

Numerical Approximation of Mean Curvature Flow

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- 1 Introduction
 - Parkinson's Disease
 - Level Sets Method
- 2 Existence and unicity
 - Viscosity solutions
 - Energy estimate
- 3 Numerical Analysis
 - Numerical IMEX Method
 - Parallel Splitting Algorithm
- 4 Segmentation Model
 - Chan and Vese Model
 - Numerical Results
- 5 Conclusions and Future Work

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What is it?

- **Degenerative disorder** of the central nervous system that **affects the control of muscles** and so may affect movement, speech and posture
- Caused by insufficient formation and action of **dopamine**

Parkinson disease

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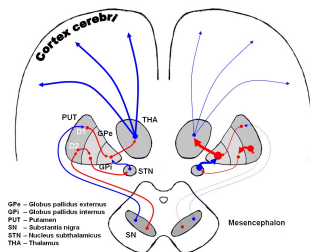
Diagnoses

- There are currently no blood or laboratory tests that have been proven to help in diagnosing the disease
- 75% of clinical diagnoses of Parkinson disease are confirmed at autopsy

Parkinson disease

Pathology

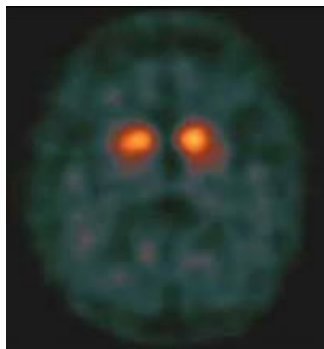
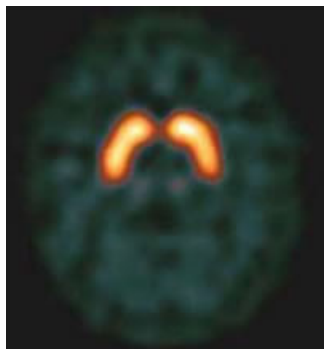
- The symptoms of Parkinson disease result from the **loss of dopaminergic cells** and subsequent loss of melanin
- The neurons project to the **striatum** and their loss leads to alteration in the activity of neural circuits within the basal ganglia



Credits: Wikipedia, the free encyclopedia

Parkinson disease

- **Decreased dopamine activity** in the basal ganglia, a pattern which aids in diagnosing Parkinson disease
- PET and SPECT images may help

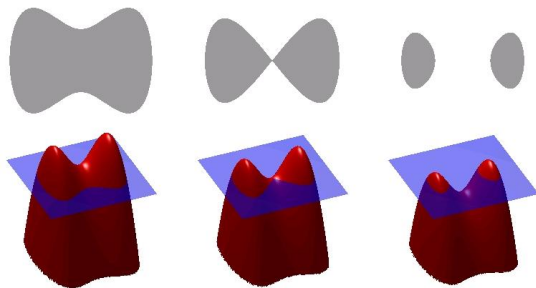


Credits: European Parkinson's Disease Association

Level Sets Method

$\Gamma(t)$ is implicitly represented by the zero level set of a higher dimension function ϕ :

$$\Gamma(t) = \{(x, y) \in \Omega : \phi(t, x, y) = 0\}$$



Credits: Oleg Alenxandrov, en.wikipedia.org

- Notions of interior and exterior of a curve are immediate
- Union and division of curves are automatic

Level Sets Method

- Evolve the curve in the direction of the normal with speed $v \Leftrightarrow$ solving a PDE

$$\frac{\partial \phi}{\partial t} = v |\nabla \phi|, \quad \phi(0, x, y) = \phi_0(x, y)$$

with suitable boundary conditions

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- For $\phi \in C^{2,1}([0, T] \times \Omega)$ and $\nabla \phi \neq 0$ in a neighborhood of $\Gamma(t)$:

$$(\text{IBVP}) \quad \begin{cases} \phi_t = |\nabla \phi| \operatorname{div} \left(\frac{\nabla \phi}{|\nabla \phi|} \right) & (x, y) \in \Omega, t > 0 \\ \phi(0, x, y) = \phi_0(x, y) & (x, y) \in \bar{\Omega} \\ \phi(t, x, y) = 0 & (x, y) \in \partial\Omega, t > 0 \end{cases}$$

with Ω a bounded open set of \mathbb{R}^2 and $\phi_0 \in C(\Omega)$.

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Viscosity solution (Evans & Spruck, 1991)

Let $\phi \in C([0, T] \times \Omega) \cap L^\infty([0, T] \times \Omega)$.

- ϕ is a viscosity sub-solution (**super-solution**) of (IBVP) if for all $v \in C^2([0, T], \Omega)$, $\phi - v$ has a local maximum (**minimum**) in (t_0, x_0, y_0) then $(\nabla\phi(t_0, x_0, y_0) \neq 0)$

$$v_t(t_0, x_0, y_0) \leq (\geq) |\nabla\phi(t_0, x_0, y_0)| \operatorname{div} \left(\frac{\nabla\phi(t_0, x_0, y_0)}{|\nabla\phi(t_0, x_0, y_0)|} \right)$$

- ϕ is a viscosity solution of (IBVP) if it simultaneously a viscosity sub and super-solution.

Results

- Under certain conditions, the viscosity solution of (IBVP) exists and its unique (Evans & Spruck, 1991)
- The curves $\Gamma(t)$ are independent of the initial choice ϕ_0 (Evans & Spruck, 1991)
- The following stability result holds (Caselles *et al.*, 1993)

$$\sup_{0 \leq s \leq t} \|\phi(s) - \hat{\phi}(s)\|_{L^\infty} \leq \|\phi_0 - \hat{\phi}_0\|_{L^\infty} \quad \forall t \in [0, T]$$

Motion by Mean Curvature

Determine ϕ from the initial boundary value problem (IBVP):

$$\begin{cases} \frac{\phi_t}{|\nabla\phi|} = \nabla^T \left(\frac{\nabla\phi}{|\nabla\phi|} \right) & (x, y) \in \Omega, t > 0 \\ \phi(0, x, y) = \phi_0(x, y) & (x, y) \in \bar{\Omega} \\ \phi(t, x, y) = 0 & (x, y) \in \partial\Omega, t > 0 \end{cases}$$

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Theorem

$$\underbrace{\|\phi(t)\|_{L^1} + \|\nabla\phi(t)\|_{L^1}}_{\|\phi(t)\|_{W^{1,1}}} \leq C \underbrace{(\|\phi_0\|_{L^1} + \|\nabla\phi_0\|_{L^1})}_{\|\phi_0\|_{W^{1,1}}}$$

Proof:

- $\|\nabla\phi(t)\|_{L^1} + \int_0^t \int_{\Omega} \frac{\phi_t^2}{|\nabla\phi|} dx dy ds = \|\nabla\phi_0\|_{L^1}$ (Walkington, 1996)
- $\|\nabla\phi(t)\|_{L^1} \leq \|\nabla\phi_0\|_{L^1}$
- Poincaré inequality in L^1 : $\|\phi\|_{L^1} \leq C^* \|\nabla\phi\|_{L^1}$

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IMEX Method: time discretization

$$\frac{1}{|\nabla\phi^n|} \frac{\phi^{n+1} - \phi^n}{\Delta t} = \operatorname{div} \left(\frac{\nabla\phi^{n+1}}{|\nabla\phi^n|} \right)$$

- $t^n = n\Delta t$, $n = 0, \dots, N$, with $t_0 = 0$ and $t_N = T$
- $\phi^n \approx \phi(n\Delta t, x, y)$, $\forall (x, y) \in \Omega$

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Theorem

$$\|\phi^{n+1}\|_{W^{1,1}} \leq C \|\phi^n\|_{W^{1,1}} \quad \forall n = 0, 1, \dots, N-1$$

Proof:

- Multiply the equation by $\phi^{n+1} - \phi^n$ with respect to the L^2 inner product and integrate by parts
- $\|\nabla\phi^{n+1}\|_{L^1} \leq \|\nabla\phi^n\|_{L^1} \quad \forall n = 0, 1, \dots, N-1$
- Poincaré inequality in L^1

IMEX Method: full discretization

$$\frac{1}{|\nabla_h \phi_{ij}^n|} \frac{\phi_{ij}^{n+1} - \phi_{ij}^n}{\Delta t} = D_x^+ \left(\frac{D_x^- \phi_{ij}^{n+1}}{|\nabla_h \phi_{ij}^n|} \right) + D_y^+ \left(\frac{D_y^- \phi_{ij}^{n+1}}{|\nabla_h \phi_{ij}^n|} \right)$$

- $\bar{\Omega}_h =$ grid in $\bar{\Omega}$ with space step h
- Finite differences: D_x^+, D_y^+ (forward); D_x^-, D_y^- (backward)
- $\phi_{ij}^n \approx \phi(n\Delta t, x_i, y_j)$, $\forall (x_i, y_j) \in \bar{\Omega}_h$
- $|\nabla_h \phi_{ij}^n| = \sqrt{(D_x^- \phi_{ij}^n)^2 + (D_y^- \phi_{ij}^n)^2}$
- Norm in the discrete $W^{1,1}$ space

$$\|\phi\|_{1,1} = \sum_{i,j} h^2 |\phi_{ij}| + \sum_{i,j} h^2 |\nabla_h \phi_{ij}|$$

Theorem

$$\|\phi^{n+1}\|_{1,1} \leq C \|\phi^n\|_{1,1} \quad \forall n = 0, \dots, N-1$$

Proof:

- Multiply

$$\frac{1}{|\nabla_h \phi_{ij}^n|} \frac{\phi_{ij}^{n+1} - \phi_{ij}^n}{\Delta t} = D_x^+ \left(\frac{D_x^- \phi_{ij}^{n+1}}{|\nabla_h \phi_{ij}^n|} \right) + D_y^+ \left(\frac{D_y^- \phi_{ij}^{n+1}}{|\nabla_h \phi_{ij}^n|} \right)$$

by $\phi_{ij}^{n+1} - \phi_{ij}^n$ with respect to the discrete L^2 inner product and use summation by parts

- $\sum_{ij} h^2 |\nabla_h \phi_{ij}^{n+1}| \leq \sum_{ij} h^2 |\nabla_h \phi_{ij}^n| \quad \forall n = 0, \dots, N-1$
- Discrete Poincaré inequality in ℓ^1 : $\sum_{ij} h^2 |\phi_{ij}^n| \leq C^* \sum_{ij} h^2 |\nabla_h \phi_{ij}^n|$

Parallel Splitting Algorithm

$$\frac{\partial \phi}{\partial t} = A\phi + f(t) \quad \text{in } \Omega \times [0, T], \quad \phi(0) = \phi_0$$

- $A = A_1 + A_2 + \dots + A_m$ and $f = f_1 + f_2 + \dots + f_m$
- A is time independent

Algorithm (Lu, Neittaanmaki and Tai, 1992)

At each level time $n = 0, \dots, N - 1$ compute:

$$\textcircled{1} \quad \frac{\phi^{n+\frac{k}{2m}} - \phi^n}{m\Delta t} = A_k \phi^{n+\frac{k}{2m}} + f_k \left(\left(n + \frac{1}{2}\right)\Delta t \right) \quad k = 1, \dots, m$$

$$\textcircled{2} \quad \phi^{n+1} = \frac{1}{m} \sum_{k=1}^m \phi^{n+\frac{k}{2m}}$$

Parallel Splitting Algorithm

Consider $A_1 \phi^{n+\frac{1}{4}} = D_x^+ \left(\frac{D_x^- \phi_{ij}^{n+\frac{1}{4}}}{|\nabla \phi_{ij}^n|} \right)$ and $A_2 \phi^{n+\frac{1}{2}} = D_x^+ \left(\frac{D_y^- \phi_{ij}^{n+\frac{1}{2}}}{|\nabla \phi_{ij}^n|} \right)$

- Construction of A_1

$$\frac{\phi^{n+\frac{1}{4}} - \phi^n}{2\Delta t} = A_1 \phi^{n+\frac{1}{4}} \Leftrightarrow$$

$$\frac{1}{|\nabla_h \phi_{ij}^n|} \frac{\phi_{ij}^{n+\frac{1}{4}} - \phi_{ij}^n}{2\Delta t} = \frac{\phi_{i-1,j}^{n+\frac{1}{4}}}{h^2 |\nabla_h \phi_{i,j}^n|} - \frac{2}{h^2} \phi_{ij}^{n+\frac{1}{4}} \left(\frac{1}{|\nabla_h \phi_{i+1,j}^n|} + \frac{1}{|\nabla_h \phi_{i,j}^n|} \right) + \frac{\phi_{i+1,j}^{n+\frac{1}{4}}}{h^2 |\nabla_h \phi_{i+1,j}^n|}$$

- A_1 is tridiagonal and diagonally dominant with

$$a_{i,i-1} = \frac{1}{h^2}, \quad a_{i,i} = -\frac{2}{h^2} \left(\frac{|\nabla_h \phi_{i,j}^n|}{|\nabla_h \phi_{i+1,j}^n|} + 1 \right), \quad a_{i,i+1} = \frac{|\nabla_h \phi_{i,j}^n|}{h^2 |\nabla_h \phi_{i+1,j}^n|}$$

- A similar construction can be made for A_2

Parallel Splitting Algorithm

Algorithm

At each level time $n = 0, \dots, N - 1$ compute:

- 1 Compute $|\nabla_h \phi_{ij}^n| = \sqrt{(D_x^- \phi_{ij}^n)^2 + (D_y^- \phi_{ij}^n)^2}$
- 2 Construct A_1 and A_2
- 3 Solve

$$(I - 2\Delta t A_1)\phi^{n+\frac{1}{4}} = \phi^n \quad \text{and} \quad (I - 2\Delta t A_2)\phi^{n+\frac{1}{2}} = \phi^{n+\frac{1}{4}}$$

- 4
$$\phi^{n+1} = \frac{\phi^{n+\frac{1}{4}} + \phi^{n+\frac{1}{2}}}{2}$$

Parallel Splitting Algorithm

Theorem (Stability)

The algorithm is unconditionally stable in the $\|\cdot\|_\infty$ norm.

Proof:

- $I - 2\Delta t A_1$ and $I - 2\Delta t A_2$ are M-matrices
- $\exists c_1, c_2 \geq 0$: $\|(I - 2\Delta t A_1)^{-1}\|_\infty \leq c_1$ and $\|(I - 2\Delta t A_2)^{-1}\|_\infty \leq c_2$
- $\|\Phi^{n+1}\|_\infty \leq \frac{1}{2} \left(\|\Phi^{n+\frac{1}{4}}\|_\infty + \|\Phi^{n+\frac{1}{2}}\|_\infty \right) \leq \frac{1}{2} (c_1 + c_2) \|\Phi^n\|_\infty$

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Theorem (Convergence)

If $(-A_k)$, $k = 1, \dots, m$, are irreducible M-matrices, then the algorithm is convergent of first order in Δt .

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CV model (Chan and Vese, 2001)

- Find the curve that minimizes:

$$F(c_1, c_2, \phi) = \mu \int_{\Omega} \delta_0(\phi) |\nabla \phi| dx dy + \nu \int_{\Omega} H(\phi) dx dy \\ + \lambda_1 \int_{\Omega} |u_0 - c_1|^2 H(\phi) dx dy + \lambda_2 \int_{\Omega} |u_0 - c_2|^2 (1 - H(\phi)) dx dy$$

- It reduces to the resolution of a PDE:

$$\frac{\partial \phi}{\partial t} = \delta_0(\phi) \left(\mu \operatorname{div} \left(\frac{\nabla \phi}{|\nabla \phi|} \right) - \nu - \lambda_1 (u_0 - c_1)^2 + \lambda_2 (u_0 - c_2)^2 \right)$$

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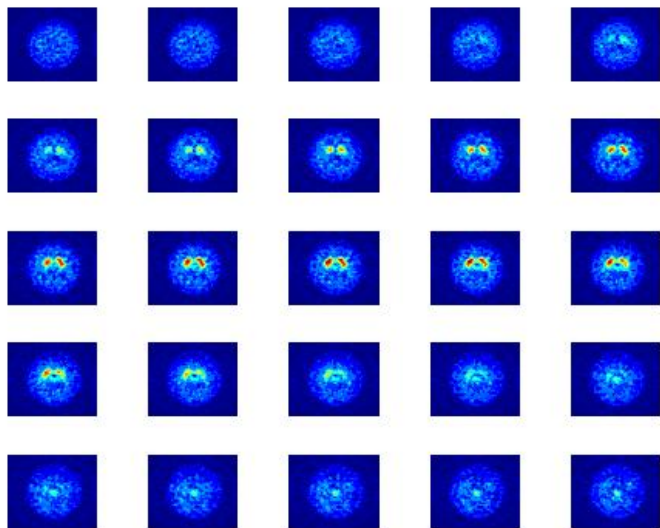
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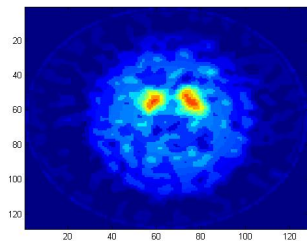
All the previous results could be generalized for

$$\phi_t = g(\phi) \operatorname{div} \left(\frac{\nabla \phi}{|\nabla \phi|} \right) \quad (x, y) \in \Omega, \quad t > 0, \quad g > 0$$

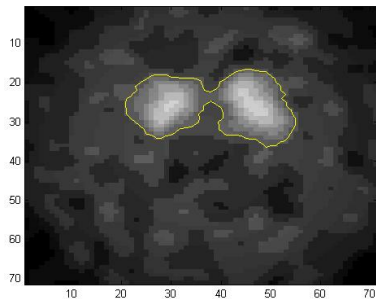
SPECT Images (given by IBILI)



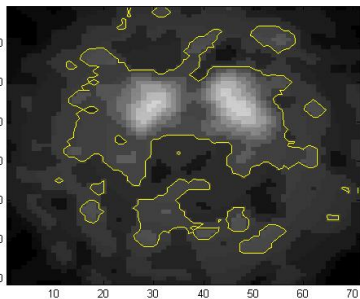
Numerical Results



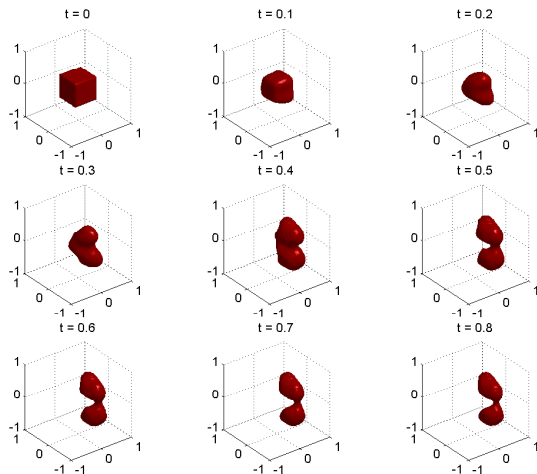
Time = 20 sec



Time = 20 sec

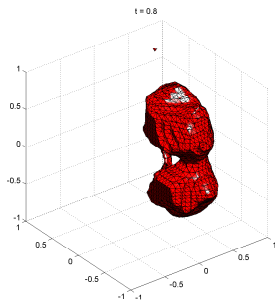
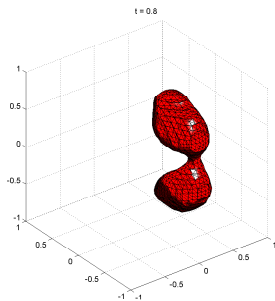


Numerical Results



Evolution of the zero level set in the iteration for $\mu = 0.05$

Numerical Results



Results of segmentation algorithm for $\mu = 0.05$ (left) and $\mu = 0.001$ (right)

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Conclusions and Future Work

Conclusions

- IMEX method with good stability properties
- A parallel splitting algorithm

Conclusions and Future Work






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Future Work

- Higher order splitting
- Optical Coherence Tomography (OCT) images

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