

Asymptotics behaviour of structured population models and growth fragmentation models arising in biology

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Motivation I: Metastatic spreading

Metastases: a major cause of death in cancer

- Metastatic state of the patient is often difficult to evaluate, as micro-tumors are hardly detectable from imagery.

Questions

- Can we design a new ‘in silico’ metastatic index?
- Can we infer the metastatic aggressivity from biomarkers?

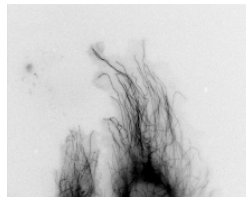
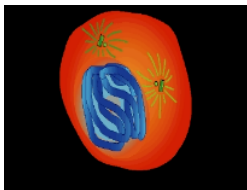
Mathematic tools

- **McKendrick-Von Foerster equation** for a simple emission
- **Growth-fragmentation equation** for general emission

Motivation II: Microtubules

Microtubules: a therapeutic target in oncology

- MTs play a crucial role in **cell division**, in **cell migration**
- ↪ MTs are a favorite target of **Microtubule Targeting Agents (MTAs)**, successfully used as **antimitotic** more recently as antiangiogenic agent or antimigratory agent in cancer treatments.
- MTs are **polymers highly dynamic**.



Questions

- Can we model the effect of MTAs on the MT dynamical instabilities?
- Can we better understand the low dose effect of MTAs?

Mathematical tools

- Complex models using **Growth-fragmentation equations**

Renewal equation vs growth-fragmentation equation

Perthame, Transport equation in biology, 2006

McKendrick-Von Foerster equation or renewal equation

$$\begin{cases} \frac{\partial \rho}{\partial t} + \frac{\partial \rho}{\partial a} = -D(a)\rho(t, a), & t > 0, a > 0 \\ \rho(t, 0) = \int_0^\infty B(y) \rho(t, y) dy, & t > 0 \\ \rho(0, a) = \rho_0(a), & a > 0 \end{cases}$$

Typically, ρ is the density of a population structured by the age a and B is the birth rate and D the death rate.

Growth-fragmentation equation

$$\begin{cases} \frac{\partial \rho}{\partial t} + \frac{\partial}{\partial x}(g(x)\rho) = -B(x)\rho(t, x) + \int_0^\infty B(y)k(x, y)\rho(t, y) dy, & t > 0, x > 0 \\ \rho(t, 0) = 0, & t > 0 \\ \rho(0, x) = \rho_0(x), & x > 0 \end{cases}$$

Typically, ρ is the density of a cell population structured by its size x and B is the division rate and $k(x, y)$ is the probability that the division of a cell of size y leads to a cell of size x .

1 The McKendrick-Von Foerster equation

- Model for mitosis - structuration by age
- Model of metastases - single cell emission

2 Growth-fragmentation equation

- Model for Mitosis - structuration by size
- Model of metastases - emission by cluster
- Model for prion disease
- Model for the MTs dynamical instabilities

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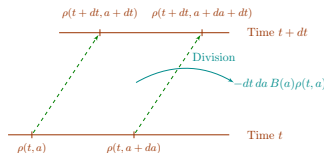
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Mitosis - structuration by age

Population of cells structured by age that divide at a rate B giving 2 cells of age 0.

$$(*) \begin{cases} \frac{\partial \rho}{\partial t} + \frac{\partial \rho}{\partial a} = -B(a)\rho(t, a), & t > 0, a > 0 \\ \rho(t, 0) = 2 \int_0^\infty B(y) \rho(t, y) dy, & t > 0 \\ \rho(0, a) = \rho_0(a), & a > 0 \end{cases}$$

- $\rho(t, a)$ density at time t with an age a
- $B(a)$ is the division rate



Theorem

Assume that $B \in L^\infty(\mathbb{R}^+)$ with $B > 0$ and $\int_0^\infty B(y) dy = \infty$, then $(*)$ admits a unique solution $\rho \in \mathcal{C}(\mathbb{R}^+; L^1(\mathbb{R}^+, \phi(a) da))$ and if there exists $\mu_0 > 0$ such that $2B(a) \geq \mu_0 \frac{\phi(a)}{\phi(0)}$ then

$$\int_0^\infty |e^{-\lambda_0 t} \rho(t, a) - \bar{\rho}_0 N(a)| \phi(a) da \xrightarrow[t \rightarrow \infty]{} 0$$

where (λ_0, N, ϕ) are the eigenelements associated to the problem and $\bar{\rho}^0 = \int_0^\infty \phi(a) \rho^0(a) da$.

Mitosis - Structuration by age

$$\rho(t, \cdot) \sim e^{\lambda_0 t} \bar{\rho}_0 N(\cdot)$$

$$\frac{\partial \rho}{\partial t} + \frac{\partial \rho}{\partial a} = -B(a)\rho(t, a), \quad \rho(t, 0) = 2 \int_0^\infty B(y) \rho(t, y) dy, \quad \rho(0, a) = \rho_0(a)$$

The eigenvalue problem

► Sketch of proof

■ Eigenvalue problem:

$$\lambda_0 N(a) + N'(a) + B(a)N(a) = 0, \quad N(0) = 2 \int_0^\infty B(a)N(a) da \quad (*)$$

■ Adjoint problem

$$\lambda_0 \phi(a) - \phi'(a) + B(a)\phi(a) = 2\phi(0)B(a) \quad (**)$$

↪ If B is a positive continuous function $\exists!(N, \phi, \lambda_0)$ taking positive values solution to $(*) - (**) such that$

$$\int_0^\infty N(a) da = \int_0^\infty \phi(a)N(a) da = 1$$

Mitosis - Structuration by age

$$\frac{\partial \rho}{\partial t} + \frac{\partial \rho}{\partial a} = -B(a)\rho(t, a), \quad \rho(t, 0) = 2 \int_0^\infty B(y) \rho(t, y) dy, \quad \rho(0, a) = \rho_0(a)$$

Eigenelements

$$\begin{cases} \lambda_0 N(a) + N'(a) = -B(a)N(a), & N(0) = \int_0^\infty B(a)N(a) da & (*) \\ \lambda_0 \phi(a) - \phi'(a) + B(a)\phi(a) = \phi(0)B(a) & & (**) \end{cases}$$

Method of generalised entropy

► Sketch of proof



$$\int_0^\infty \phi(a) e^{-\lambda_0 t} \rho(t, a) da = \int_0^\infty \phi(a) \rho^0(a) da := \bar{\rho}^0$$

- Let $m(t, a) = e^{-\lambda_0 t} \frac{\rho(t, a)}{N(a)}$, then for all convex function \mathcal{H}

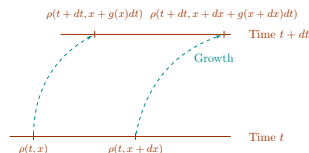
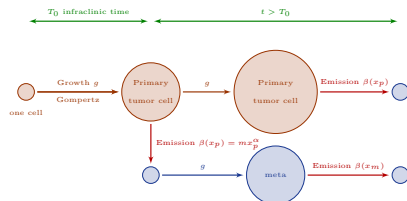
$$\frac{d}{dt} \int_0^\infty \phi(a) N(a) \mathcal{H}(m(t, a)) da := \Delta \leq 0$$

and applied it for $\mathcal{H}(m) = |m - \bar{\rho}_0|$. If $\exists \mu_0 > 0, \forall a \in \mathbb{R}^+, \chi(a) := 2 \frac{\phi(0)B(a)}{\phi(a)} \geq \mu_0$ then $\Delta \leq -\mu_0 \int \phi N \mathcal{H}(m)$.

Metastase model

The original model of metastases

Iwata & al (2000)



■ $\rho(t, x)$ density of metastases at time t of size x .

A transport equation for the growth of metastases

$$\partial_t \rho(x, t) + \partial_x (g(x) \rho(x, t)) = 0, \quad t > 0, \quad x > 1$$

A boundary condition for the emission

$$g(1)\rho(t, 1) = \underbrace{\beta(x_p(t))}_{\text{Emission by the primary tumor: } f(t)} + \underbrace{\int_1^b \beta(x) \rho(t, x) dx}_{\text{Emission by the metastases}}, \quad t > 0$$

Growth law

$$x'_p = g(x_p) \text{ with } g(x) = ax \ln \left(\frac{b}{x} \right) \rightsquigarrow \text{Gompertz law}$$

$$\begin{cases} \partial_t \rho(x, t) + \partial_x (g(x) \rho(x, t)) = 0, & t > 0, x > 1 \\ g(1) \rho(t, 1) = f(t) + \int_1^b \beta(x) \rho(t, x) dx \end{cases}$$

Existence and uniqueness

Barbolosi, Benabdallah, FH, Verga 2008

- If $\rho_0 \in L^1(1, b)$, there exists a unique weak solution $\rho \in \mathcal{C}([0, \infty[; L^1(1, b))$.
- Existence of strong solution for more regular ρ_0 and compatibility condition between ρ_0 and $\beta(x_p(0))$.

Asymptotic behaviour

Barbolosi, Benabdallah, FH, Verga 2008

- There exists (λ_0, V, ϕ) and $\gamma > 0$ such that

$$\left\| e^{-\lambda_0 t} \rho(t) - \bar{\rho}_0 V \right\|_{L^1_{\phi}(1, b)} \leq e^{-\gamma t} \|\rho_0\|_{L^1_{\phi}(1, b)} + \int_0^t e^{-\lambda_0 \tau} |f(\tau)| d\tau.$$

↪ Proof thanks to the semi-group approach

$$\begin{cases} \partial_t \rho(x, t) + \partial_x (g(x) \rho(x, t)) = 0, & t > 0, x > 1 \\ g(1) \rho(t, 1) = f(t) + \int_1^b \beta(x) \rho(t, x) dx \end{cases}$$

Inverse problem

Hartung, 2015

- The observables $F_f(t) = \int_1^b f(x) \rho(t, x) dt$ are solution of a Volterra equation

$$F_f(t) = [f(x_p) * \beta(x_p)](t) + [F_f * \beta(x_p)](t)$$

Theorem

If $F_f \in \mathcal{C}^1$, $F_f(0) = 0$ and $F_f + f \in \mathcal{C}^1$, $F_f + f(0) \neq 0$, then β can be identified from $F_f(t)$ and x_p .

Metastases model

$$\begin{cases} \partial_t \rho(x, t) + \partial_x (g(x) \rho(x, t)) = 0, & t > 0, x > 1 \\ g(1) \rho(t, 1) = f(t) + \int_1^b \beta(x) \rho(t, x) dx \end{cases}$$

Confrontation to the data

■ Extension on the model

$$\begin{cases} \frac{\partial}{\partial t} \rho(t, x) + \frac{\partial}{\partial x} [g_m(x) \rho(t, x)] = 0, & x \in [1, b), t \geq 0 \\ g_m(1) \rho(t, 1) = \int_1^b \beta(x) \rho(t, x) dx + \beta(x_p(t)) \\ \rho(0, x) = 0, \end{cases}$$

where g_p and g_m are one of the classical growth speed:

Gompertz model (1825)	$g(x) = ax \ln\left(\frac{b}{x}\right)$
Hybrid Gompertz (HG)	$g(x) = \min\left(a_{in vitro}, ax \ln\left(\frac{b}{x}\right)\right)$
Logistic model (1838)	$g(x) = ax \left(1 - \frac{x}{b}\right)$
Von Bertalanffy (1949)	$g(x) = ax \left(\left(\frac{x}{b}\right)^{-\frac{1}{3}} - 1\right)$
West& al (1997)	$g(x) = ax \left(\left(\frac{x}{b}\right)^{-\frac{1}{4}} - 1\right)$
Hybrid West (HW)	$g(x) = \min\left(a_{in vitro}, ax \left(\left(\frac{x}{b}\right)^{-\frac{1}{4}} - 1\right)\right)$

Hartung & al, 2014



■ Use SAEM algorithm



■ Good estimates for

- HG
- HW

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Growth fragmentation equation

$$\begin{cases} \frac{\partial \rho}{\partial t} + \frac{\partial}{\partial x}(g(x)\rho) = -B(x)\rho(t, x) + \int_0^\infty B(y)k(x, y)\rho(t, y) dy, & t > 0, x > 0 \\ \rho(t, 0) = 0, & t > 0 \\ \rho(0, x) = \rho_0(x), & x > 0 \end{cases}$$

- ρ is the density of a population structured by a variable (trait) x at time t
- g is the growth rate
- B is the total division/fragmentation rate
- $k(x, y)$ is the fragmentation kernel: rate at which individuals of trait x are obtained from an individual of trait y .

Growth fragmentation equation

$$\begin{cases} \partial_t \rho + \partial_x(g(x)\rho) = -B(x)\rho(t, x) + \int_x^{+\infty} B(y)k(x, y)\rho(t, y) dy, \\ \rho(t, 0) = 0, \rho(0, x) = \rho_0(x) \end{cases}$$

Properties of the kernel

- No fragmentation to a bigger size: $k(x, y) = 0$ if $x > y$
- Conservation of the total size: $\int_0^y xk(x, y) dx = y$
- For division into a fixed number p of pieces: $\int_0^y k(x, y) dx = p$

Classical examples

- Division into 2 cells of equal size - equal mitosis

$$\begin{cases} \partial_t \rho + \partial_x(g(x)\rho) = -B(x)\rho(t, x) + 4B(2x)\rho(t, 2x), & x > 0, t > 0, \\ \rho(t, 0) = 0, \rho(0, x) = \rho_0(x) \end{cases}$$

with $k(x, y) = 2\delta_{x=\frac{y}{2}}$, so that $\int_0^y k(x, y) dy = 2$.

- Division into 2 cells with different sizes

$$\begin{cases} \partial_t \rho + \partial_x(g(x)\rho) = -B(x)\rho(t, x) + 2 \int_x^{+\infty} B(y)\kappa(x, y)\rho(t, y) dy, \\ \rho(t, 0) = 0, \rho(0, x) = \rho_0(x) \end{cases}$$

here $k(x, y) = 2\kappa(x, y)$

Growth fragmentation equation

$$\begin{cases} \partial_t \rho + \partial_x (g(x) \rho) = -B(x) \rho(t, x) + \int_x^{+\infty} B(y) k(x, y) \rho(t, y) dy, \\ \rho(t, 0) = 0, \rho(0, x) = \rho_0(x) \end{cases}$$

Properties of the kernel

- No fragmentation to a bigger size: $k(x, y) = 0$ if $x > y$
- Conservation of the total size: $\int_0^y x k(x, y) dx = y$
- For division into a fixed number p of pieces: $\int_0^y k(x, y) dx = p$

Classical examples

- Renewal equation: $k(x, y) = \frac{1}{2}(\delta(x=0) + \delta(x=y))$
- Autosimilar case: $k(x, y) = \frac{1}{y} \kappa_0\left(\frac{x}{y}\right)$ with $\int_0^1 s \kappa_0(s) ds = 1$.
 - \rightsquigarrow general mitosis: $\kappa_0 = \delta_r + \delta_{1-r}$, $r \in [0, \frac{1}{2}]$
 - \rightsquigarrow homogeneous fragmentation: $\kappa_0(s) = (1 + \alpha)(s^\alpha + s^{1-\alpha})$, $\alpha > -1$

Growth fragmentation equation

$$\begin{cases} \partial_t \rho + \partial_x (g(x)\rho) = -B(x)\rho(t, x) + \int_x^{+\infty} B(y)k(x, y)\rho(t, y) dy, \\ \rho(t, 0) = 0, \rho(0, x) = \rho_0(x) \end{cases}$$

Many references on this topic, *e.g.*

- **Perthame, 2007:** Study for $g = 1$ of the eigenvalue problem via the Krein-Rutman problem. Hints for the proof of convergence.
- **Doumic-Gabriel, 2013:** existence of a solution to the eigenvalue problem (direct and dual) given with many details for the case $\int \kappa(x, y)dy = 2$ and for B and g general.
- **Gabriel & al, 2021:** Asymptotic behaviour $\rho(t, x) \sim e^{\lambda t} N(x)$ for quite general assumption on k and B using a probabilistic approach namely **Harry's theorem**.

Growth fragmentation equation

$$\begin{cases} \partial_t \rho + \partial_x (g(x)\rho) = -B(x)\rho(t, x) + \int_x^{+\infty} B(y)k(x, y)\rho(t, y) dy, \\ \rho(t, 0) = 0, \rho(0, x) = \rho_0(x) \end{cases}$$

Results from Gabriel & al, 2021

Assumptions (H_*)

- Assumptions on the kernel.

- Autosimilar kernel such that $\kappa_0(s) \geq \underline{c} > 0$ and $\int_0^1 \kappa_0 < \infty$.
- $\kappa_0 = 2\delta_{\frac{1}{2}}$ (can be relax)

- Assumptions on the growth term :

- $\int_0^1 \frac{1}{g} < \infty$
- Assumption on H defined by $H(z) = \int_0^z \frac{1}{g} < \infty$ eg $H < \infty$ on \mathbb{R}^+ , H invertible, H^{-1} does not grow too fast

- Assumptions on the relation between B and g

- $\int_0^1 \frac{B}{g} < \infty$, $\lim_0 \frac{x B(x)}{g(x)} = 0$, $\lim_{+\infty} \frac{x B(x)}{g(x)} = +\infty$

Growth fragmentation equation

$$\begin{cases} \partial_t \rho + \partial_x (g(x)\rho) = -B(x)\rho(t, x) + \int_x^{+\infty} B(y)k(x, y)\rho(t, y) dy, \\ \rho(t, 0) = 0, \rho(0, x) = \rho_0(x) \end{cases}$$

Results from Gabriel & al, 2021

Theorem

1 Under assumptions (H_*) , the eigenvalue problem

$$\begin{cases} -(gN)' - BN + \int_x^\infty B(y)k(x, y)N(y) dy = \lambda_0 N, (gN)(0) = 0, \int N = 1 \\ -g\phi' - B\phi + \int_x^\infty B(y)k(x, y)\phi(y) dy = \lambda_0 \phi, \int N\phi = 1 \end{cases}$$

admits a unique solution (λ_0, N, ϕ) .

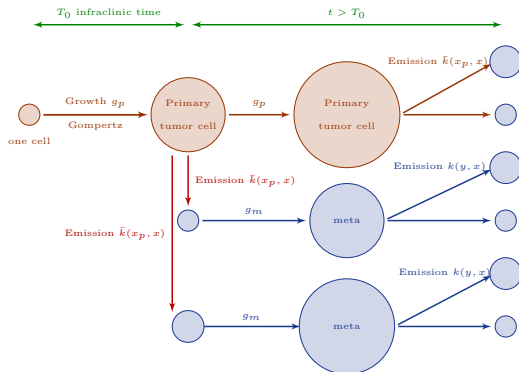
2 If $\|\rho_0\|_V < \infty$,

$$\|e^{-\lambda_0 t} \rho(t, \cdot) - \bar{\rho}_0 N\|_V \leq C e^{-\gamma t} \|\rho_0 - \bar{\rho}_0 N\|_V, \forall t \geq 0$$

where V is a weight depending on the data.

General emission of metastases

Each tumor (primary or secondary) can emit several tumors of different size !



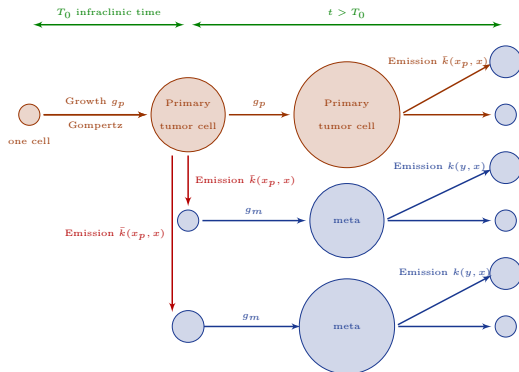
Characterisation of the emission

- ▶ $\beta(x)$ emission rate
- ▶ $k(y, x)$ probability for a tumor of size x to emit a metastase of size y .

↪ a growth-fragmentation equation with source term

General emission of metastases

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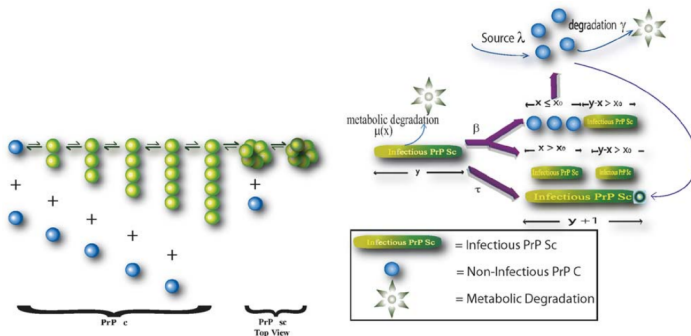
$$\frac{\partial}{\partial t} \rho(t, x) + \frac{\partial}{\partial x} [g_m(x) \rho(t, x)] = \bar{k}(x, x_p(t)) - \beta(x) \rho(t, x) + \int_x^{+\infty} \beta(y) k(x, y) \rho(t, y) dy$$

Few results on this equation and still open questions on this equation !

Model for prion disease

Prion diseases, a family of progressive neurodegenerative disorders

The prion protein can appear in two forms: the normal one PrP^c and the infectious one PrP^{Sc} . They form polymers also called amyloid fibers.



A polymer of size x can divide into two polymers of size y and $x - y$, as soon as a polymer has a too small size it depolymerizes.

Model for prion disease

Prion diseases, a family of progressive neurodegenerative disorders

- A polymer of size x can divide into two polymers of size y and $x - y$, as soon as a polymer has a too small size it depolymerizes.
- Polymerisation speed depends on the total number of PrP^c available.

$$\begin{cases} \partial_t \rho + \partial_x (\tau V(t) \rho) = -(\mu + \beta(x)) \rho(t, x) + 2 \int_x^\infty \beta(y) k(x, y) \rho(t, y) dy \\ V'(t) = \lambda - \gamma V(t) - \tau V(t) \int_0^\infty \rho(t, x) dx + 2 \int_0^{x_0} x \int_{x_0}^\infty \beta y k(x, y) u(y, t) dy dx \end{cases}$$

- $\rho(t, x)$ density of polymers at time t of size x .
- $V(t)$ number of PrP^c monomers
- $\beta(x) = \bar{\beta}x$ division rate
- $k(y, x) = \frac{1}{x}(x_0 < x)(y < x)$ probability that a polymer of size x divides into a polymer of size y .
- μ elimination rate

Greer et al (2006)

Model for prion disease

Prion diseases, a family of progressive neurodegenerative disorders

$$\begin{cases} \partial_t \rho + \partial_x (\tau V(t) \rho) = -(\mu + \beta(x)) \rho(t, x) + 2 \int_x^\infty \beta(y) k(x, y) \rho(t, y) dy \\ V'(t) = \lambda - \gamma V(t) - \tau V(t) \int_0^\infty \rho(t, x) dx + 2 \int_0^{x_0} x \int_{x_0}^\infty \beta y k(x, y) u(y, t) dy dx \end{cases}$$

Macroscopic asymptotic behaviour: derivation of an ODE system in

- Moments of ρ : $U(t) = \int_{x_0}^\infty \rho(t, x) dx$, $P(t) = \int_{x_0}^\infty x \rho(t, x) dx$
- $V(t)$ number of PrP^c monomers

$$\begin{cases} U' = \beta P - \mu U - 2\beta x_0 U \\ V' = \lambda - \gamma V - \tau V U + \beta x_0^2 U \\ P' = \tau V U - \mu P - \beta x_0^2 U \end{cases}$$

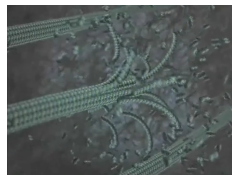
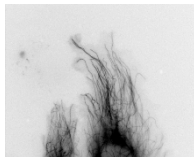
↪ Existence of a stable of a disease steady state if $(\bar{\beta} \lambda \tau / \gamma)^{\frac{1}{2}} > x_0 \bar{\beta} + \mu$.

- **Open problem:** long time behaviour of the density ρ .

Microtubule dynamical instabilities

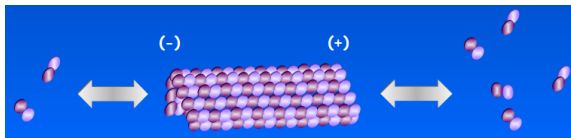
MT in the cell

- MTs are part of the cytoskeleton.
- MTs are characterized by their instabilities.



Protein structure

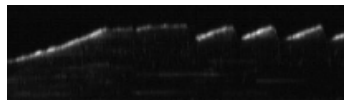
- Each MT is a long (up to $50\mu\text{m}$) hollow cylinder of 25nm diameter built from about 13 protofilaments.
- Each protofilament is composed by an assembly of $\alpha|\beta$ tubulin dimers.
- The assembly is polarized with different dynamics at the + end (highly dynamic) or - end (fixed in cells).
- Dimers can be in two energy states :
 - GTP : Guanosine triphosphate - active form
 - GDP : Guanosine diphosphate - inactive form



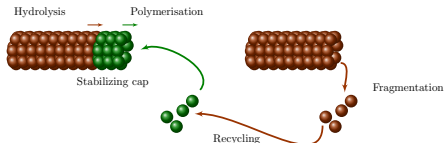
Dynamics of one MT at its + end

Dimers of tubulin

- Dimers can be in two energy states :
 - GTP : Guanosine triphosphate - active form
 - GDP : Guanosine diphosphate - inactive form
- Dimers can be polymerized or not. In fine,
 - GTP polymerized in MTs
 - GDP polymerized in MTs
 - Free GTP
 - Free GDP
- Biological observations:
 - Existence of a GTP-stabilizing cap
 - Disparition of the cap at the catastrophe



■ Four reactions



MTs dynamical instabilities

A structured population approach as in Hinow et al. (2009)

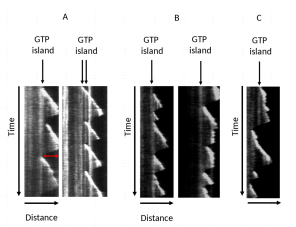
- 1 $u(t, z, x)$ density of MT in polymerisation
 - t time, x length, z length of the cap.
- 2 $v(t, x)$ density of the population of MT in depolymerisation
 - t time, x length.
- 3 $p = p(t)$ Free GTP tubulin
 - t time.
- 4 $q = q(t)$ Free GDP tubulin
 - t time.

↪ Two transport equations (for both polymerisation and depolymerisation) coupled to two ODEs.

↪ Several extensions

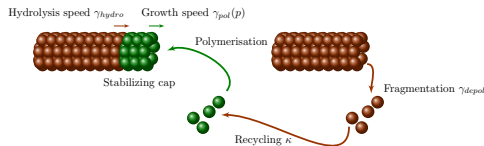
Barlukova PHD

New issue for the depolymerisation: ↪ fragmentation process



MTs dynamical instabilities

FH, M. Tournus, D. White, JTB (2017)



Equation for u

$$\partial_t u + \gamma_{pol}(p(t)) \partial_x u + (\gamma_{pol}(p(t)) - \gamma_{hydro}) \partial_z u = 0$$

Equation for v

$$\partial_t v = -R(t)u(t, 0, x) + \gamma_{depol} \left(- \int_0^x k(x, \tilde{x}) v(t, x) d\tilde{x} + \int_x^\infty k(\tilde{x}, x) v(t, \tilde{x}) d\tilde{x} \right)$$

Equation for p

$$\frac{d}{dt} p = -\gamma_{pol}(p(t)) \int_0^\infty \int_0^x u(t, z, x) dz dx + \kappa q$$

Equation for q

$$\frac{d}{dt} q = \gamma_{depol} \int_0^\infty \int_0^x (x - \tilde{x}) k(x, \tilde{x}) v(t, x) d\tilde{x} dx - \kappa q$$

MTs dynamical instabilities

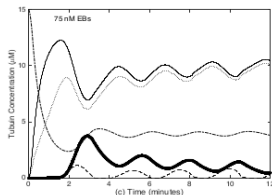
Macroscopic level

- 1 $M_u : t \mapsto \int_0^\infty \int_0^x xu(t, z, x) dz dx$ Total amount of MT in polymerisation
- 2 $M_v : t \mapsto \int_0^\infty xv(t, x) dx$ Total amount of MT in depolymerisation
- 3 $p = p(t)$ Free GTP tubulin
- 4 $q = q(t)$ Free GDP tubulin

\rightsquigarrow Conservation of the tubulin

$$M_u(t) + M_v(t) + p(t) + q(t) = Cte$$

Asymptotic behaviour at the macroscopic level



\rightsquigarrow Damped oscillations at the macroscopic level !

Simplified models to understand the asymptotics

- The population of polymer represented by $w : \rightsquigarrow w(t, x)$

- The model reduces to evolution of w, p, q

- Model should nevertheless reflect

- The role of the balance between hydrolysis and growth rate.

- $\gamma_{pol}(p(t)) < \gamma_{hydro} \Rightarrow$ period of catastrophe

- $\gamma_{pol}(p(t)) > \gamma_{hydro} \Rightarrow$ period of rescue

We introduce a threshold $\rightsquigarrow p_h$ such that $\gamma_{pol}(p_h) = \gamma_{hydro}$

- $p < p_h \Rightarrow$ period of catastrophe

- $p > p_h \Rightarrow$ period of rescue

Simplified models to understand the asymptotics

Equation for w

$$\partial_t w + \gamma_{pol}(p(t)) \partial_x w = \\ + \gamma_{depol}(p(t) < p_h) \left(- \int_0^x k(\tilde{x}, x) w(t, x) d\tilde{x} + \int_x^\infty k(x, \tilde{x}) w(t, \tilde{x}) d\tilde{x} \right)$$

Equation for p

$$\frac{d}{dt} p = -\gamma_{pol}(p(t)) \int_0^\infty \int_0^x w(t, z, x) dz dx + \kappa q$$

Equation for q

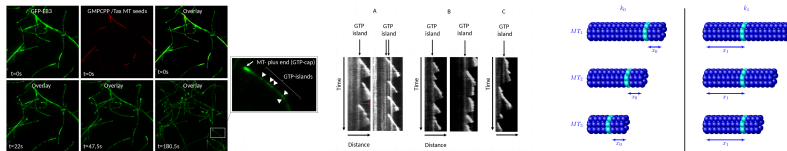
$$\frac{d}{dt} q = \gamma_{depol}(p(t) < p_h) \int_0^\infty \int_0^x (x - \tilde{x}) k(\tilde{x}, x) w(t, x) d\tilde{x} dx - \kappa q$$

Simplified models to understand the asymptotics

The fragmentation terms

$$-\gamma_{depol} \int_0^x k(\tilde{x}, x) w(t, x) d\tilde{x} + \gamma_{depol} \int_x^\infty k(x, \tilde{x}) w(t, \tilde{x}) d\tilde{x}$$

with $k(\tilde{x}, x)$ the probability for a MT of size x to reach the size $\tilde{x} < x$
 Two types of kernel identified from the experiments



- $k_0(y, x) = G(y - x)$: depolymerisation length is almost fixed
- $k_1(y, x) = G(x)$: size of the MTs after a depolymerisation is almost fixed

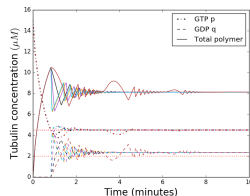
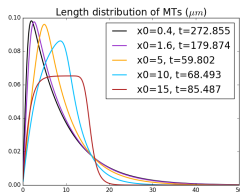
here $G(z) = \frac{1}{\sigma\sqrt{2\pi}} \exp \frac{-(z-x_0)^2}{2\sigma^2}$, $x_0 > 0$, $\sigma > 0$

Properties

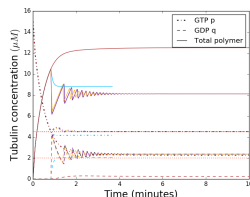
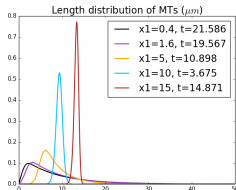
↪ Reduction to ODE system is impossible

Simplified models to understand the asymptotics

Asymptotics for the kernel k_0



Asymptotics for the kernel k_1



↪ Rapid convergence at the macroscopic level, slow convergence of the distribution profil

Simplified models to understand the asymptotics

The most simplified model

Equation for w

$$\begin{aligned} \partial_t w + \gamma_{pol}(p(t)) \partial_x w &= \psi(x) \mathcal{N}(p(t)) \\ + \underbrace{\beta(p(t))}_{\sim \gamma_{depol}(p(t) < p_h)} &\left(- \int_0^x k(x, \tilde{x}) w(t, x) d\tilde{x} + \int_x^\infty k(\tilde{x}, x) w(t, \tilde{x}) d\tilde{x} \right) \end{aligned}$$

Equation for p

$$\begin{aligned} \frac{d}{dt} p &= -\gamma_{pol}(p(t)) \int_0^\infty \int_0^x w(t, z, x) dz dx - \bar{\mathcal{N}}(p(t)) \\ &+ \beta(p(t)) \int_0^\infty \int_0^x (x - \tilde{x}) k(x, \tilde{x}) w(t, x) d\tilde{x} dx \end{aligned}$$

↪ **Wellposedness of the system** with conservation properties

$$\int_0^\infty x w(t, x) dx + p(t) = \int_0^\infty x w(0, x) dx + p(0) := M_1^0$$

↪ **Numerical observations** $p(t) \rightarrow p^\infty$, $w(t, \cdot) \rightarrow W$ for large time **FH**,
Tournus, White, 2017

↪ **Existence and uniqueness of the asymptotic profile** (W, p^∞)

↪ **Convergence** **Work in progress with M. Potomkin, S. D. Ryan, M. Tournus**

Conclusion

- Transport equations with eventually fragmentation terms are a powerful tool to model biological issues.
- Advertisement
 - ↪ **Summer school on domain decomposition method for optimal control problems** September 5-9, 2022
 - Part of the chair Jean Morlet hold by Martin Gander



Thank you for your attention !

$$\lambda_0 N(a) + N'(a) = -B(a)N(a), \quad N(0) = 2 \int_0^\infty B(a)N(a) da \quad (*)$$

We have $N(a) = N(0)e^{-\int_0^a (\lambda_0 + B(s)) ds}$ with

$$N(0) = 2 \int_0^\infty B(a)N(a) da = 2N(0) \int_0^\infty B(a)e^{-\lambda_0 a} da$$

\rightsquigarrow Existence of $N \Leftrightarrow$ Existence of λ_0 such that $F(\lambda_0) = 1$ where

$$F(\lambda) = 2 \int_0^\infty B(a)e^{-\int_0^a (\lambda + B(s)) ds} da.$$

If $B \in L^\infty$ with $\int_0^\infty B = +\infty$, F is a decreasing function and

$$\lim_{\lambda \rightarrow 0} F(\lambda) = 2 \text{ and } \lim_{\lambda \rightarrow \infty} F(\lambda) = 0$$

Therefore, there exists a unique (λ_0, N) solution of $(*)$ such that $\int_0^\infty N(a) da = 1$.

The parameter λ_0 is called the **the Malthus parameter**.

$$\lambda_0 N(a) + N'(a) = -B(a)N(a), \quad N(0) = 2 \int_0^\infty B(a)N(a) da \quad (*)$$

Adjoint problem

$$\lambda_0 \phi(a) - \phi'(a) + B(a)\phi(a) = 2\phi(0)B(a) \quad (**)$$

To find the adjoint problem, multiply $(*)$ by ϕ and integrate

$$0 = \int_0^\infty (\lambda_0 N + N' + BN)\phi da = \int_0^\infty N(\lambda_0 \phi - \phi' + B)\phi da - \phi(0)N(0) = \int_0^\infty N(a)(\lambda_0 \phi - \phi' + B - 2B\phi(0)) da$$

The solution of $(**)$ is given by

$$\phi(a) = 2\phi(0) \int_a^\infty B(a') e^{-\int_a^{a'} (\lambda + B(s)) ds} da' \quad \text{with } \phi(0) \text{ such that } \int_0^\infty N\phi = 1.$$

■ Conservation properties

$$\Psi(t) = \int_0^\infty \phi(a) e^{-\lambda_0 t} \rho(t, a) da = \int_0^\infty \phi(a) \rho^0(a) da := \bar{\rho}^0$$

Indeed,

$$\begin{aligned} \frac{d}{dt} \Psi(t) &= \int_0^\infty \phi e^{-\lambda_0 t} (-\lambda_0 \rho + \partial_t \rho) da = \int_0^\infty \phi e^{-\lambda_0 t} (-(\lambda_0 + B)\rho - \partial_a \rho) da \\ &= e^{-\lambda_0 t} \left(\int_0^\infty \rho (-(\lambda_0 + B)\psi + \phi') da - \rho(t, 0)\phi(0) \right) \\ &= e^{-\lambda_0 t} \phi(0) \left(\int_0^\infty 2\rho B - \rho(t, 0) \right) = 0 \end{aligned}$$

- Let $m(t, a) = e^{-\lambda_0 t} \frac{\rho(t, a)}{N(a)}$, then for all convex function \mathcal{H}

$$\frac{d}{dt} \int_0^\infty \phi(a) N(a) \mathcal{H}(m(t, a)) da := \Delta \leq 0$$

Indeed,

$$\partial_t m + \partial_a m = e^{-\lambda_0 t} \frac{(-\lambda_0 \rho + \partial_t \rho + \partial_a \rho) N - N' \rho}{N^2} = e^{-\lambda_0 t} \frac{(-\lambda_0 N - B N - N') \rho}{N^2} = 0$$

with

$$m(t, 0) = \int_0^\infty m(t, a) d\mu(a), \quad d\mu(a) = 2 \frac{B(a) N(a)}{N(0)}$$

$$\frac{d}{da} \phi N = -2\phi(0) B(a) N(a)$$

Thus, for $\bar{m}(t, a) = \phi(a) N(a) \mathcal{H}(m(t, a))$, we have

$$\partial_t \bar{m}(t, a) + \partial_a \bar{m}(t, a) = -\chi(a) \bar{m}(t, a) \quad \text{with} \quad \chi(a) = 2\phi(0) \frac{B(a)}{\phi(a)}$$

and thus

$$\begin{aligned} \frac{d}{dt} \int_0^\infty \bar{m}(t, a) da &= \bar{m}(t, 0) - \int_0^\infty \chi(a) \bar{m}(t, a) da \\ &= \phi(0) N(0) \mathcal{H}(m(t, 0)) - \int_0^\infty 2\phi(0) B(a) N(a) \mathcal{H}(m(t, a)) da \\ &= \phi(0) N(0) \left(\mathcal{H} \left(\int_0^\infty m(t, a) d\mu(a) \right) - \int_0^\infty \mathcal{H}(m(t, a)) d\mu(a) \right) \leq 0 \end{aligned}$$

- If $\exists \mu_0 > 0$ such that $\forall a \in \mathbb{R}^+$, $\chi(a) := \frac{2\phi(0)B(a)}{\phi(a)} \geq \mu_0$ for a $\mathcal{H}(m) = |m - \bar{\rho}|$ we have $\Delta \leq -\mu_0 \int \phi N \mathcal{H}(m)$.
Recall that

$$m(t, a) = e^{-\lambda t} \frac{\rho(t, a)}{N(a)} \text{ and } \bar{\rho}^0 = \int_0^\infty e^{-\lambda t} \rho(t, a) \phi(a) da$$

so if $\tilde{m}(t, a) = \phi(a)N(a)(m(t, a) - \bar{\rho}^0)$ we have $\int_0^\infty \tilde{m}(t, a) da = 0$
Now,

$$\begin{aligned} \tilde{m}(t, 0) &= \phi(0)(e^{-\lambda t} \rho(t, 0) - N(0)\bar{\rho}^0) \\ &= 2\phi(0) \left(\int_0^\infty B(a)e^{-\lambda t} \rho(t, a) - \bar{\rho}^0 \int_0^\infty B(a)N(a) da \right) \\ &= 2\phi(0) \left(\int_0^\infty B(a)N(a)(m(t, a) - \bar{\rho}^0) \right) = \int_0^\infty \chi(a)\tilde{m}(t, a) da \end{aligned}$$

The entropy estimate then gives ($\tilde{m} = |\tilde{m}|$)

$$\begin{aligned} \frac{d}{dt} \int_0^\infty |\tilde{m}(t, a)| da &= |\tilde{m}(t, 0)| - \int_0^\infty \chi |\tilde{m}| = \left| \int_0^\infty \chi \tilde{m} \right| - \int_0^\infty \chi |\tilde{m}| \\ &= \left| \int_0^\infty (\chi - \mu_0) \tilde{m} \right| - \int_0^\infty \chi |\tilde{m}| \leq -\mu_0 \int_0^\infty |\tilde{m}| \end{aligned}$$

Properties of the fragmentation kernels

◀ Return

$$k(x, y) = B(x)\kappa(x, y) \text{ with } \int \kappa(x, y) dy = 1, \kappa(x, y) = 0 \text{ if } y > x$$

The kernel $k_0(x, y) = G(x - y)(x > y)$ with $\int_0^\infty G < +\infty$

$$B(x) = \int_0^x G(x - y) dy = \int_0^x G(y) dy, \int_x^\infty B(y)\kappa(y, x) dy = \int_x^\infty G(y - x) dy = \int_0^\infty G(z) dz < \infty$$

The kernel $k_1(x, y) = G(y)(x > y)$ with $\int_0^\infty G < +\infty$

$$B(x) = \int_0^x G(y) dy, \int_x^\infty B(y)\kappa(y, x) dy = \int_x^\infty G(y) dy < \infty$$

In both cases, G is a non negative function with

$$B(x) \leq B_M \text{ if } \int_0^\infty G(y) dy < +\infty$$

B is an increasing function such that $B(0) = 0$,

$$\exists x_- > 0 \text{ such that } B(x) \geq B_m > 0 \forall x > x_- \text{ if } \int_0^\infty G(y) dy \neq 0$$