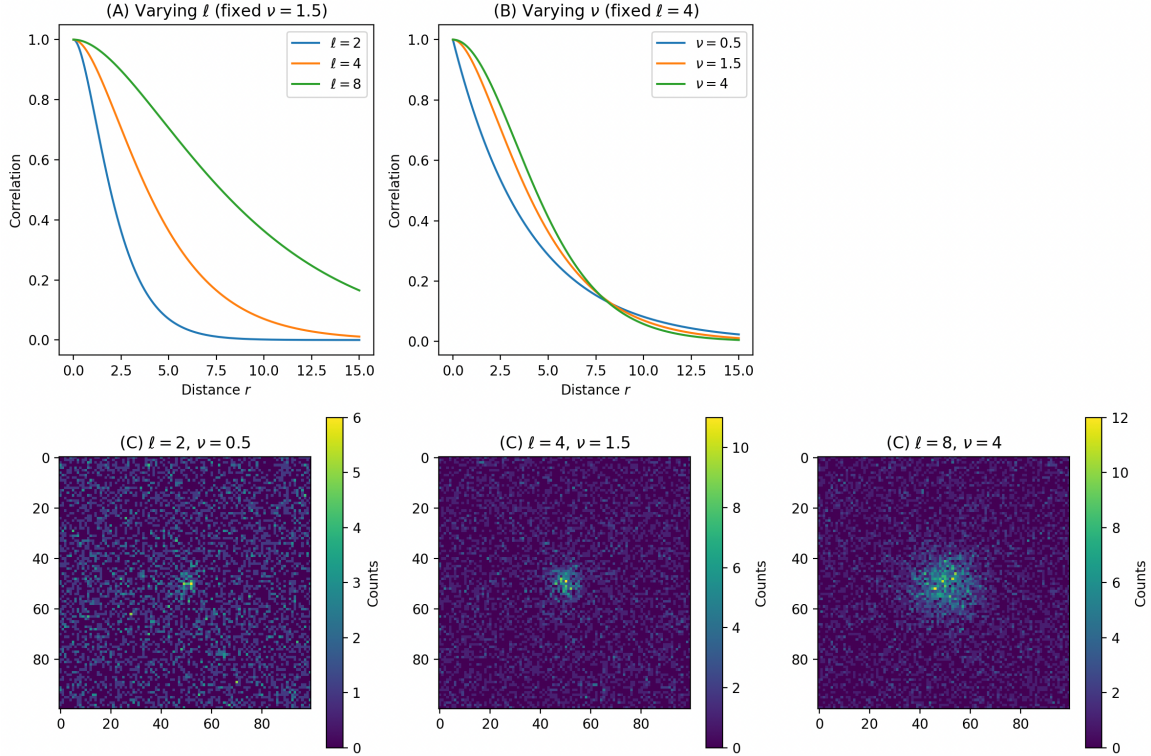
$$\mu_g(\mathbf{s}) = \mu_{\text{baseline},g} + \sum_{m \in H_g} A_m \kappa_\nu(\|\mathbf{s} - \mathbf{h}_m\|)$$

where  $\mu_{\text{baseline},g}$  is the baseline expression,  $\{\mathbf{h}_m\}$  are the hotspot coordinates, and  $\{A_m\}$  are the corresponding amplitudes that determine the peak expression intensity.

To model technical artifacts, we introduced zero inflation and stochasticity. First, the latent mean  $\mu_g(\mathbf{s})$  was set to zero with a probability  $p_{\text{drop}}$  to mimic technical dropout. The final observed integer count  $X_g(\mathbf{s})$  was then generated by sampling from a Poisson distribution:

$$X_g(\mathbf{s}) \sim \text{Poisson}(\mu_g(\mathbf{s}))$$

The resulting mean  $\mu_g(\mathbf{s})$  can be zero either due to dropout or if the spot is sufficiently distant from all hotspots.



**Figure 1 | Simulation of spatial transcriptomics data with tunable Matérn parameters.** (A) Effect of varying the correlation length scale  $\ell$  (with  $\nu = 1.5$  fixed). Larger  $\ell$  values yield broader spatial correlation. (B) Effect of varying the smoothness parameter  $\nu$  (with  $\ell = 4$  fixed). Larger  $\nu$  values produce smoother spatial fields. (C) Simulated  $100 \times 100$  spatial expression maps for different  $(\ell, \nu)$  combinations, showing hotspot locations (bright central pixels) and the decay of expression intensity. Hotspots are generated by convolving their locations with a precomputed Matérn kernel  $\kappa_\nu(r) = \frac{2^{1-\nu}}{\Gamma(\nu)} \left(\frac{\sqrt{2\nu}r}{\ell}\right)^\nu K_\nu\left(\frac{\sqrt{2\nu}r}{\ell}\right)$ , followed by zero inflation and Poisson sampling.

## Simulation Parameterization

**Spatial Expression Parameters:** The smoothness  $\nu$ , characteristic length scale  $\ell$ , and kernel cutoff  $c_{\text{cut}}$  were set to 1.5, 2.5, and  $10^{-3}$ , respectively. The baseline expression

$\mu_{\text{baseline},g}$  and amplitude was 1.0 and 4.0. The dropout rate  $p_{\text{drop}}$  can be set by users. For power evaluation, one can set one or more hotspot was placed at grid location (30,30). For Type I error evaluation, no hotspots were included.