

Are fish Poisson? About the neurogenesis of the zebrafish using spatial point process analysis

Felix Cheysson

Joint work with Nicolas Dray, Laure Mancini (Institut Pasteur),
Udi Binshtok, David Sprinzak (Tel-Aviv University),
and more

Université Gustave Eiffel, CNRS, UMR 8050, **LAMA**.

Statistiques au sommet de Rochebrune

March 26th 2024

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When a zebra loves a fish very much, then...

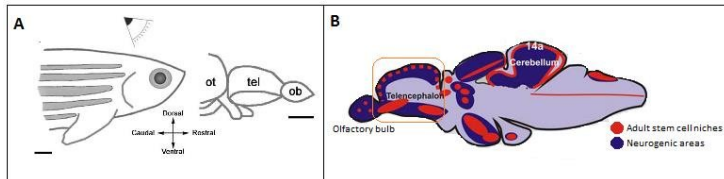


The zebrafish: a classic vertebrate model in biology

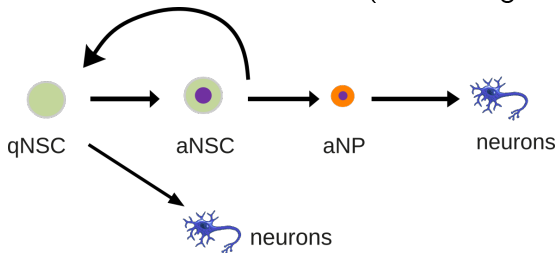


The zebrafish neurogenesis

- Development of Neural Stem Cells (NSCs) in the zebrafish dorsal pallium of the telencephalon.



- Homeostasis of Neural Stem Cells (NSCs) is maintained through renewal and differentiation mechanisms (Than-trong et al., 2020).



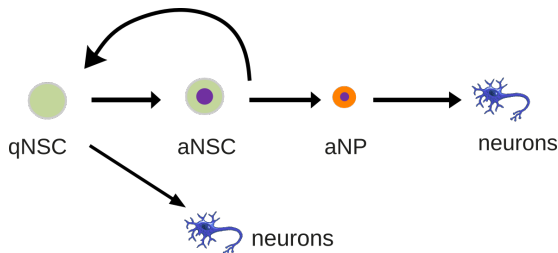
Live intravital imaging

- The Casper mutant ($roy^{-/-}; nacre^{-/-}$) (White et al., 2008).



- Multicolor fluorescence and harmonics multiphoton microscopy (Dray et al., 2015).
 - Fluorescence: to detect NSC markers, and markers for activation events.
 - Harmonics: provide persistent landmarks for longitudinal imaging.
 - Video.

The BBQ: Big Biological Question



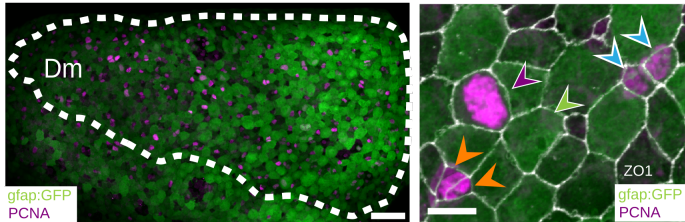
What spatial and temporal regulation mechanisms explain the homeostasis of the cell population and spatial organisation of the pallium?

Our approach:

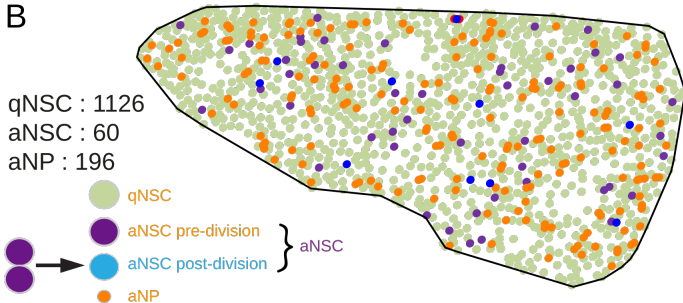
- Spatial statistics to study the dependence between cell divisions.
- Biological experiments to (in-)validate the hypothesised signalling pathways.

Cells as a marked point process

A



B



- Consider a point process \mathbf{X} as a locally finite random set on \mathbb{R}^2 with density ρ .
- Second-order density $\rho^{(2)}$: for any f measurable,

$$\mathbb{E} \left[\sum_{x \in \mathbf{X}} \sum_{y \in \mathbf{X}, y \neq x} f(x, y) \right] = \int \int f(x, y) \rho^{(2)}(dx, dy).$$

- Define $g(x, y) = \rho^{(2)}(x, y) / \rho(x)\rho(y)$.
- Assuming stationarity of \mathbf{X} , $g(x, y) = g(x - y)$, its reduced second moment measure is

$$\mathcal{K}(A) = \int_A g(x) dx, \quad \text{for } A \text{ Borel set.}$$

- Assuming isotropy of \mathbf{X} , Ripley's K-function:

$$K(r) = \mathcal{K}(b(x, r)) \quad \left(= \frac{1}{\beta} \mathbb{E} \left[|\mathbf{X} \cap b(x, r) \setminus \{x\}| \right] \right).$$

- Example: for the Poisson process, then $g(x) = 1$ and $K(r) = \pi r^2$.
- Estimator for the K-function from a window W :

$$\hat{K}(r) = \hat{\beta}^{-1} \sum_{\substack{x \in \mathbf{X} \cap W \\ y \in \mathbf{X} \cap W, x \neq y}} w(x, y)^{-1} \frac{\mathbb{1}\{|x - y| \leq r\}}{|\mathbf{X} \cap W|},$$

where $w(x_i, x_j)$ provides an *edge correction*.

- L-function $\hat{L}(r) = (\hat{K}(r)/\pi)^{1/2}$ stabilises the variance.
- Generally, explicit formulas for the mean and variance of $\hat{K}(r)$ unavailable (unless Poisson, see Lang and Marcon, 2013).

In the marked temporal case

- Assume the processes $\mathbf{X}_j^t = (\mathbf{X}_j^1, \dots, \mathbf{X}_j^T)$ are stationary (in time and space), and isotropic.
- Study interactions between types of cells through Ripley's function:

$$\widehat{K}_{ij}^t(r) = (\widehat{\beta}_i \widehat{\beta}_j |W|)^{-1} \sum_s \sum_{\substack{x \in \mathbf{X}_i^s \cap W \\ y \in \mathbf{X}_j^{s+t} \cap W, x \neq y}} w(x, y)^{-1} \mathbb{1}\{|x - y| \leq r\}.$$

- Under *random labelling*, each process \mathbf{X}_i^t can be seen as a random thinning of the marked process \mathbf{X}^t , so that $K_{ij}^t(r) = K^t(r) = K(r)$ for any i, j .
- Inference usually achieved through Monte Carlo tests, or normal approximations when available.

A test of independence between point processes

- We want to test the dependence between NSCs in their different states (quiescent, qNSCs; activated, aNSCs; progenitors, aNP).
- Due to cellular constraints, Poisson hypothesis is wrong: need to test dependence between processes under another null hypothesis.
- Simulation envelopes (= fluctuation envelopes) can be computed under random labelling:

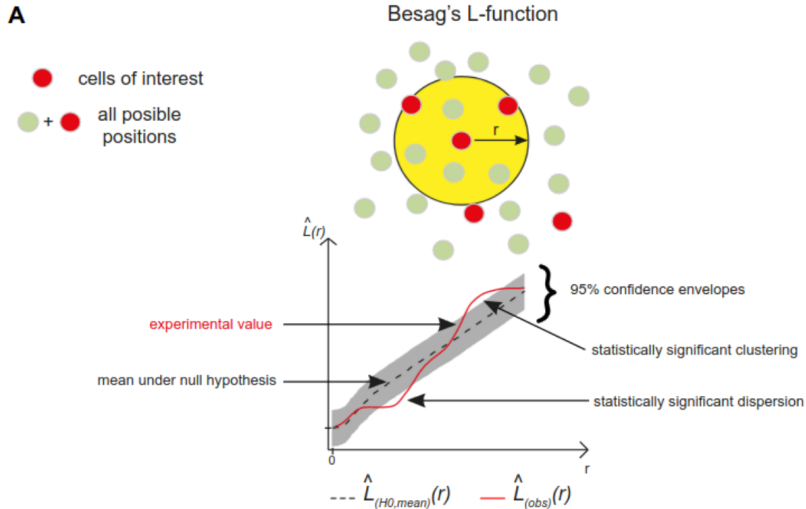
$$\mathcal{H}_0 : \forall i, \mathbf{X}_i^t \text{ is an independent random thinning of } \mathbf{X}^t.$$

- Idea: permutation test.
 - Simulate samples $(\tilde{\mathbf{X}}_{(1)}^t, \dots, \tilde{\mathbf{X}}_{(m)}^t)$ under random labelling.
 - Under \mathcal{H}_0 , \mathbf{X}^t has the same distribution as any $\tilde{\mathbf{X}}_{(k)}^t$.
 - Conclude via any test statistic $f(\mathbf{X}^t)$ (e.g. Ripley or L-function):

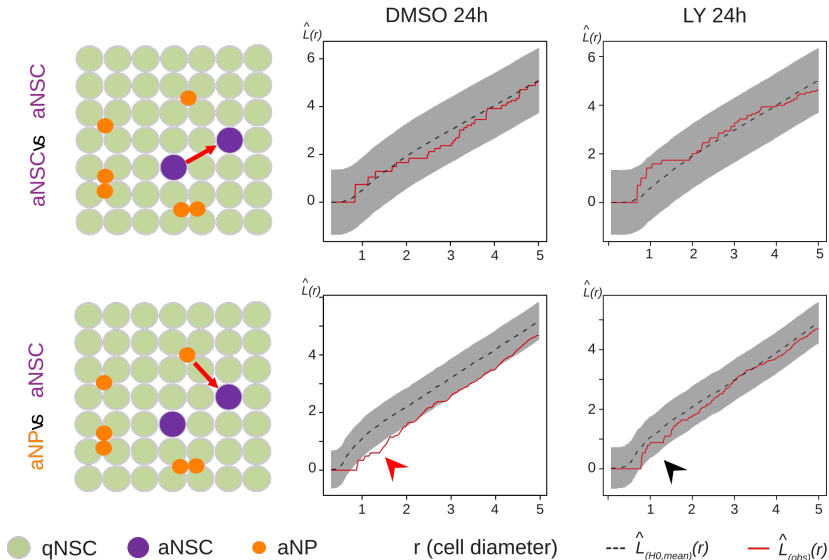
$$\mathbb{P}_{\mathcal{H}_0}(f(\mathbf{X}^t) > f_{(k)}) = 1 - \frac{k}{m+1},$$

where $f_{(k)}$ denotes the k -th largest of the simulated values $f(\mathbf{X}_{(k)}^t)$.

Pictures are worth a thousand words



Are fish Poisson?

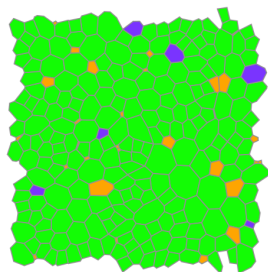


An experimental testing of inhibitive local feedback

- aNP-mediated feedback inhibition on aNSCs?
- Notch3 signaling (= a pathway using the receptors Notch3 on qNSC) promotes NSC quiescence, and aNPs express the Notch ligand DeltaA.
- Experiment in zebrafish:
 - Decrease Notch signaling through a short treatment.
 - Abolition of the local inhibition verified by testing (cf. LY).
 - qNSCs for fish under the treatment rapidly undergo rapid division.

A lattice model for neurogenesis

- Construction of a lattice model (system of EDPs + rules for lattice construction) to simulate and explore different mechanisms of interaction in NSCs (Video).
- Driven by data, and verified showing that the empirical statistical behaviour of the model is similar to that of the data.
- Neurons cannot be seen on live imaging of the pallium: explore spatial distribution of neurons on the simulations.



qNSC



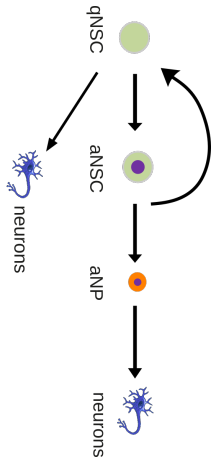
aNSc



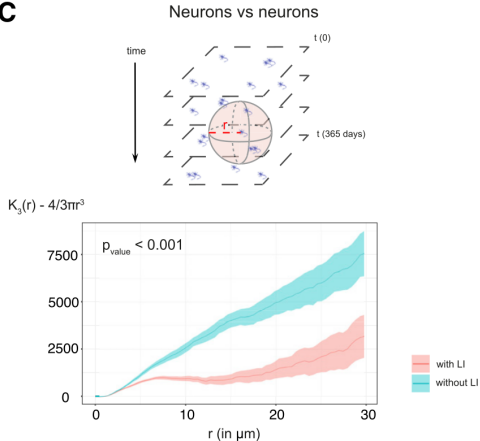
aNP / divided aNP

Lateral inhibition homogenises neurons in the pallium

- Comparison through two-sample permutation tests between simulations with and without lateral inhibition.
- Lateral inhibition supports more homogeneous neurogenesis output.







C



Thank you for your attention.



For Further Reading I

-  Dray, Nicolas et al. (2015). “Large-scale live imaging of adult neural stem cells in their endogenous niche”. In: *Development* 142.20, pp. 3592–3600. ISSN: 1477-9129. DOI: [10.1242/dev.123018](https://doi.org/10.1242/dev.123018).
-  Lang, Gabriel and Eric Marcon (2013). “Testing randomness of spatial point patterns with the riplely statistic”. In: *ESAIM - Probability and Statistics* 17, pp. 767–788. ISSN: 12623318. DOI: [10.1051/ps/2012027](https://doi.org/10.1051/ps/2012027). arXiv: [1006.1567](https://arxiv.org/abs/1006.1567).
-  Than-trong, Emmanuel et al. (2020). “Lineage hierarchies and stochasticity ensure the long-term maintenance of adult neural stem cells”. In: *Science Advances* 6.18, eaaz5424. DOI: [10.1126/sciadv.aaz5424](https://doi.org/10.1126/sciadv.aaz5424).
-  White, Richard Mark et al. (2008). “Transparent Adult Zebrafish as a Tool for In Vivo Transplantation Analysis”. In: *Cell Stem Cell* 2.2, pp. 183–189. ISSN: 19345909. DOI: [10.1016/j.stem.2007.11.002](https://doi.org/10.1016/j.stem.2007.11.002).