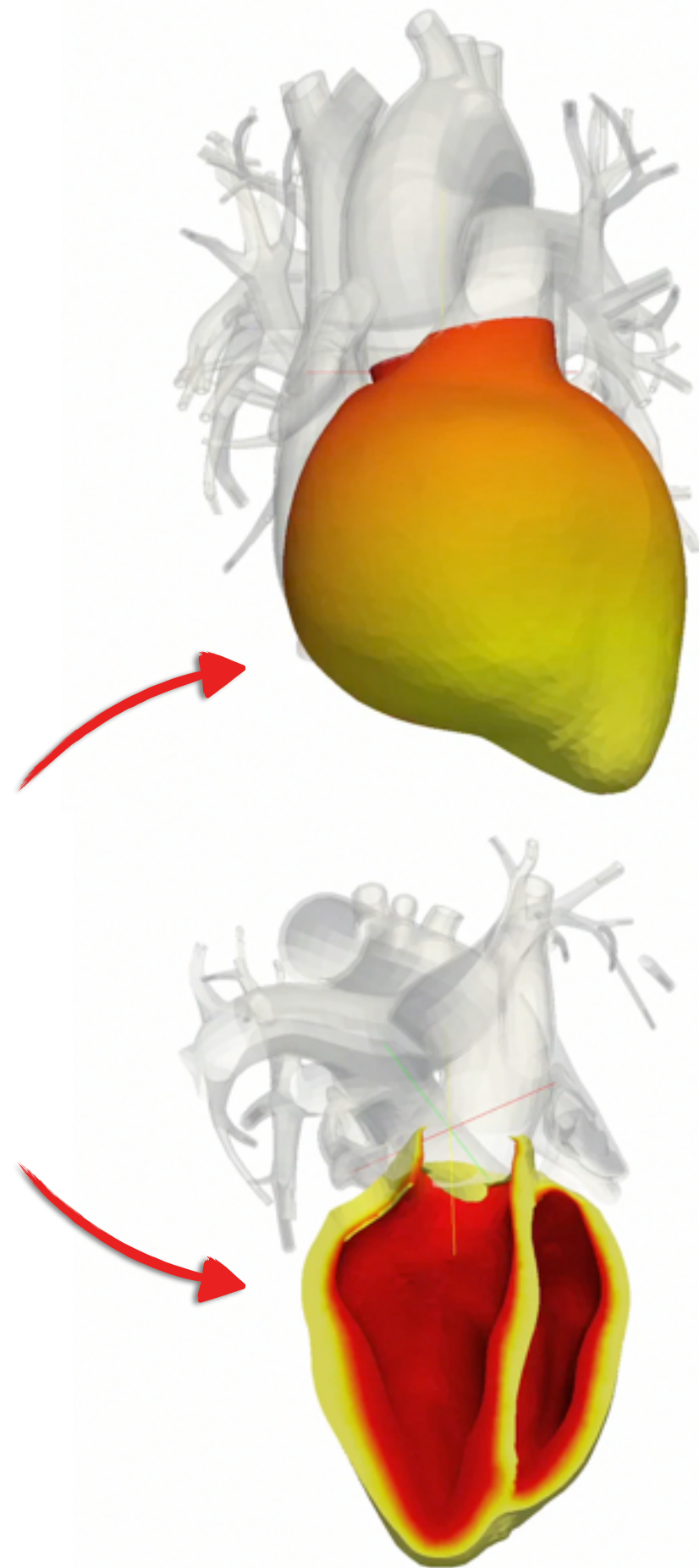
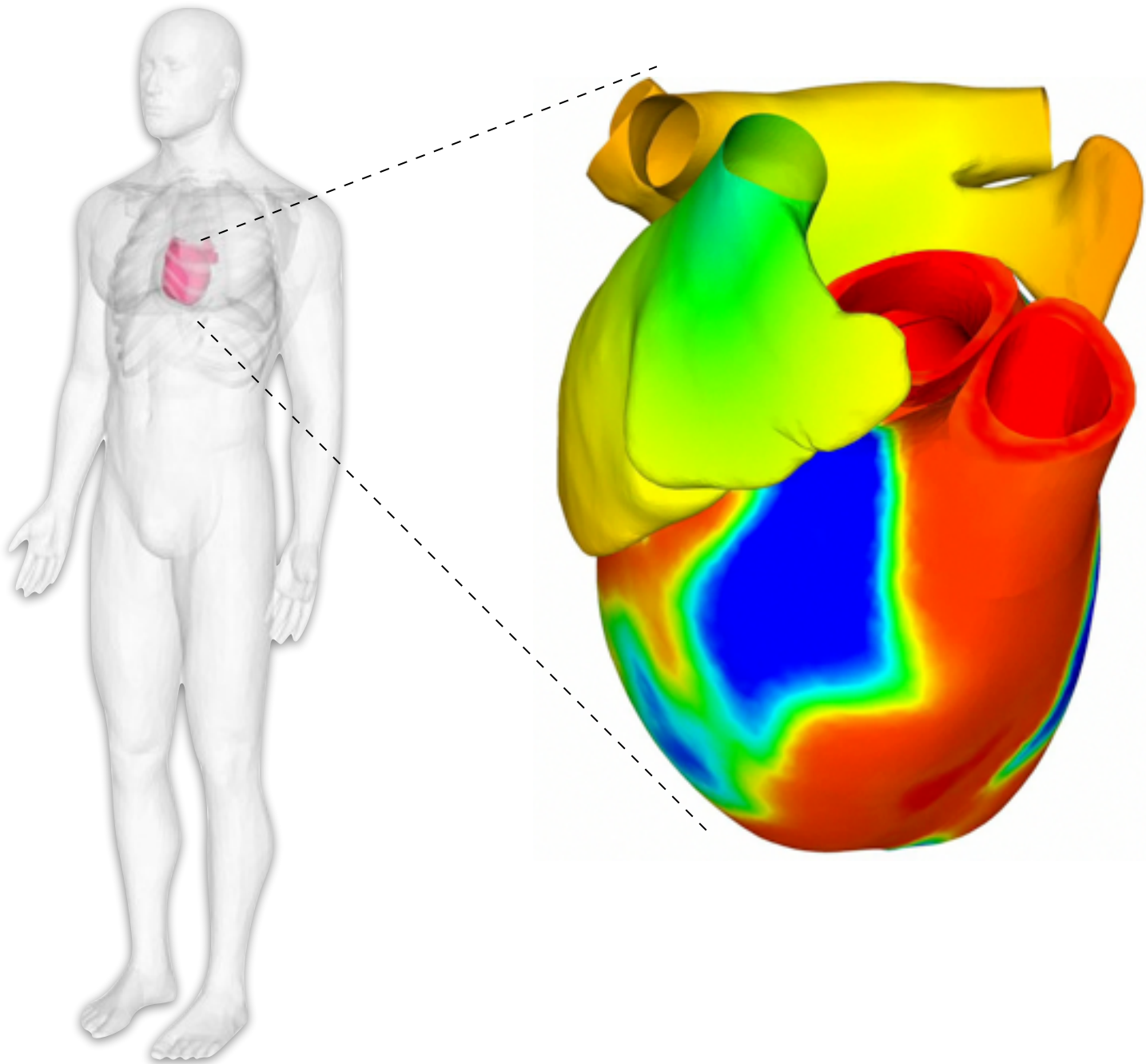




Observers based on front data for PDE models of propagative phenomena.

*P. Moireau, Inria Saclay - Ile de France, Team MΞDISIM
— Joint work with D. Chapelle, A. Collin, S. Imperiale*

Electrophysiology



From electrophysiology to reaction-diffusion

- Asymptotic Bidomain model

$$\begin{cases} A_m \left(C_m \frac{\partial V_m}{\partial t} + I_{ion}(V_m, \dots) \right) - \operatorname{div}(\vec{\sigma}_i \cdot \vec{\nabla} V_m) = \operatorname{div}(\vec{\sigma}_i \cdot \vec{\nabla} u_e) + I_{app}, & \text{in } \mathcal{B} \times (0, T), \\ \operatorname{div}((\vec{\sigma}_i + \vec{\sigma}_e) \cdot \vec{\nabla} u_e) = -\operatorname{div}(\vec{\sigma}_i \cdot \vec{\nabla} V_m), & \text{in } \mathcal{B} \times (0, T), \end{cases}$$

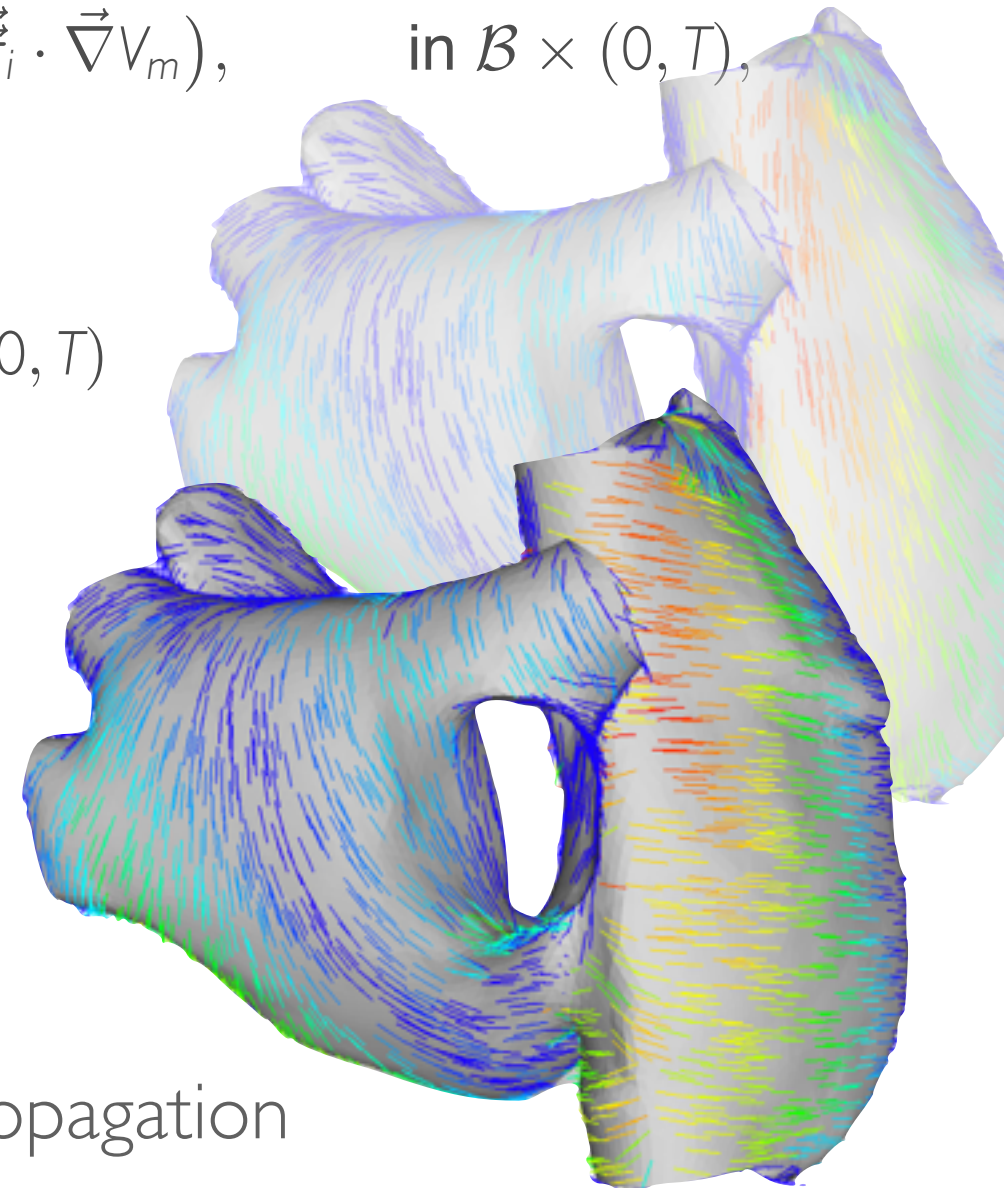
- Monodomain model

$$A_m \left(C_m \frac{\partial V_m}{\partial t} + I_{ion}(V_m, \dots) \right) - \operatorname{div}(\vec{\sigma} \cdot \vec{\nabla} V_m) = I_{app}, \quad \text{in } \mathcal{B} \times (0, T)$$

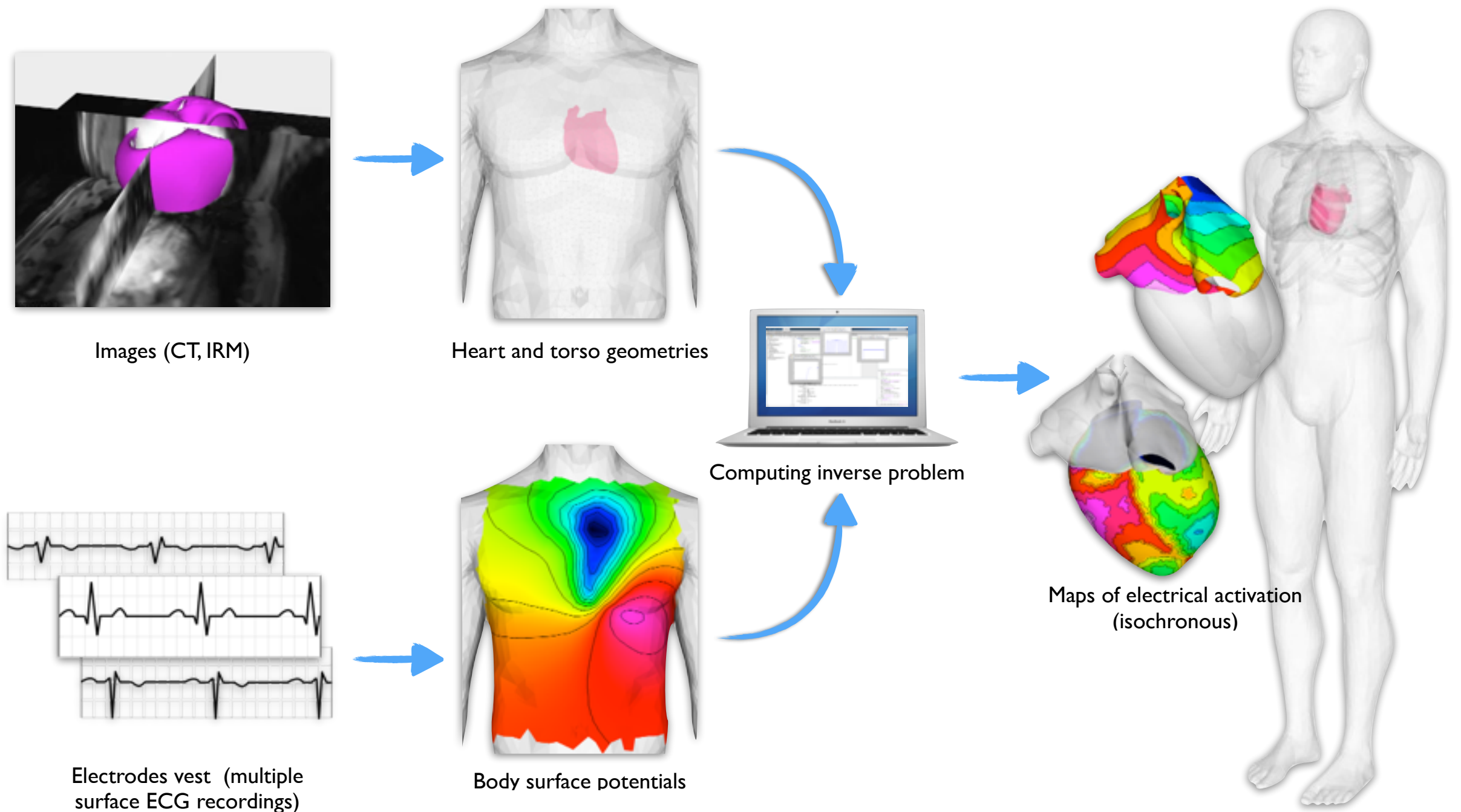
- Reaction diffusion system

$$\begin{cases} \partial_t u - \underline{\nabla} \cdot (D \underline{\nabla} u) = f(u), & \text{in } \mathcal{B} \times (0, T), \\ (D \underline{\nabla} u) \cdot \underline{n} = 0, & \text{on } \partial \mathcal{B} \times (0, T), \\ u(\underline{x}, 0) = u_0(\underline{x}), & \text{in } \mathcal{B}. \end{cases}$$

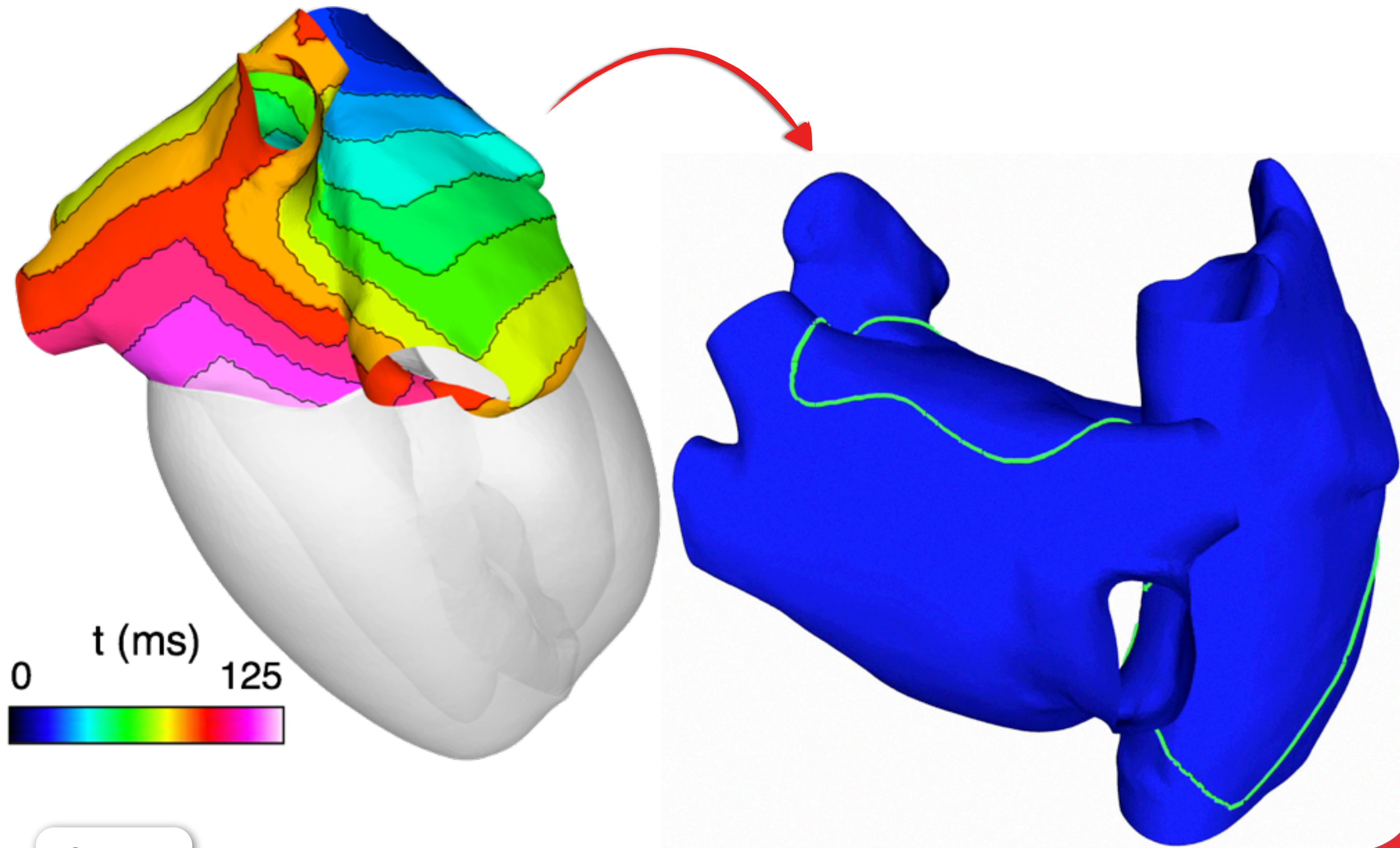
- Other application fields: Tumor growth, wildfire propagation



Electrocardiographic Imaging



From activation map to front propagation



Motivation

Traveling wave propagation phenomena modelled by reaction-diffusion systems to be compared with front propagation observations.

Other examples: Wild Fire propagation, Tumour growth, Chemotaxis

Objective: Data assimilation, namely state and parameter estimation, using observer

Outline

- Introduction to observer theory
- Observer for reaction-diffusion system based on front observations
 - ▶ Formulation
 - ▶ Mathematical analysis
 - ▶ Discretization & Numerical illustrations
 - ▶ Adaptive observers
 - ▶ Perspectives

Introduction to observer theory

Observer: definition

- Class of dynamical systems

$$\begin{cases} \dot{y}_{|\zeta,\omega}(t) = A(y_{|\zeta,\omega}(t), t) + B(t)\omega(t), & t \in \mathbb{R}^+ \\ y_{|\zeta,\omega}(0) = y_\diamond + \zeta \end{cases}$$

- Trajectory $\{y_\square(t), t \in \mathbb{R}^+\}$

- Observations

$$z(t) = C(y_\square(t), t) + \chi(t) \quad (\forall y, D(y, t) = z(t) - C(y, t))$$

- Observer or (state) estimator

$$\begin{cases} \dot{\hat{y}}(t) = A(\hat{y}, t) + G(\hat{y}, t)D(z, C(\hat{y}, t)), & t \in \mathbb{R}^+ \\ \hat{y}(0) = y_\diamond \end{cases}$$

- Objective here $\tilde{y}(t) = y_\square(t) - \hat{y}(t)$

$$\forall t, \omega(t) = 0, \chi(t) = 0 \Rightarrow \tilde{y}(t) \xrightarrow[t \rightarrow +\infty]{} 0$$

or at least $\exists$ contracting map $\beta \mid \|\tilde{y}(t)\| \leq \beta(\|\tilde{y}(s)\|, t - s), \quad \forall t \geq s.$

Observer: the wave example

- Wave equation $y(t) = (u(t) \quad \dot{u}(t))^T \in H^1(\Omega) \times L^2(\Omega)$

$$\begin{cases} \ddot{u} - c^2 \Delta u = f, & \text{in } \Omega \\ u = g, & \text{on } \partial\Omega \end{cases}$$

$$u(0) = u_\diamond + \zeta_u, \quad \dot{u}(0) = v_\diamond + \zeta_v$$

- Observations $z = \dot{u}_\omega$

- Observer

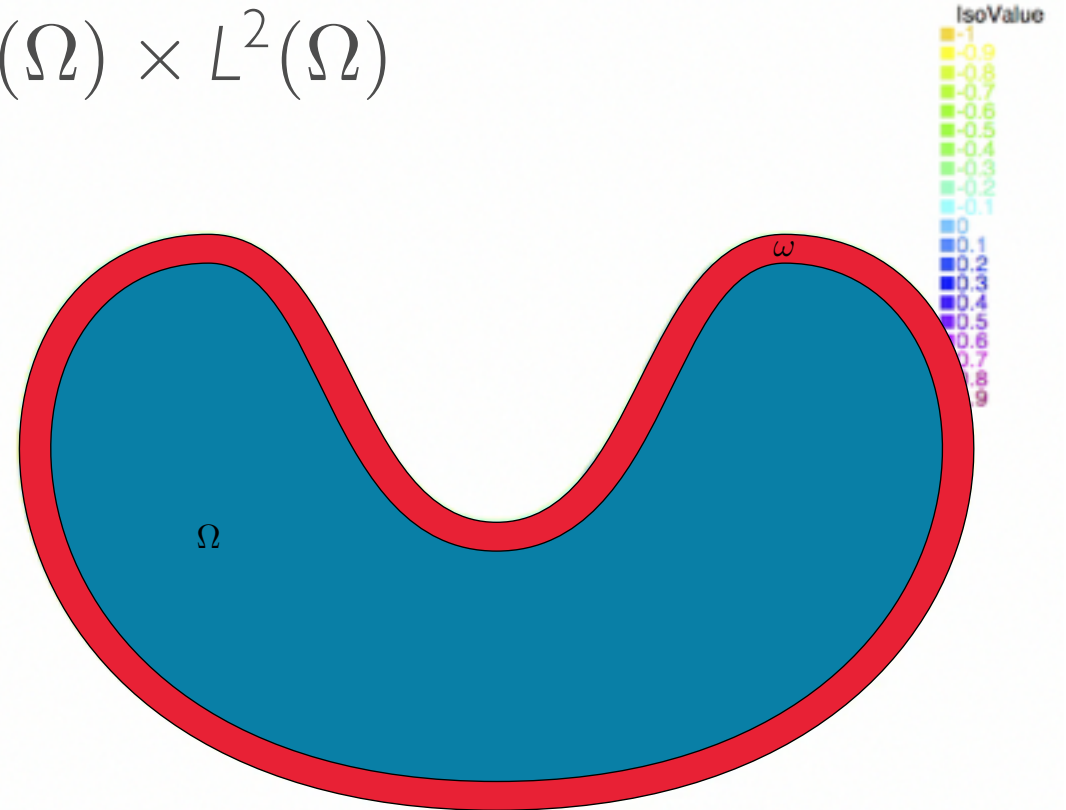
$$\begin{cases} \ddot{\hat{u}} - c^2 \Delta \hat{u} = f + \gamma(z - \chi_\omega \dot{\hat{u}}), & \text{in } \Omega \\ \hat{u} = g, & \text{on } \partial\Omega \end{cases}$$

$$\hat{u}(0) = u_\diamond, \quad \dot{\hat{u}}(0) = v_\diamond$$

- Error

$$\begin{cases} \ddot{\tilde{u}} + \chi_\omega \dot{\tilde{u}} - c^2 \Delta \tilde{u} = 0 & \text{in } \Omega \\ \tilde{u} = 0, & \text{on } \partial\Omega \end{cases}$$

$$\tilde{u}(0) = \zeta_u, \quad \dot{\tilde{u}}(0) = \zeta_v$$



Theorem. The observer converges to the target trajectory.
When the GCC is satisfied the convergence is exponential.

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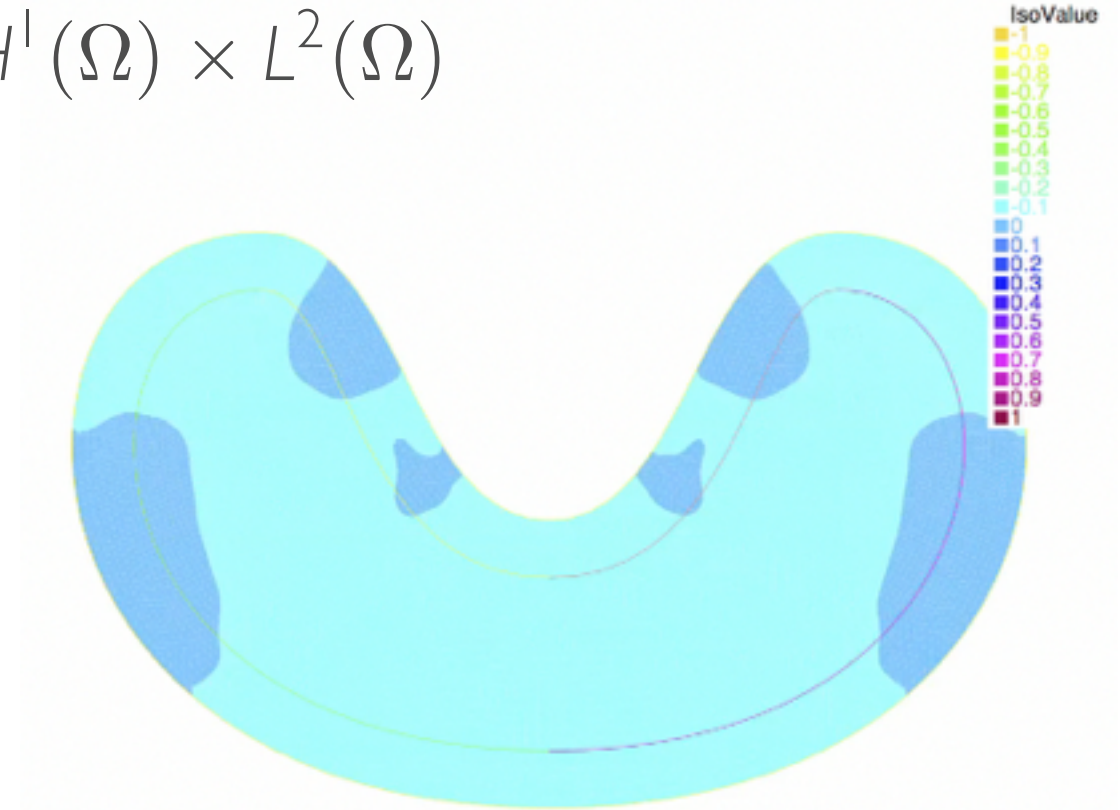
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Observer: the Saint-Venant system example

- Saint-Venant system

$$\begin{cases} \partial_t h + \partial_x hu = 0, & \text{in } \Omega \\ \partial_t(hu) + \partial_x(hu^2 + \frac{gh^2}{2}) = 0, & \text{in } \Omega \end{cases}$$

$$h(0) = h_\diamond + \zeta_h, \quad u(0) = u_\diamond + \zeta_u$$

- Observations $z = h$

- Observer

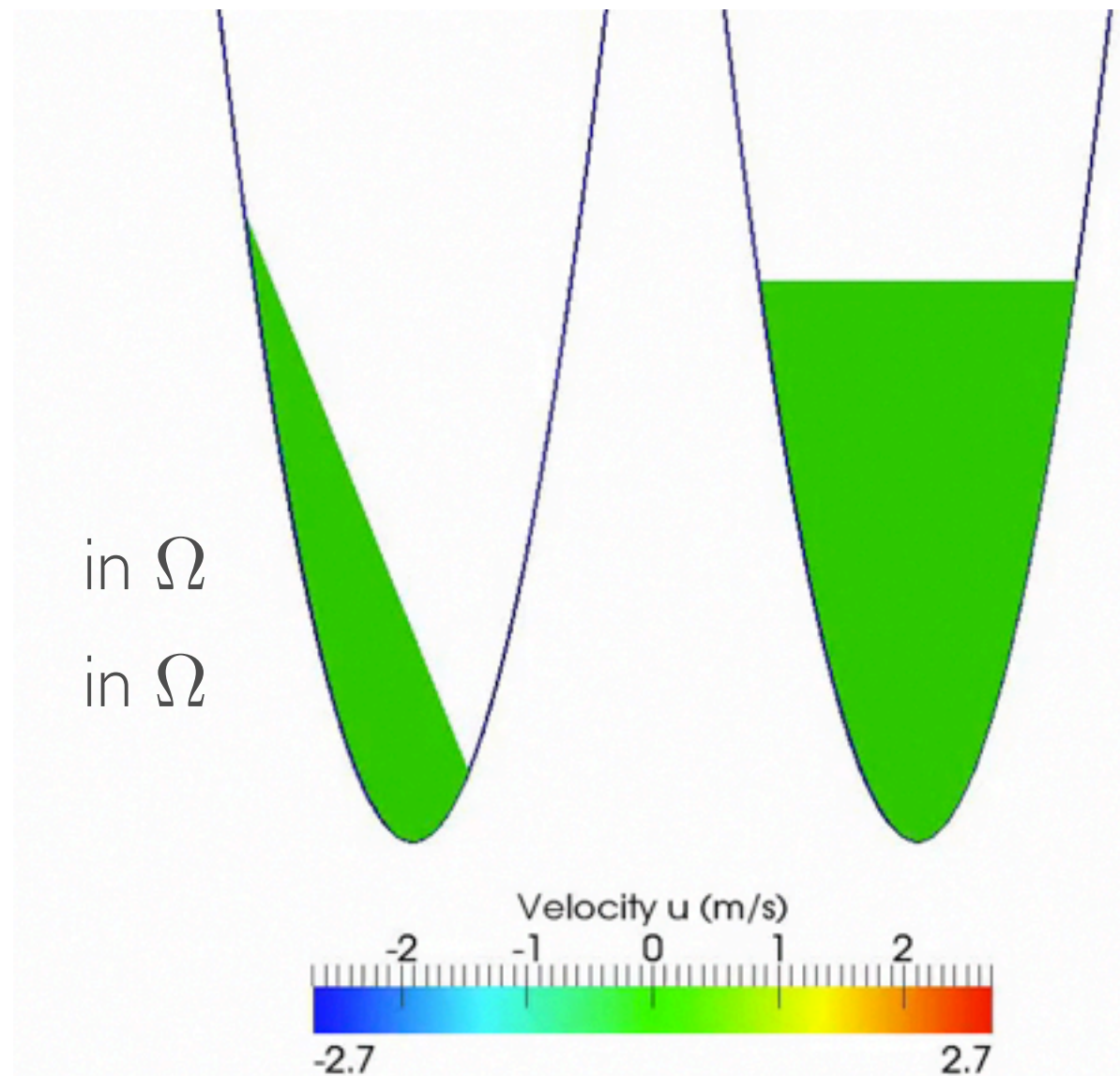
$$\begin{cases} \partial_t \hat{h} + \partial_x \hat{h} \hat{u} = \gamma(z - \hat{h}), & \text{in } \Omega \\ \partial_t(\hat{h} \hat{u}) + \partial_x(\hat{h} \hat{u}^2 + \frac{g \hat{h}^2}{2}) = \gamma \hat{u}(z - \hat{h}), & \text{in } \Omega \end{cases}$$

$$\hat{h}(0) = h_\diamond, \quad \hat{u}(0) = u_\diamond$$

- Error (non-linear)

$$\partial_t \hat{\mathcal{E}} + \partial_x \left(\hat{u} \left(\hat{\mathcal{E}} + g \frac{\hat{h}^2}{2} \right) \right) \leq \gamma(\mathcal{E} - \hat{\mathcal{E}})$$

or use kinetic representation



A.C. Boulanger, P. Moireau, B. Perthame, J. Sainte-Marie Data Assimilation for hyperbolic conservation laws — A Luenberger observer approach based on a kinetic description. Comm Math Sci 2015

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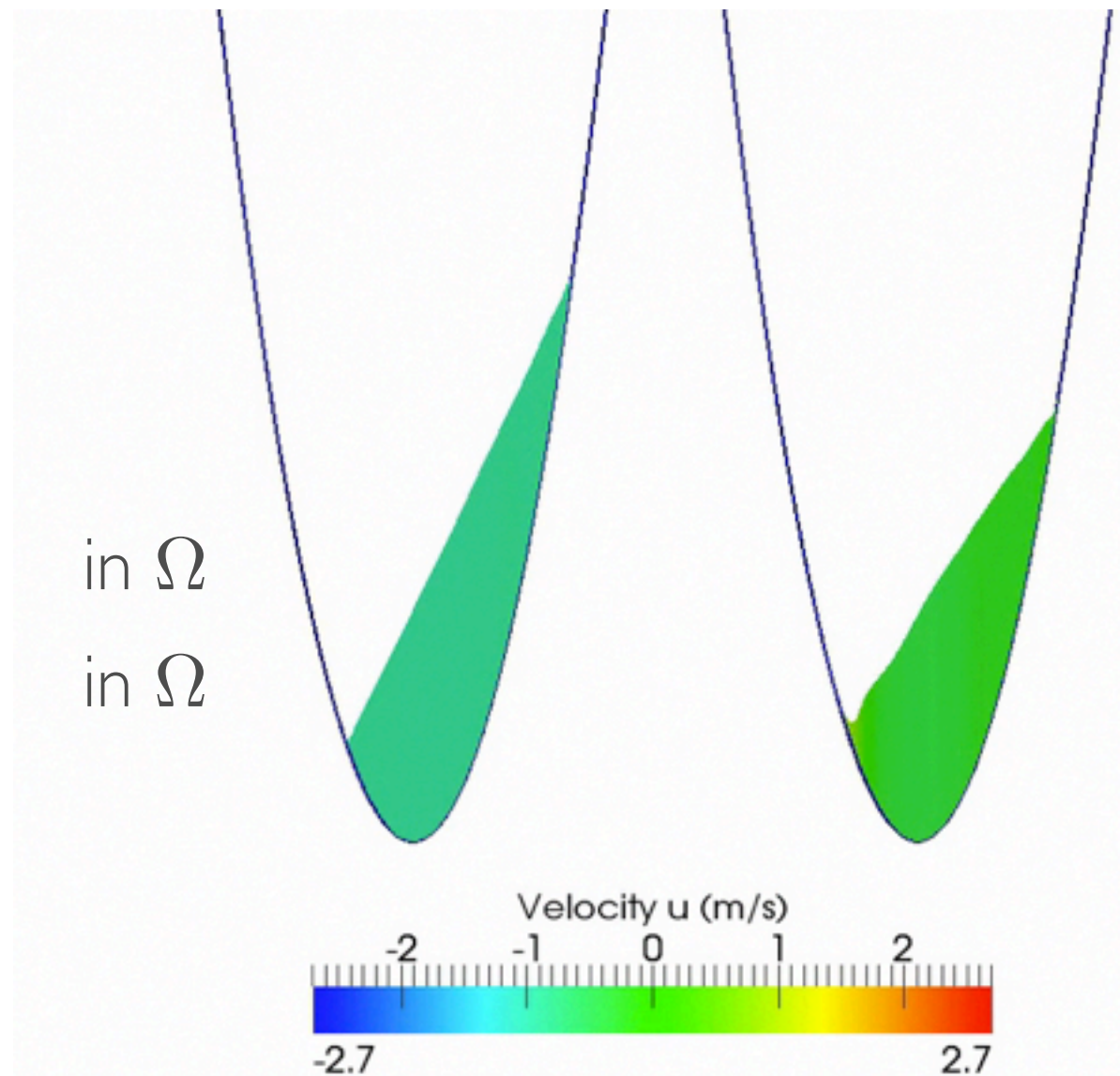
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- Observer

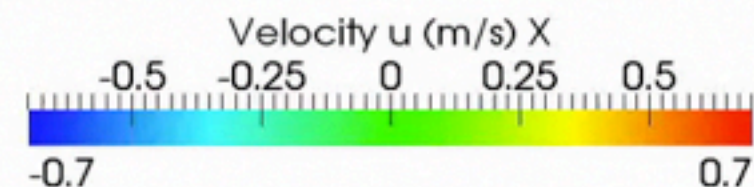
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- Observer

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Observer for reaction-diffusion system based on front observations

Part I: Observer formulation

The "depolarization" front

- Reaction diffusion model

$$\begin{cases} \partial_t u - \underline{\nabla} \cdot (D \underline{\nabla} u) &= kf(u), & \text{in } \mathcal{B} \times (0, T), \\ (D \underline{\nabla} u) \cdot \underline{n} &= 0, & \text{on } \partial \mathcal{B} \times (0, T), \\ u(\underline{x}, 0) &= u_0(\underline{x}), & \text{in } \mathcal{B}. \end{cases}$$

- We define c_{th} as the depolarisation constant, namely the depolarisation area is given by $\Omega_u(t) = \{\underline{x} \in \mathcal{B} \mid u(\underline{x}, t) > c_{th}\}$

- We can represents by a level-set $\phi_u(t)$

$$\Omega_u(t) = \{\underline{x} \in \mathcal{B} \mid \phi_u(t) > 0\}$$

$$\Gamma_u(t) = \{\underline{x} \in \mathcal{B} \mid \phi_u(t) = 0\}$$

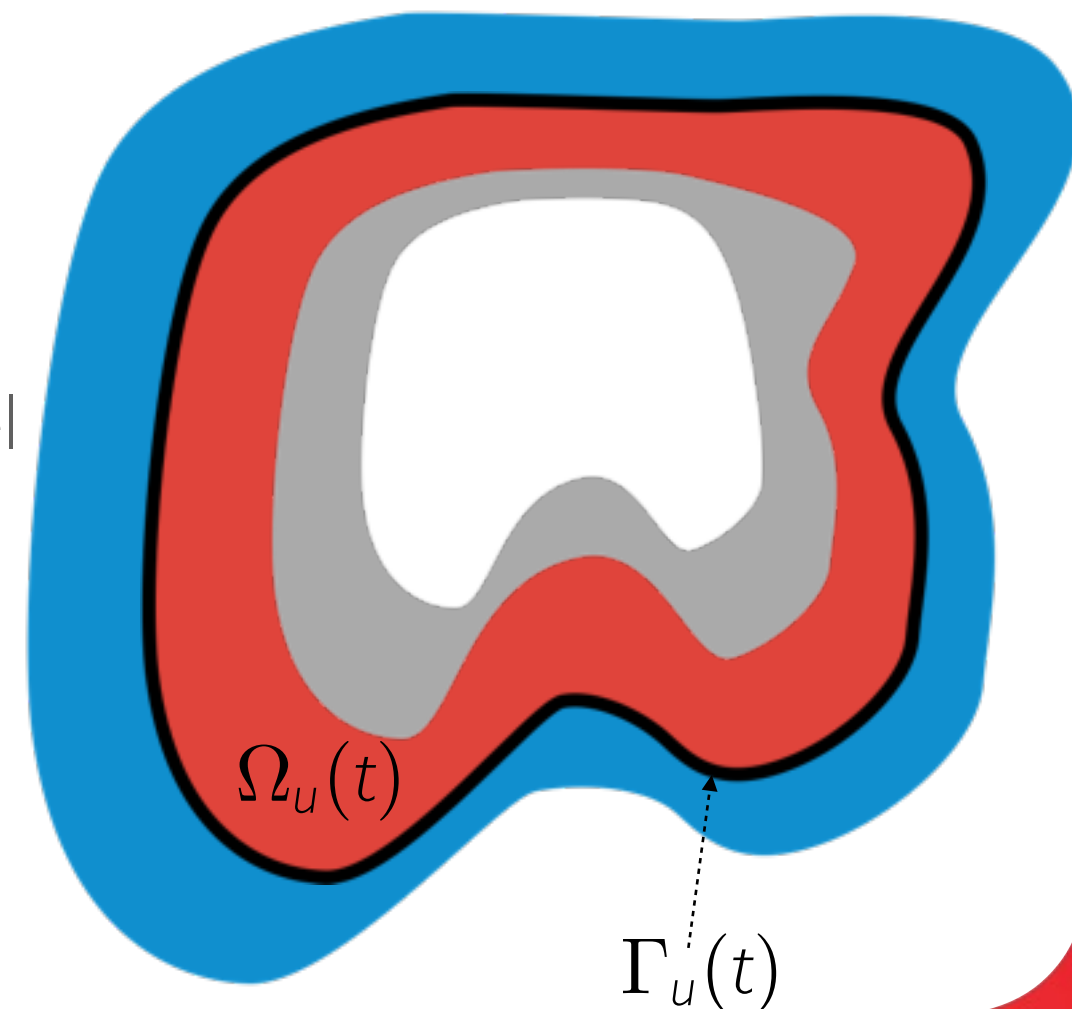
- And this level-set is solution of an Eikonal model

$$\dot{\phi}_u = |\underline{\nabla} \phi_u| \left(D \underline{\nabla} \cdot \left(\frac{\underline{\nabla} \phi_u}{|\underline{\nabla} \phi_u|} \right) + \sqrt{Dk} c_0 \right),$$

$$\text{in } \mathcal{B} \times (0, T)$$



Keener, An eikonal-curvature equation for action potential propagation in myocardium. 1991.



Eikonal equation from asymptotic analysis

- We define a time-dependent coordinate system (ξ_1, ξ_2, ξ_3) designed to follow the front over time

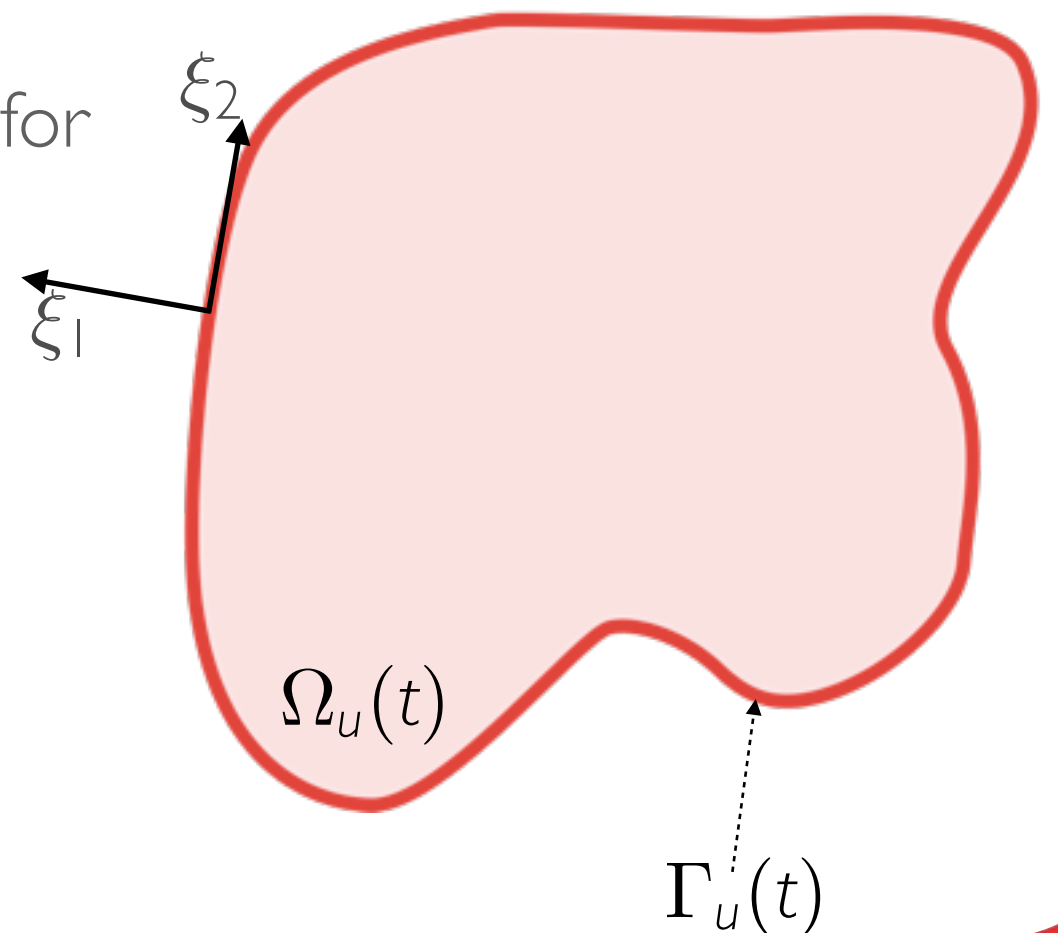
$$\partial_t u = \sigma a_{il} a_{ik} \frac{\partial^2 u}{\partial \xi_l \partial \xi_k} + \sigma \frac{\partial a_{il}}{\partial x_i} \frac{\partial u}{\partial \xi_l} + a_{jk} (\partial_t x_j) \frac{\partial u}{\partial \xi_k} + kf, \quad a_{ij} = \frac{\partial \xi_i}{\partial x_j}$$

- The front is sharp $O(\epsilon) \Rightarrow a_{il} = O(\epsilon^{-1})$,

$$\sigma |\underline{a}|^2 \frac{\partial^2 u}{\partial \xi_1^2} + \left(\sigma \underline{\nabla}_{\underline{x}} \cdot \underline{a} + \underline{a} \cdot \partial_t \underline{x} \right) \frac{\partial u}{\partial \xi_1} + kf(u) = 0.$$

which is a second order ODE where we seek for decreasing solutions $u'' + c_0 u' + f(u) = 0$

$$\begin{cases} kc_0 = \sigma \underline{\nabla}_{\underline{x}} \cdot \underline{a} + \underline{a} \cdot \partial_t \underline{x} \\ \underline{a} = -\sqrt{\frac{k}{\sigma}} \frac{\underline{\nabla}_{\underline{x}} \phi_u}{|\underline{\nabla}_{\underline{x}} \phi_u|} \\ \underline{\nabla} \phi_u \cdot \partial_t \underline{x} + \partial_t \phi_u = 0 \end{cases}$$



Observer for the Eikonal equation

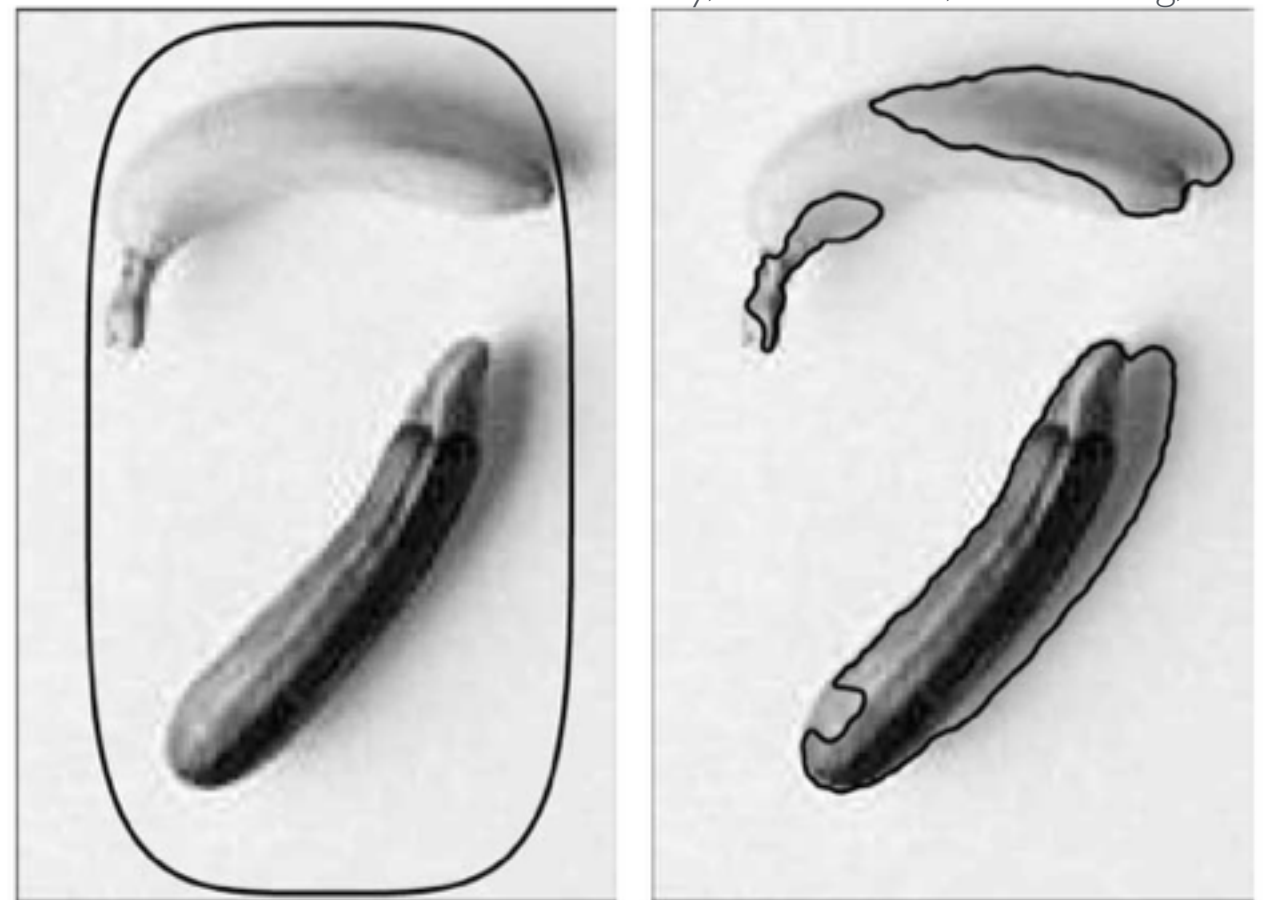
- In image processing theory (in segmentation) there are Eikonal models associated with data-fitting terms.

- For instance using the Mumford-Shah functional or its Chan-Vese version

$$\mathcal{J}(\phi) = \alpha|\Omega_\phi| + \beta|\Gamma_\phi| + \gamma \int_{\Omega_\phi} (z - C_1(\phi))^2 dx + \gamma \int_{\mathcal{B} \setminus \Omega_\phi} (z - C_2(\phi))^2$$

$$\left| \begin{array}{l} C_1 = \frac{1}{|\Omega_\phi|} \int_{\Omega_\phi} z dx \\ C_2 = \frac{1}{|\mathcal{B} \setminus \Omega_\phi|} \int_{\mathcal{B} \setminus \Omega_\phi} z dx \end{array} \right.$$

Courtesy, Hintermüller, M. and Ring, W.



Observer for the Eikonal equation

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where

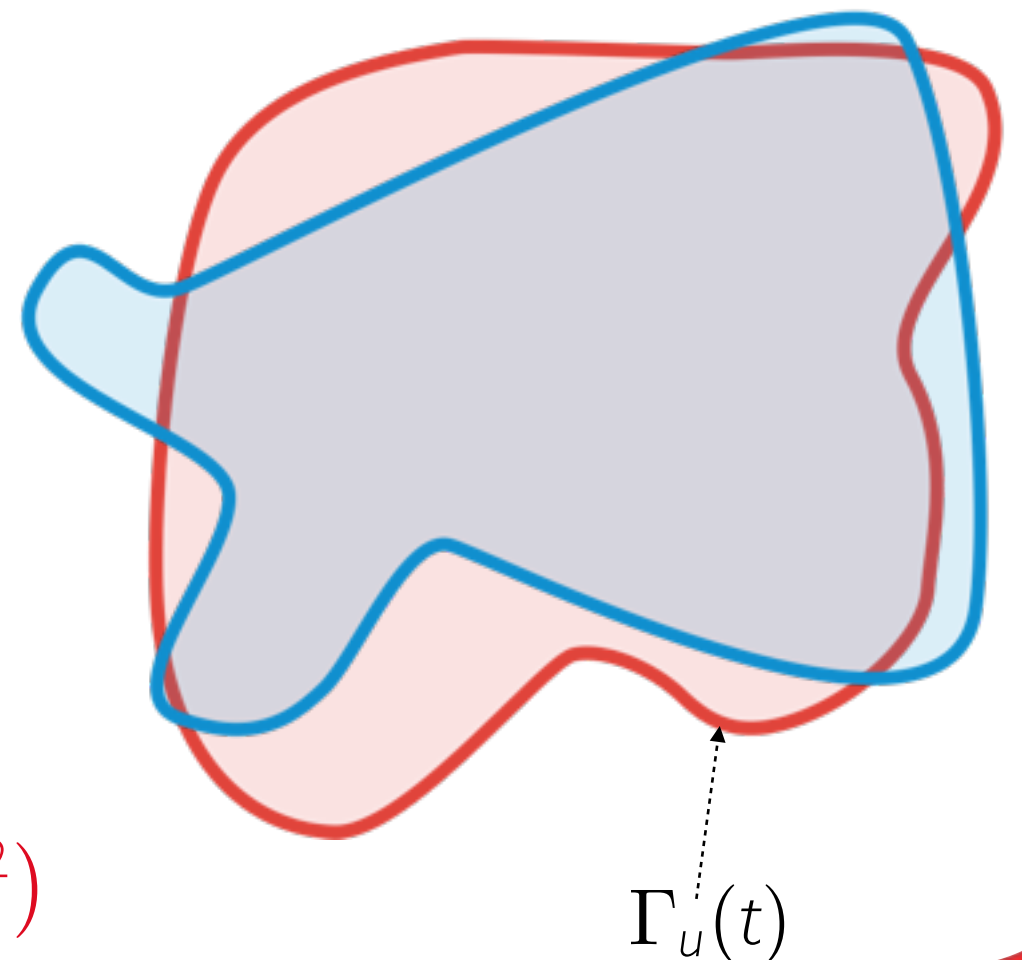
$$\begin{cases} C_1 = \frac{1}{|\Omega_\phi|} \int_{\Omega_\phi} z dx \\ C_2 = \frac{1}{|\mathcal{B} \setminus \Omega_\phi|} \int_{\mathcal{B} \setminus \Omega_\phi} z dx \end{cases}$$

- Minimization by gradient projection method

$$\dot{\phi}_u = -\Upsilon \nabla \mathcal{J}(\phi_u)$$

- Eikonal observer

$$\begin{aligned} \dot{\phi} = & -\alpha \delta(\phi) + \beta \delta(\phi) \underline{\nabla} \cdot \left(\frac{\underline{\nabla} \phi}{|\underline{\nabla} \phi|} \right) \\ & + \gamma \delta(\phi) (-(z - C_1)^2 + (z - C_2)^2) \end{aligned}$$



Observer for the Reaction-diffusion model

- Find a correction on the reaction-diffusion system so that the asymptotic analysis leads to the Eikonal observer

$$\dot{\hat{u}} = \underline{\nabla} \cdot (D \underline{\nabla} \hat{u}) + kf(\hat{u}) + \gamma \delta(\phi_{\hat{u}}) \frac{1}{|\underline{\nabla} \phi_{\hat{u}}|} \frac{\underline{\nabla} \phi_{\hat{u}}}{|\underline{\nabla} \phi_{\hat{u}}|} \cdot \underline{\nabla} u (-(z - C_1(\Omega_{\hat{u}}))^2 + (z - C_2(\Omega_{\hat{u}}))^2)$$

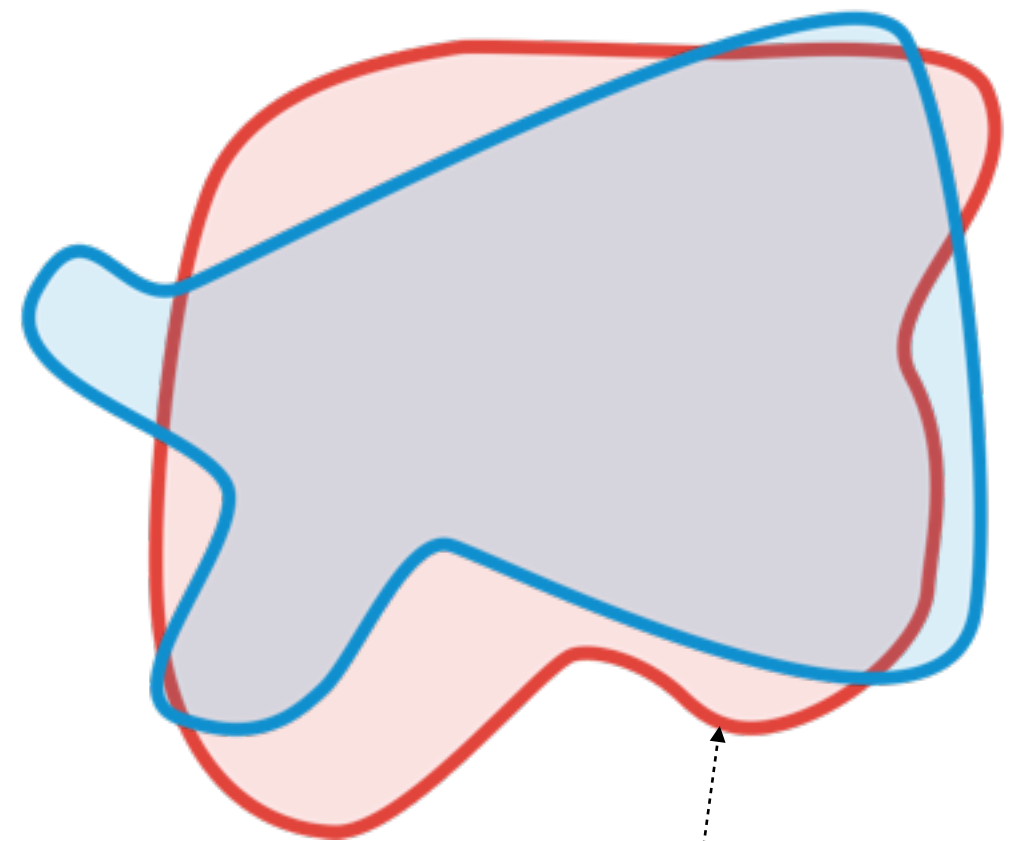
$$\gamma(|\underline{\nabla} \hat{\phi}_{\hat{u}}|) = \frac{\gamma}{|\underline{\nabla} \hat{\phi}_{\hat{u}}|}$$

- Eikonal observer

$$\dot{\hat{\phi}} = \sqrt{kDc_0} |\underline{\nabla} \hat{\phi}_{\hat{u}}| + |\underline{\nabla} \hat{\phi}_{\hat{u}}| \underline{\nabla} \cdot \left(\frac{\underline{\nabla} \hat{\phi}_{\hat{u}}}{|\underline{\nabla} \hat{\phi}_{\hat{u}}|} \right)$$

$$+ \gamma(|\underline{\nabla} \hat{\phi}_{\hat{u}}|) \delta(\hat{\phi}_{\hat{u}}) (-(z - C_1)^2 + (z - C_2)^2)$$

$\Gamma_u(t)$



Observer for the Reaction-diffusion model

- Find a correction on the reaction-diffusion system so that the asymptotic analysis leads to the Eikonal observer

$$\begin{aligned}\dot{\hat{u}} = & \underline{\nabla} \cdot (D \underline{\nabla} \hat{u}) + kf(\hat{u}) \\ & + \gamma \delta(\hat{u} - c_{th}) \left(-(z - C_1(\Omega_{\hat{u}}))^2 + (z - C_2(\Omega_{\hat{u}}))^2 \right)\end{aligned}$$

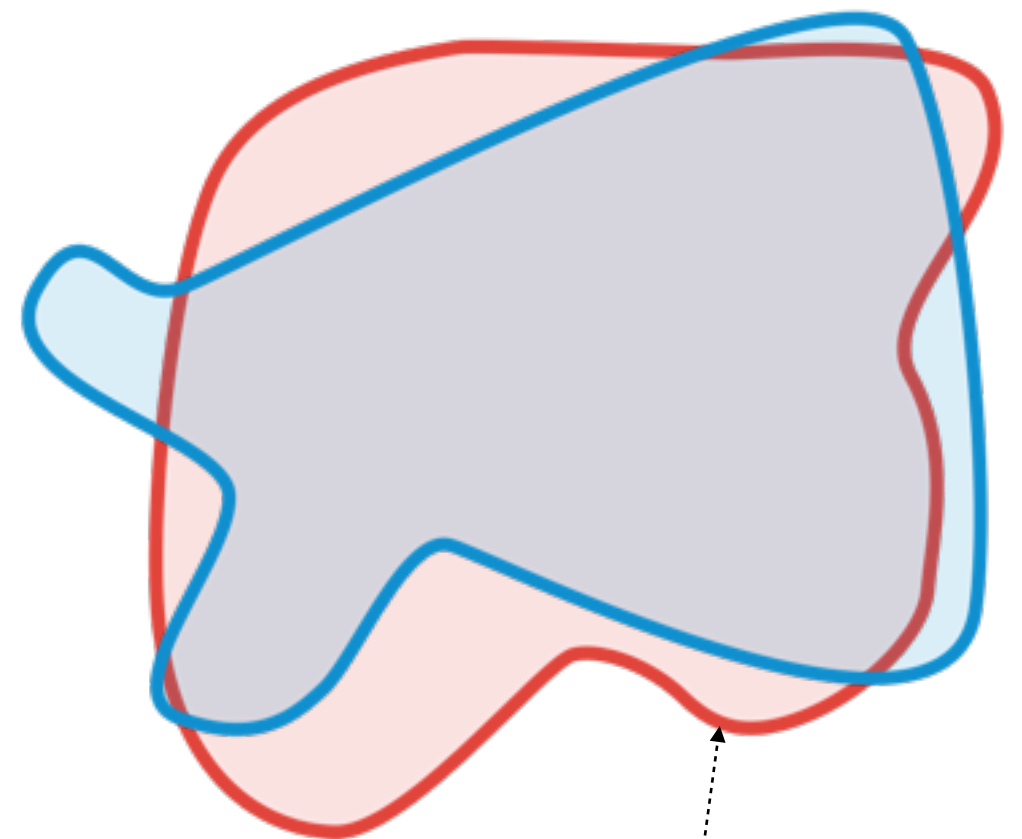
$$\phi_{\hat{u}} = \hat{u} - c_{th}$$

- Eikonal observer

$$\dot{\hat{\phi}} = \sqrt{kDc_0} |\underline{\nabla} \hat{\phi}_{\hat{u}}| + |\underline{\nabla} \hat{\phi}_{\hat{u}}| \underline{\nabla} \cdot \left(\frac{\underline{\nabla} \hat{\phi}_{\hat{u}}}{|\underline{\nabla} \hat{\phi}_{\hat{u}}|} \right)$$

$$+ \gamma (|\underline{\nabla} \hat{\phi}_{\hat{u}}|) \delta(\hat{\phi}_{\hat{u}}) \left(-(z - C_1)^2 + (z - C_2)^2 \right)$$

$\Gamma_u(t)$



Observer for reaction-diffusion system based on front observations

Part II: Elements of analysis

Error Analysis (I)

- System

$$\begin{cases} \partial_t u - \underline{\nabla} \cdot (D \underline{\nabla} u) &= kf(u), & \mathcal{B} \times (0, T), \\ (D \underline{\nabla} u) \cdot \underline{n} &= 0, & \partial \mathcal{B} \times (0, T), \\ u(\underline{x}, 0) &= u_\diamond(\underline{x}) + \zeta(\underline{x}), & \mathcal{B}. \end{cases}$$

- Observer

$$\begin{cases} \partial_t \hat{u} - \underline{\nabla} \cdot (D \underline{\nabla} \hat{u}) &= kf(\hat{u}) \\ + \gamma \hat{\delta}(\Gamma_{\hat{u}}, \underline{x}) \frac{1}{|\underline{\nabla} \hat{u}|} \left(-(z - C_1(\Omega_{\hat{u}}))^2 + (z - C_2(\Omega_{\hat{u}}))^2 \right), & \mathcal{B} \times (0, T), \\ (D \underline{\nabla} \hat{u}) \cdot \underline{n} &= 0, & \partial \mathcal{B} \times (0, T), \\ \hat{u}(\underline{x}, 0) &= u_\diamond(\underline{x}), & \mathcal{B}. \end{cases}$$

- Error ?

$$\begin{cases} \partial_t \tilde{u} - \underline{\nabla} \cdot (D \underline{\nabla} \tilde{u}) &= k(f(u) - f(\hat{u})) \\ + \gamma \hat{\delta}(\Gamma_{\hat{u}}, \underline{x}) \frac{1}{|\underline{\nabla} \hat{u}|} \left((z - C_1(\Omega_{\hat{u}}))^2 - (z - C_2(\Omega_{\hat{u}}))^2 \right), & \mathcal{B} \times (0, T), \\ (D \underline{\nabla} \tilde{u}) \cdot \underline{n} &= 0, & \partial \mathcal{B} \times (0, T), \\ \tilde{u}(\underline{x}, 0) &= \zeta(\underline{x}), & \mathcal{B}. \end{cases}$$

- or Linearized error around the target trajectory ?

Error Analysis (2)

- Error in variational form

$$\int_{\mathcal{B}} \partial_t \tilde{u} v \, dx = - \int_{\mathcal{B}} D \vec{\nabla} \tilde{u} \cdot \vec{\nabla} v \, dx + \int_{\mathcal{B}} k(f(u) - f(\hat{u})) v \, dx$$

$$+ \gamma \int_{\Gamma_{\hat{u}}} \frac{1}{|\vec{\nabla} \hat{u}|} \left((z - C_1(\Omega_{\hat{u}}))^2 - (z - C_2(\Omega_{\hat{u}}))^2 \right) v \, d\Gamma$$

Stabilizing term ?

- Energy

$$\int_{\mathcal{B}} \partial_t \tilde{u}^2 \, dx = - \int_{\mathcal{B}} D \vec{\nabla} \tilde{u} \cdot \vec{\nabla} \tilde{u} \, dx + \int_{\mathcal{B}} k(f(u) - f(\hat{u})) v \, dx$$

$$+ \gamma \int_{\Gamma_{\hat{u}}} \frac{1}{|\vec{\nabla} \hat{u}|} \left((z - C_1(\Omega_{\hat{u}}))^2 - (z - C_2(\Omega_{\hat{u}}))^2 \right) \tilde{u} \, d\Gamma$$

Dissipative ?

- Dissipative property at least around the target trajectory

$$\mathcal{Q}_{\hat{u}}(\tilde{u}) = \int_{\Gamma_{\hat{u}}} \frac{\gamma}{|\vec{\nabla} \hat{u}|} \left((z - C_1(\Omega_{\hat{u}}))^2 - (z - C_2(\Omega_{\hat{u}}))^2 \right) \tilde{u} \, d\Gamma.$$

Shape optimisation based computations

Assuming that the contour is closed

$$\begin{aligned}
 (d_{\hat{u}} \mathcal{Q}_{\hat{u}}(\chi); \psi) &= \int_{\Gamma_{\hat{u}}} \kappa \gamma(|\underline{\nabla} \hat{u}|) \left((z_u - C_1(\Omega_{\hat{u}}))^2 - (z_u - C_2(\Omega_{\hat{u}}))^2 \right) \chi \frac{\psi}{|\underline{\nabla} \hat{u}|} d\Gamma \\
 &\quad - \int_{\Gamma_{\hat{u}}} \gamma'(|\underline{\nabla} \hat{u}|) \left((z_u - C_1(\Omega_{\hat{u}}))^2 - (z_u - C_2(\Omega_{\hat{u}}))^2 \right) \chi \partial_n \psi d\Gamma \\
 &\quad - \int_{\Gamma_{\hat{u}}} \gamma'(|\underline{\nabla} \hat{u}|) \partial_n^2 \hat{u} \left((z_u - C_1(\Omega_{\hat{u}}))^2 - (z_u - C_2(\Omega_{\hat{u}}))^2 \right) \chi \frac{\psi}{|\underline{\nabla} \hat{u}|} d\Gamma \\
 &\quad + 2 \int_{\Gamma_{\hat{u}}} \gamma(|\underline{\nabla} \hat{u}|) \partial_n z_u (C_2(\Omega_{\hat{u}}) - C_1(\Omega_{\hat{u}})) \chi \frac{\psi}{|\underline{\nabla} \hat{u}|} d\Gamma \\
 &\quad - 2 \frac{1}{|\Omega_{\hat{u}}|} \int_{\Gamma_{\hat{u}}} (z_u - C_1(\Omega_{\hat{u}})) \frac{\psi}{|\underline{\nabla} \hat{u}|} d\Gamma \times \int_{\Gamma_{\hat{u}}} \gamma(|\underline{\nabla} \hat{u}|) (z_u - C_1(\Omega_{\hat{u}})) \chi d\Gamma \\
 &\quad - 2 \frac{1}{|\mathcal{B} \setminus \overline{\Omega_{\hat{u}}}|} \int_{\Gamma_{\hat{u}}} (z_u - C_2(\Omega_{\hat{u}})) \frac{\psi}{|\underline{\nabla} \hat{u}|} d\Gamma \times \int_{\Gamma_{\hat{u}}} \gamma(|\underline{\nabla} \hat{u}|) (z_u - C_2(\Omega_{\hat{u}})) \chi d\Gamma \\
 &\quad + \int_{\Gamma_{\hat{u}}} \gamma(|\underline{\nabla} \hat{u}|) \left((z_u - C_1(\Omega_{\hat{u}}))^2 - (z_u - C_2(\Omega_{\hat{u}}))^2 \right) \partial_n \chi \frac{\psi}{|\underline{\nabla} \hat{u}|} d\Gamma,
 \end{aligned}$$



Delfour. Zolésio.
Shapes and Geometry.
2nd edition. **2011.**



Hintermüller, Ring. *An Inexact Newton-CG-Type Active Contour Approach for the Minimization of the Mumford-Shah Functional.* **2004.**

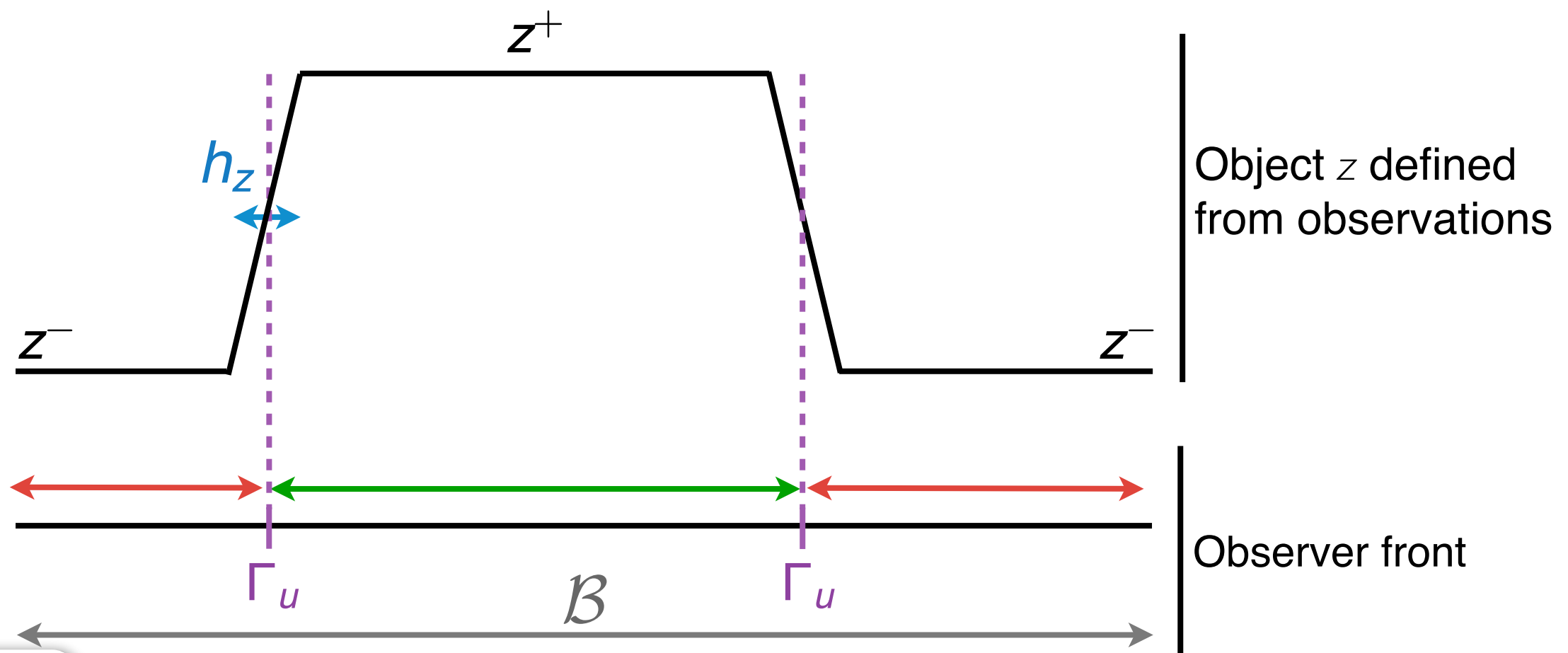
$$\begin{aligned}
 (d_{\hat{u}} \mathcal{Q}_u(\tilde{u}); \tilde{u}) &= 2(C_2(\Omega_u) - C_1(\Omega_u)) \int_{\Gamma_u} \gamma(|\underline{\nabla} u|) \partial_n z_u \frac{\tilde{u}}{|\underline{\nabla} u|} \tilde{u} d\Gamma \\
 &\quad - 2 \left(\frac{C_2(\Omega_u) - C_1(\Omega_u)}{2} \right)^2 \left(\frac{1}{|\Omega_u|} + \frac{1}{|\mathcal{B} \setminus \overline{\Omega_u}|} \right) \int_{\Gamma_u} \frac{\tilde{u}}{|\underline{\nabla} u|} d\Gamma \times \int_{\Gamma_u} \gamma(|\underline{\nabla} u|) \tilde{u} d\Gamma
 \end{aligned}$$

Stabilization property

Proposition. If the condition

$$\frac{1}{h_z} \geq \frac{1}{4} \left(\frac{1}{|\Omega_u|} + \frac{1}{|\mathcal{B} \setminus \overline{\Omega_u}|} \right) \left(\int_{\Gamma_u} \gamma(|\nabla u|) |\vec{\nabla} u| d\Gamma \times \int_{\Gamma_u} \frac{1}{\gamma(|\nabla u|) |\vec{\nabla} u|} d\Gamma \right)^{\frac{1}{2}}$$

is satisfied then the observer effect is energy-decreasing in the linearized error



Stabilization property

Proposition. If the contrast condition

$$\frac{1}{h_z} \geq \frac{1}{4} \left(\frac{1}{|\Omega_u|} + \frac{1}{|\mathcal{B} \setminus \overline{\Omega_u}|} \right) |\Gamma_u|$$

is satisfied then the observer effect is energy-decreasing in the linearized error

• Proof

$$\left(\int_{\Gamma_u} \gamma(|\underline{\nabla} u|) |\underline{\nabla} u| d\Gamma \times \int_{\Gamma_u} \frac{1}{\gamma(|\underline{\nabla} u|) |\underline{\nabla} u|} d\Gamma \right)^{\frac{1}{2}} \geq \int_{\Gamma_u} \frac{\gamma(|\underline{\nabla} u|) |\underline{\nabla} u|}{\gamma(|\underline{\nabla} u|) |\underline{\nabla} u|} d\Gamma = |\Gamma_u|,$$

and equality with

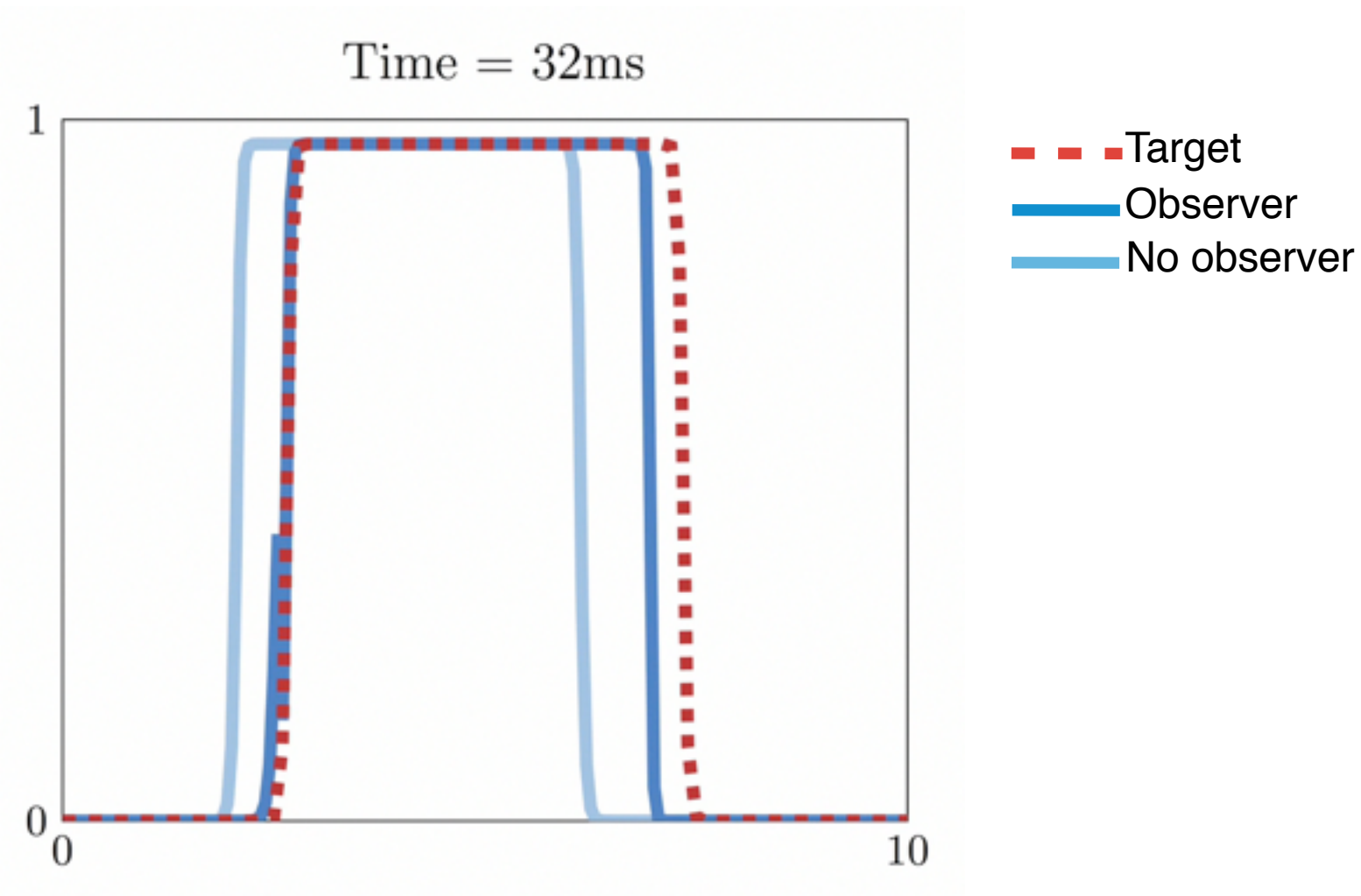
$$\gamma(|\underline{\nabla} u|) = \frac{1}{|\underline{\nabla} u|}.$$

Observer for reaction-diffusion system based on front observations

Part III: Numerical illustrations

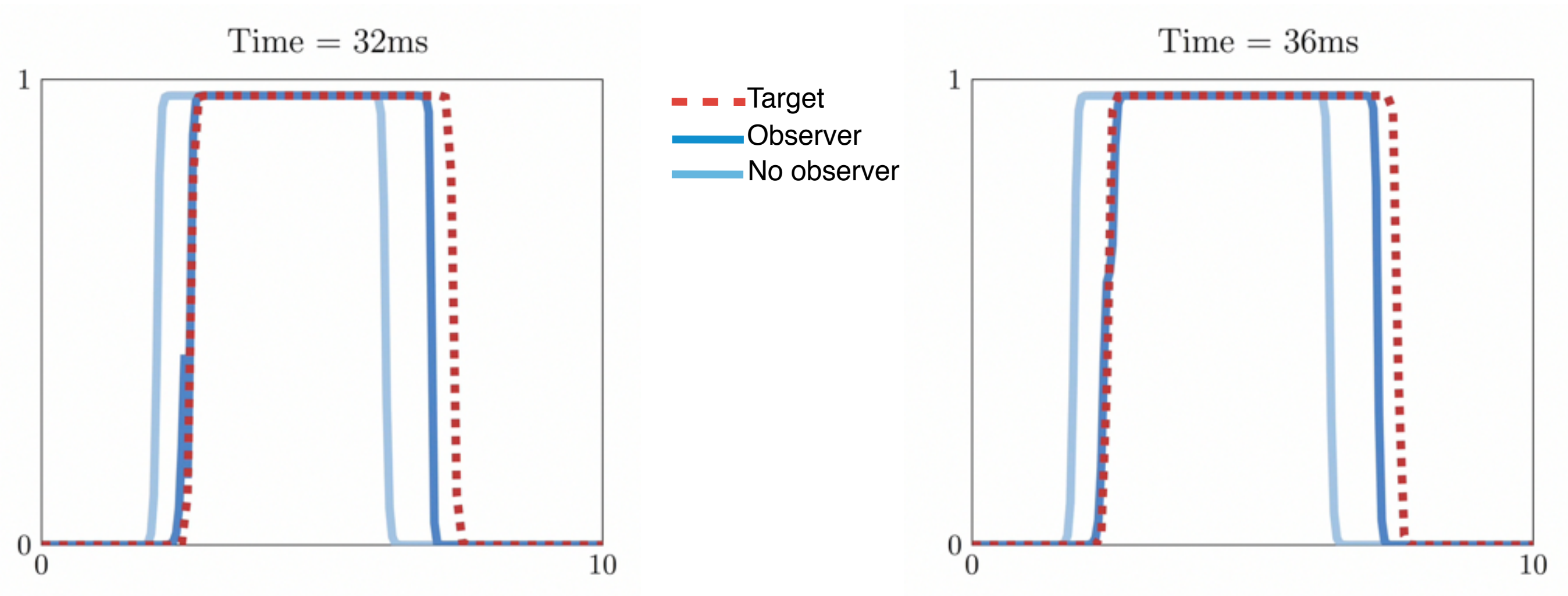
ID Case

- Reaction diffusion system of Mitchell-Schaeffer type (similar to Fitzugh-Nagumo)



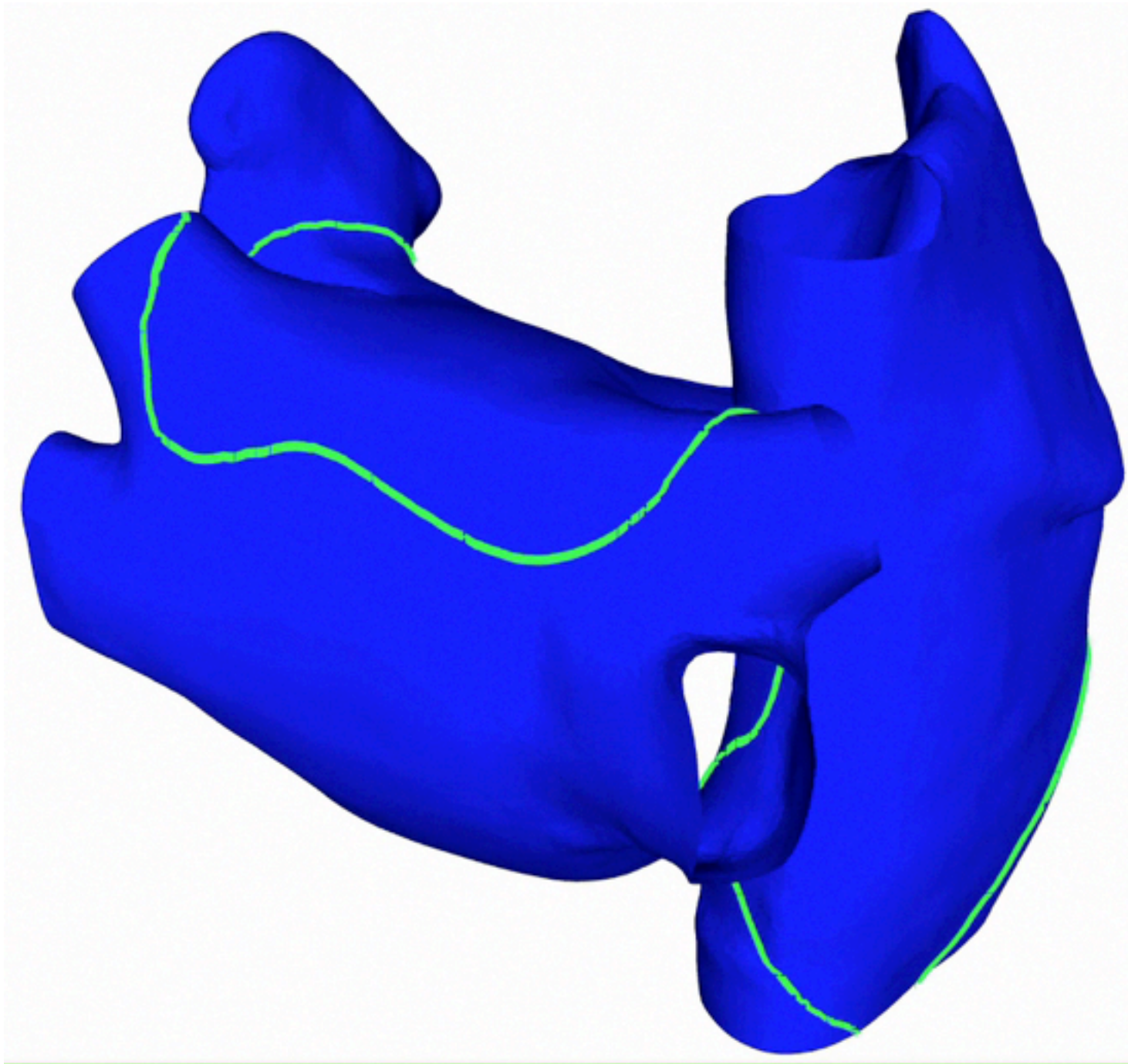
ID Case

- Reaction diffusion system of Mitchell-Schaeffer type (similar to Fitzugh-Nagumo)
- When time-sampling the data



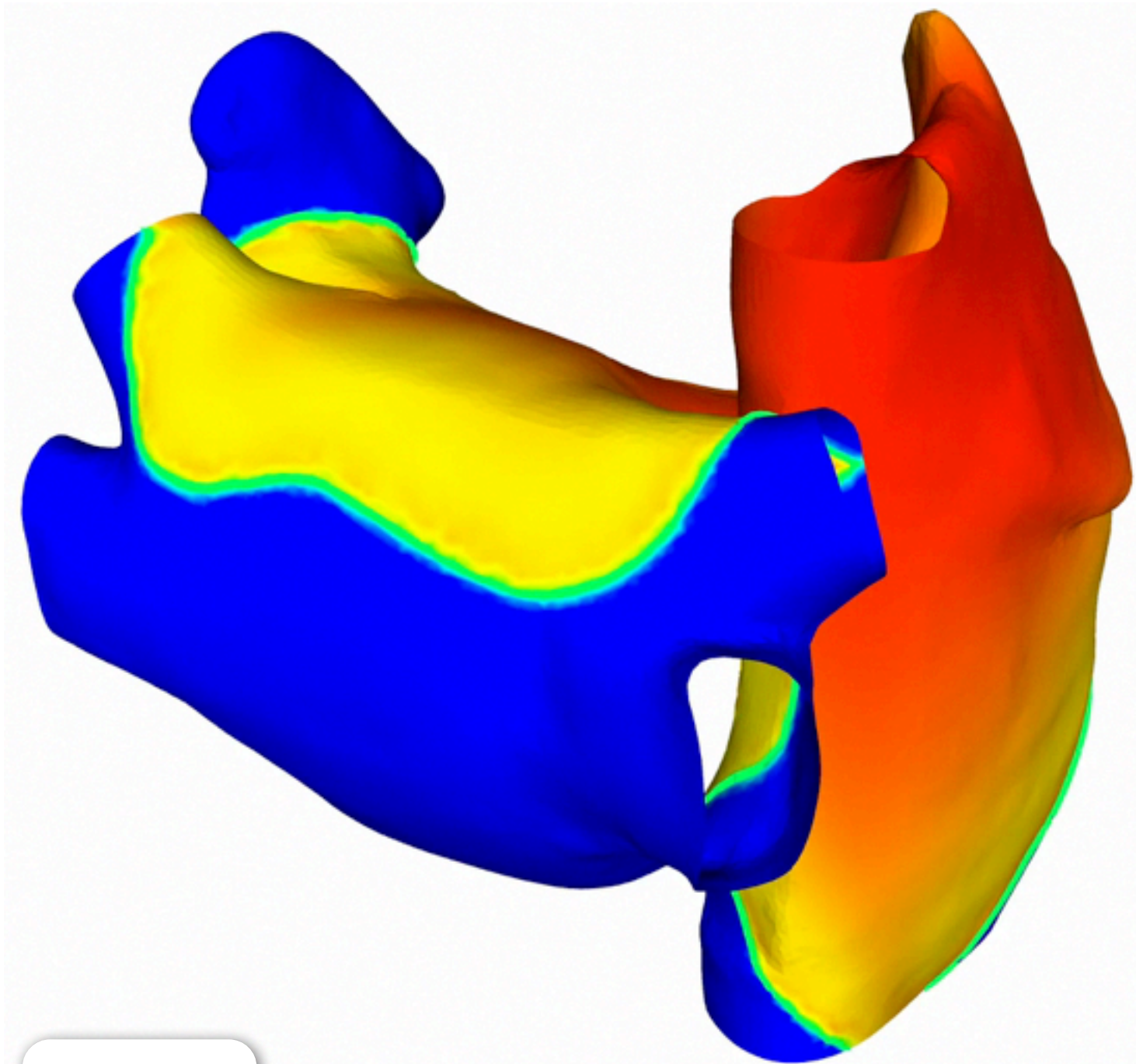
3D (Surfacic) Case

- Surface bidomain model with Courtemanche-Ramirez-Nattel ionic model specifically adapted to the atria (12 ionic currents and 20 other variables)



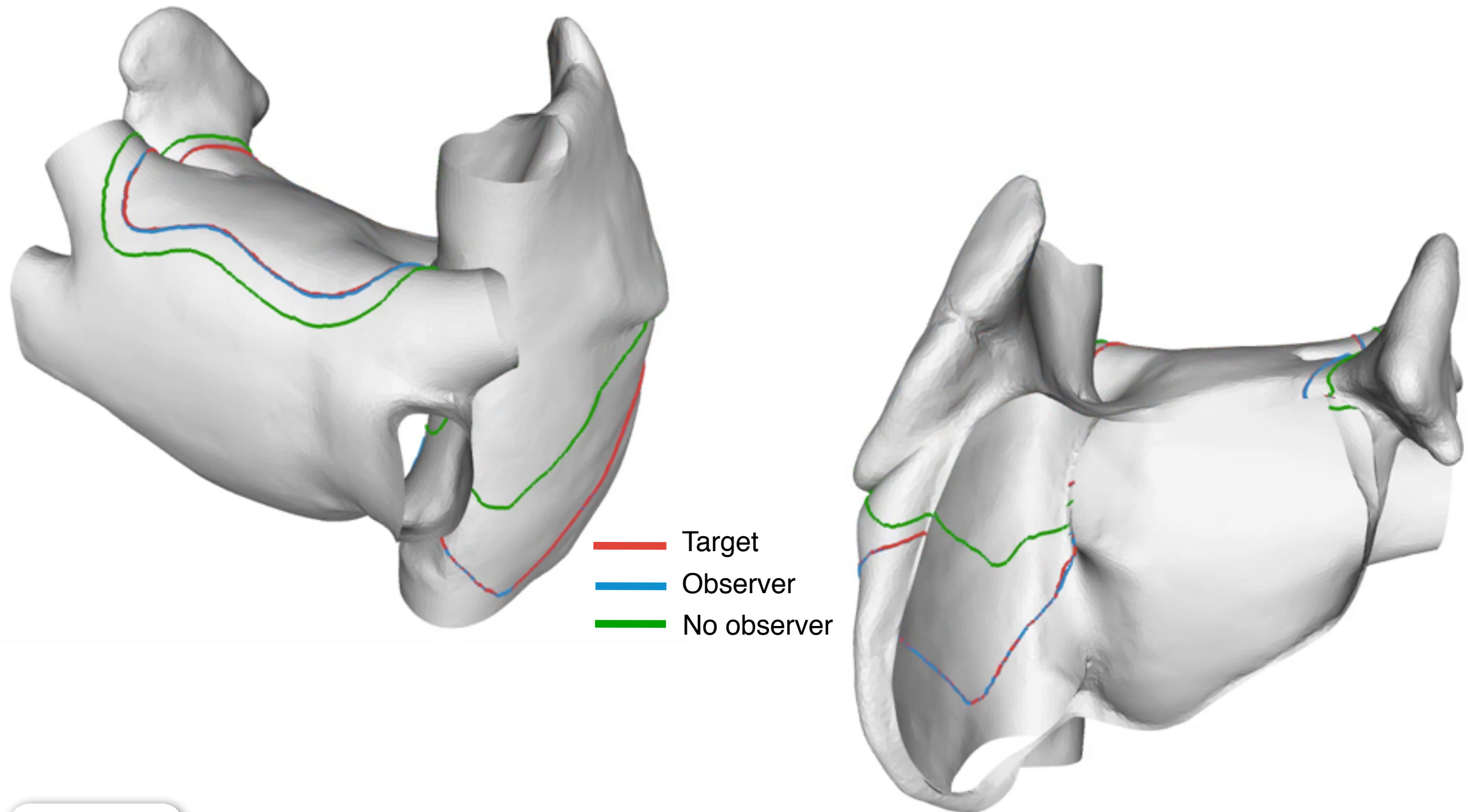
3D (Surfacic) Case

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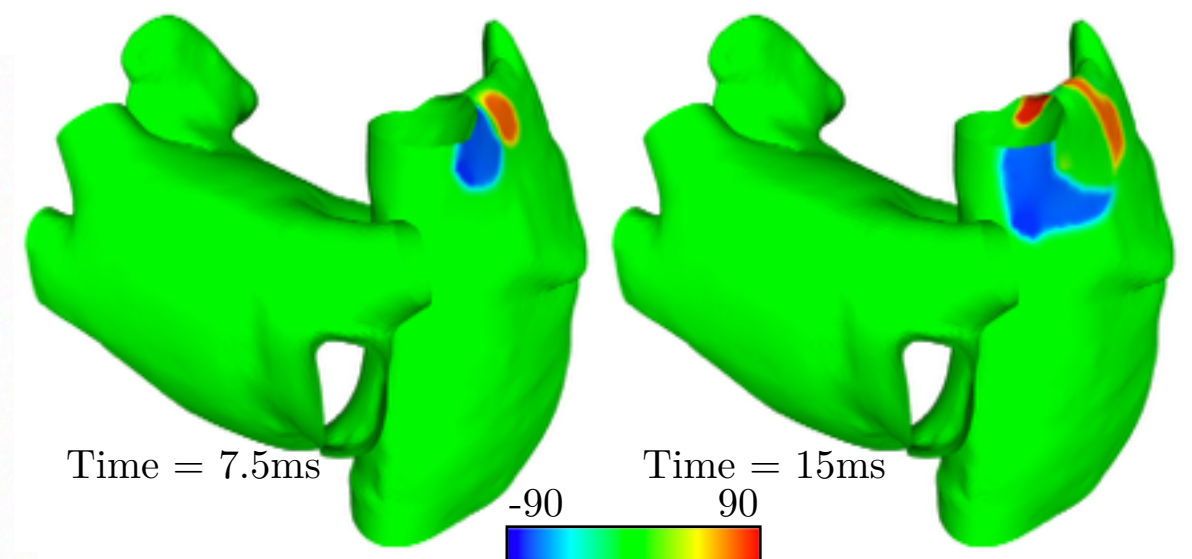
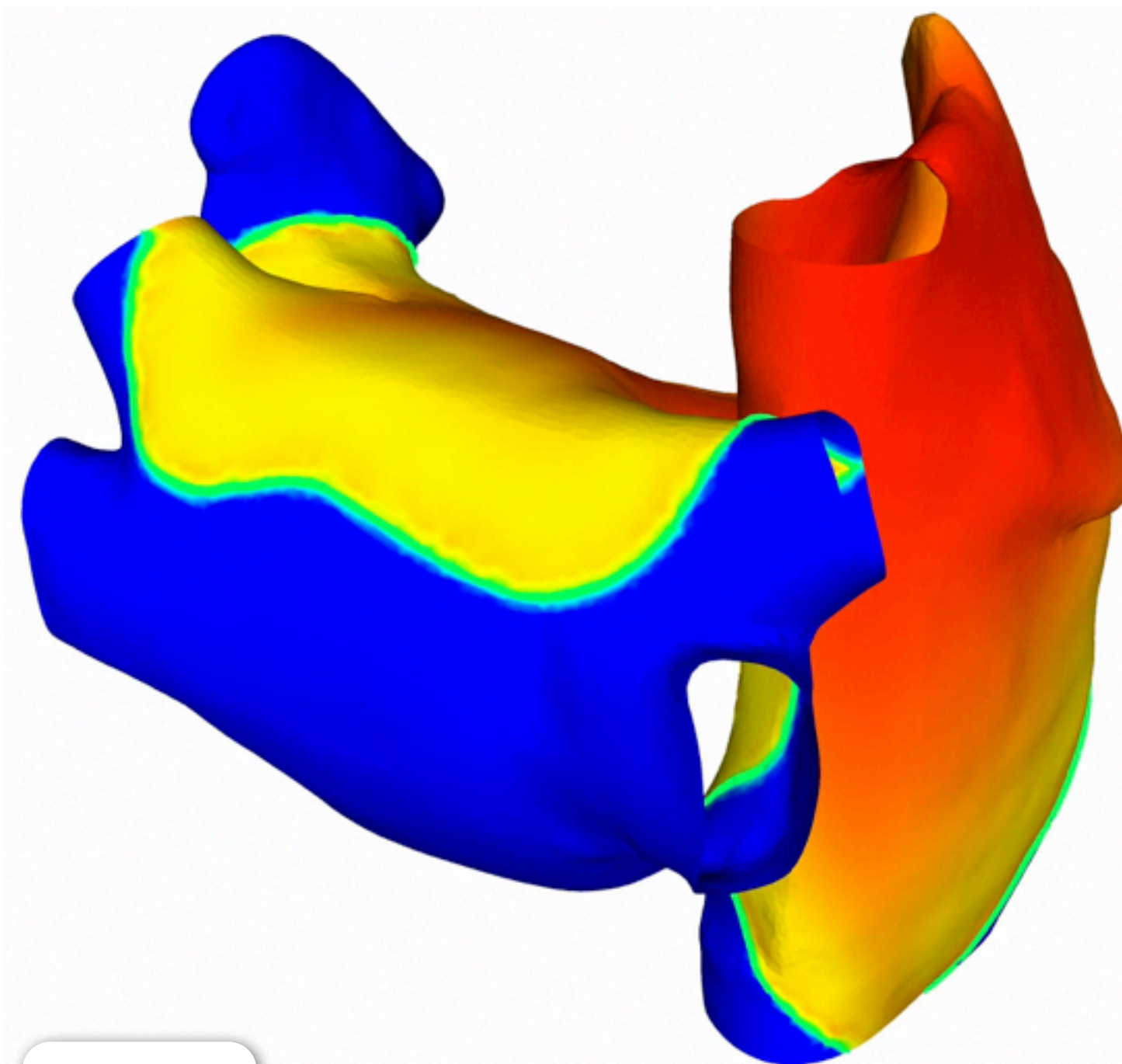
3D (Surfacic) Case

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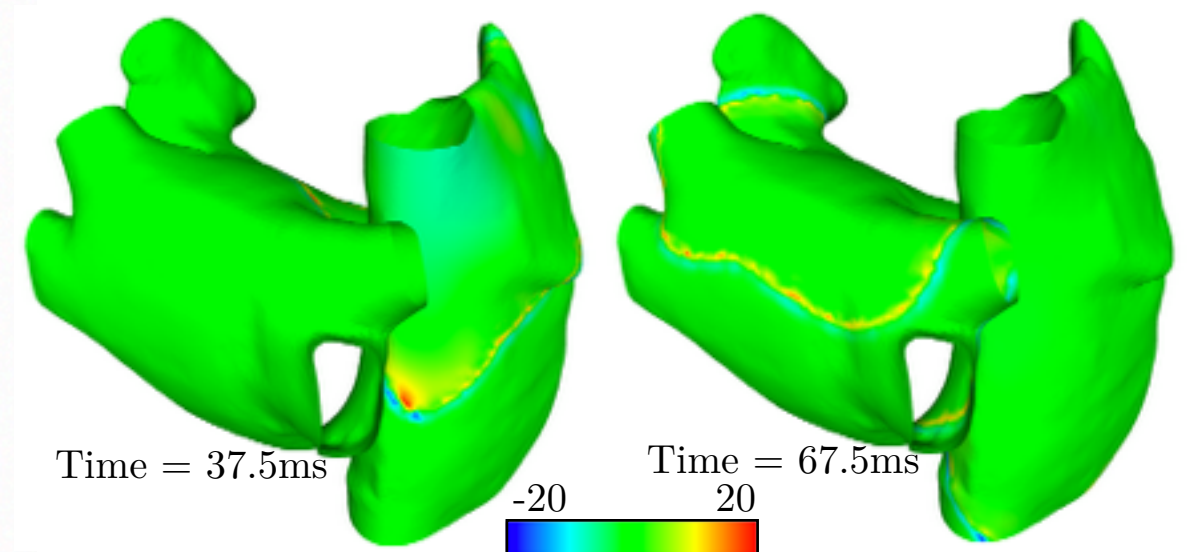


3D (Surfacic) Case

- Surface bidomain model with Courtemanche-Ramirez-Nattel ionic model specifically adapted to the atria (12 ionic currents and 20 other variables)



Errors L^∞ - norm



Observer for reaction-diffusion system based on front observations

Part IV: Adaptive Observer for Parameter identification

RoEKF applied to the parameter space

Principle: Apply a Kalman-based filter of reduced rank to the parameters space where the uncertainties are now concentrated.

- System
$$\begin{cases} \dot{y} = A(y, \vartheta) \\ \dot{\vartheta} = 0 \end{cases}$$
- Observer
$$\begin{cases} \dot{\hat{y}} = A(\hat{y}, \hat{\vartheta}) + \gamma dC^*(z - C(\hat{y})) + L\hat{\vartheta} \\ \dot{\hat{\vartheta}} = U^{-1}L^*dC^*M(z - C(\hat{y})) \\ \dot{U} = L^*dC^*M dC L \\ \dot{L} = (\partial_y A(\hat{y}, \hat{\vartheta}) - \gamma dC^* dC)L + \partial_{\vartheta} A(\hat{y}, \hat{\vartheta}) \end{cases}$$
- Conditions of application
 - ✓ Use only discrepancies $dC^*(z - C(\hat{y})) = -dD^* D(y)$
 - ✓ Define a discrepancy D (with corresponding gain as dD^*)
$$D = H(\phi_u)(z - C_1(\phi_u)) + (I - H(\phi_u))(z - C_2(\phi_u))$$
$$dD^* = \delta(\phi_u)((z - C_1(\phi_u)) - (z - C_2(\phi_u)))$$

Mathematical Analysis

- Only after linearization of the error $(\tilde{y}, \tilde{\vartheta})$ small $\Rightarrow (\delta\tilde{y}, \delta\tilde{\vartheta})$
- After a change of variable $(\delta\tilde{y}, \delta\tilde{\vartheta}) \rightarrow (\eta = \delta\tilde{y} - L\delta\tilde{\vartheta}, \delta\tilde{\vartheta})$
the system of the linearised error satisfied **Asymptotically Stable**

$$\begin{cases} \dot{\eta} = (\partial_y A - \gamma \, dC^* \, dC) \eta \\ \dot{\delta\tilde{\vartheta}} = -U^{-1} L^* \, dC^* M \, dC L \, \delta\tilde{\vartheta} - U^{-1} \, dC^* M \, dC \, \eta \end{cases}$$

- Bibliography **(Exponentially) Stable if the system is identifiable**



P. Moireau, D. Chapelle, P. Le Tallec Joint state and parameter estimation for distributed mechanical systems *Comput. Methods Appl. Mech. Engrg.* 197 659–677 2008

- This strategy unifies two existing types of results



Q. Zhang, A. Clavel Adaptive observer with exponential forgetting factor for linear time varying systems, *Decision and Control*, 2001.



DT Pham, J Verron, and L Gourdeau Singular evolutive Kaiman filters for data assimilation in oceanography, *Comptes Rendus de l'Académie des Sciences-Series IIA-Earth and Planetary Science*, 326(4), 255–260 1996

Discretized version with RoUKF

- Splitting time-discretization scheme

► Sampling

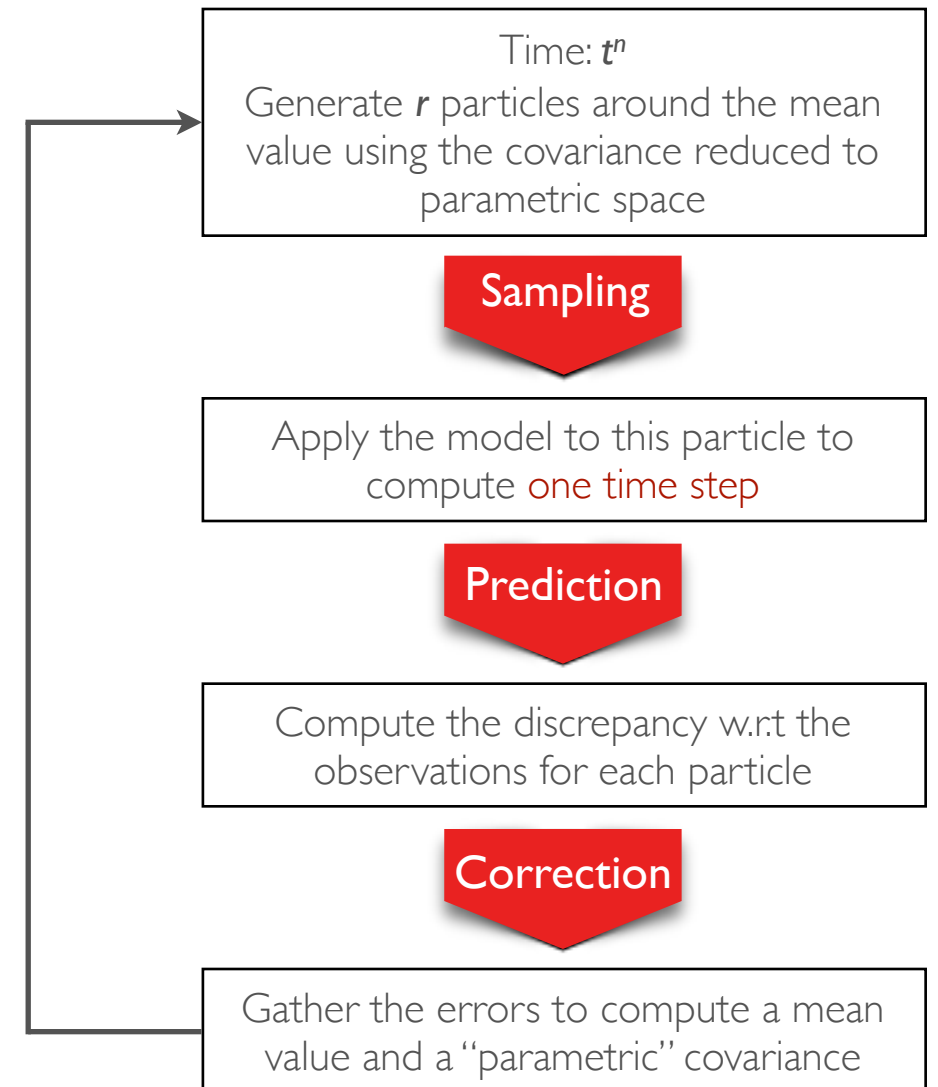
$$\begin{cases} \Lambda_n = \sqrt{U_n^{-1}} \\ \hat{Y}_{n|n}^{(i)} = \hat{Y}_{n|n} + L_n^Y \Lambda_n^T l^{(i)}, \quad 1 \leq i \leq r \\ \hat{\theta}_{n|n}^{(i)} = \hat{\theta}_n^+ + L_n^\theta \Lambda_n^T l^{(i)}, \quad 1 \leq i \leq r \end{cases}$$

► Prediction

$$\begin{cases} \hat{Y}_{n+1|n}^{(i)} = A_{n+1|n}(\hat{Y}_{n|n}^{(i)}, \hat{\theta}_{n|n}^{(i)}), \quad 1 \leq i \leq r \\ \hat{Y}_{n+1|n} = E_\alpha(\hat{Y}_{n+1|n}^*) = \sum_i \alpha_i \hat{Y}_{n+1|n}^{(i)} \\ \hat{\vartheta}_{n+1|n} = \hat{\vartheta}_{n|n} \end{cases}$$

► Correction

$$\begin{cases} L_{n+1}^Y = [\hat{Y}_{n+1|n}^*] D_\alpha[l^*]^T \\ L_{n+1}^\theta = [\hat{\theta}_{n+1|n}^*] D_\alpha[l^*]^T \\ \Gamma_{n+1}^{(i)} = D(Y_{n+1}^{(i)-}, t_{n+1}) \\ \Delta_{n+1} = [\Gamma_{n+1}^*] D_\alpha[l^*]^T \\ U_{n+1} = Id + \Delta_{n+1}^T M_{n+1} \Delta_{n+1} \\ \hat{Y}_{n+1|n+1} = \hat{Y}_{n+1|n} + L_{n+1}^Y U_{n+1}^{-1} \Delta_{n+1}^T M_{n+1} (E_\alpha(\Gamma_{n+1}^*)) \\ \hat{\theta}_{n+1|n+1} = \hat{\theta}_{n+1|n} + L_{n+1}^\theta U_{n+1}^{-1} \Delta_{n+1}^T M_{n+1} (E_\alpha(\Gamma_{n+1}^*)) \end{cases}$$

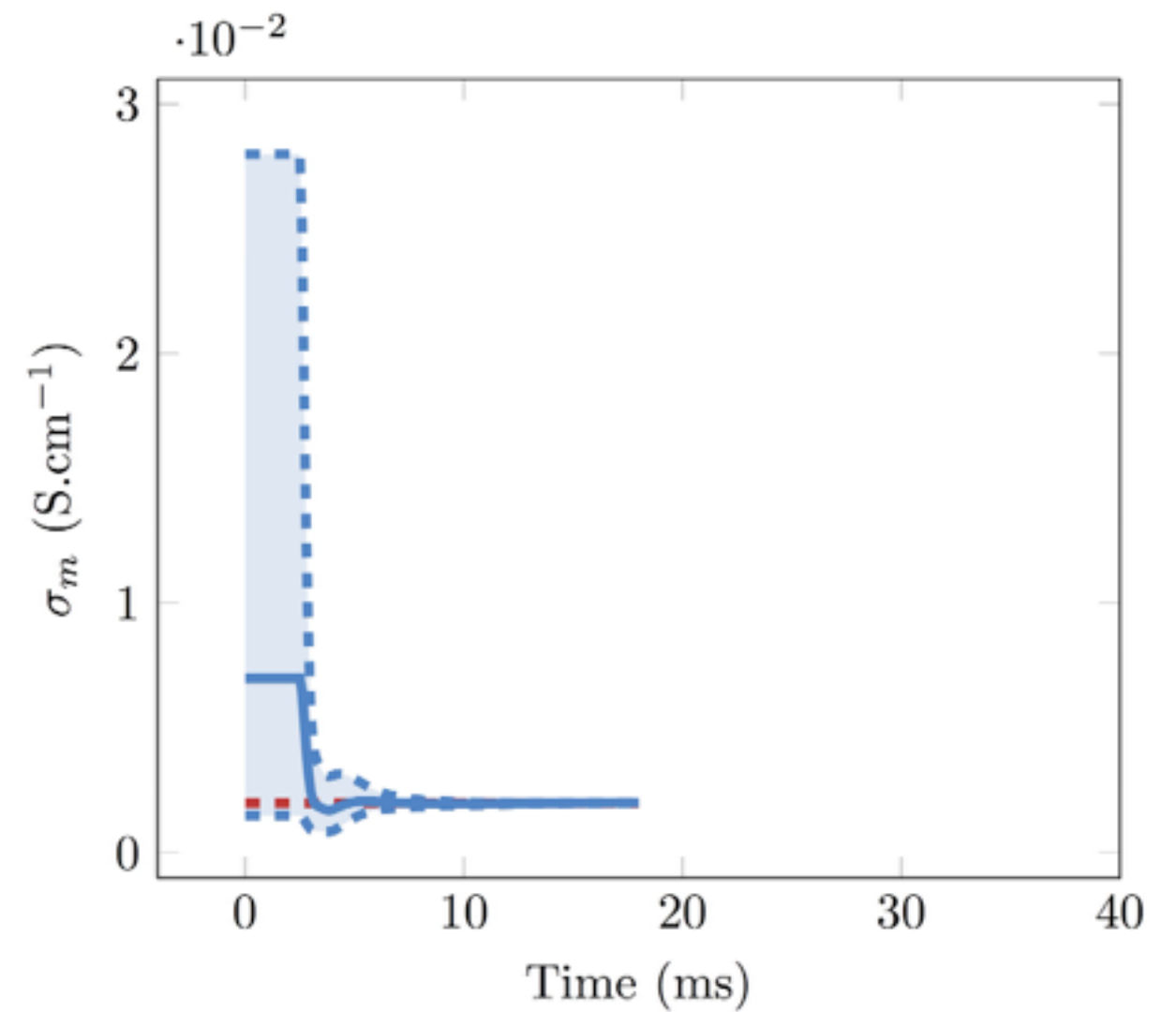
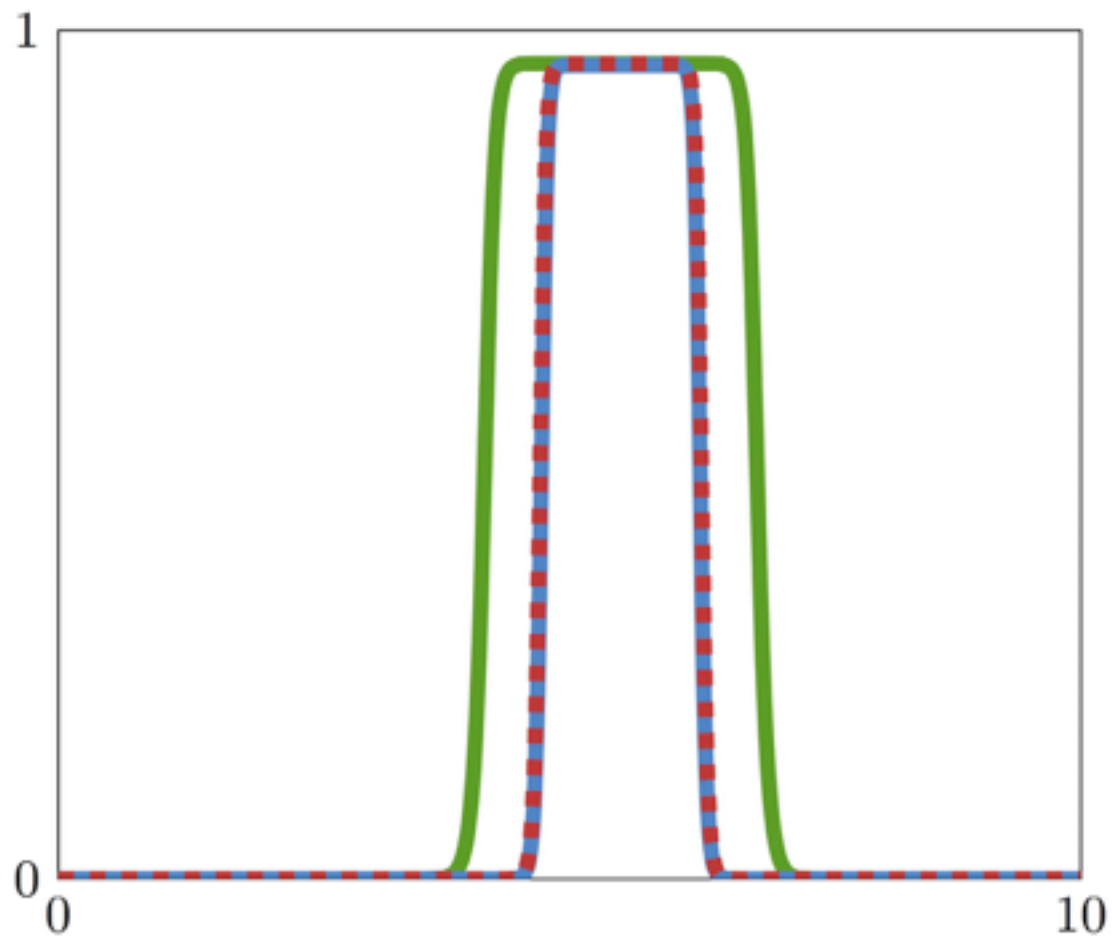


P. Moireau, D. Chapelle, Reduced-Order Unscented Kalman Filtering With Application to Parameter Identification in Large-Dimensional Systems - ESAIM COCV 2011

ID Case

- Isotropic and spacewise-constant diffusion tensor

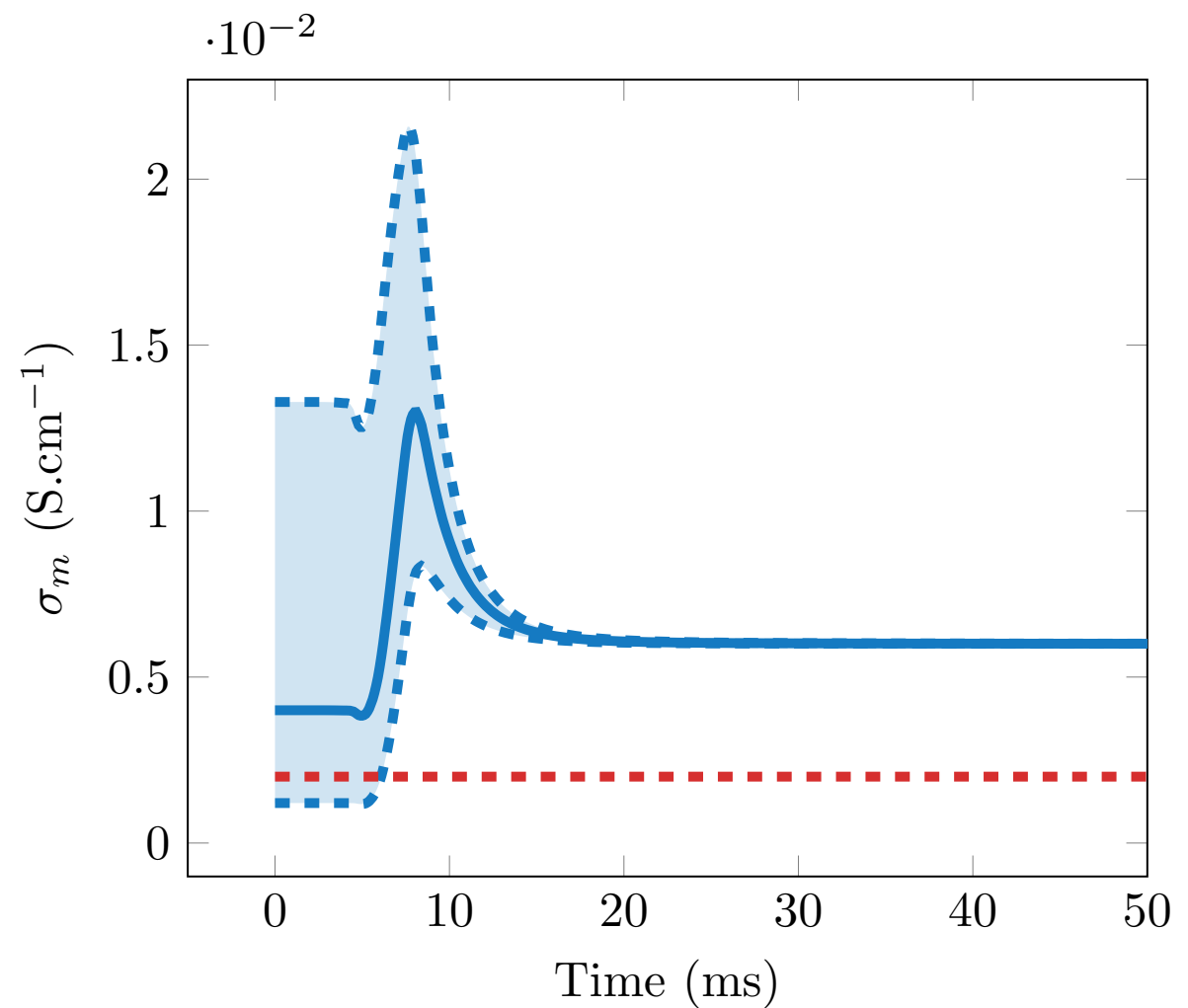
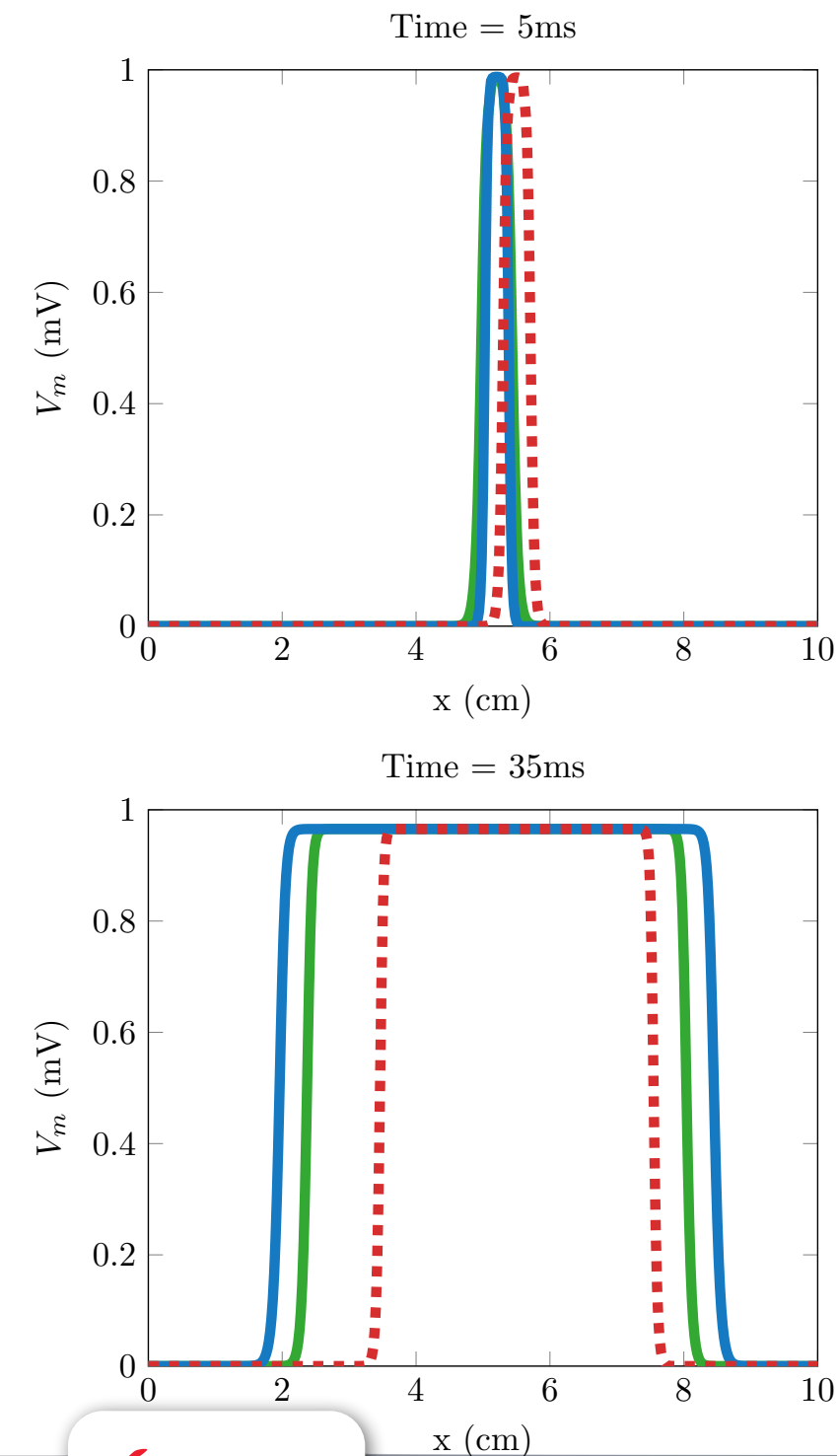
Time = 15ms



ID Case

- Importance of the **state** estimator during identification

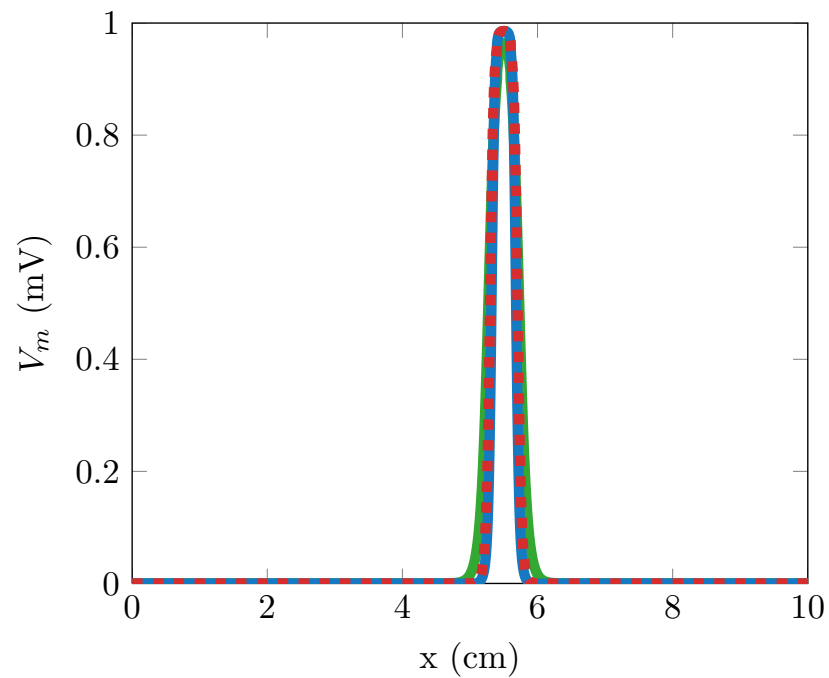
➔ Even for a small error on the initial condition



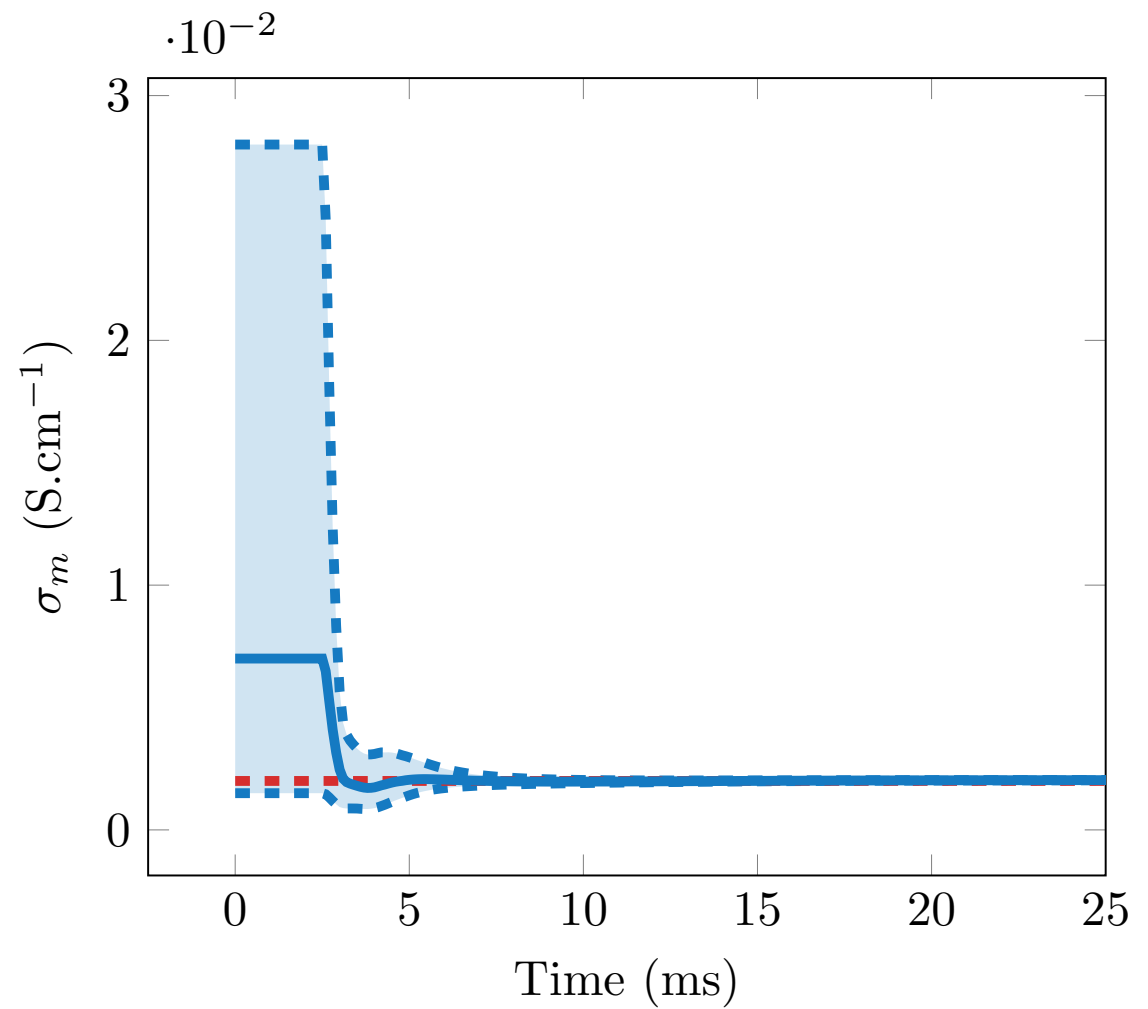
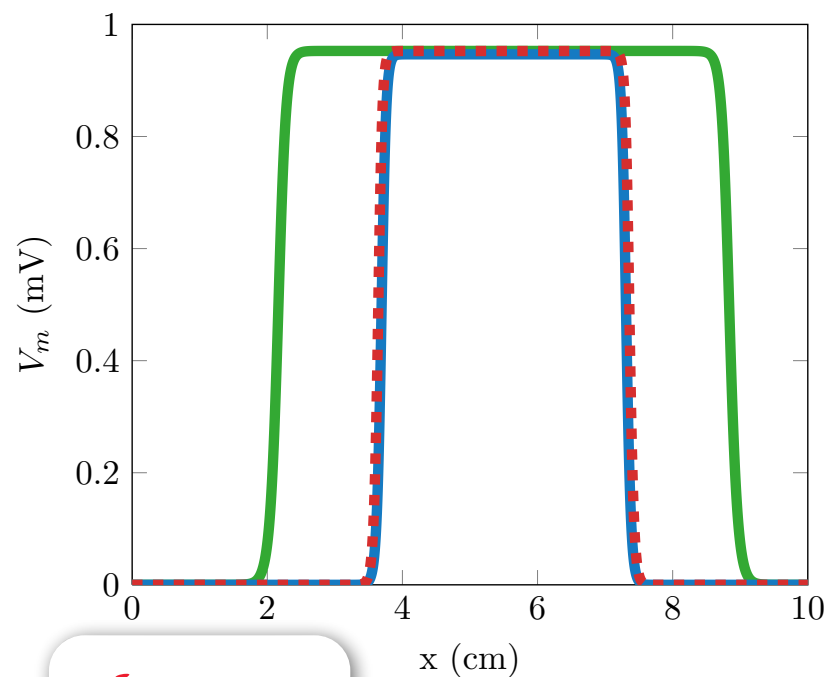
ID Case

- Isotropic and spacewise-constant diffusion tensor
 - Good initial condition - **Wrong parameter**

Time = 5ms



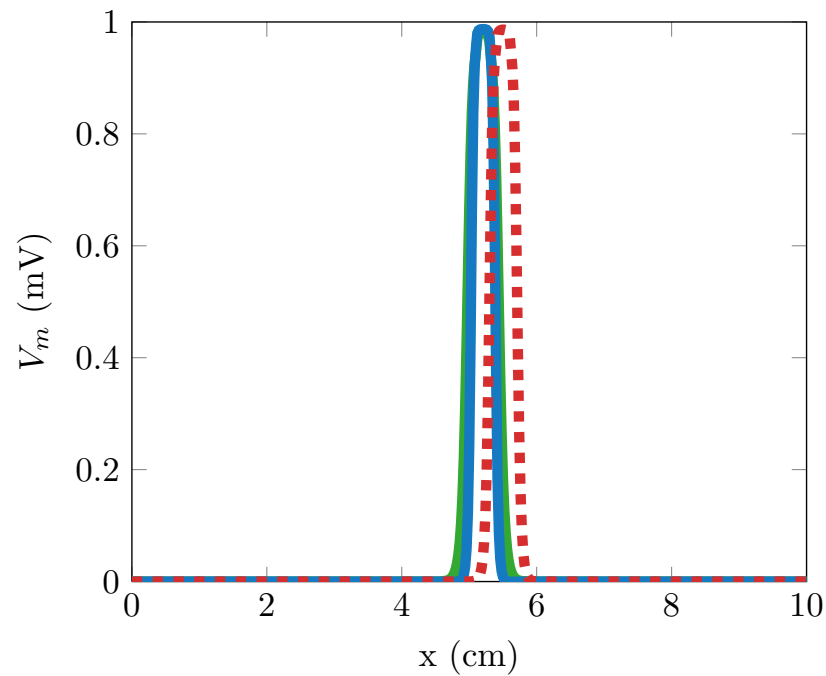
Time = 35ms



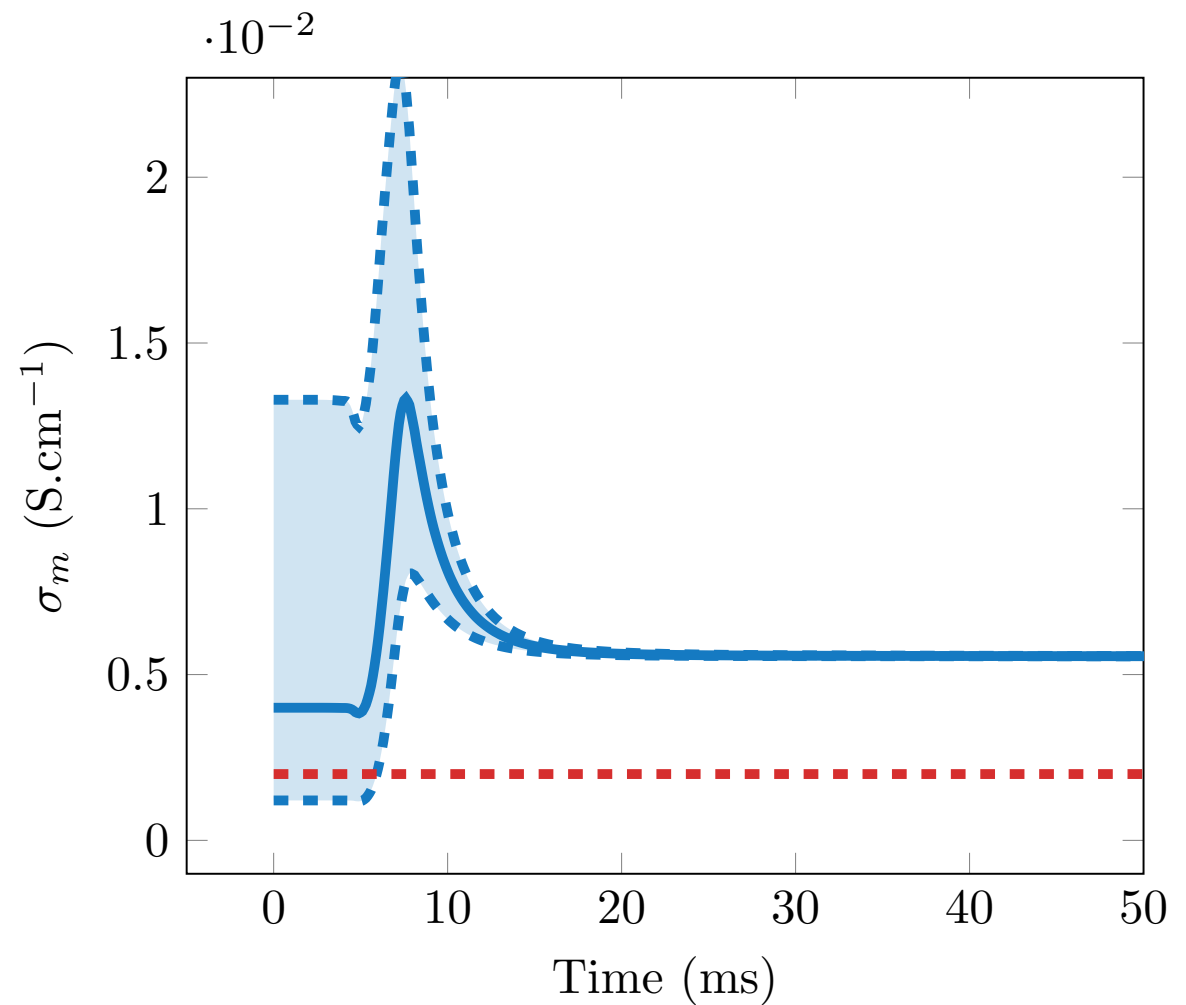
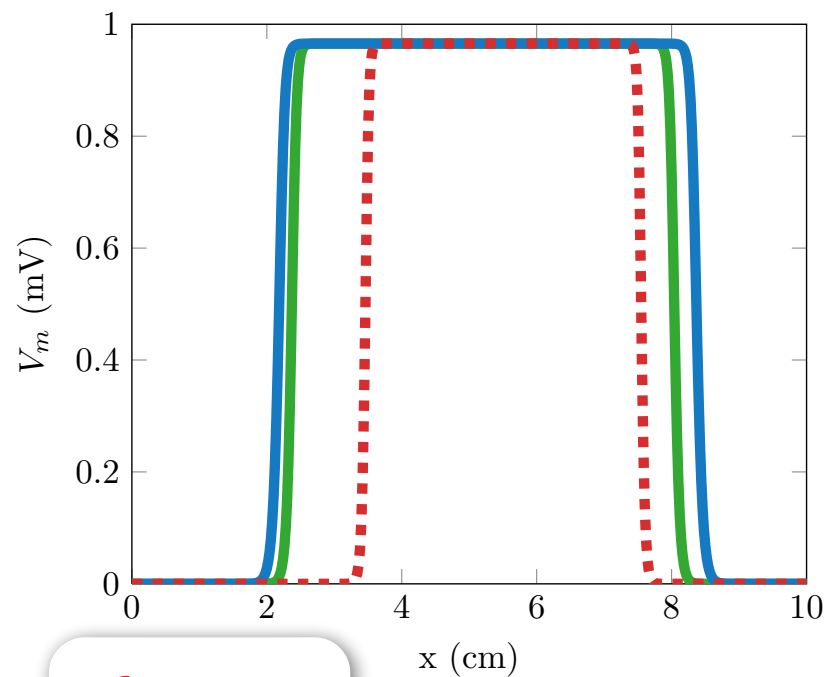
ID Case

- Isotropic and spacewise-constant diffusion tensor
 - ▶ ~~Good~~ initial condition - **Wrong parameter**

Time = 5ms



Time = 35ms

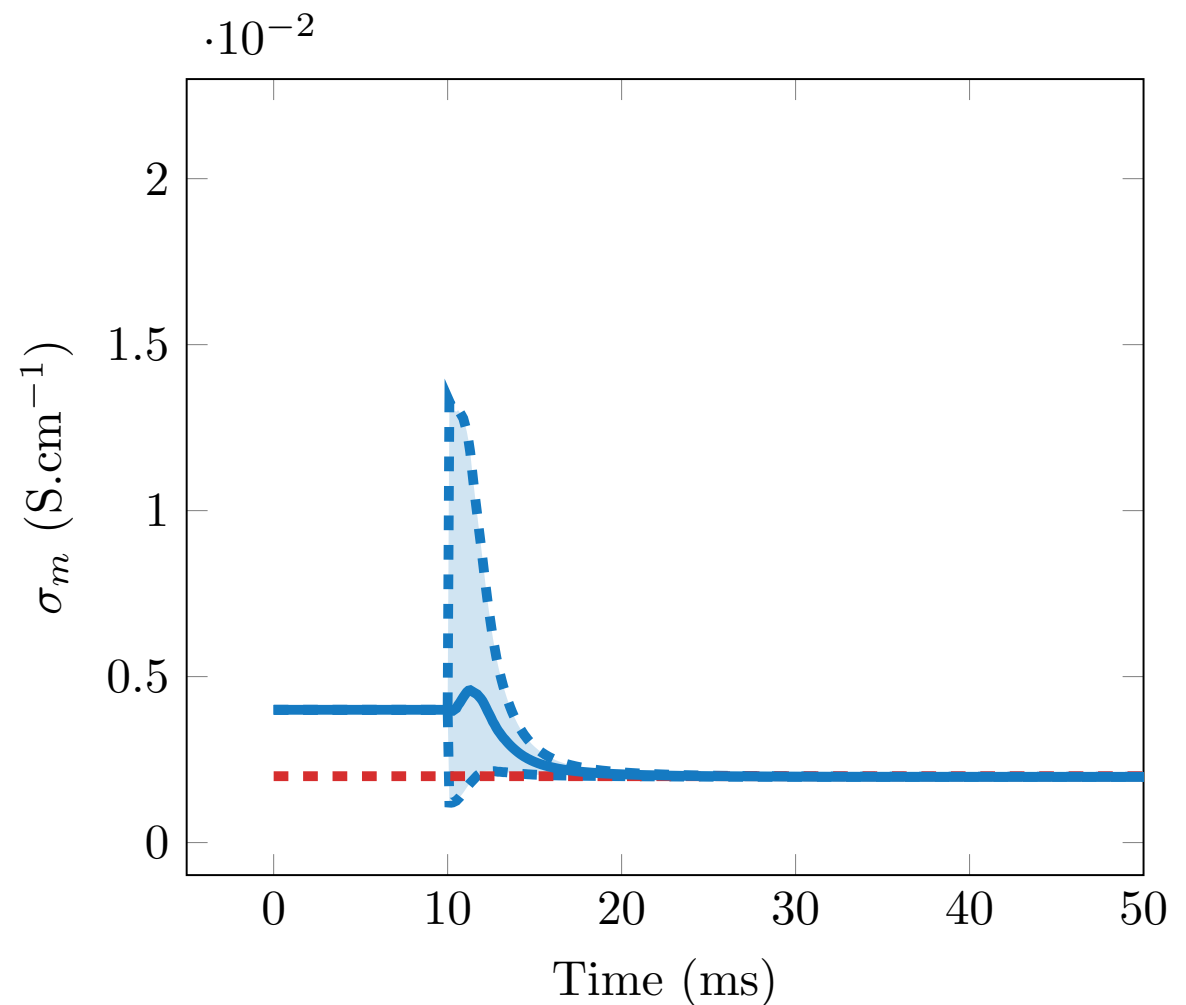
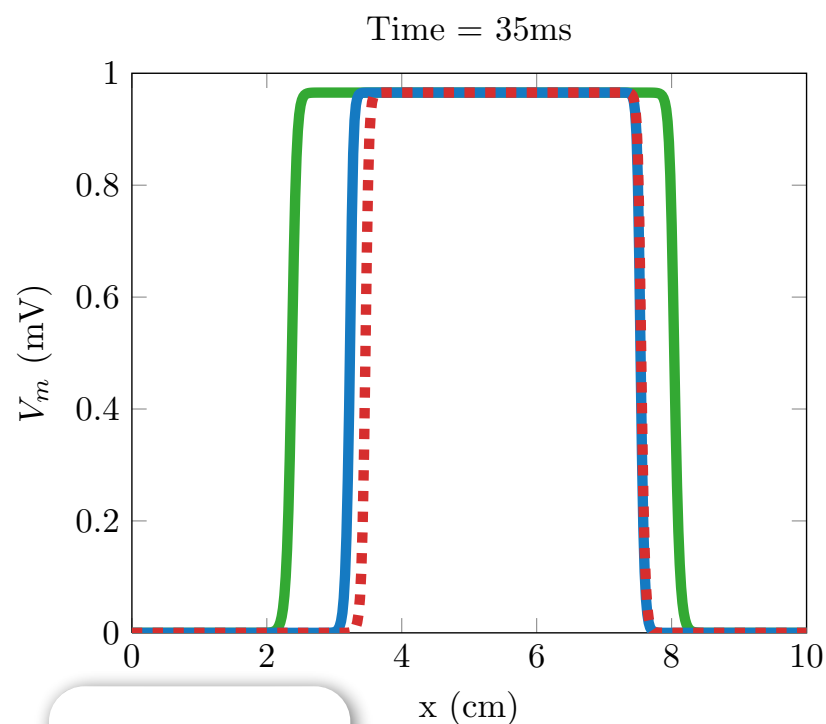
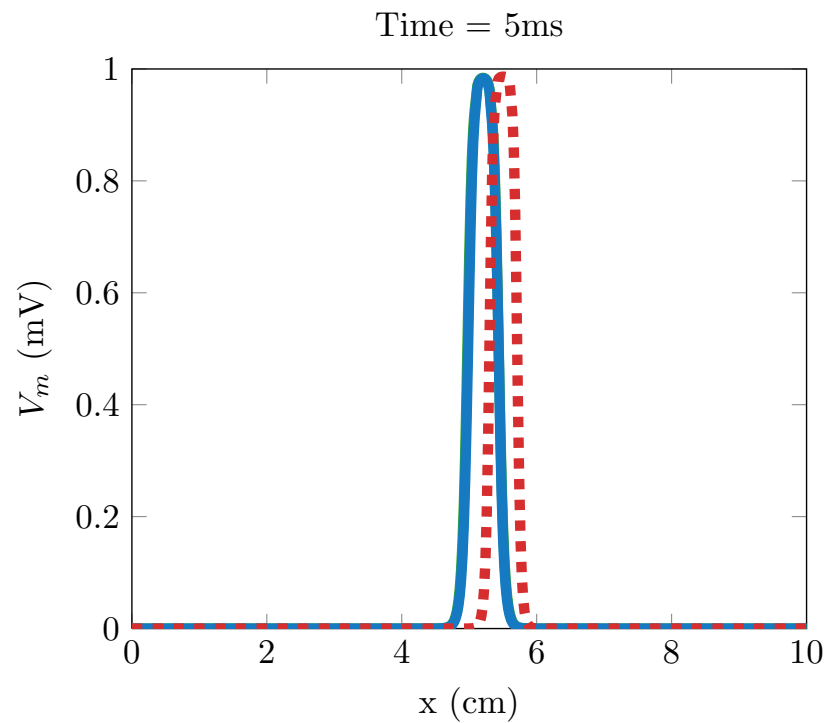


ID Case

- Isotropic and spacewise-constant diffusion tensor

► ~~Good~~ initial condition - **Wrong parameter**

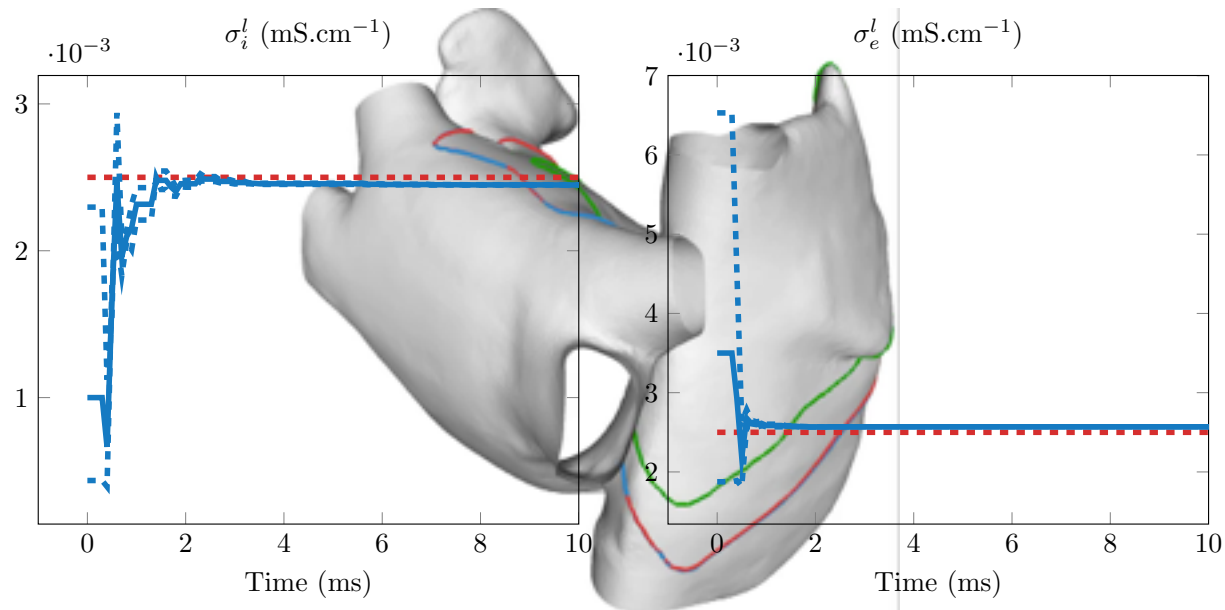
➔ Delaying the joint state and parameter observer



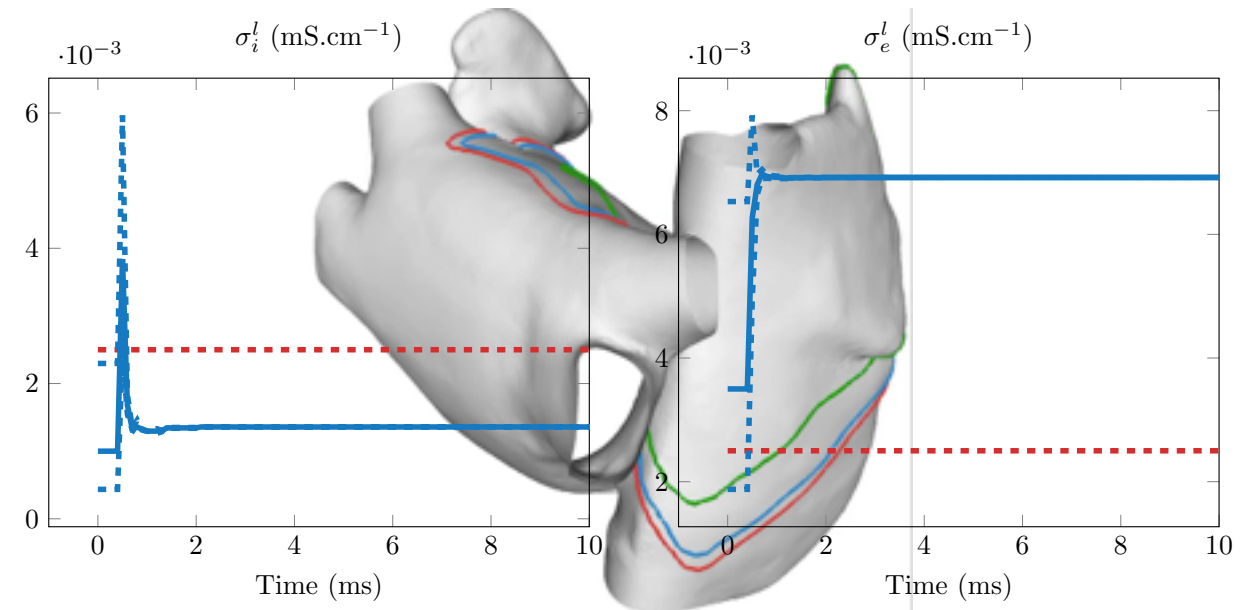
3D (Surfacic) Case

- First tests

- Good initial condition

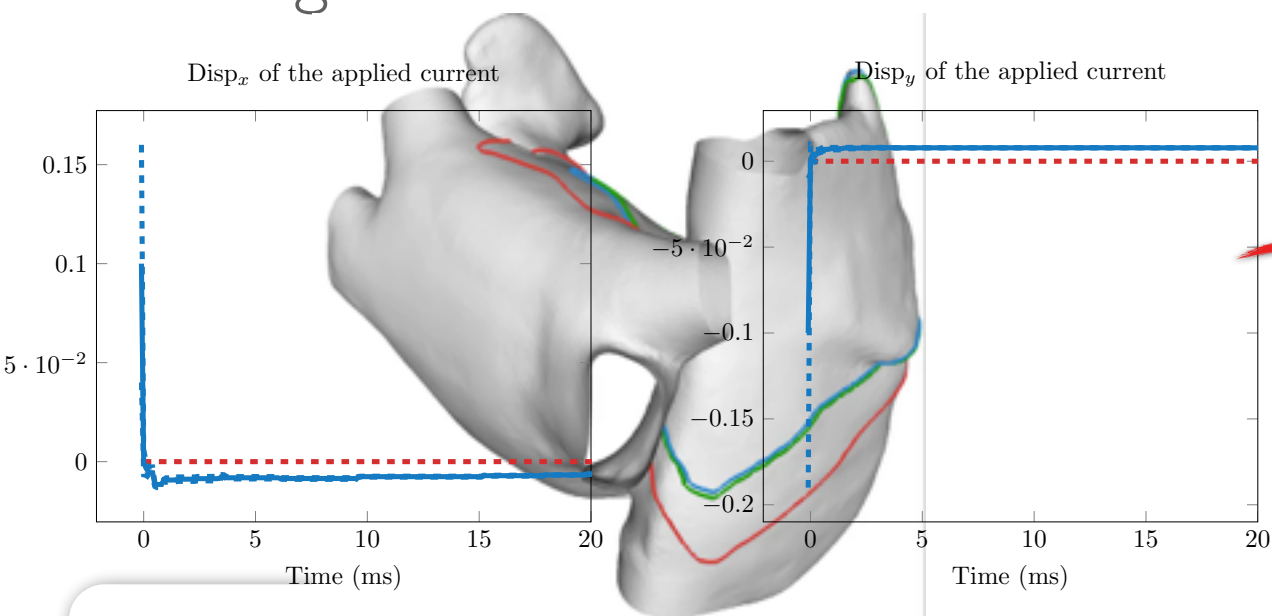


- ~~Good initial condition~~

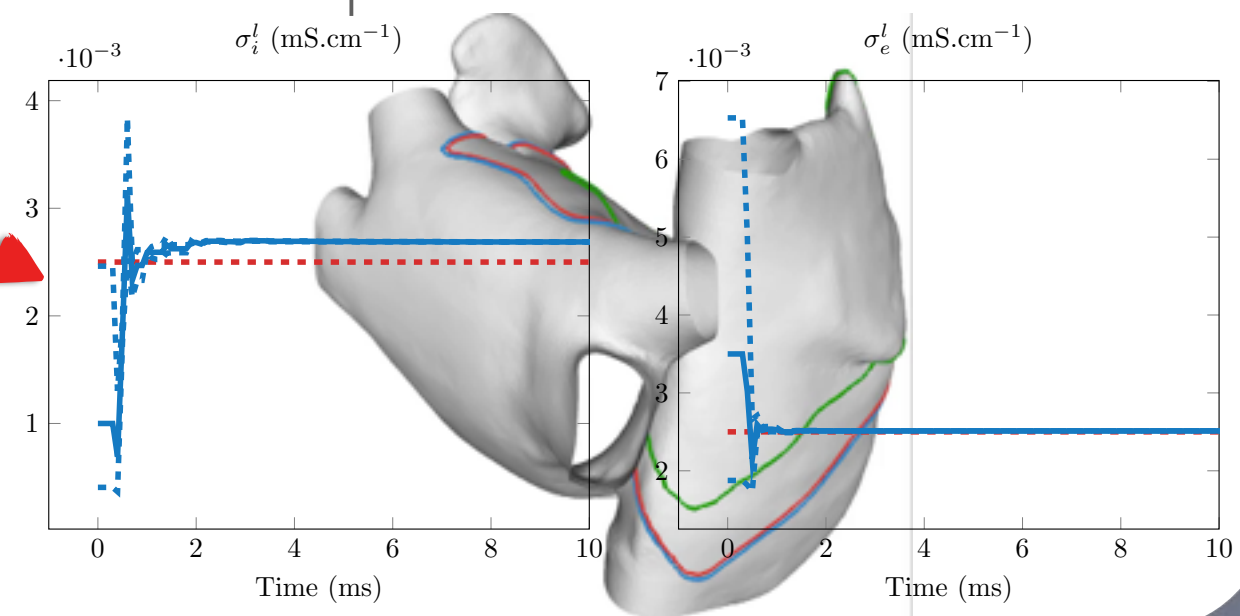


- Complete strategy

- Stage of initial condition identification



- Real parameter identification



Observer for propagating phenomena based on front observations

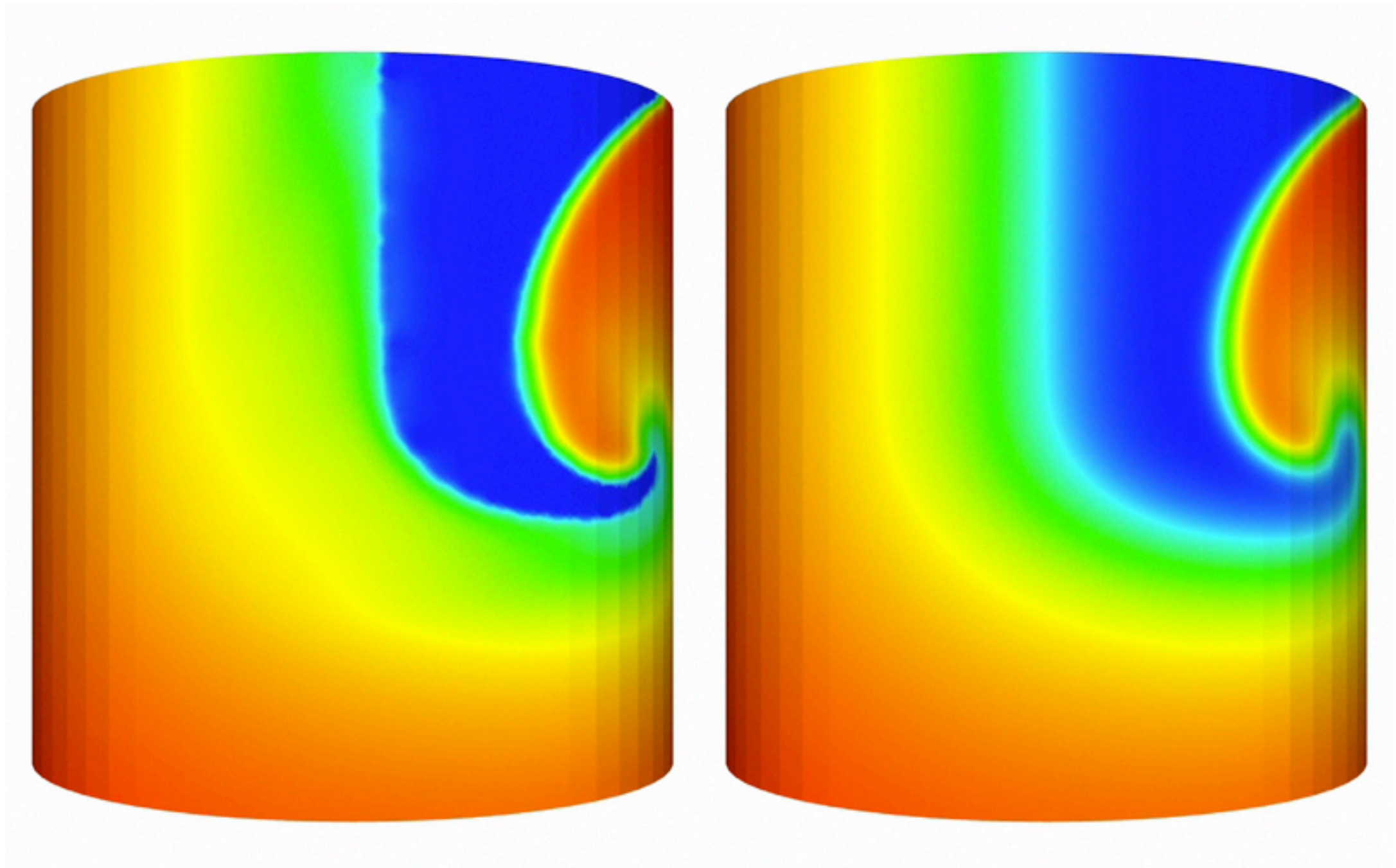
Part V: Perspectives

A — *Complete analysis*

- Finalize the actual analysis
 - ▶ Existence of the observer
 - ▶ Observability condition associated to the linearised error
- Via the Mumford Shah functional
 - ▶ Consider a stationary pattern
 - ▶ Define a functional associated with our observer
 - ▶ Use the classical Mumford Shah theory to define a minimizer and show that the observer converges to it

B — Incorporate more information

- Incorporate topological gradient associated to the Chan-Vese data fitting term



- Use multiple phase to handle multiple isocontours

C — Wave equation case

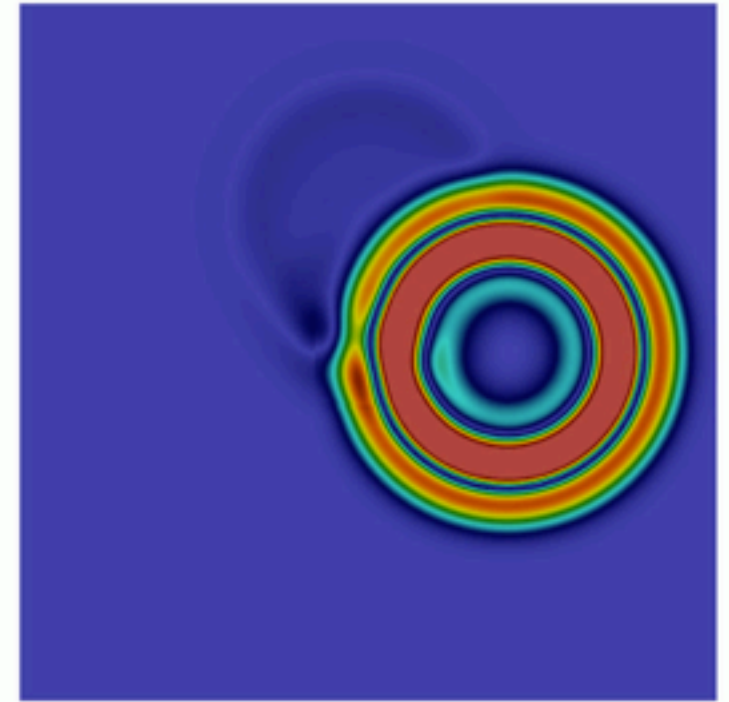
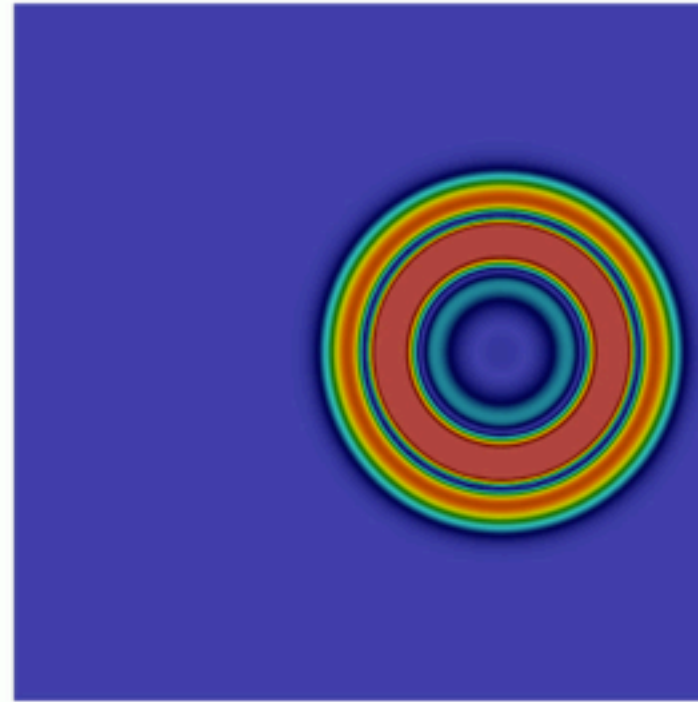
- Wave equation with no b.c.

$$\ddot{u} - c^2 \Delta u = 0, \quad \text{in } \mathcal{B}$$

- Special first order form

$$\begin{cases} \dot{u} + c \underline{\nabla} \cdot \underline{v} = 0, & \text{in } \mathcal{B} \\ \dot{\underline{v}} + c \underline{\nabla} u = 0, & \text{in } \mathcal{B} \end{cases}$$

with $\mathcal{E} = \int_{\mathcal{B}} u^2 + v^2 dx$



- Observer with a partial measurement $u|_{\omega}$

$$\begin{cases} \dot{\hat{u}} + c \underline{\nabla} \cdot \hat{\underline{v}} = \gamma \chi_{\omega} (z - \hat{u}|_{\omega}), & \text{in } \mathcal{B} \\ \dot{\hat{\underline{v}}} + c \underline{\nabla} \hat{u} = 0, & \text{in } \mathcal{B} \end{cases}$$

- Observer with a front observation ?

$$\begin{cases} \dot{\hat{u}} + c \underline{\nabla} \cdot \hat{\underline{v}} = \gamma \delta_{\Gamma_{\hat{u}}} \frac{1}{|\underline{\nabla} \hat{u}|} \left(- (z - C_1(\phi_{\hat{u}}))^2 + (z - C_2(\phi_{\hat{u}}))^2 \right), & \text{in } \mathcal{B} \\ \dot{\hat{\underline{v}}} + c \underline{\nabla} \hat{u} = 0, & \text{in } \mathcal{B} \end{cases}$$

To be continued ...



**ANALYSE ASYMPTOTIQUE EN
ÉLECTROPHYSIOLOGIE CARDIAQUE.
APPLICATIONS À LA MODÉLISATION ET À
L'ASSIMILATION DE DONNÉES.**

THÈSE DE DOCTORAT

Présentée par

Annabelle COLLIN

pour obtenir le grade de

**DOCTEUR DE
L'UNIVERSITÉ PIERRE ET MARIE CURIE - Paris VI**

Spécialité : MATHÉMATIQUES APPLIQUÉES

Soutenue publiquement le ?????????? devant le jury composé de :

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