Stochastic Model of Prion's Disease: Nucleation-Polymerisation Model Part II

YVINEC Romain

April, 22nd, 2010

YVINEC Romain Stochastic Model of Prion's Disease: Nucleation-Polymerisatio

Experimental heterogeneity

Heterogeneity of the structure Heterogeneity of the dynamics Seeding experiments Model proposed

Modelling each reaction

Precise reactions step and parameters Deterministic rate equations Stochastic algorithm Reproducing the experimental variability

Further...

4 B K 4 B K

Heterogeneity of the structure Heterogeneity of the dynamics Seeding experiments Model proposed

Experiments



Nucleation-Polymerisation were performed by an initially population of recombinant Prion protein PrP^{c} upon different conditions (pH, temperature, concentration, solution) Here are represented several experiment performed upon two different conditions. The *ThT* fluroescence is used as a measurable quantity which is correlated to the global mass of the polymer during experiments were performed.

YVINEC Romain Stochastic Model of I

Stochastic Model of Prion's Disease: Nucleation-Polymerisation

Heterogeneity of the structure Heterogeneity of the dynamics Seeding experiments Model proposed

Simoid shape



Four quantities appear to be characteristic of the dynamics.

- F_{max} is the maximal fluorescent value reached.
- $\frac{1}{\tau}$ is the maximal growth rate during the polymerization, which is achieved at $t = T_i$.
- T_{lag} is the waiting time before the polymerization starts truly

All of these quantities can be measured on each experimental curves.

(ロ) (同) (E) (E) (E)

Heterogeneity of the structure Heterogeneity of the dynamics Seeding experiments Model proposed

Relation between Fmax and Ti and τ



(a) Relation between *Fmax* and *Ti*

(b) Relation between *Fmax* and τ

イロト イポト イヨト イヨト

The maximum value does not correlate with the remain quantity of monomers at the end of the experiment, and does not correlate neither with the lag time and the maximum growth rate This proves that no "off-pathway"

Heterogeneity of the structure Heterogeneity of the dynamics Seeding experiments Model proposed

Observed heterogeneity of the structure



This lead to conjecture to an heterogeneity of the final structure itself of the polymer. The electron microscopy analysis give some clue to this heterogeneity and the relation between the size of polymer with its fluorescence response to ThT confirm it.

Heterogeneity of the structure Heterogeneity of the dynamics Seeding experiments Model proposed

Variability in Lag Time and Growth rate



These result can also be interpreted as heterogeneity of nucleation and subsequent polymer formation. It can be suggested that different nucleus generated structurally different polymers each exhibiting specific polymerisation dynamics. $\langle \Box \rangle + \langle \overline{\Box} \rangle + \langle \overline{\Box}$

Heterogeneity of the structure Heterogeneity of the dynamics Seeding experiments Model proposed

Seeding



Seeding experiments explain the infectiousity of the prior disease. But the nucleation is predominant in the formation of prior amyloids. The conformational change has to take place during the nucleation process more than during the polymerisation process.

イロト イポト イヨト イヨト

æ

Heterogeneity of the structure Heterogeneity of the dynamics Seeding experiments Model proposed

Model



YVINEC Romain Stochastic Model of Prion's Disease: Nucleation-Polymerisatio

Precise reactions step and parameters Deterministic rate equations Stochastic algorithm Reproducing the experimental variability

Variables

• M =: (inactive) normal monomer

YVINEC Romain Stochastic Model of Prion's Disease: Nucleation-Polymerisatio

・ロト ・回ト ・ヨト ・ヨト

Precise reactions step and parameters Deterministic rate equations Stochastic algorithm Reproducing the experimental variability

Variables

- M =: (inactive) normal monomer
- $f_1 =:$ (active) abnormal monomer (misfolded protein)

・ロト ・回ト ・ヨト ・ヨト

Precise reactions step and parameters Deterministic rate equations Stochastic algorithm Reproducing the experimental variability

Variables

- ► *M* =: (inactive) normal monomer
- $f_1 =:$ (active) abnormal monomer (misfolded protein)
- f_i , i = 2..n 1, =: oligomer (unstable species)

Precise reactions step and parameters Deterministic rate equations Stochastic algorithm Reproducing the experimental variability

Variables

- ► *M* =: (inactive) normal monomer
- $f_1 =:$ (active) abnormal monomer (misfolded protein)
- ▶ f_i , i = 2..n 1, =: oligomer (unstable species)
- u_i , $i \ge n =:$ polymer (stable species)

(ロ) (同) (E) (E) (E)

Precise reactions step and parameters Deterministic rate equations Stochastic algorithm Reproducing the experimental variability

Variables

- ► *M* =: (inactive) normal monomer
- $f_1 =:$ (active) abnormal monomer (misfolded protein)
- f_i , i = 2..n 1, =: oligomer (unstable species)

•
$$u_i, i \ge n =:$$
 polymer (stable species)

n: is the size of the nucleus.

ヘロン 人間 とくほど くほとう

Precise reactions step and parameters Deterministic rate equations Stochastic algorithm Reproducing the experimental variability

Nucleation

$$M \quad \stackrel{1}{\underset{c_{0}}{\longleftarrow}} \quad f_{1} \text{ (spontaneous conformation)} \qquad (1)$$

$$f_{1} + f_{1} \quad \stackrel{c_{f}}{\underset{c_{f}}{\longleftarrow}} \quad f_{2} \text{ (dimerization)} \qquad (2)$$

$$\vdots$$

$$f_{k-1} + f_{1} \quad \stackrel{c_{f}}{\underset{c_{f}}{\longleftarrow}} \quad f_{k} \text{ ((k)-mer formation)} \qquad (3)$$

$$\vdots$$

$$f_{n-1} + f_{1} \quad \stackrel{c_{f}}{\longrightarrow} \quad u_{n} \text{ (nucleus formation)} \qquad (4)$$

YVINEC Romain Stochastic Model of Prion's Disease: Nucleation-Polymerisatio

Precise reactions step and parameters Deterministic rate equations Stochastic algorithm Reproducing the experimental variability

Polymerization

$$\begin{array}{rcl} u_{i}+f_{1} & \xrightarrow{k_{pf}} & u_{i+1}, \ (i \geq n), \ (\text{elongation}) & (5) \\ u_{i} & \xrightarrow{\frac{k_{pf}}{d_{2}}} & u_{i-1}+f_{1}, \ (i \geq n), \ (\text{shortening}) & (6) \\ u_{i} & \xrightarrow{k_{b}} & u_{i-j}+u_{j}, \ (i \geq n), \ (\text{polymer break}) & (7) \\ \text{and} \ u_{k} & \xrightarrow{\infty} & k * f_{1}, \ \text{if } k \ \leq n-1, \ (\text{oligomer instability}) \end{array}$$

・ロト ・回ト ・ヨト ・ヨト

Precise reactions step and parameters Deterministic rate equations Stochastic algorithm Reproducing the experimental variability

Parameters

 c₀: equilibrium constant between inactive monomers M and active monomers f₁. We suppose (fast variable)

$$M = c_0 * f_1$$

・ロン ・回と ・ヨン ・ヨン

Precise reactions step and parameters Deterministic rate equations Stochastic algorithm Reproducing the experimental variability

Parameters

 c₀: equilibrium constant between inactive monomers M and active monomers f₁. We suppose (fast variable)

$$M=c_0*f_1$$

c_f: elongation constant in nucleation steps

Precise reactions step and parameters Deterministic rate equations Stochastic algorithm Reproducing the experimental variability

Parameters

▶ c₀: equilibrium constant between inactive monomers M and active monomers f₁. We suppose (fast variable)

$$M = c_0 * f_1$$

- c_f: elongation constant in nucleation steps
- ► *d*₁: equilibrium constant in nucleation steps

Precise reactions step and parameters Deterministic rate equations Stochastic algorithm Reproducing the experimental variability

Parameters

▶ c₀: equilibrium constant between inactive monomers M and active monomers f₁. We suppose (fast variable)

$$M = c_0 * f_1$$

- c_f: elongation constant in nucleation steps
- d₁: equilibrium constant in nucleation steps
- k_{pf}: elongation constant in polymerization steps

Precise reactions step and parameters Deterministic rate equations Stochastic algorithm Reproducing the experimental variability

Parameters

▶ c₀: equilibrium constant between inactive monomers M and active monomers f₁. We suppose (fast variable)

$$M=c_0*f_1$$

- c_f: elongation constant in nucleation steps
- d₁: equilibrium constant in nucleation steps
- k_{pf}: elongation constant in polymerization steps
- ▶ k_b: fragmentation rate (independent of the length of the filaments)

・ロン ・回と ・ヨン ・ヨン

Precise reactions step and parameters Deterministic rate equations Stochastic algorithm Reproducing the experimental variability

Parameters

▶ c₀: equilibrium constant between inactive monomers M and active monomers f₁. We suppose (fast variable)

$$M=c_0*f_1$$

- c_f: elongation constant in nucleation steps
- d₁: equilibrium constant in nucleation steps
- k_{pf}: elongation constant in polymerization steps
- ▶ k_b: fragmentation rate (independent of the length of the filaments)
- M(0): initial quantity of monomers.

・ロト ・回ト ・ヨト ・ヨト

Precise reactions step and parameters Deterministic rate equations Stochastic algorithm Reproducing the experimental variability

Parameters

▶ c₀: equilibrium constant between inactive monomers M and active monomers f₁. We suppose (fast variable)

$$M = c_0 * f_1$$

- c_f: elongation constant in nucleation steps
- d₁: equilibrium constant in nucleation steps
- *k_{pf}*: elongation constant in polymerization steps
- ▶ k_b: fragmentation rate (independent of the length of the filaments)
- M(0): initial quantity of monomers.
- n: nucleus size.

ヘロン 人間 とくほど くほとう

 Outline
 Precise reactions step and parameters

 Experimental heterogeneity
 Deterministic rate equations

 Modelling each reaction
 Stochastic algorithm

 Further...
 Reproducing the experimental variability



イロン イヨン イヨン

Precise reactions step and parameters Deterministic rate equations Stochastic algorithm Reproducing the experimental variability

General Behavior



Nucleation-Polymerisation in a deterministic simulation

YVINEC Romain Stochastic Model of Prion's Disease: Nucleation-Polymerisatio

<ロ> <部> <き> <き>

Precise reactions step and parameters Deterministic rate equations Stochastic algorithm Reproducing the experimental variability

General Behavior: nucleation dynamics



イロト イロト イヨト イヨト

Precise reactions step and parameters Deterministic rate equations Stochastic algorithm Reproducing the experimental variability

Influences of parameters

► c₀: Changes drastically the Lag time and the Growth rate. If c₀ increases, less active monomers will be available, resulting in increasing the time of nucleation and slowing down also the polymerization. It has also by the same way an influence on the final equilibrium.

Precise reactions step and parameters Deterministic rate equations Stochastic algorithm Reproducing the experimental variability

Influences of parameters

- ► c₀: Changes drastically the Lag time and the Growth rate. If c₀ increases, less active monomers will be available, resulting in increasing the time of nucleation and slowing down also the polymerization. It has also by the same way an influence on the final equilibrium.
- ▶ *c_f*: Only changes the Lag time.

Precise reactions step and parameters Deterministic rate equations Stochastic algorithm Reproducing the experimental variability

Influences of parameters

- ► c₀: Changes drastically the Lag time and the Growth rate. If c₀ increases, less active monomers will be available, resulting in increasing the time of nucleation and slowing down also the polymerization. It has also by the same way an influence on the final equilibrium.
- c_f: Only changes the Lag time.
- d₁: Influences the nucleation time. Can also limit the polymerisation if the oligomers don't disappears quickly enough (ie if d1 high, which is unlikely to be).

(ロ) (同) (E) (E) (E)

Precise reactions step and parameters Deterministic rate equations Stochastic algorithm Reproducing the experimental variability

Influences of parameters

▶ k_{pf} and k_b: Influence both polymerization and nucleation in the same way. Change also the final equilibrium.



Precise reactions step and parameters Deterministic rate equations Stochastic algorithm Reproducing the experimental variability

Influences of parameters

► M(0) and n: M(0) reduces the lag time, while n increases it. Up to a constant, Tlag follows the same relation with M(0), whatever the nucleus size is. In agreement with experimental date. We have

$$log(T_{lag}) \simeq -1.07 * log(M(0)) \tag{8}$$



YVINEC Romain

Stochastic Model of Prion's Disease: Nucleation-Polymerisatio

Precise reactions step and parameters Deterministic rate equations Stochastic algorithm Reproducing the experimental variability

Influences of parameters

► M(0) and n: M(0) speeds up the polymerization, while n slightly decreases it. This contradicts experiments.



Precise reactions step and parameters Deterministic rate equations Stochastic algorithm Reproducing the experimental variability

Some key factor: general scaling laws

• $M_{tot} * k_{pf} \gg k_b$ guarantee an initially growing polymer.

・ロト ・回ト ・ヨト ・ヨト

Precise reactions step and parameters Deterministic rate equations Stochastic algorithm Reproducing the experimental variability

Some key factor: general scaling laws

- $M_{tot} * k_{pf} \gg k_b$ guarantee an initially growing polymer.
- Tlag versus V: For large M(0), they are clearly inversely correlated. This contradicts experiments.



() < </p>

Precise reactions step and parameters Deterministic rate equations Stochastic algorithm Reproducing the experimental variability

experimental variability



Figure: Several polymerisation experiments

YVINEC Romain Stochastic Model of Prion's Disease: Nucleation-Polymerisatio

Precise reactions step and parameters Deterministic rate equations Stochastic algorithm Reproducing the experimental variability

Stochastic algorithm - global dynamic



・ロン ・回と ・ヨン・

Precise reactions step and parameters Deterministic rate equations Stochastic algorithm Reproducing the experimental variability

Break events or Secondary nucleation pathway

 One single nucleation event then elongation and break events can explain the sigmoid form



Precise reactions step and parameters Deterministic rate equations Stochastic algorithm Reproducing the experimental variability

Stochasticity in the oligomer

 One can have many (small) fluctuations in the oligomer species, before reaching nucleation



YVINEC Romain Stochastic Model of Prion's Disease: Nucleation-Polymerisatio

Precise reactions step and parameters Deterministic rate equations Stochastic algorithm Reproducing the experimental variability

Influences of parameters - stochastic algorithm

 c₀: if it increases, the relative proportion of normal and abnormal monomer will increase variability of the Lag Time. Indeed, larger c₀ will result in less active monomer available (while keeping the total value of monomer constant) making the nucleation event a rare event with high variability.

<ロ> (四) (四) (三) (三) (三)

Precise reactions step and parameters Deterministic rate equations Stochastic algorithm Reproducing the experimental variability

Influences of parameters - stochastic algorithm

- c₀: if it increases, the relative proportion of normal and abnormal monomer will increase variability of the Lag Time. Indeed, larger c₀ will result in less active monomer available (while keeping the total value of monomer constant) making the nucleation event a rare event with high variability.
- c_f: if the elongation rate of the nucleation steps decreases, it will increase the variability of the Lag Time. Smaller c_f gives less probability to the aggregation event, and then increases the fluctuations.

(ロ) (同) (E) (E) (E)

Precise reactions step and parameters Deterministic rate equations Stochastic algorithm Reproducing the experimental variability

Influences of parameters - stochastic algorithm

- c₀: if it increases, the relative proportion of normal and abnormal monomer will increase variability of the Lag Time. Indeed, larger c₀ will result in less active monomer available (while keeping the total value of monomer constant) making the nucleation event a rare event with high variability.
- c_f: if the elongation rate of the nucleation steps decreases, it will **increase the variability of the Lag Time**. Smaller c_f gives less probability to the aggregation event, and then increases the fluctuations.
- d1: The equilibrium constant of the nucleation step has no evident effect on variability.

(ロ) (同) (E) (E) (E)

Precise reactions step and parameters Deterministic rate equations Stochastic algorithm Reproducing the experimental variability

Influences of parameters - stochastic algorithm

k_{pf}: if it increases, the elongation constant will **increase the** variability of the growth rate. Large *k_{pf}* appears to give higher variability on the growth rate. Almost no influence on the lag time.

Precise reactions step and parameters Deterministic rate equations Stochastic algorithm Reproducing the experimental variability

Influences of parameters - stochastic algorithm

- *k_{pf}*: if it increases, the elongation constant will **increase the** variability of the growth rate. Large *k_{pf}* appears to give higher variability on the growth rate. Almost no influence on the lag time.
- *k_b*: the fragmentation rate influences on the respective numbers of nucleation events/ break events. Small *k_b* appears to give higher variability on the growth rate. Almost no influence on the lag time.

・ロト ・回ト ・ヨト ・ヨト

Precise reactions step and parameters Deterministic rate equations Stochastic algorithm Reproducing the experimental variability

Influences of parameters - example of k_{pf}





(b) Influence of k_{pf} on T_{lag}

Precise reactions step and parameters Deterministic rate equations Stochastic algorithm Reproducing the experimental variability

Evolution of the Lag Time and the Growth rate on the stochastic algorithm, for different populaton size



For both, large population decrease the variability, but it much more obvious for the lag time than the growth rate.

Precise reactions step and parameters Deterministic rate equations Stochastic algorithm Reproducing the experimental variability

Relation between the Lag Time and the Growth rate on the stochastic algorithm



Figure: Relation between lag time and growth rate (log scale)

Precise reactions step and parameters Deterministic rate equations Stochastic algorithm Reproducing the experimental variability

Distribution of the Lag time on the stochastic algorithm



YVINEC Romain Stochastic Model of Prion's Disease: Nucleation-Polymerisatio

-

Precise reactions step and parameters Deterministic rate equations Stochastic algorithm Reproducing the experimental variability

Distribution of the Growth rate on the stochastic algorithm



Distribution of the maximal growth rate $1/\tau$ in stochastic simulations

YVINEC Romain Stochastic Model of Prion's Disease: Nucleation-Polymerisatio

Precise reactions step and parameters Deterministic rate equations Stochastic algorithm Reproducing the experimental variability

Coexsitence of two structure



Figure: One example of coexistence of two structures

Precise reactions step and parameters Deterministic rate equations Stochastic algorithm Reproducing the experimental variability

Coexsitence of two structure



Figure: Reproducing experimental variability

YVINEC Romain Stochastic Model of Prion's Disease: Nucleation-Polymerisation

< ∃⇒



Improving the algorithm and precise fitting with data.

イロン 不同と 不同と 不同と

æ



- Improving the algorithm and precise fitting with data.
- Analyse study, generic scaling laws.

Conclusion

- Improving the algorithm and precise fitting with data.
- Analyse study, generic scaling laws.
- Mathematical analysis of the stochasticity.

Conclusion

- Improving the algorithm and precise fitting with data.
- Analyse study, generic scaling laws.
- Mathematical analysis of the stochasticity.
- Interaction between structures.