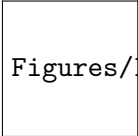


Modelisation and numerical simulation of the Smoluchowski equation in perforated domains

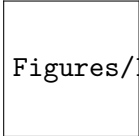
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15th of April



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Outline

- 1 Introduction
- 2 The Smoluchowski equation
- 3 Mathematical model for the aggregation of $A\beta$
- 4 Finite elements
- 5 Results
- 6 Conclusions and outlook

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Introduction

The development of Alzheimer's disease (AD)

- The agglomeration of β -Amyloid peptide ($A\beta$) : process associated with the development of AD.
- In some neurons, imbalance between production and clearance of $A\beta$ during aging.
- At elevated levels : produces pathological aggregates
→ accumulation of deposits known as **senile plaques**.

Smolushowski equation

- system of partial differential equations
- describes the evolving densities of diffusing particles that are prone to coagulate in pairs

Outline

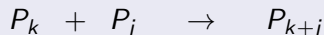
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Smoluchowski equation in polymerization 1/2

Binary reactions

- P_k , for $k \in \mathbb{N}$: polymer of size k , a set of k identical particle (monomers).
- With time, if 2 polymers are very close, they are likely to merge \rightarrow creation of a new polymer
- By convention, only binary reactions

Coalescence of polymers size k and j :



Restrictions of the model studied

- Assumptions : aggregation results only from Brownian movement or diffusion (thermal coagulation).
- We neglect other effects such as : multiple coagulation, condensation, fragmentation ...

Smoluchowski equation in polymerisation 2/2

Discrete diffusive coagulation equations

$$\frac{\partial u_i}{\partial t}(t, x) - d_i \Delta_x u_i(t, x) = Q_i(u) \quad \text{in } [0, T] \times \Omega \quad (1)$$

$u_i(t, x) \geq 0$ (for $i \geq 1$) \rightarrow **concentration of i -clusters**
 clusters with i identical elementary particles.
 $d_i \rightarrow$ **diffusion coefficient** of an i -cluster, $d_i > 0 \forall i \geq 1$.

Coagulation term

$$Q_i(u) = Q_{g,i}(u) - Q_{l,i}(u) \quad i \geq 1 \quad (2)$$

Coagulation term

$a_{i,j} \rightarrow$ **coagulation rates**, non nonnegative constants such that $a_{i,j} = a_{j,i}$ coefficient that represents the reaction in which an $i + j$ -cluster is formed from an i and a j -cluster.

Gain

$$Q_{g,i} = \frac{1}{2} \sum_{j=1}^{i-1} a_{i-j,j} u_{i-j} u_j \quad (3)$$

Describes the creation of polymers of size i by coagulation of polymers of size j and $i - j$ (clusters of size $\leq i - 1$).

Loss

$$Q_{l,i} = u_i \sum_{j=1}^{\infty} a_{i,j} u_i u_j \quad (4)$$

Describes the depletion (reduction) of polymers of size i after coalescence with other polymers.

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Parameters of the microscopic model

Domain

- $\Omega_0 \subset \mathbb{R}^3$: smooth bounded domain representing a portion of **cerebral tissue**
- $\bar{\Omega}_j \subset \Omega_0$ with $j = 1 \dots N$: regular regions representing N **neurons**
- $\bar{\Omega}_i \cap \bar{\Omega}_j \neq \emptyset$ if $i \neq j$

$$\Omega := \Omega_0 \setminus \bigcup_{j=1}^N \bar{\Omega}_j$$

Concentration u

- $u = (u_1, \dots, u_M) : M \in \mathbb{N}$ and $u_j = u_j(t, x), t \geq 0, x \in \Omega$
- $u_j(t, x)$ with $1 \leq j < M - 1$: concentration of an $A\beta$ assembly of j monomers.
- u_M : takes into account aggregation of more than $M-1$ monomers (M = maximum size of an assembly).

$A\beta$'s concentration in the monomeric form

$$\left\{ \begin{array}{l} \frac{\partial u_1}{\partial t}(t, x) - d_1 \Delta_x u_1(t, x) - Q_{l,1} = 0 \\ \nabla_x u_1 \cdot n = 0 \\ \nabla_x u_1 \cdot n = \Psi_j \\ u_1(0, x) = U_1 \geq 0 \end{array} \right. \quad \begin{array}{l} \text{on } \partial\Omega_0 \\ \text{on } \partial\Omega_j, j = 1 \dots N \end{array} \quad (5)$$

- No $Q_{gain} \rightarrow$ we cannot form a monomer by coagulation.
- Homogeneous Neumann condition on $\partial\Omega_0 \rightarrow$ artificial isolation of the cerebral from its environment
- Non-homogeneous Neumann condition on $\partial\Omega_j, j = 1 \dots N \rightarrow$ production of $A\beta$ in the monomeric form at the level of neuron membranes
- $\Psi_j \neq 0 \rightarrow$ only neurons affected by the disease are taken into account

$A\beta$'s concentration in assemblies

For assemblies of size $1 < m \leq M - 1$:

$$\left\{ \begin{array}{l} \frac{\partial u_m}{\partial t}(t, x) - d_m \Delta_x u_m(t, x) - Q_{l,m} = Q_{g,m} \\ \nabla_x u_m \cdot n = 0 \\ \nabla_x u_m \cdot n = 0 \\ u_m(0, x) = 0 \end{array} \right. \begin{array}{l} \text{on } \partial\Omega_0 \\ \text{on } \partial\Omega_j, j = 1 \dots N \end{array} \quad (6)$$

For assemblies of size M :

$$\left\{ \begin{array}{l} \frac{\partial u_M}{\partial t}(t, x) - d_M \Delta_x u_M(t, x) = Q_{g,M} \\ \nabla_x u_M \cdot n = 0 \\ \nabla_x u_M \cdot n = 0 \\ u_M(0, x) = 0 \end{array} \right. \begin{array}{l} \text{on } \partial\Omega_0 \\ \text{on } \partial\Omega_j, j = 1 \dots N \end{array} \quad (7)$$

Special case : assembly size M

(takes into account all assemblies with size $\geq M$)

No Q_{loss} and $Q_{g,M} = \frac{1}{2} \sum_{\substack{j+k \geq M \\ k < M \\ j < M}} a_{j,k} u_j u_k$

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Weak formulation 1/2

For $m = 1$, non-homogeneous Neumann condition :

$$\int_{\Omega} \frac{\partial u_1}{\partial t} \cdot v - d_1 \int_{\Omega} \Delta u_1 \cdot v = \int_{\Omega} Q_1 v \quad \text{on } \Omega$$

Then, applying the Green theorem :

$$\begin{aligned} \int_{\Omega} \Delta u_1 \cdot v &= - \int_{\Omega} \nabla u_1 \cdot \nabla v + \int_{\partial\Omega} \underbrace{(\partial_n u_1)}_{=\Psi_j \text{ on } \partial\Omega_j} v \\ \Rightarrow \int_{\Omega} \frac{\partial u_1}{\partial t} \cdot v + d_1 \int_{\Omega} \nabla u_1 \cdot \nabla v &= \int_{\Omega} Q_1 v + \int_{\partial\Omega_j} \Psi_j v \quad \text{on } \Omega \end{aligned}$$

Weak formulation 2/2

For $m = 2 \dots M$, homogeneous Neumann condition :

$$\int_{\Omega} \frac{\partial u_m}{\partial t} \cdot v + d_m \int_{\Omega} \nabla u_m \cdot \nabla v = \int_{\Omega} Q_m v \quad \text{on } \Omega$$

Final weak formulation :

$$\begin{cases} \text{Finding } u_m \in V \text{ such that :} \\ a(u_m, v) = l_m(v) \forall v \in V \end{cases} \quad (8)$$

with :

$$\begin{aligned} a(u_m, v) &= \int_{\Omega} \frac{\partial u_m}{\partial t} + d_m \int_{\Omega} \nabla u_m \cdot \nabla v \quad \text{for } m=1 \dots M \\ l_m(v) &= \begin{cases} \int_{\Omega} Q_1 v + \int_{\partial\Omega_{j=1 \dots N}} \Psi_j v, & \text{if } m = 1 \\ \int_{\Omega} Q_m v & \text{otherwise} \end{cases} \end{aligned} \quad (9)$$

Time discretisation

Implicit Euler scheme to approximate $\frac{\partial u_m}{\partial t}$ considering $u_m^n = u_m(t_n)$:

$$\frac{\partial u_m^{n+1}}{\partial t} \approx \frac{u_m^{n+1} - u_m^n}{\Delta t} \quad \text{with } \Delta t = t_{n+1} - t_n$$

Replacing $\frac{\partial u_m^{n+1}}{\partial t}$ in the weak formulation :

$$\begin{aligned} a(u_m, v) &= \int_{\Omega} u_m^{n+1} \cdot v + \Delta t d_m \int_{\Omega} \nabla u_m^{n+1} \cdot \nabla v \\ I_m(v) &= \begin{cases} \int_{\Omega} u_m^n \cdot v + \Delta t \int_{\Omega} Q_1 v + \Delta t \int_{\partial\Omega_{j=1\dots N}} \Psi_j v, & \text{if } m = 1 \\ \int_{\Omega} u^n \cdot v + \Delta t \int_{\Omega} Q_m v & \text{otherwise} \end{cases} \end{aligned}$$

(10)

Galerkin Method

We consider a space V_h of finite dimension and define a base $(\phi_i)_{i=1..N_s}$ such that : $u_{h,m} = \sum_{i=1}^{N_s} x_{i,m} \phi_i$

Replacing u_m by u_h, m in the weak formulation, we obtain the following linear system :

For $m=1..M$

$$(M + \Delta t d_m K) X_m^{n+1} = \Delta t B^n + M X_m^n$$

with :

- M : the mass matrix
- K : the rigidity matrix
- B^n : the coagulation term at time t_n

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Only diffusion

Parameters :
Figure 1

Result with one neuron and $N=??$

Parameters :

$U_0 =$, $\Psi =$, $d_m =$, $a_i, j =$ Figure ...

Convergence studies

Computational domain: $\Omega = [-0.5; 0.5] \times [-0.5; 0.5] \times [0; 1]$

Initial condition:

$$\begin{bmatrix} u \\ v \\ w \\ \eta \end{bmatrix} = \begin{bmatrix} \cos\left(\pi x - \frac{\pi}{4}\right) \sin\left(\pi y - \frac{\pi}{4}\right) \\ -\sin\left(\pi x - \frac{\pi}{4}\right) \cos\left(\pi y - \frac{\pi}{4}\right) \\ 0 \\ 1 \end{bmatrix}$$

Convergence studies: divergence free

$\mathcal{O}2$ (N=1)	$\mathcal{O}(L_1)$		$\mathcal{O}(L_2)$		$\mathcal{O}(L_\infty)$	
8.94E-03	9.3167E-02	-	4.7585E-02	-	2.0916E-02	-
5.36E-03	3.4633E-02	1.9	1.7711E-02	1.9	7.7751E-03	1.9
3.83E-03	1.7973E-02	1.9	9.1864E-03	2.0	4.0131E-03	2.0
2.98E-03	1.0984E-02	2.0	5.6113E-03	2.0	2.4419E-03	2.0
$\mathcal{O}3$ (N=2)	$\mathcal{O}(L_1)$		$\mathcal{O}(L_2)$		$\mathcal{O}(L_\infty)$	
8.94E-03	6.3636E-03	-	3.3567E-03	-	1.7099E-03	-
5.36E-03	1.4233E-03	2.9	7.4921E-04	2.9	3.7861E-04	3.0
3.83E-03	5.2460E-04	3.0	2.7610E-04	3.0	1.3905E-04	3.0
2.98E-03	2.4831E-04	3.0	1.3062E-04	3.0	6.5434E-05	3.0
$\mathcal{O}4$ (N=3)	$\mathcal{O}(L_1)$		$\mathcal{O}(L_2)$		$\mathcal{O}(L_\infty)$	
8.94E-03	1.3029E-03	-	6.8714E-04	-	4.5625E-04	-
5.36E-03	1.8287E-04	3.8	9.5818E-05	3.9	6.5249E-05	3.8
3.83E-03	4.9274E-05	3.9	2.5704E-05	3.9	1.7244E-05	4.0
2.98E-03	1.8384E-05	3.9	9.5611E-06	3.9	6.3443E-06	4.0

Convergence studies: NON-divergence free

$\mathcal{O}2$ (N=1)	$\mathcal{O}(L_1)$		$\mathcal{O}(L_2)$		$\mathcal{O}(L_\infty)$	
8.94E-03	9.3301E-02	-	4.7647E-02	-	2.1148E-02	-
5.36E-03	3.4648E-02	1.9	1.7716E-02	1.9	7.8523E-03	1.9
3.83E-03	1.7975E-02	1.9	9.1864E-03	2.0	4.0512E-03	2.0
2.98E-03	1.0984E-02	2.0	5.6106E-03	2.0	2.4647E-03	2.0
$\mathcal{O}3$ (N=2)	$\mathcal{O}(L_1)$		$\mathcal{O}(L_2)$		$\mathcal{O}(L_\infty)$	
8.94E-03	9.6634E-03	-	4.9852E-03	-	2.6242E-03	-
5.36E-03	2.1862E-03	2.9	1.1255E-03	2.9	6.0363E-04	3.0
3.83E-03	8.1014E-04	3.0	4.1625E-04	3.0	2.2272E-04	3.0
2.98E-03	3.8443E-04	3.0	1.9725E-04	3.0	1.0511E-05	3.0
$\mathcal{O}4$ (N=3)	$\mathcal{O}(L_1)$		$\mathcal{O}(L_2)$		$\mathcal{O}(L_\infty)$	
8.94E-03	1.4103E-03	-	7.2876E-04	-	3.9854E-04	-
5.36E-03	1.9756E-04	3.8	1.0153E-05	3.9	5.6671E-05	3.8
3.83E-03	5.3161E-05	3.9	2.7238E-05	3.9	1.4968E-05	4.0
2.98E-03	1.9817E-05	3.9	1.0134E-06	3.9	5.5062E-06	4.0

Convergence studies: divergence-free error

div-free			
$h(\Omega)$	$\epsilon_{L_2} \mathcal{O}(2)$	$\epsilon_{L_2} \mathcal{O}(3)$	$\epsilon_{L_2} \mathcal{O}(4)$
8.94E-03	0.0000E-00	5.1383E-16	1.3726E-15
5.36E-03	0.0000E-00	6.0425E-16	1.1620E-15
3.83E-03	0.0000E-00	4.7411E-16	9.9785E-16
2.98E-03	0.0000E-00	4.5329E-16	7.7298E-16

NON div-free			
$h(\Omega)$	$\epsilon_{L_2} \mathcal{O}(2)$	$\epsilon_{L_2} \mathcal{O}(3)$	$\epsilon_{L_2} \mathcal{O}(4)$
8.94E-03	8.9549E-02	6.7913E-04	1.3485E-04
5.36E-03	1.0477E-03	1.4317E-04	1.5366E-05
3.83E-03	4.9002E-04	5.0082E-05	4.2957E-06
2.98E-03	2.8251E-04	2.2798E-05	1.6770E-06

Nonlinear convective terms

Horizontal plane $x - y$

$$\begin{aligned}(u, v, w) &= (x, y, x) \times (0, 0, 1) \\ \eta &= 1\end{aligned}$$

Nonlinear convective terms

Vertical plane $x - z$

$$\begin{aligned}(u, v, w) &= (x, y, x) \times (0, 1, 0) \\ \eta &= 1\end{aligned}$$

Horizontal viscous terms

Finite volume discretization

$$\nabla_i^2 \mathbf{u}_k = \frac{1}{|P_i|} \int_{P_i} \nabla \cdot \nabla_h \mathbf{u}_k dA = \frac{1}{|P_i|} \int_{\partial P_i} \nabla_h \mathbf{u}_k \cdot \mathbf{n}_j ds.$$

Viscous numerical flux (Gassner et al., *JCP*, 2017):

$$\nabla_h \mathbf{u}_k \cdot \mathbf{n}_j = \frac{1}{2} (\nabla \mathbf{u}_k^+ + \nabla \mathbf{u}_k^-) \cdot \mathbf{n}_j + \frac{1}{\delta_j \sqrt{\frac{1}{2}\pi}} (\mathbf{u}_k^+ - \mathbf{u}_k^-)$$

Horizontal viscous terms

First problem of Stokes ($\nu^h = 10^{-3}$)

Figure: Horizontal velocity distribution for the first problem of Stokes at time $t = 0.3$ s (left) and $t = 0.5$ s. The initial flow velocity is $(u,v,w)=(1,0,0)$ with $\eta = 1$.

Horizontal viscous terms

First problem of Stokes ($\nu^h = 10^{-3}$)

Figure: Horizontal velocity distribution for the first problem of Stokes at time $t = 0.3$ s (left) and $t = 0.5$ s. The initial flow velocity is $(u,v,w)=(1,0,0)$ with $\eta = 1$. Different order of accuracy are shown.

Nonhydrostatic flow

Computational domain: $\Omega = [0; 10]^3$

$$(u, v, w) = 0$$

$$\eta = 0.02x - 0.1$$

Nonhydrostatic flow

Computational domain: $\Omega = [0; 10]^3$

$$(u, v, w) = 0$$

$$\eta = 0.02x - 0.1$$

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Conclusions and outlook

Conclusions

- high order reconstruction for the velocity field on unstructured Voronoi meshes;
- decoupled horizontal and vertical divergence-free reconstruction procedure;
- applications to hydrostatic and nonhydrostatic flows.

Conclusions and outlook

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- high order reconstruction for the velocity field on unstructured Voronoi meshes;
- decoupled horizontal and vertical divergence-free reconstruction procedure;
- applications to hydrostatic and nonhydrostatic flows.

Outlook

- sediment transport;
- turbulence model.

Thank you!

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