# Polymerization Models for Prion

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BANG day, 22 september 2009



## Outline

- First Model
- The PMCA
  - Principle and Model
  - Two Optimization Problems
- New model
  - Presentation
  - Numerical Study

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# Biological Model

The pathogenic agent is an abnormal protein present as aggregates :

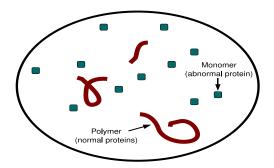


Fig.: Infected cell.



# Equations

Model of Greer *et al.* (2006), Calvez *et al.* (2008), Doumic *et al.* (2009) :

$$\begin{cases} \frac{dV(t)}{dt} = \lambda - V(t) \left[ \gamma + \int_0^\infty \tau(x) u(x, t) \, dx \right], \\ \frac{\partial}{\partial t} u(x, t) = -V(t) \frac{\partial}{\partial x} \left( \tau(x) u(x, t) \right) - \left[ \mu(x) + \beta(x) \right] u(x, t) \\ + 2 \int_x^\infty \beta(y) \kappa(x, y) \, u(y, t) \, dy, \\ u(0, t) = 0, \\ u(x, 0) = u_0(x). \end{cases}$$

V(t): quantity of monomers at time t, u(x,t): quantity of polymers of size x at time t.

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- Incubation during which the polymerization is promoted
- Sonication which increases a lot the fragmentation

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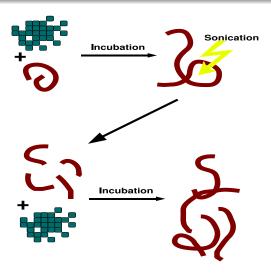


Fig.: PMCA principle.



We model the sonication multiplying the fragmentation  $\beta(x)$  by a parameter  $\alpha(t)$ :

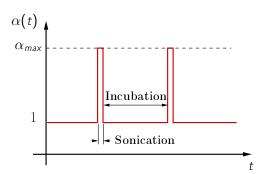
$$\frac{\partial}{\partial t}u(x,t) = -\frac{\partial}{\partial x}(\tau(x)u(x,t)) - \beta(x)u(x,t) + 2\int_{x}^{\infty}\beta(y)\kappa(x,y)u(y,t)\,dy,$$

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#### $\alpha$ constant

(with V. Calvez and M. Doumic)

For a constant sonication  $\alpha(t) \equiv \alpha$ , there exists a principal eigenvalue  $\lambda_{\alpha}$  for the equation :

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### Theorem : Existence of $\alpha_{opt}$ (V. Calvez, M. Doumic, P. G.)

If the polymerization rate au satisfies

- $\tau(0) = 0$  and  $\tau$  is convex in a neighborhood of 0,
- $\tau(x) = o(x)$  when x tends to  $+\infty$ ,

Then there exists a  $\alpha_{opt}$  such that  $\lambda_{\alpha} \leq \lambda_{\alpha_{opt}}$  for all  $\alpha > 0$ .



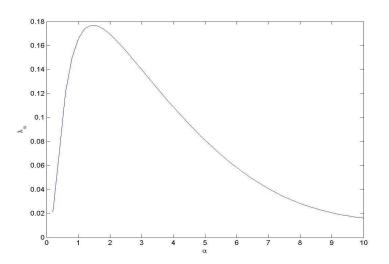


Fig.: Existence of an optimal  $\alpha$ .

For a given outlook T, we search a control  $\alpha:[0,T]\to [1,\alpha_{max}]$  which maximizes the mass obtained at the final time T. For this we come back to a discrete in size model :

$$\begin{cases} \frac{du_i}{dt} = -\tau_i(u_i - u_{i-1}) - \alpha(t)\beta_i u_i + 2\sum_{j=i+1}^n \alpha(t)\beta_j \kappa_{i,j} u_j, \\ u_i(0) = u_i^0, \end{cases}$$

for all  $1 \le i \le n$ . Then we search to optimize  $\sum_{i=1}^{n} iu_i(T)$ .

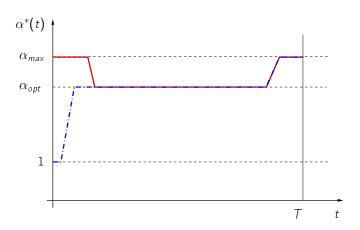


Fig.: Profils of  $\alpha^*(t)$  when  $1 < \alpha_{opt} < \alpha_{max}$ .

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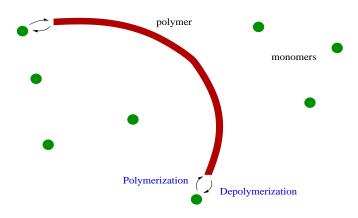


Fig.: Polymerization-depolymerization

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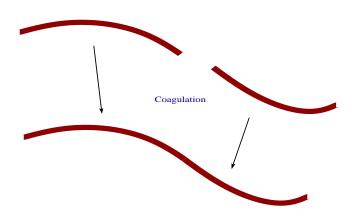


Fig.: Coagulation of two polymers

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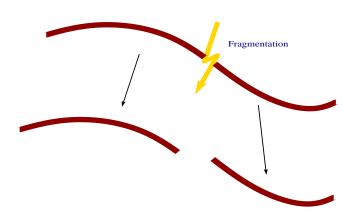


Fig.: Fragmentation of a polymer

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# Simulations (comparison WENO/Upwind)

with L.M. Tine

