CEMRACS 2018 project: Mathematical modelling of cell aggregation and segregation.

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Biological context

Cells of the same type can regroup into regions \Rightarrow spatial organisation. Cell segregation and border sharpening in two-species systems:

INTERFACE

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Research



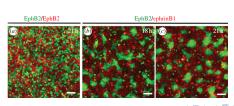
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Working hypothesis: inter(heterotypic) and intra(homotypic) species repulsion control cell segregation and border sharpening. They have more influence than inter- or intraspecies adhesion.

Goal: to understand the mechanisms of morphogenesis.





How to model?

Several mathematical models and differents approaches have been proposed for cell segregation.

Macroscopic model

- ullet Continuous approach o analysis tools
- Theoretical framework to link the solutions to the model parameters

BUT

- Loss of info about cell-interactions
- No info about number of clusters size and popolation size

Microscopic model

- Agent-based models: simplicity and flexibility
- Precision of the modeling
- Link with experimental data

BUT

- Computational expensive
- Theoretically harder



Microscopic framework

Individual Based Model for particles interacting through repulsion interactions:

$$\begin{cases} dX_i^A = -\mu \nabla_{X_i^A} W^A(X^A, X^B) dt + \sqrt{2D_A} dB_i, & \forall i \in \{1, \dots, N_A\} \\ dX_i^B = -\mu \nabla_{X_\ell^A} W^B(X^A, X^B) dt + \sqrt{2D_B} dB_\ell, & \forall \ell \in \{1, \dots, N_B\} \end{cases}$$
(1)

- $\mu > 0$ is the constant mobility coefficient,
- B_i is a 2-dimensional Brownian motion $B_i = (B_i^1, B_i^2)$ of intensity $D_A, D_B > 0$ respectively for species A and B,
- W^S total energy of the S-type particle, $S \in \{A, B\}$, defined as:

$$W^{A}(X^{A}, X^{B}) = \underbrace{\sum_{k_{1}=1}^{K_{AA}} \Phi^{AA}(X_{i(k_{1})}^{A} - X_{j(k_{1})}^{A})}_{k_{1}=1} + \underbrace{\sum_{k_{3}=1}^{K_{AB}} \Phi^{AB}(X_{i(k_{3})}^{A} - X_{\ell(k_{3})}^{B})}_{k_{2}=1},$$

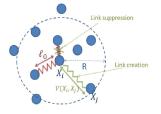
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sum over all pairwise link potentials acting on particles A

$$W^{B}(X^{A}, X^{B}) = \sum_{k_{2}=1}^{K_{BB}} \Phi^{BB}(X^{B}_{\ell(k_{2})} - X^{B}_{m(k_{2})}) + \sum_{k_{3}=1}^{K_{AB}} \Phi^{BA}(X^{B}_{\ell(k_{3})} - X^{A}_{\ell(k_{3})})$$
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Case: Hookean interaction potential

We suppose that the homotypic (AA,BB) species links and heterotypic (AB,BA) act as a springs of equilibrium length R between the particles that it is also detection radius for the interaction.



Case of Hookean springs

$$\Phi^{ST}(x) = \frac{\nu_c^{ST}}{\nu_d^{ST}} \frac{\kappa^{ST}}{2} \begin{cases} (|x| - R)^2, & \text{for } |x| \le R \\ 0, & \text{for } |x| > R \end{cases}$$

with ν_c^{ST}, ν_d^{ST} Poisson processes frequencies and κ^{ST} interaction/repulsion intensity.

- Each particle can link/unlink with its neighbors located in a ball of radius R
- Links are not permanent: created and supressed via random processes
- Linking/unlinking processes are very fast



Logistic growth term

We add a growth process to the microscopic model as follows:

- Cell of type S divide into 2 cells with probability β_S and die with probability δ_S at each time step. $S \in \{A, B\}$
- Birth and death processes depend on the local density of individuals
- Birth occurs at distance r

$$\beta_{\mathcal{S}}(X_i) = b_0^{\mathcal{S}} - (b_0^{\mathcal{S}} - \theta) \left(\frac{\mathcal{N}_0}{N^*}\right), \quad \delta_{\mathcal{S}}(X_i) = d_0^{\mathcal{S}} + (\theta - d_0^{\mathcal{S}}) \left(\frac{\mathcal{N}_0}{N^*}\right) \tag{2}$$

Parameters:

- $\mathcal{N}_0 = \mathcal{N}_{R_0}(X_i^S)$: number of cells (of both population) at distance R_0 of the cell located in X_i^S
- N^* is the maximal number of cell in a radius R_0 allowing cell division.
- $m{ heta}$, constant coefficient that assures the randomness at the population N^* .



Micro macro



Form micro to macro

Fock space

- Main difficult the varying size of the cell population
- Introduction of the Fock Space:
- Probability space of all the possible states of the particle system

$$X_k := [x_1, x_2, \dots, x_k]$$

 $\mathbb{P}_k(X_k,t)dX_k = \Pr\{k \text{ cells, with one cell in } dx_1, \text{ another in } dx_2 \text{ etc. } \}$



Form micro to macro

Reduced distribution function

• The density or concentration of cells is define by:

$$f(x,t) = \langle \sum_{p=1}^{k} \delta(x - x_p) \rangle.$$

which becomes

$$f(x,t) = \sum_{k=1}^{\infty} k \int \mathbb{P}_k(x, X_{k-1}, t) dX_{k-1},$$

• We define a master equation for the Probability $\mathbb{P}_k(X_k,t)$ time evolution:



Master Equation

$$\begin{split} \mathbb{P}_{k}(X_{k}, t+\tau) &= \int \mathbb{W}_{k}(X_{k}, t+\tau | X_{k}^{'}, t) \mathbb{P}_{k}(X_{k}^{'}, t) dX_{k}^{'} \\ &+ \tau \sum_{i=1}^{k-1} \beta(X_{i}) \mathbb{BP}_{k-1} - \tau \left[\sum_{i=1}^{k} (\beta(X_{i}) + \delta(X_{i})) \right] \mathbb{P}_{k}(X_{k}, t) \\ &+ \tau \int \sum_{k=1}^{k+1} \beta(X_{i}) \mathbb{P}_{k} + 1(X_{k+1}, t) dX_{i} \end{split}$$

with $\mathbb{W}_k(X_k, t + \tau | X_k^{'}, t)$ the transition probability from a state $X_k^{'}$ to a state X_k and

$$\mathbb{BP}_{k-1} = \frac{2}{k(k-1)} \sum_{1 \le p < q \le k} \delta_{pq} \mathbb{P}_{k-1}(X_{k|p}, t)$$
(3)





Master Equation

Using Kramers-Moyal expansion on the term $\int \mathbb{W}_k(X_k, t + \tau | X_{\iota}', t) \mathbb{P}_k(X_{\iota}', t) dX_{\iota}'$:

$$\begin{split} \mathbb{P}_k(X_k, t + \tau) &= \left[1 + \sum_{|\alpha| > 0} (-1)^{|\alpha|} \partial^{\alpha} \left(D^{|\alpha|}(X_k, t) \tau + O(\tau^2) \right) \right] \mathbb{P}_k(X_k, t) \\ &+ \tau \sum_{i=1}^{k-1} \beta(X_i) \mathbb{BP}_{k-1} - \tau \left[\sum_{i=1}^k (\beta(X_i) + \delta(X_i)) \right] \mathbb{P}_k(X_k, t) \\ &+ \tau \int \sum_{k=1}^{k+1} \beta(X_i) \mathbb{P}_{k+1}(X_{k+1}, t) dx_i. \end{split}$$



Master Equation

the coefficients $D^{|\alpha|}$ are the so-called **Kramers-Moyal** expansion coefficients with:

$$D^{|\alpha|}(X_k,t) = \frac{1}{\alpha!} \lim_{\tau \to 0} \frac{1}{\tau} \langle (\xi(t+\tau) - X_k)^{\alpha} \rangle |_{\xi(t) = X_k}.$$

Using the definition:

$$f(x,t) = \sum_{k=1}^{\infty} k \int \mathbb{P}_k(x, X_{k-1}, t) dX_{k-1},$$

we can deduce the reduced equation on f(x,t) by summing and integrating the master equation.



Macroscopic framework

Macroscopic model should provide an approximation of the agent-based model:

$$\begin{cases} \partial_{t}f^{A} = \nabla \cdot \underbrace{\left(f^{A}\nabla_{x}(\boldsymbol{\Phi}^{AA}*f^{A}) + f^{A}\nabla_{x}(\boldsymbol{\Phi}^{AB}*f^{B})\right)}_{interaction\ potential} + \underbrace{D_{A}\Delta_{x}f^{A}}_{diffusion} + \underbrace{\nu_{b}^{A}f^{A}\left(1 - \frac{f^{A} + f^{B}}{f^{*}}\right)}_{logistic\ term} \\ \partial_{t}f^{B} = \nabla \cdot \left(f^{B}\nabla_{x}(\boldsymbol{\Phi}^{BB}*f^{B}) + f^{B}\nabla_{x}(\boldsymbol{\Phi}^{BA}*f^{A})\right) + D_{B}\Delta_{x}f^{B} + \nu_{b}^{B}f^{B}\left(1 - \frac{f^{A} + f^{B}}{f^{*}}\right) \end{cases}$$

- f*: carrying capacity of the environment
- ν_b^A, ν_b^B growth rates

Remark: f^A , f^B play the same role in logistic term



Analysis of the macroscopic model

We recall macroscopic equations for f^A and f^B :

$$\begin{cases} \partial_t f^A = \nabla \cdot (f^A \nabla_x (\Phi^{AA} * f^A) + f^A \nabla_x (\Phi^{AB} * f^B)) + D_A \Delta_x f^A + \nu_b^A f^A \left(1 - \frac{f^A + f^B}{f^*}\right) \\ \partial_t f^B = \nabla \cdot (f^B \nabla_x (\Phi^{BB} * f^B) + f^B \nabla_x (\Phi^{BA} * f^A)) + D_A \Delta_x f^A + \nu_b^B f^B \left(1 - \frac{f^A + f^B}{f^*}\right) \end{cases}$$
(4)

Linearization around constant steady states \bar{f}^A, \bar{f}^B and Fourier transform:

$$\partial_{t}\begin{pmatrix} \hat{f}^{A} \\ \hat{f}^{B} \end{pmatrix} = \underbrace{\begin{pmatrix} -|y|^{2}(2\pi\bar{f}^{A}\hat{\Phi}^{AA}(y) + D_{A}) - \nu_{b}^{A}\frac{\bar{f}^{A}}{f^{*}} & -|y|^{2}2\pi\bar{f}^{A}\hat{\Phi}^{AB}(y) - \nu_{b}^{A}\frac{\bar{f}^{A}}{f^{*}} \\ -|y|^{2}\bar{f}^{B}\hat{\Phi}^{BA}(y) - \nu_{b}^{B}\frac{\bar{f}^{B}}{f^{*}} & -|y|^{2}(2\pi\bar{f}^{B}\hat{\Phi}^{BB}(y) + D_{B}) - \nu_{b}^{B}\frac{\bar{f}^{B}}{f^{*}} \end{pmatrix}}_{M(y)} \begin{pmatrix} \hat{f}^{A} \\ \hat{f}^{B} \end{pmatrix}$$

The constant steady states will be unstable if:

•
$$\nu^{B} \frac{\bar{f}^{A}}{\bar{f}^{*}} (\bar{f}^{A} 2\pi \hat{\Phi}^{AA} + D_{A} - \bar{f}^{A} 2\pi \hat{\Phi}^{AB}) < \nu^{A} \frac{\bar{f}^{A}}{\bar{f}^{*}} (\bar{f}^{B} 2\pi \hat{\Phi}^{BB} + D_{B} - \bar{f}^{B} 2\pi \hat{\Phi}^{BA}).$$

We want to focus on the ratio of homo- and hetero-typic species repulsion. We introduce a parameter $s \in \mathbb{R}$ s.t.: $\kappa^{ST} = s \tilde{\kappa}^{ST}$



Analysis of the macroscopic model

We find critical value s_L^* related to instability:

$$s_{L}^{*} = \frac{(24D_{A} + c'^{AA}\bar{f}^{A})\nu_{b}^{B}\bar{f}^{B} + (24D_{B} + c'^{BB}\bar{f}^{B})\nu_{b}^{A}\bar{f}^{A}}{\nu_{b}^{B}\bar{f}^{B}c'^{AB}\bar{f}^{A} + \nu_{b}^{A}\bar{f}^{A}c'^{BA}\bar{f}^{B}}$$

with \bar{f}^A and \bar{f}^B constant steady states and $c'^{ST} = \frac{2\pi\kappa^{ST} \nu_c^{ST} R^4}{\nu_c^{ST}}$, $S, T \in \{A, B\}$.

The constant steady states are unstables if $s>s_{\scriptscriptstyle L}^*$.

To simplify notation and since $\bar{f}^B = f^* - \bar{f}^{\bar{A}}$, we obtain:

$$F(\bar{f}^A) = \frac{\beta(\bar{f}^A)^2 + \alpha \bar{f}^A + \gamma}{\varepsilon(\bar{f}^A)^2 + \delta \bar{f}^A},$$

with parameters:

$$\alpha = 24D_{B}\nu_{b}^{A} - 24D_{A}\nu_{b}^{B} + c'^{AA}\nu_{b}^{B}f^{*} + c'^{BB}\nu_{b}^{A}f^{*}, \quad \beta = -c'^{AA}\nu_{b}^{B} - c'^{BB}\nu_{b}^{A},$$

$$\gamma = 24D_{A}\nu_{b}^{B}f^{*}, \quad \delta = c'^{AB}\nu_{b}^{B}f^{*} + c'^{BA}\nu_{b}^{A}f^{*}, \quad \varepsilon = -\nu_{b}^{B}c'^{AB} - \nu_{b}^{A}c'^{BA}.$$
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Logistic vs. no logistic

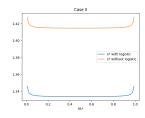
We compare critical values related to instability:

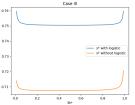
Logistic growth model

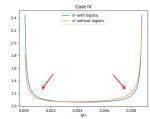
No logistic model

$$s_L^* = \frac{(24D_A + c'^{AA})\nu_b^B \bar{f}^B + (24D_B + c'^{BB})\nu_b^A \bar{f}^A}{\nu_b^B \bar{f}^B \tilde{c}'^{AB} + \nu_b^A \bar{f}^A \bar{c}'^{BA}},$$

$$s_{C}^{*} = \sqrt{\frac{576}{\tilde{c}^{\prime AB}\tilde{c}^{\prime BA}}\left(D_{A} + \frac{c^{\prime AA}}{24}\right)\left(D_{B} + \frac{c^{\prime BB}}{24}\right)}$$









Numerical simulations



Numerical simulations

Micro Macro



blabla



Thank you!

