Mathematical modeling of a growing tissue

Individual Based Models

Aggregation Models

Mechanical Models

Free Boundary Models

Refs : [Perthame Quiros Vazquez 2014] (power pressure law), Sophie Hecht's PhD works 2017-2020 (singular pressure law); **Sophie Hecht's course**

IBM N->+0 Large ropulation AM Localisation limit inviscitu Poorcy law Brinkman law

$$\partial_t n +
abla \cdot (nv) = nG(p),$$
 Growt $v = -
abla p,$ Darcy $p = p(n) = \epsilon rac{n}{1-n},$ Presse

n(t,x)cell density of the tissuev(t,x)velocity of the tissuep(n)pressure due to congestionGpressure-dependent growth function

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- th and migration
- 's law
- ure law (singular)



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 $< \partial_{+}e^{\nu} - \nabla(e^{\nu}\nabla\psi * e^{\nu}), \Psi > =$ 3) Incompressible limit We convoider the following model $\partial \mu m - \nabla (m \nabla p) = m G(p) (n)$ with D=T(n) -Dp



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Viscosity and rotational

Using Darcy's law



we necessarily have a null rotational !



Viscosity and rotational

Using Darcy's law



we necessarily have a null rotational !

To reproduce the rotational movements observed in the embryo, and take into account the (differential) viscosities of the tissues, we choose instead the **Brinkman form**:

 $-eta\Delta v$ -





$$+v = -\nabla p$$

Multi-tissue models

$$\left\{egin{aligned} &\partial_t n_1 +
abla \cdot (n_1 v_1) = n_1 G_1(p_1), \ &\partial_t n_2 +
abla \cdot (n_2 v_2) = n_2 G_2(p_2), \ &-eta_1 \Delta v_1 + v_1 = -
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abla p_2, \ &p_1 = p_\epsilon (n_1 + n_2), \ &p_2 = p_\epsilon (n_1 + n_2), \ &p_2 = p_\epsilon (n_1 + n_2), \ &p_\epsilon (n) = \epsilon rac{n}{1-n}. \end{array}
ight.$$

- n_i : density of each tissue
- v_i : velocity of each tissue
- p_i : pressure of each tissue
- β_i : viscosity of each tissue
- p_{ϵ} : congestion pressure, parametrized by $\epsilon > 0$
- G_i : growth fonctions



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mixing of the two tissues ?

?

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- v_i : velocity of each tissue
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Propagated segregation: starting from initial segregated tissues

we get that the tissues remain segregated for all times

- $n_1^{ini} n_2^{ini} = 0$
- $n_1(t, \cdot)n_2(t, \cdot) = 0, t > 0.$

Propagated segregation: starting from initial segregated tissues

we get that the tissues remain segregated for all times

This property is proved for equal viscosities, and numerically observed for different viscosities.



- $n_1^{ini} n_2^{ini} = 0$
- $n_1(t, \cdot)n_2(t, \cdot) = 0, t > 0.$





$$\begin{cases} \partial_t n_1 + \nabla \cdot (n_1 v_1) = n_1 G_1(p_1), \\ \partial_t n_2 + \nabla \cdot (n_2 v_2) = n_2 G_2(p_2), \\ -\beta_1 \Delta v_1 + v_1 = -\nabla p_1, \\ -\beta_2 \Delta v_2 + v_2 = -\nabla p_2, \\ p_1 = p_{\epsilon}(n_1 + n_2), \\ p_2 = p_{\epsilon}(n_1 + n_2), \\ p_2 = p_{\epsilon}(n_1 + n_2), \\ p_{\epsilon}(n) = \epsilon \frac{n}{1-n}. \end{cases}$$

Segregation is **propagated** from the initial data (**passive segregation**).

- n_i : density of each tissue
- v_i : velocity of each tissue
- p_i : pressure of each tissue
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Mechanical model: origin of segregation ?



origin?

Cells of NT and PSM types migrate from the PZ towards the two other tissues

- If the segregation is only propagated, then, what is its
- What happens at the interface with the PZ, which is a mixed zone?
- Biological hypothesis: could it be that some active segregation is at play?

We keep our first model with *passive segregation* but also build a model with *active segregation* in order to test this hypothesis.



$$\begin{cases} \partial_t n_1 + \nabla \cdot (n_1 v_1) + \alpha \nabla \cdot (n_1 \nabla (\Delta n_1)) \\ \partial_t n_2 + \nabla \cdot (n_2 v_2) + \alpha \nabla \cdot (n_2 \nabla (\Delta n_2)) \\ -\beta_1 \Delta v_1 + v_1 = -\nabla p_1, \\ -\beta_2 \Delta v_2 + v_2 = -\nabla p_2, \end{cases}$$

$$p_1 = p_{\epsilon}(n_1 + n_2) + n_2 q_m(n_1 n_2), \\ p_2 = p_{\epsilon}(n_1 + n_2) + n_1 q_m(n_1 n_2), \\ p_{\epsilon}(n) = \epsilon \frac{n}{1-n}, \\ q_m(r) = \frac{m}{m-1}((1+r)^{m-1} - 1). \end{cases}$$

Active segregation

Ref : [Chertock Degond Hecht Vincent 2019] (bubble effect)

 $)) = n_1 G_1(p_1),$ $)) = n_2 G_2(p_2),$

- n_i : density of each tissue
- v_i : velocity of each tissue
- p_i : pressure of each tissue
- β_i : viscosity of each tissue
- p_{ϵ} : congestion pressure, parametrized by $\epsilon > 0$
- G_i : growth fonctions
- q_m : repulsion pressure, parametrized by m > 0
- $\alpha > 0$: regularization parameter





Numerical simulation and comparison with data

Model 1 (fluid, passive segregation)















0¹⁰

8

6

2

0.8

4

10

, ,

Comparison between models: densities



Fluid model, Passive segregation



Fluid model, Active segregation



Comparison between models: rotational



Fluid model, Passive segregation



Fluid model, Active segregation



Comparison between models: rotational



Fluid model, Passive segregation



Fluid model, Active segregation



Comparison between models



Our model suggests that some active segregation is at play.







Diving deeper into the biological data

One step forward towards biology: - we add injection (duly *quantified*) - other parameters taken from the biological literature

Parameter	Source	Value	Unit
NT viscosity (β1)	Michaut 2018	10 ⁴ - 10 ⁵	Pa·s
PSM viscosity (β ₂)	Mahadevan 2017, Michaut 2018	10 ⁴ - 10 ⁵	Pa·s
Proliferation rate of the NT (g ₁)	Bénazéraf 2017	1/10.83	1/ hour
Proliferation rate of the PSM (g ₂)	Bénazéraf 2017	1/8.75	1/ hour
Injection rate from the PZ into the NT (KNT)	Measured experimentally	5	Cells/ hour/ 10µr
Injection rate from the PZ into the PSM (KPSM)	Measured experimentally	10	Cells/ hour/ 10µr
Total maximal density (n _{max})	Bénazéraf 2017	2800	Cells/ 100 µm ³
Pressure sensing (ϵ)	Estimated free parameter	1	Pa
Tissue friction with surroundings (µ)	Mahadevan 2017, Michaut 2018	1012 - 1013	Pa • s/m²
Growth sensitivity near the maximal density (a)	Estimated free parameter	8	Dimensionless
Width of injection zone (δ)	Measured experimentally	214	μm



Diving deeper into the biological data

One step forward towards biology:

- we add injection (duly *quantified*)
- other parameters taken from the biological literature

Validation of the model:

- with « natural » parameters, compared with wild type data (« healthy » embryo) - with pathological parameters, compared with experiments in vivo (proliferation suppressed, ...) In particular, the model reproduces well the differential elongation and the sliding between tissues.







Diving deeper into the biological data

One step forward towards biology:

- we add injection (duly *quantified*)
- other parameters taken from the biological literature

Validation of the model:

- with « natural » parameters, compared with wild type data (« healthy » embryo)

(differential) tissue proliferation in the embryogenesis.





- with pathological parameters, compared with experiments in vivo (proliferation suppressed, ...) In particular, the model reproduces well the differential elongation and the sliding between tissues.
- Output of the model: we perform a sensitivity analysis based on relevant outputs characterizing shapes and velocities of the tissues. This analysis reveals the underestimated importance of





Score:= density, pressure, tissue width (mid AP axis), velocities, elongation rate





Mathematical analysis: a few words

$$\begin{cases} \partial_t n_1 + \nabla \cdot (n_1 v_1) + \alpha \nabla \cdot (n_1 \nabla (\Delta n_1)) \\ \partial_t n_2 + \nabla \cdot (n_2 v_2) + \alpha \nabla \cdot (n_2 \nabla (\Delta n_2)) \\ -\beta_1 \Delta v_1 + v_1 = -\nabla p_1, \\ -\beta_2 \Delta v_2 + v_2 = -\nabla p_2, \end{cases}$$

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Ref : [Degond Hecht Romanos Trescases 2022]

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Can we compute the incompressible limit?

 $\epsilon \rightarrow$ 0, $m \rightarrow \infty$ and $\alpha \rightarrow$ 0

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Formally, YES.

Simulation before and at the limit



Ref : [Degond Hecht Romanos Trescases 2022]



Geometrical model, Passive segregation

Analysis at the incompressible limit

Under simplifying assumptions (in a bounded domain in 2D, stationary case, linearly decreasing growth function, smooth subdomains, given segregation pressure in L^2), and under an ellipticity condition, we prove:

- well-posedness (weak solution in H^{1})
- smoothness inside each subdomain
- equivalence with a transmission problem
- pressure jump at the interface (when the segregation pressure is null)

Ref : [Degond Hecht Romanos Trescases 2022]



Transmission problem

$$on \ \Omega_{\mathbf{1}} : (T_{\mathbf{1}}) \begin{cases} -\beta_{\mathbf{1}} \Delta v_{\mathbf{1}} + v_{\mathbf{1}} - \frac{\mathbf{1}}{g_{\mathbf{1}}} \nabla \nabla \cdot v_{\mathbf{1}} = \mathbf{0}, \\ -\beta_{\mathbf{2}} \Delta v_{\mathbf{2}} + v_{\mathbf{2}} - \frac{\mathbf{1}}{g_{\mathbf{1}}} \nabla \nabla \cdot v_{\mathbf{1}} = \mathbf{0}, \end{cases}$$

on
$$\Omega_2$$
: (T_2)

$$\begin{cases}
-\beta_1 \Delta v_1 + v_1 - \frac{1}{g_2} \nabla \nabla \cdot v_2 = \mathbf{0}, \\
-\beta_2 \Delta v_2 + v_2 - \frac{1}{g_2} \nabla \nabla \cdot v_2 = \mathbf{0},
\end{cases}$$

on
$$\Omega^{c}$$
: (T_{3})
$$\begin{cases} -\beta_{1}\Delta v_{1} + v_{1} = \mathbf{0}, \\ -\beta_{2}\Delta v_{2} + v_{2} = \mathbf{0}. \end{cases}$$

on
$$\Gamma_{\mathbf{1}}$$
: $(C_{\mathbf{1}})$

$$\begin{cases}
\beta_{\mathbf{1}}[(\nabla v_{\mathbf{1}})_{\Omega_{\mathbf{1}}} - (\nabla v_{\mathbf{1}})_{\Omega}c] \cdot \vec{\nu} = [p_{\mathbf{1}}^{\star} - \frac{\mathbf{1}}{g_{\mathbf{1}}}(\nabla \cdot v_{\mathbf{1}})_{\Omega_{\mathbf{1}}}] \\
\beta_{\mathbf{2}}[(\nabla v_{\mathbf{2}})_{\Omega_{\mathbf{1}}} - (\nabla v_{\mathbf{2}})_{\Omega}c] \cdot \vec{\nu} = [p_{\mathbf{1}}^{\star} - \frac{\mathbf{1}}{g_{\mathbf{1}}}(\nabla \cdot v_{\mathbf{1}})_{\Omega_{\mathbf{1}}}] \\
(v_{\mathbf{1}})_{\Omega_{\mathbf{1}}} = (v_{\mathbf{1}})_{\Omega}c, \quad (v_{\mathbf{2}})_{\Omega_{\mathbf{1}}} = (v_{\mathbf{2}})_{\Omega}c,
\end{cases}$$

on
$$\Gamma_{2}$$
: (C_{2})

$$\begin{cases}
\beta_{1}[(\nabla v_{1})_{\Omega_{2}} - (\nabla v_{1})_{\Omega^{c}}] \cdot \vec{\nu} = [p_{2}^{\star} - \frac{1}{g_{2}}(\nabla \cdot v_{2})_{\Omega_{2}}] \\
\beta_{2}[(\nabla v_{2})_{\Omega_{2}} - (\nabla v_{2})_{\Omega^{c}}] \cdot \vec{\nu} = [p_{2}^{\star} - \frac{1}{g_{2}}(\nabla \cdot v_{2})_{\Omega_{2}}] \\
(v_{1})_{\Omega_{2}} = (v_{1})_{\Omega^{c}}, \quad (v_{2})_{\Omega_{2}} = (v_{2})_{\Omega^{c}},
\end{cases}$$

$$on \ \Gamma: (C_{\mathbf{3}}) \begin{cases} \beta_{\mathbf{1}}[(\nabla v_{\mathbf{1}})_{\Omega_{\mathbf{1}}} - (\nabla v_{\mathbf{1}})_{\Omega_{\mathbf{2}}}] \cdot \vec{\nu} = [(p_{\mathbf{1}}^{\star} - p_{\mathbf{2}}^{\star}) + \frac{\mathbf{1}}{g_{\mathbf{2}}}(\nabla \cdot v_{\mathbf{2}})_{\Omega_{\mathbf{2}}}) - \frac{\mathbf{1}}{g_{\mathbf{1}}}(\nabla \cdot v_{\mathbf{1}})_{\Omega_{\mathbf{1}}}]\vec{\nu}, \\ \beta_{\mathbf{2}}[(\nabla v_{\mathbf{2}})_{\Omega_{\mathbf{1}}} - (\nabla v_{\mathbf{2}})_{\Omega_{\mathbf{2}}}] \cdot \vec{\nu} = [(p_{\mathbf{1}}^{\star} - p_{\mathbf{2}}^{\star}) + \frac{\mathbf{1}}{g_{\mathbf{2}}}(\nabla \cdot v_{\mathbf{2}})_{\Omega_{\mathbf{2}}}) - \frac{\mathbf{1}}{g_{\mathbf{1}}}(\nabla \cdot v_{\mathbf{1}})_{\Omega_{\mathbf{1}}}]\vec{\nu}, \\ (v_{\mathbf{1}})_{\Omega_{\mathbf{1}}} = (v_{\mathbf{1}})_{\Omega_{\mathbf{2}}}, \quad (v_{\mathbf{2}})_{\Omega_{\mathbf{1}}} = (v_{\mathbf{2}})_{\Omega_{\mathbf{2}}}, \quad v_{\mathbf{1}} \cdot \vec{\nu} = v_{\mathbf{2}} \cdot \vec{\nu}. \end{cases}$$

Ref : [Degond Hecht Romanos Trescases 2022]



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The rigorous incompressible limit

For the passive segregation model:

- recently proven for viscous tissues *in the whole domain* refs: [Perthame Vauchelet 2015] (one tissue); [Dębiec Schmidtchen 2020] and [Dębiec Perthame Schmidtchen Vauchelet 2021] (multi-species, *same viscosities*)

- still open in a bounded domain
- still open for different viscosities

The active segregation model is another challenge..!



Conclusion

Summary

- We developed multi-tissue mechanical models with *passive* or *enforced* segregation - We derived from it (formally) a geometrical model
- We analyzed the geometrical model: stationary state, pressure jump
- We simulated the model and compared it with biological data

Biological conclusions

- Suggests the existence of a segregation force
- Highlights the underestimated role of differential tissue proliferation rates



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Perspectives

- More tissues, 3D, ...
- Diverse species
- Link with microscopic models
- Rigorous incompressible limit















Thank you for your attention

Numerical schemes

Mechanical model (1): Finite volume semi-implicit scheme - Matlab

Mechanical model (2):

Finite volume semi-implicit scheme, with a relaxation method (order reduction) - Matlab

Geometrical model: Finite elements - Freefem++





