Stochastic coagulation-fragmentation models for the study of protein aggregation phenomena

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Amyloid diseases and Becker-Döring model

Numerical results

Coarse-graining

Open problems (for me)

Outline

Amyloid diseases and Becker-Döring model

Numerical results

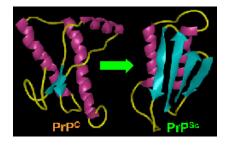
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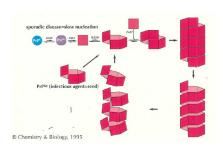
Motivation Numerical results Coarse-graining Open problem

Protein accumulation in amyloid by nucleation-polymerization

Misfolding



Prusiner model for prion

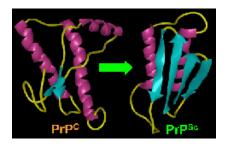


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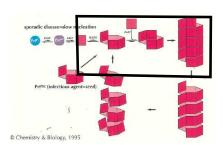
Mechanism of Prion Propagation: Amyloid Growth Occurs by Monomer Addition Motivation Numerical results Coarse-graining Open problem

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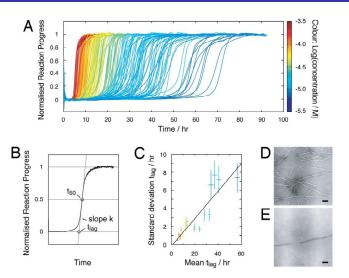
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Mechanism of Prion Propagation:

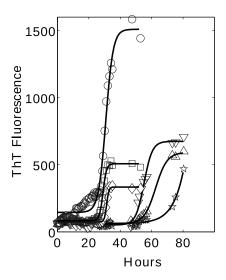
Amyloid Growth Occurs by Monomer Addition

Times series of in-vitro spontaneous polymerization

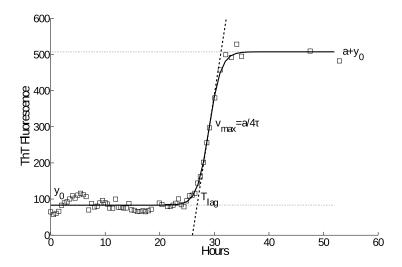


Xue et al. PNAS (2008) Eugene el al. hal-01205549 (2015)

Times series of *in-vitro* spontaneous polymerization

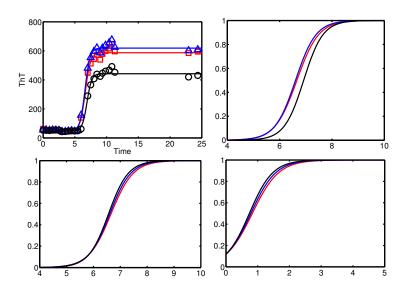


Quantification of experiment

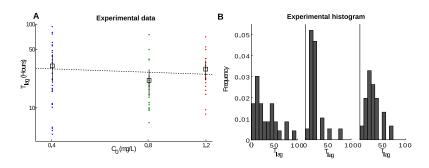


Motivation Numerical results Coarse-graining Open problem

Quantification of experiment



Statistics of lag time



Reversible one-step agregation

$$C_i + C_1 \xrightarrow[q_{i+1}]{p_i} C_{i+1} \qquad (1)$$







The nucleation time is given by the following First Passage Time,

$$T^{N,M} := \inf\{t \ge 0 : C_N(t) = 1 \mid C_i(t=0) = M\delta_{i=1}\}.$$
 (2)

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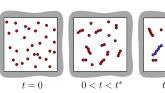
$$T^{N,M} := \inf\{t \ge 0 : C_N(t) = 1 \mid C_i(t=0) = M\delta_{i=1}\}.$$
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What are the dependencies of the nucleation time with respect to the model parameters?

total mass : M; nucleus size : N aggregation rates : p_i , $i \ge 1$ fragmentation rates : q_i , $i \ge 2$

Reversible one-step agregation

$$C_i + C_1 \xrightarrow{\stackrel{p_i}{\longleftarrow}} C_{i+1} \qquad (1)$$



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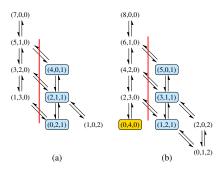
• What is the nucleation time for very large initial quantity M and nucleus size N?

$$\lim_{M,N\to\infty}T^{N,M}$$

In data, $M \approx 10^{10} - 10^{15}$, Size of (observed) polymers $\approx 10^3 - 10^6$, N = ?

Reversible one-step agregation

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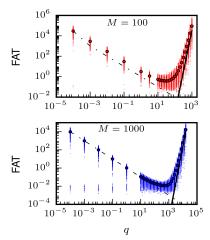
Amyloid diseases and Becker-Döring mode

Numerical results

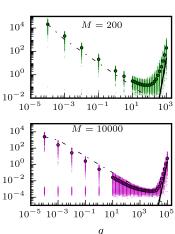
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Non-monotonous w.r.t reaction rate



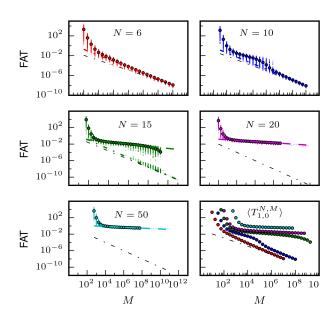
Bimodal for 'small' fragmentation rate



N=10, $p_1 = 0.5$, $p_k = 1$ and $q_k \equiv q$ for $k \ge 2$,.

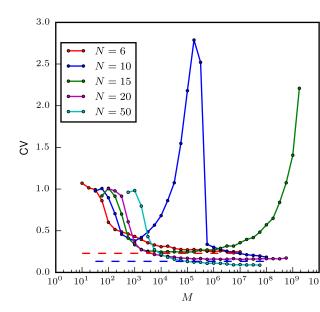
'Weak' dependency w.r.t. total monomer number M

 $p_1 = 0.5$, $p_k = 1$ and $q_k \equiv 100$ for $k \geqslant 2$,.

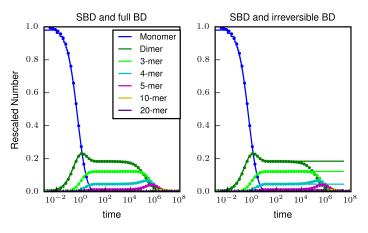


- Normalized variance nonmonotonous.
- Normalized variance non zero pour M → ∞.

 $p_1 = 0.5$, $p_k = 1$ and $q_k \equiv 100$ for $k \geqslant 2$,.



- For $M \to \infty$: deterministic trajectory.
- Metastable behavior : 'pure-aggregation'.
- Medium-large polymer formed only after a longer time



 $p_k = 1 \text{ and } q_k \equiv 1 \text{ for } k \geqslant 2, \ M = 10^5 \text{ (we plot } M^{-1}C_k(tM^{-1})).$ Detailed Analysis: cf [Wattis, J. Phys. A: Math., 35 (2002)]

Outline

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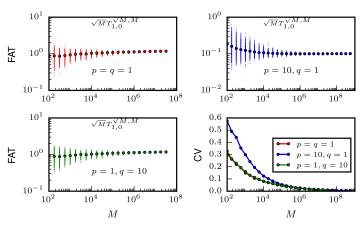
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Large nucleus $N \sim \sqrt{M}$

First case (p(0)m > q(0)): Convergence towards a deterministic value.

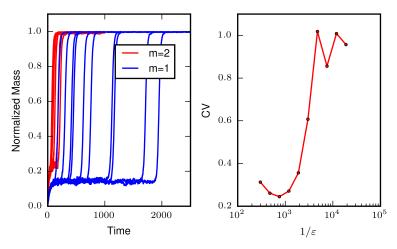


case A

case B

Large nucleus $N \sim \sqrt{M}$

• Second case (p(0)M < q(0)): Exponentially large time and 'translated' trajectory.



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A much simpler version of this model consider that a **single** aggregate may be formed at a time :

$$k \xrightarrow[q_{k+1}]{p_k(m-k\varepsilon)} k+1,$$

which converges (with time rescaling) to

$$\frac{dx}{dt} = p(x)(m-x) - q(x)$$

FPT are given by

$$\mathbb{E}\big[T_{1,0}^N\big] = \sum_{i=1}^{N-1} \sum_{j=1}^i \frac{\prod_{k=j+1}^i q_k}{\prod_{k=j}^i p_k(m-\varepsilon k)} \,.$$

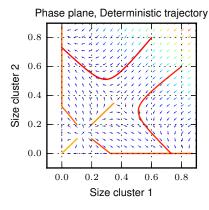
$$\int_0^1 \int_0^y \cdot \exp\left[\varepsilon^{-1} \int_z^y \ln\left(\frac{q(x)}{p(x)(m-x)}\right) dx\right].$$

Can we perform similar calculations with n clusters?

$$\begin{array}{ccc} (k_0,k_1) & \xrightarrow{\frac{\rho_{k_0}(m-(k_0+k_1)\varepsilon)}{q_{k_0+1}}} & (k_0+1,k_1)\,, \\ (k_0,k_1) & \xrightarrow{\frac{\rho_{k_1}(m-(k_0+k_1)\varepsilon)}{q_{k_1+1}}} & (k_0,k_1+1)\,, \end{array}$$

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$$\frac{dy}{dt} = p(y)(m-x-y) - q(y)$$

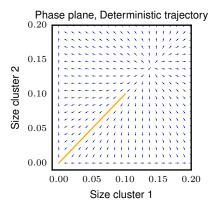


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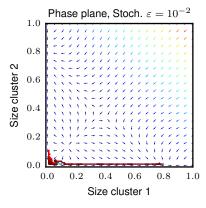


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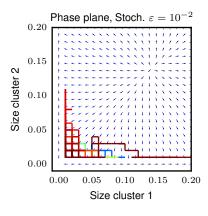


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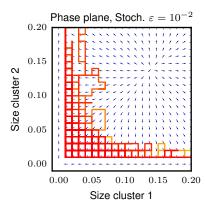


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- Extension to spatial models (diffusion)?
- ▶ Data fitting with $10^{10} 10^{15}$ proteins ?

Toy model with time-scale separation

$$\begin{cases} X & \xrightarrow{\gamma} & X^*, & 1.2 \\ X^* + X^* & \xrightarrow{\varepsilon \alpha} & 2Y, & \frac{1}{2} & 0.6 \\ X + Y & \xrightarrow{\varepsilon \beta} & 2Y, & \frac{1}{2} & 0.4 \\ X + Y & \xrightarrow{\varepsilon \beta} & 2Y, & \frac{1}{2} & 0.4 \\ X + Y & \xrightarrow{\varepsilon \beta} & 2Y, & \frac{1}{2} & 0.4 \\ X + Y & \xrightarrow{\varepsilon \beta} & 2Y, & \frac{1}{2} & 0.4 \\ X + Y & \xrightarrow{\varepsilon \beta} & 2Y, & \frac{1}{2} & 0.4 \\ X + Y & \xrightarrow{\varepsilon \beta} & 2Y, & \frac{1}{2} & 0.4 \\ X + Y & \xrightarrow{\varepsilon \beta} & 2Y, & \frac{1}{2} & 0.4 \\ X + Y & \xrightarrow{\varepsilon \beta} & 2Y, & \frac{1}{2} & 0.4 \\ X + Y & \xrightarrow{\varepsilon \beta} & 2Y, & \frac{1}{2} & 0.4 \\ X + Y & \xrightarrow{\varepsilon \beta} & 2Y, & \frac{1}{2} & 0.4 \\ X + Y & \xrightarrow{\varepsilon \beta} & 2Y, & \frac{1}{2} & 0.4 \\ X + Y & \xrightarrow{\varepsilon \beta} & 2Y, & \frac{1}{2} & 0.4 \\ X + Y & \xrightarrow{\varepsilon \beta} & 2Y, & \frac{1}{2} & 0.4 \\ X + Y & \xrightarrow{\varepsilon \beta} & 2Y, & \frac{1}{2} & 0.4 \\ X + Y & \xrightarrow{\varepsilon \beta} & 2Y, & \frac{1}{2} & 0.4 \\ X + Y & \xrightarrow{\varepsilon \beta} & 2Y, & \frac{1}{2} & 0.4 \\ X + Y & \xrightarrow{\varepsilon \beta} & 2Y, & \frac{1}{2} & 0.4 \\ X + Y & \xrightarrow{\varepsilon \beta} & 2Y, & \frac{1}{2} & 0.4 \\ X + Y & \xrightarrow{\varepsilon \beta} & 2Y, & \frac{1}{2} & 0.4 \\ X + Y & \xrightarrow{\varepsilon \beta} & 2Y, & \frac{1}{2} & 0.4 \\ X + Y & \xrightarrow{\varepsilon \beta} & 2Y, & \frac{1}{2} & 0.4 \\ X + Y & \xrightarrow{\varepsilon \beta} & 2Y, & \frac{1}{2} & 0.4 \\ X + Y & \xrightarrow{\varepsilon \beta} & \frac{1}{2} & 0$$

with initial condition

$$Z^{\varepsilon}(0) = (|x_0/\varepsilon|, 0, 0)$$
.

