

2025
MFEM
Community
Workshop



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A Domain-Decomposition Framework for Multiphysics Biomedical Modeling: Application to Cardiac Radiofrequency Ablation

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Outline

01.

Clinical Background

What is Radiofrequency Ablation (RFA)?

02.

Multiphysics Model

Why do we need multiphysics modeling?

What physics are we including?

How do we implement it?

03.

Domain Decomposition

How do we handle multiple domains?

Why do we need Domain Decomposition?

04.

Results

What did we find?

05.

Conclusion & Future work

How can we improve our framework?

01.

Clinical Background

Cardiac Arrhythmias, Radiofrequency Ablation and Motivation to Modeling

Cardiac RadioFrequency Ablation

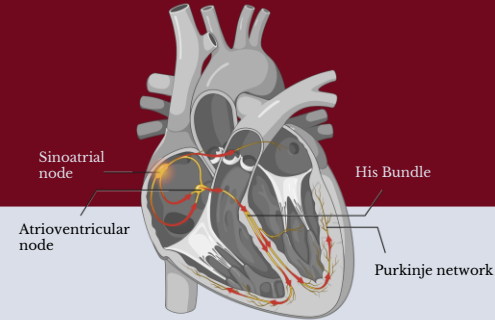
Arrhythmia

“Any problem in the rate or rhythm of a person’s heartbeat”

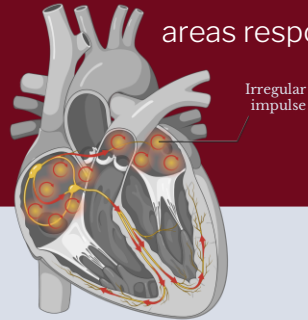
- American Heart Association

Cardiac Ablation

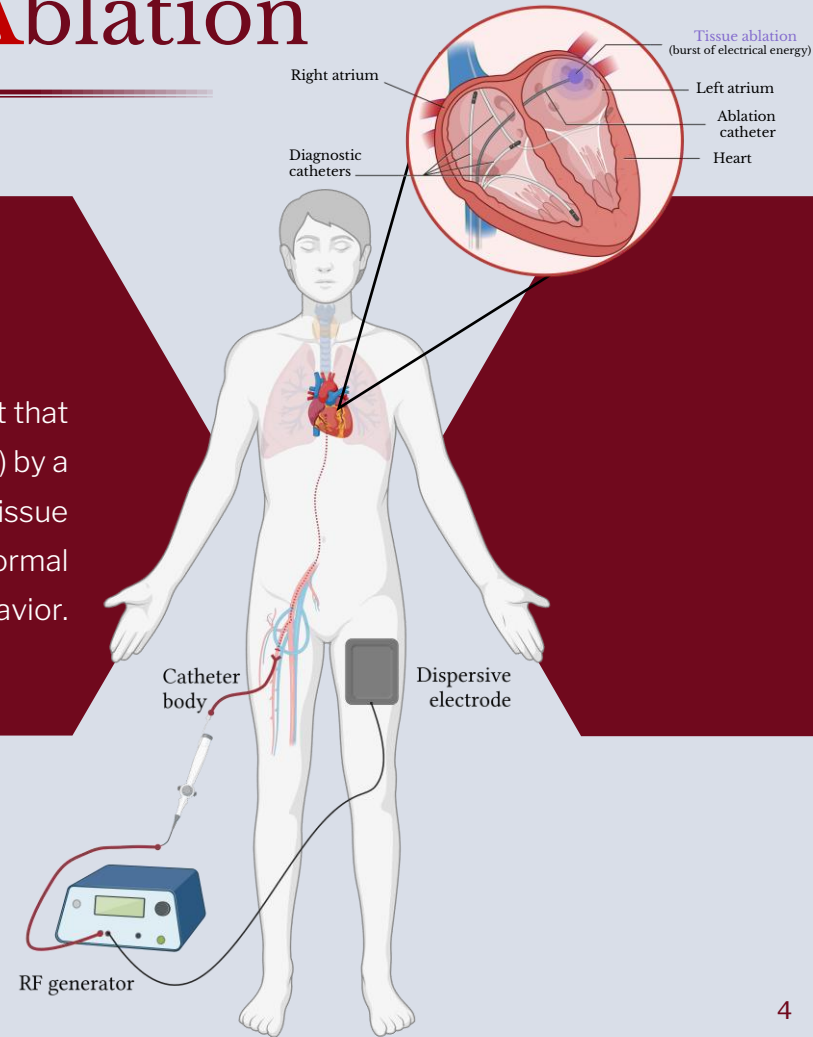
Minimally invasive treatment that delivers **energy** (heat, cold) by a catheter to destroy targeted tissue areas responsible for abnormal electrical behavior.



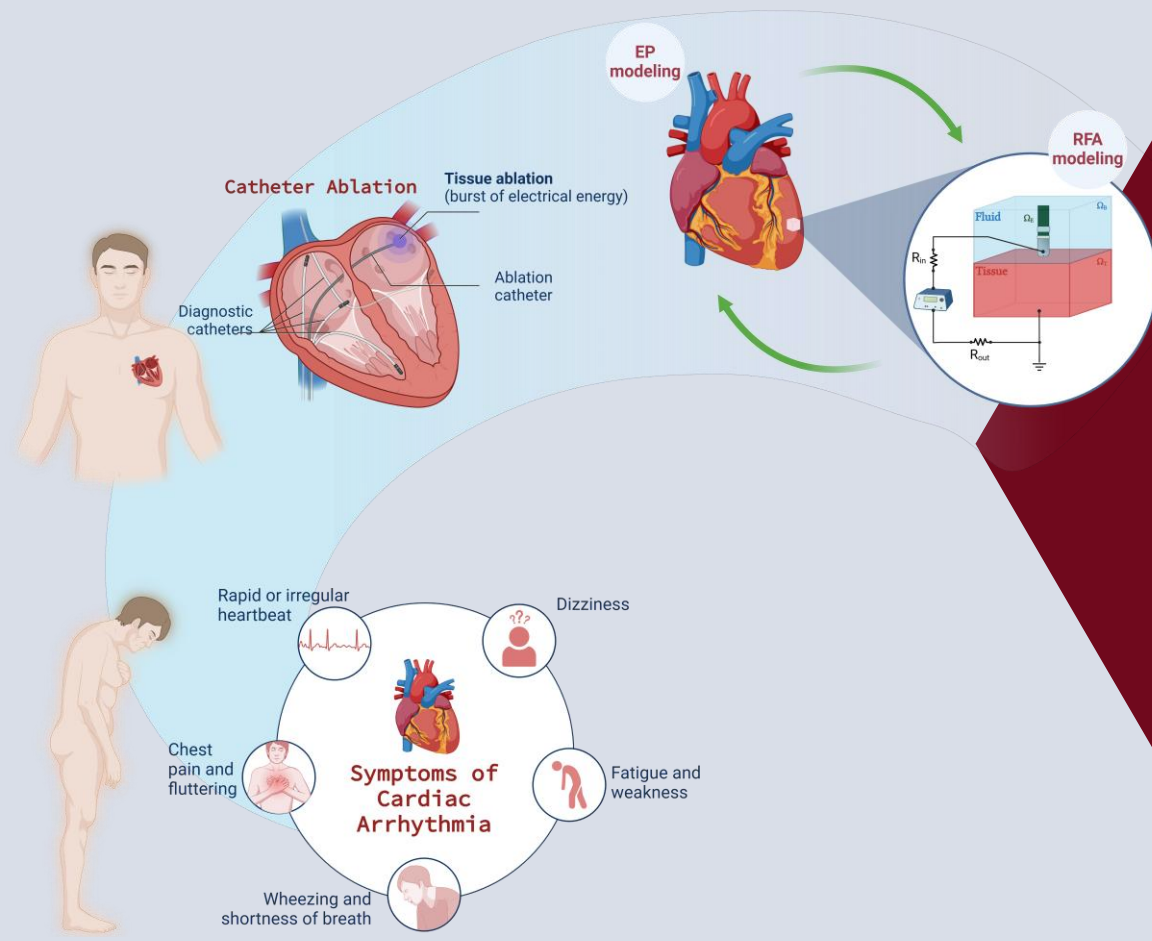
Typical Heart Rhythm



Atrial Fibrillation (AFib)



Why do we need modeling?



01.

Improve the mechanistic understanding of lesion formation

02.

Patient-tailored therapy

03.

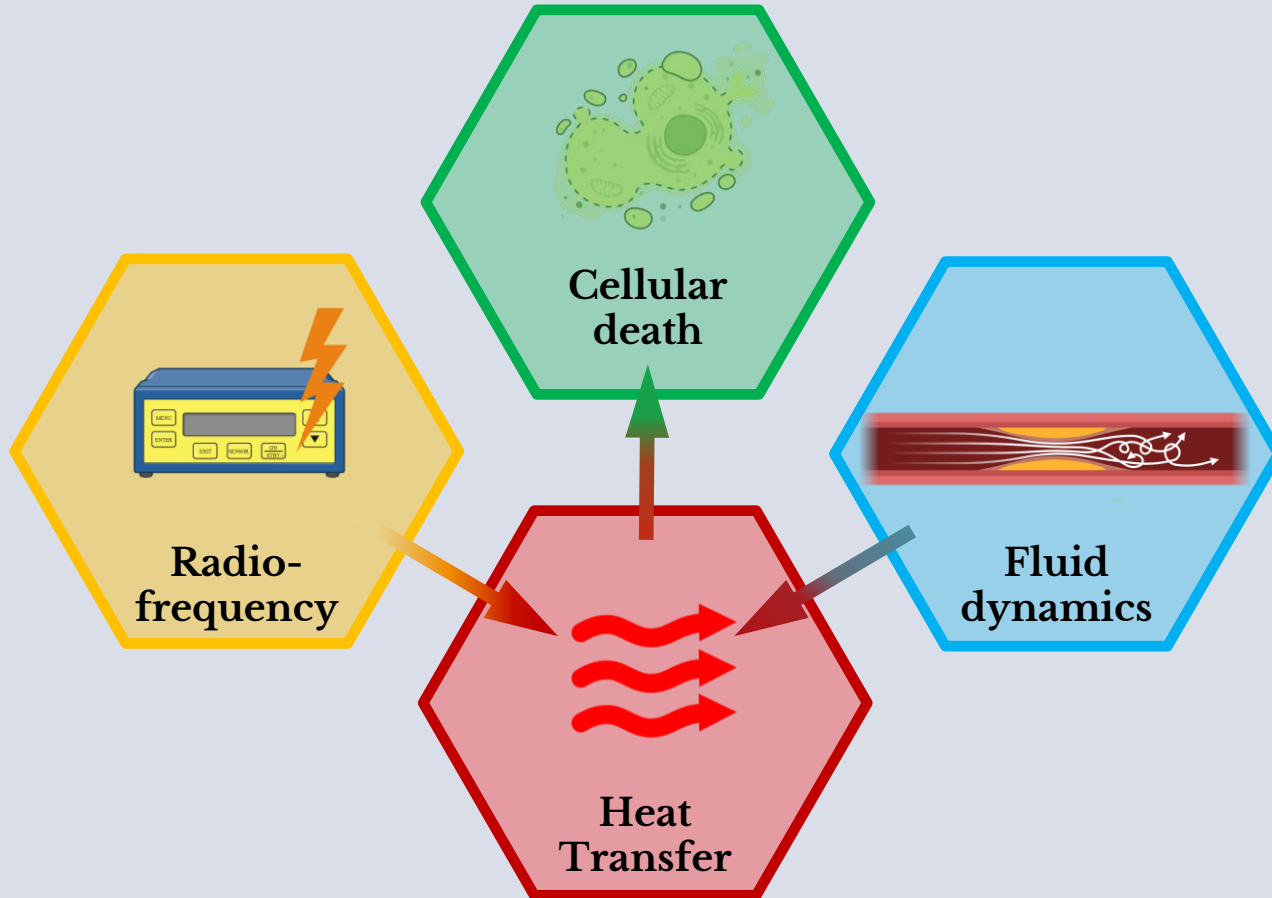
Optimal design of new devices and clinical protocols

02.

Multiphysics Model of RFA

Mathematical and computational models of RF, Bioheat, Cellular death, Fluid Dynamics

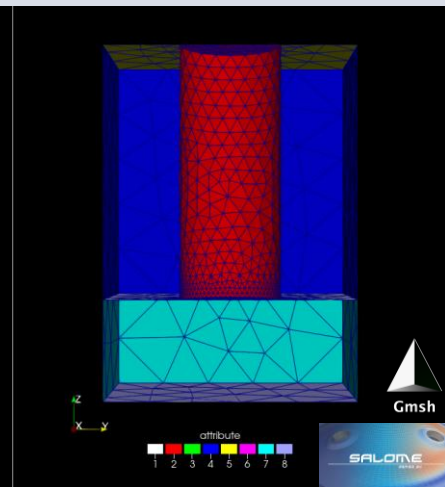
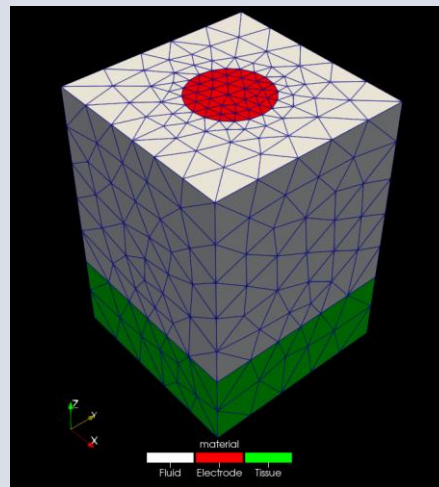
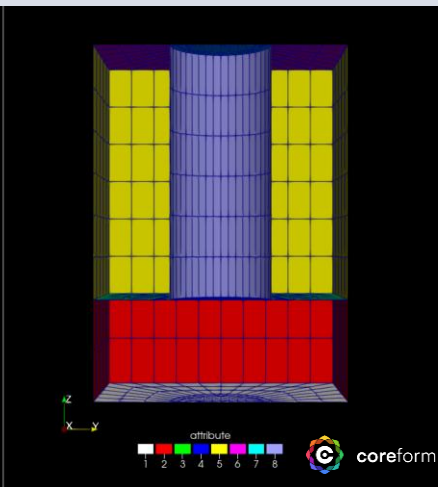
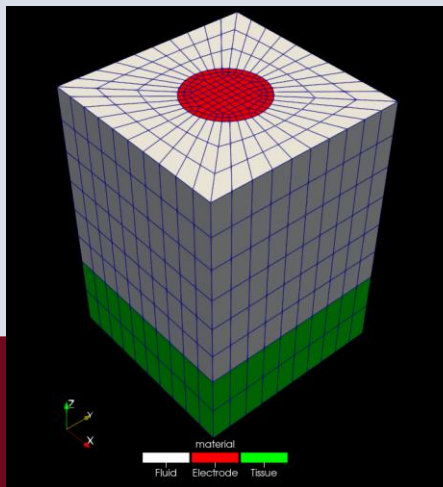
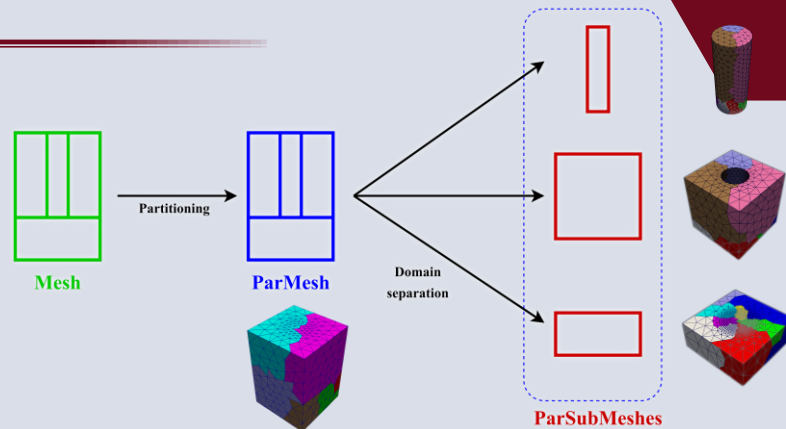
Model overview



Geometry

⬡ Idealized geometry (...for now)

- Three physical domains: Tissue, Fluid, Electrode
- Available both as *structured* and *unstructured* meshes



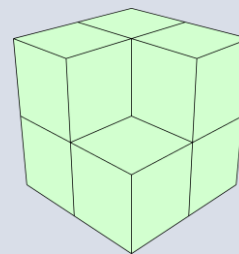
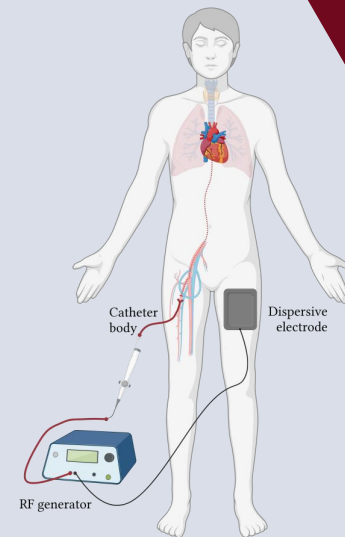
Radiofrequency source

Quasi-static Maxwell's equation

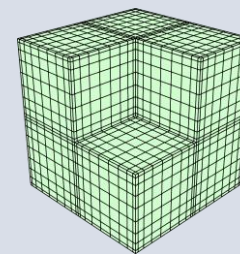
$$\nabla \cdot (\sigma \nabla \Phi) = 0 \quad \text{in } \Omega_f, \Omega_t$$

Numerical implementation

- Spatial discretization: \mathcal{Q}_4 elements
- Linear solver: CG preconditioned with Low-Order-Refined¹ (**LOR**) coupled to AMG (PA)



High-order (FE)
mesh



Low-order (FE)
refined mesh

t = Tissue
f = Fluid
e = Electrode

[1] Will Pazner, Tzanio Kolev, and Jean-Sylvain Camier. End-to-end gpu acceleration of low-order-refined preconditioning for high-order finite element discretizations. The International Journal of High Performance Computing Applications, 37(5):578–599, 2023.

[2] Pazner, W. "Efficient low-order refined preconditioners for high-order matrix-free continuous and discontinuous Galerkin methods, arXiv preprint 1908.07071 (Aug. 2019)." URL <https://arxiv.org/abs/1908.07071>

Heat Transfer

Pennes' bioheat equation

$$\rho_t c_t \frac{\partial T}{\partial t} = \nabla \cdot (\mathbf{k}_t \nabla T) + Q_m + Q_p + Q_s \quad \text{in } \Omega_t$$

$$\rho_f c_f \frac{\partial T}{\partial t} = \nabla \cdot (\mathbf{k}_f \nabla T) - \mathbf{u} \cdot \nabla T \quad \text{in } \Omega_f$$

$$\rho_e c_e \frac{\partial T}{\partial t} = \nabla \cdot (\mathbf{k}_e \nabla T) \quad \text{in } \Omega_e$$

$$Q_s = \sigma \cdot |\mathbf{E}|^2 \quad [\text{W/m}^3]$$

Q_m = metabolic (constant)

$$Q_p = -w_b \rho_b c_b (T_t - T_{bp})$$

T_{bp} constant temperature of blood perfusing the tissue

w_b is a perfusion coefficient depending on the cell viability so that $0 < w_b(N) \leq w_{b,\max}$

Hex Numerical implementation

- **Spatial discretization:** \mathcal{Q}_2 elements
- **Time discretization:** two-step third-order Singly-diagonal implicit Runge-Kutta scheme (*SDIRK23*)
- **Linear solver:** GMRES preconditioned with matrix-free Jacobi smoother (PA)

t = Tissue
f = Fluid
e = Electrode

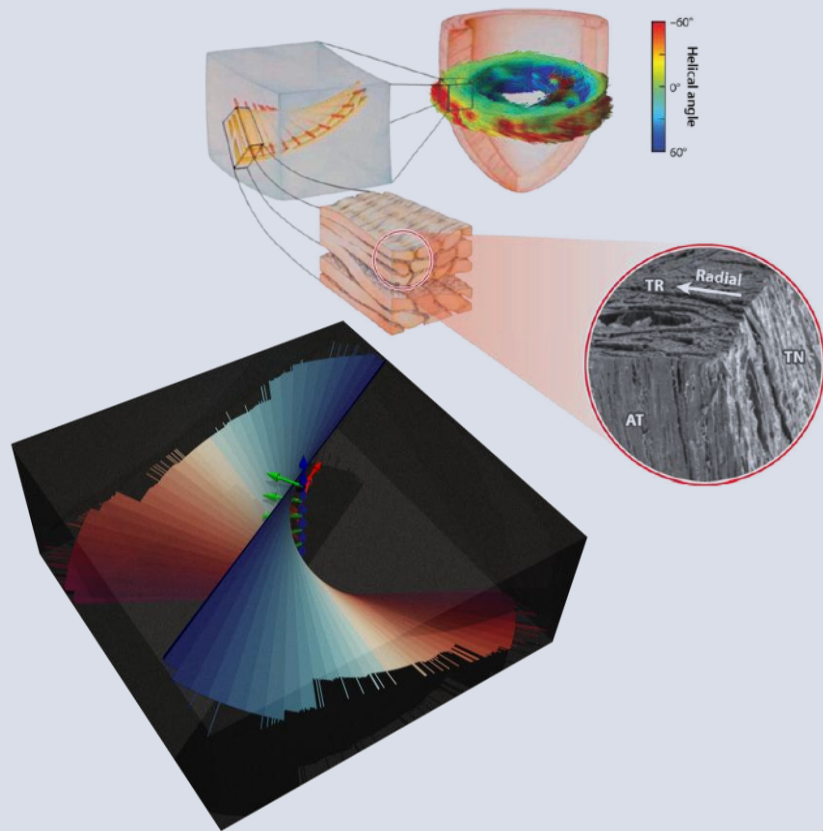
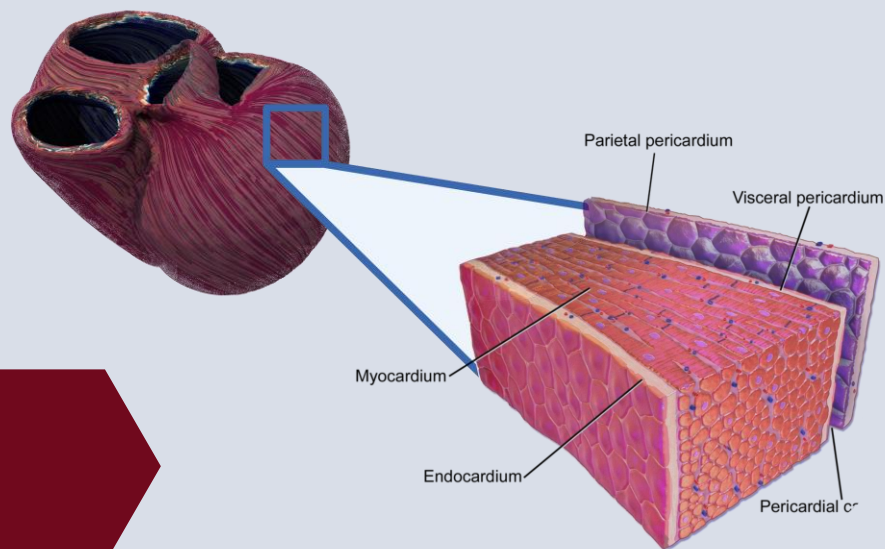
Anisotropy

⬡ Anisotropic thermo-electrical conductive properties in the tissue due to fiber microstructure

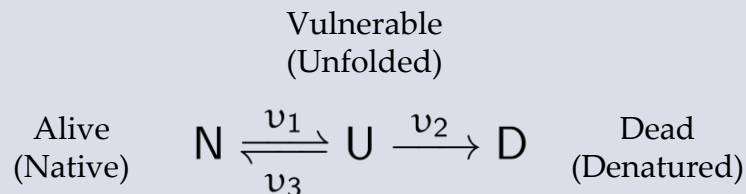
$$\sigma = \mathbf{R} \begin{bmatrix} \sigma_f & 0 & 0 \\ 0 & \sigma_{\perp}^y & 0 \\ 0 & 0 & \sigma_{\perp}^z \end{bmatrix} \mathbf{R}^T$$

$$\mathbf{k} = \mathbf{R} \begin{bmatrix} k_f & 0 & 0 \\ 0 & k_{\perp}^y & 0 \\ 0 & 0 & k_{\perp}^z \end{bmatrix} \mathbf{R}^T$$

$$\mathbf{R} = \mathbf{R}_z(\alpha) \mathbf{R}_y(\beta) \mathbf{R}_x(\gamma) :$$



Cellular death



Three-state cell-death model

$$\begin{aligned}
 \frac{dN}{dt} &= -v_1 N + v_3 U \\
 \frac{dU}{dt} &= v_1 N - v_2 U - v_3 U \quad \text{in } \Omega_t \\
 \frac{dD}{dt} &= v_2 U
 \end{aligned}$$

$$v_i(T) = A_i e^{-\Delta E/RT}$$

$$N + U + D = 1$$

Numerical implementation

- Eigenvalue method to solve $\frac{dX}{dt} = \mathcal{A} X$, $X^T = [N, U, D]^T$
 $X = P e^{\int_0^t \text{diag}(\lambda) d\tau} P^{-1} X(0)$

$$\mathcal{A} = \begin{bmatrix} -v_1 & v_3 & 0 \\ v_1 & -(v_2 + v_3) & 0 \\ 0 & v_2 & 0 \end{bmatrix} = P \Lambda P^{-1} \quad X_{n+1} = P e^{\Lambda \Delta t} P^{-1} X_n$$

- Symbolic evaluation of eigenpairs $\{\Lambda, P\}$

Incompressible Navier Stokes equation

$$\frac{\partial \mathbf{u}}{\partial t} + \mathbf{u}^* \cdot \nabla \mathbf{u} - \nabla \cdot \boldsymbol{\tau} + \nabla p = \mathbf{f}$$

$$\nabla \cdot \mathbf{u} = 0$$

$$\mathbf{u}^* = \sum_i^p (\beta_i \mathbf{u}_{n+1-i} \cdot \nabla) \mathbf{u}_{n+1} \quad \text{extrapolated velocity (semi-implicit convection)}$$

$$\boldsymbol{\tau} \quad \text{viscous stress tensor} \quad \boldsymbol{\tau} = \begin{cases} \nu \nabla \mathbf{u} & \text{Stiff strain} \\ \nu (\nabla \mathbf{u} + \nabla \mathbf{u}^T) & \text{Viscous strain} \end{cases}$$

⬡ Numerical implementation

ALgebraic splitting Time ADaptive solver for Incompressible Navier-Stokes (ALADINS)¹

- Time discretization: **BDF** up to order 3
- Spatial discretization: stable pair $\mathcal{Q}_2 - \mathcal{Q}_1$
- Time adaptivity (Pressure correction- based)
- Algebraic system:

$$\mathcal{A} = \begin{bmatrix} \mathcal{C} & \mathcal{G} \\ \mathcal{D} & \end{bmatrix},$$

$$\text{with } \mathcal{C} = \frac{\alpha}{\Delta t} \mathcal{M} + \nu \mathcal{K} + \mathcal{N}(\mathbf{u}), \quad \mathcal{G} = -\mathcal{D}^T$$

- FGMRES preconditioned with **block preconditioner** $\widehat{\mathcal{P}}$

[1] Alessandro Veneziani and Umberto Villa. Aladins: An algebraic splitting time adaptive solver for the incompressible navier–stokes equations. Journal of Computational Physics, 238:359–375, 2013.

Fluid Dynamic: Preconditioner

Navier-Stokes Block Preconditioner $\hat{\mathcal{P}}$

- Algebraic splitting pressure corrected preconditioner^{1,2}

Where we assumed :

- $\hat{\mathcal{S}}$ preconditioner for the Schur Complement $\Sigma = \mathcal{D}\mathcal{C}^{-1}\mathcal{G}$, $\hat{\mathcal{S}} = \mathcal{D} \text{diag}(\mathcal{M}_v)^{-1}\mathcal{G}$
- $\hat{\mathcal{C}}$ approximation to \mathcal{C} ,
- $\mathcal{H}_1, \mathcal{H}_2$ cheap approximation to \mathcal{C}^{-1}
- \mathcal{Q}, \mathcal{R} pressure correction matrices, depend on the chosen factorization method.

$$\mathcal{A} \approx \begin{bmatrix} \hat{\mathcal{C}} & \\ \mathcal{D} & -\hat{\mathcal{S}} \end{bmatrix} \begin{bmatrix} \mathbf{I} & \mathcal{H}_2\mathcal{G}\mathcal{R} \\ & \mathcal{Q} \end{bmatrix}$$

- Factorization method:

High-Order Yosida (HOY)

(momentum preserving)

$$\mathcal{A}_{HOY} = \begin{bmatrix} \hat{\mathcal{C}} & \\ \mathcal{D} & -\hat{\mathcal{S}} \end{bmatrix} \begin{bmatrix} \mathbf{I} & \mathcal{C}^{-1}\mathcal{G} \\ & \mathcal{Q} \end{bmatrix}$$

$$\mathcal{H}_1 = (\frac{\alpha}{\Delta t}\mathcal{M})^{-1}, \quad \mathcal{H}_2 = \mathcal{C}^{-1}$$

$$\mathcal{R} = \mathbf{I}, \quad \mathcal{Q} = (\mathcal{D}\mathcal{H}_1\mathcal{C} \mathcal{H}_1\mathcal{G})^{-1}\hat{\mathcal{S}}$$

$$\mathcal{D}_k = -\mathcal{D}(-\mathcal{H}_1\mathcal{F})^k\mathcal{G}, \quad k > 0 \quad \mathcal{F} = \nu\mathcal{K} + \mathcal{N}(\mathbf{u})$$

Pressure correction step:

$$\mathbf{y} = \mathcal{Q}^{-1}\mathbf{x} = \begin{cases} \hat{\mathcal{S}}\mathbf{z}_1 = \mathcal{D}_1\mathbf{z}_0 \\ \hat{\mathcal{S}}\mathbf{z}_2 = \mathcal{D}_2\mathbf{z}_1 + \mathcal{D}_1\mathbf{z}_0 \\ \vdots \\ \mathbf{y} = \sum_{l=0}^p \mathbf{z}_l \end{cases}$$

HOY

- Time adaptivity:

Adaptive selection of time step based on a *posteriori* pressure-correction based error estimator

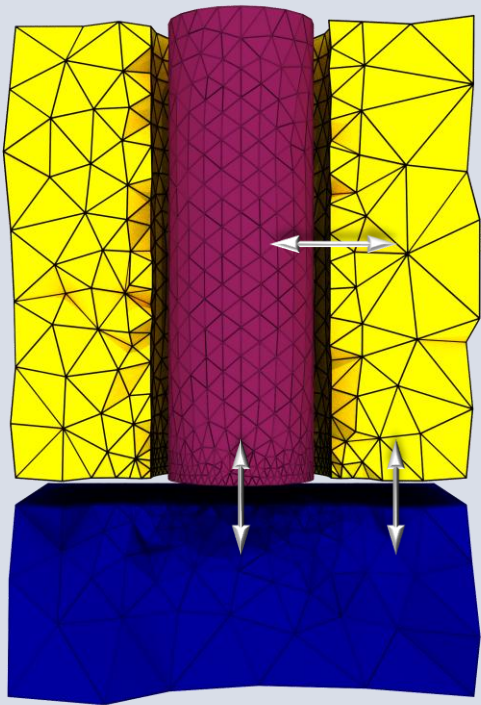
- α Safety factor for time adaptivity
- ϵ : Tolerance for time adaptivity
- z_q pressure correction of order q

$$\Delta t = \chi \Delta t_{old}$$

$$\chi = \min\left(\max\left(\alpha \cdot \left(\frac{\epsilon \Delta t_{old}}{\|\mathbf{z}_q\|}\right)^{\frac{1}{q}}, \chi_{min}\right), \chi_{max}\right)$$

[1] Alessandro Veneziani. Block factorized preconditioners for high-order accurate in time approximation of the navier-stokes equations. Numerical Methods for Partial Differential Equations, 19(4):487–510, 2003.

[2] Alessandro Veneziani and Umberto Villa. Aladins: An algebraic splitting time adaptive solver for the incompressible navier–stokes equations. Journal of Computational Physics, 238:359–375, 2013.



03.

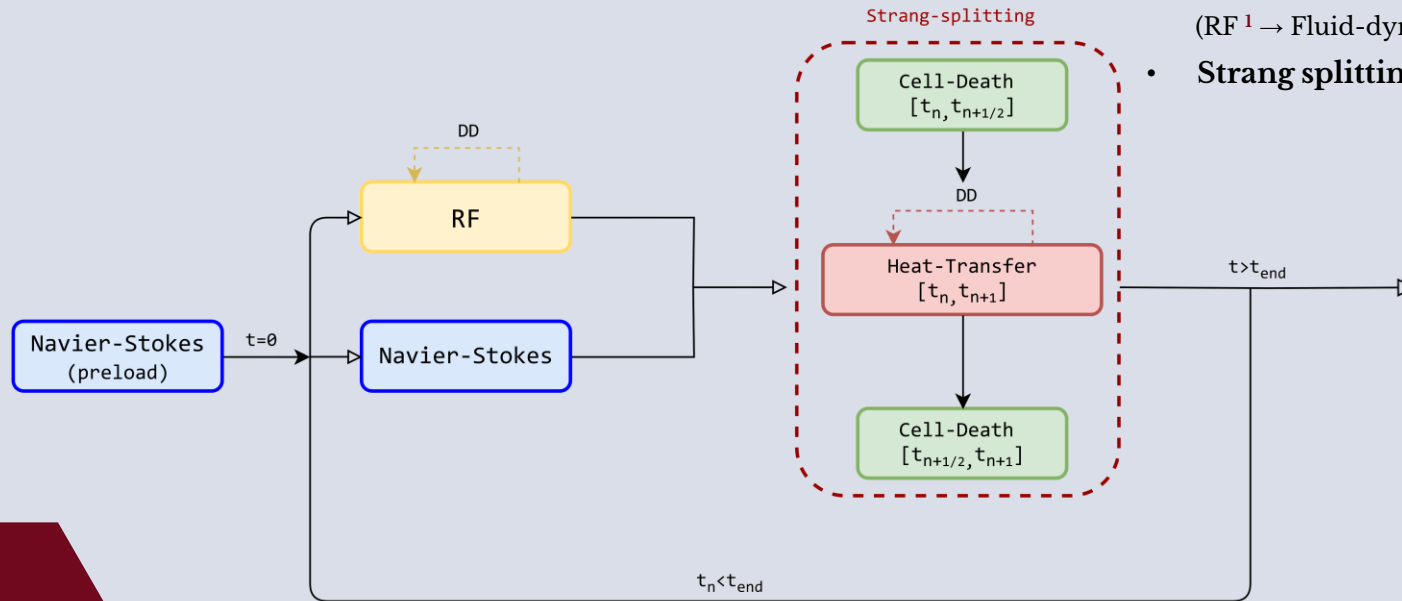
Domain Decomposition

Modeling **interface** between *tissue*, *electrode* and *fluid* domains

Two-level segregation approach

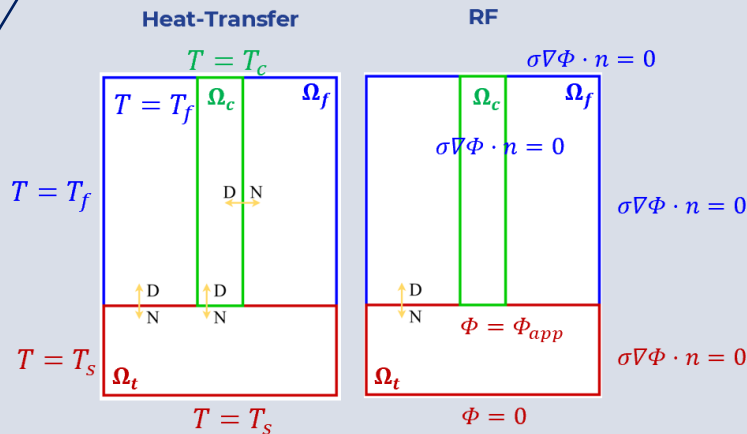
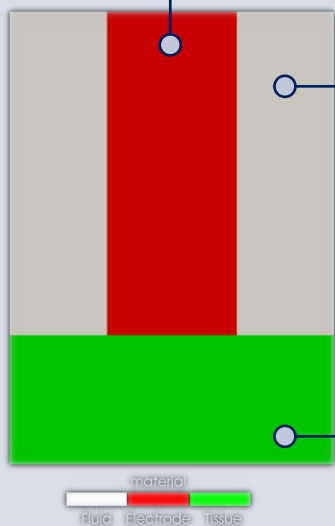
1. Multiphysics segregation

- **Staggered approach** for multiphysics
(RF ¹ → Fluid-dynamics → Heat transfer ¹ → Cell-death)
- **Strang splitting** for HT/CD problems $O(\Delta t^2)$

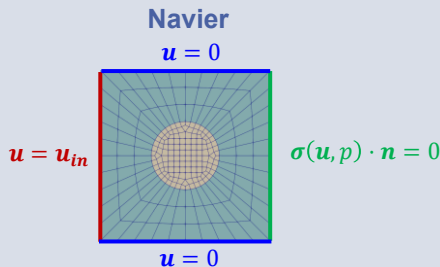


Two-level segregation approach

- Heat transfer
- RF
- Fluid dynamics
- Heat transfer (*advection-diffusion*)



- RF
- Heat transfer (*joule heating*)
- Cell-death



1. Multiphysics segregation

- Staggered approach** for multiphysics
(RF ¹ → Fluid-dynamics → Heat transfer ¹ → Cell-death)

2. Substructuring by (physical) subdomains

- Nonoverlapping domain decomposition**
Based on physical domains
- Dirichlet-Neumann, Robin-Robin coupling (DN, RR) ¹**
For transmission conditions at the physical interfaces

$$\begin{aligned}
 \text{RF} \quad & \nabla \cdot \sigma_1 \nabla \phi_1^{n+1} = f_1 \quad \text{in } \Omega_1 \\
 & \phi_1^{n+1} = \phi_2^n \quad \text{on } \Gamma_{12} \\
 & \nabla \cdot \sigma_2 \nabla \phi_2^{n+1} = f_2 \quad \text{in } \Omega_2 \\
 & \sigma_2 \nabla \phi_2^{n+1} \cdot \mathbf{n} = \sigma_1 \nabla \phi_1^{n+1} \cdot \mathbf{n} \quad \text{on } \Gamma_{12}
 \end{aligned}$$



04.

Results

Solvers verifications, Domain Decomposition convergence, RFA

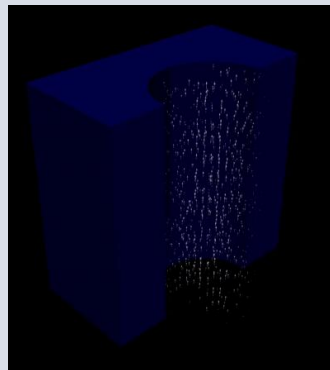
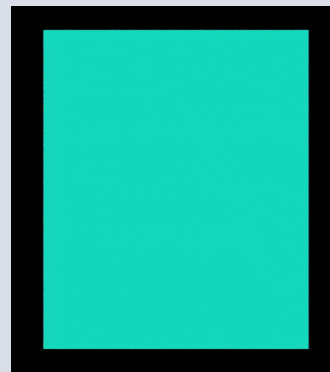
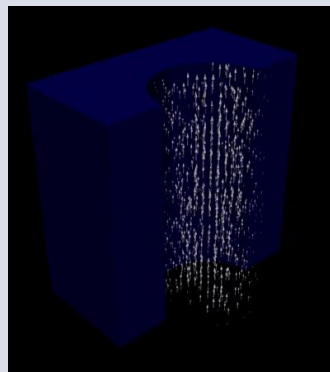
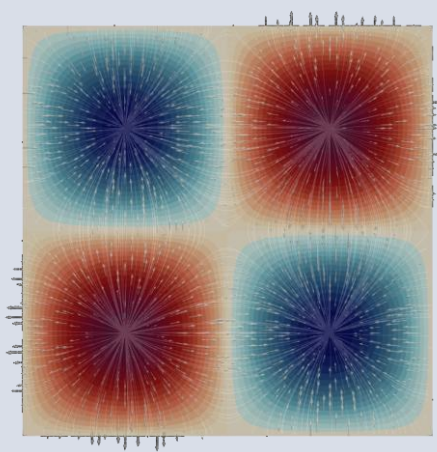
Solvers verification: Heat transfer

Convergence (MMS on squared domain) + toy problems

Advection-diffusion

Time-dependent
square plate heating

$$T(t) = \sin(k \pi x) \sin(k \pi y) + \beta t, \quad k = 2, \beta = 1.1$$



Heat-Transfer ($\sigma = 2$)

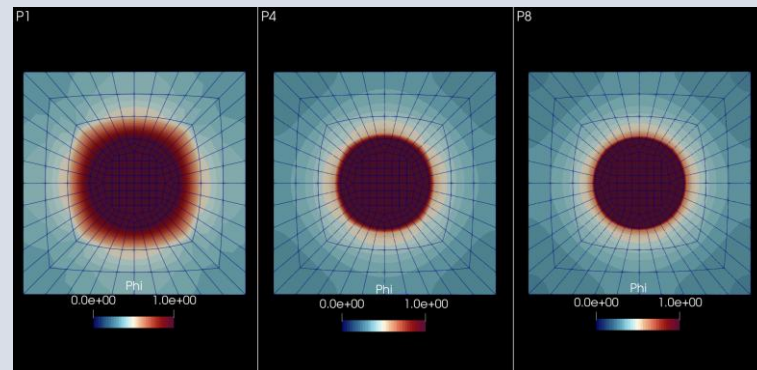
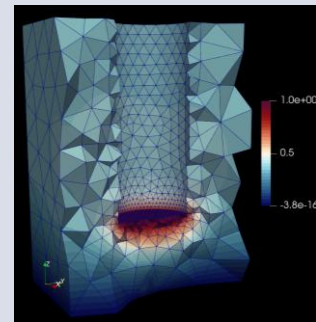
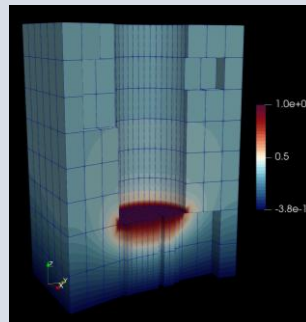
DOFs	h	L^2 error	L^2 rate	H^1 error	H^1 rate
81	0.25	0.01445	0	0.4044	0
289	0.125	0.001933	2.902	0.102	1.987
1089	0.0625	0.0002451	2.979	0.02553	1.998
4225	0.03125	3.075e-05	2.995	0.006383	2
16641	0.01562	3.847e-06	2.999	0.001596	2

Convective cooling of
sphere

Solvers verification: RF

Convergence (MMS on squared domain) + toy problems

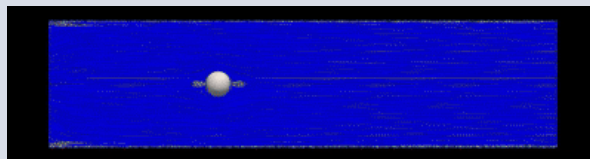
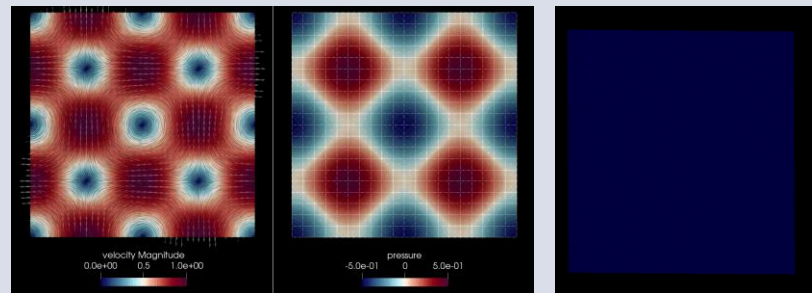
RF					
Dofs	h	L^2 error	L^2 rate	H^1 error	H^1 rate
289	0.2686	2.425×10^{-5}	0	0.001126	0
1089	0.1343	7.761×10^{-7}	4.965	7.144×10^{-5}	3.979
4225	0.06716	2.443×10^{-8}	4.99	4.478×10^{-6}	3.996
16641	0.03358	8.309×10^{-10}	4.878	2.802×10^{-7}	3.998



Solvers verification: Navier-Stokes

Convergence (MMS on squared domain) + toy problems

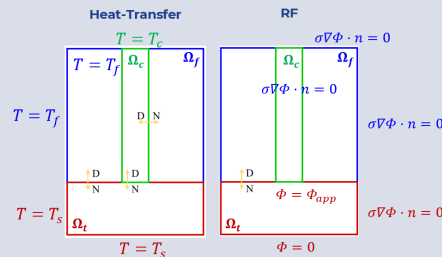
Navier-Stokes						
DOFs u	DOFs p	h	L^2 error (u)	L^2 error (p)	L^2 rate (u)	L^2 rate (p)
242	36	0.2	1.99644×10^{-2}	0.110601	0	0
882	121	0.1	5.26485×10^{-3}	0.0185581	1.92297	2.57525
3362	441	0.05	7.36945×10^{-4}	0.00379613	2.83676	2.28945
13122	1681	0.025	5.4351×10^{-5}	9.36223×10^{-4}	3.76118	2.0196
51842	6561	0.0125	4.3819×10^{-6}	2.99535×10^{-4}	3.63268	1.64413



Domain Decomposition convergence

Subiteration convergence

Dirichlet-Neumann scheme exhibits **linear convergence** for Heat/RF problems.

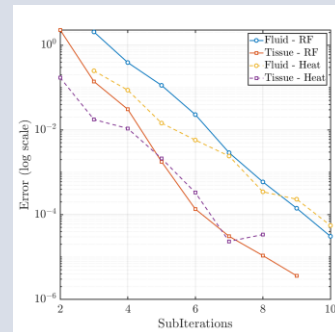


h-refinement

Consistent results for DD convergence across different levels of **mesh refinement**. For RF, RR conditions did not lead to significant improvements (but we did not analyze the optimal parameters derived from Fourier Analysis)

p-refinement

Slower convergence rates for the DD algorithm



h-Refinement

Ref	Dofs RF	Dofs H	It-RF	It-Heat
0	90k	19k	10	7.4 ± 0.70
1	673k	150k	9	7.4 ± 0.52
2	5M	1.1M	9	7.7 ± 0.67

(a) Iteration count for DD convergence ($tol = 10^{-6}$) after **h-refinement** (uniform refinement). For heat problem, iterations taken over the first 10 timesteps. H = Heat, RF = RadioFrequency.

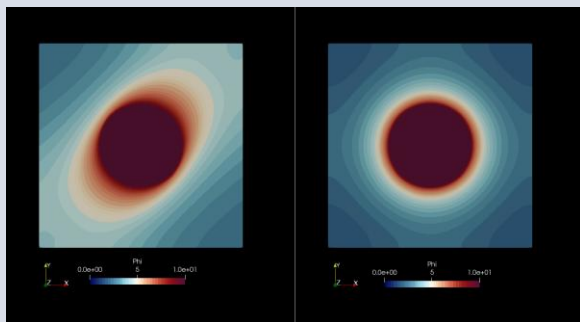
p-Refinement

Order (H-RF)	Dofs RF	Dofs H	It-RF	It-Heat
$Q_1 - Q_2$	13k	3k	10	7.5 ± 0.53
$Q_2 - Q_4$	90k	15k	10	7.4 ± 0.70
$Q_4 - Q_8$	673k	150k	15	12.9 ± 2.92

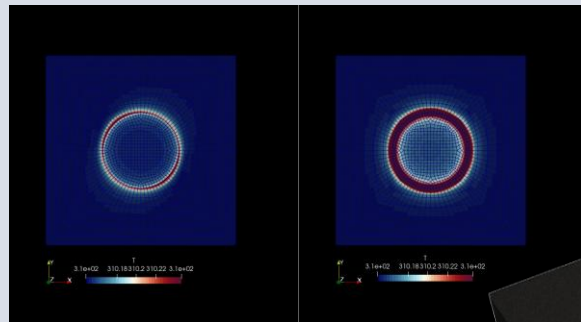
(b) Iteration count for DD convergence ($tol = 10^{-6}$) after **p-refinement**. For heat problem, iterations taken over the first 10 timesteps. H = Heat, RF = RadioFrequency.

Anisotropy

RF, tissue only

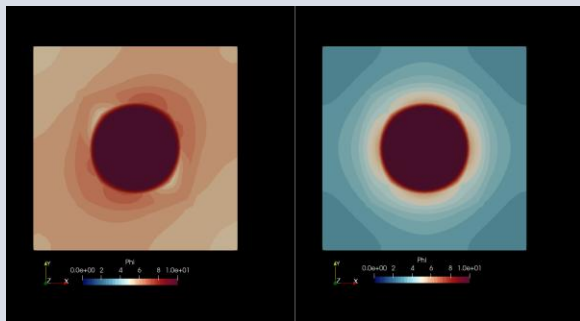


Electric potential

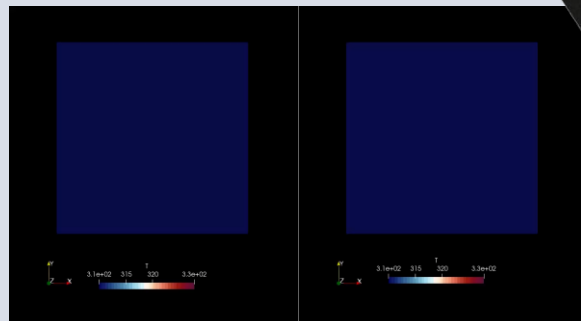


Electric field

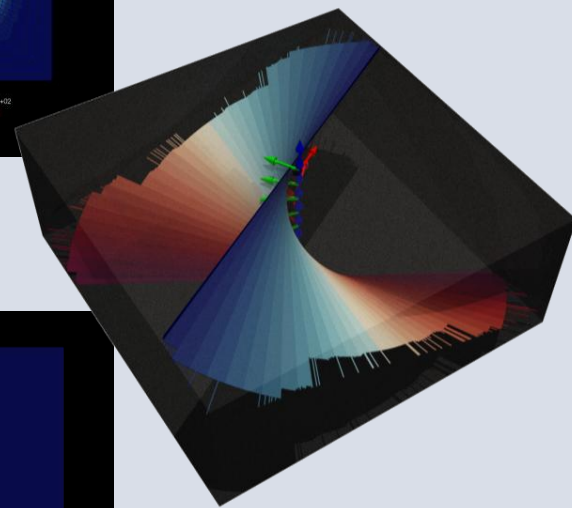
RFA, three domains



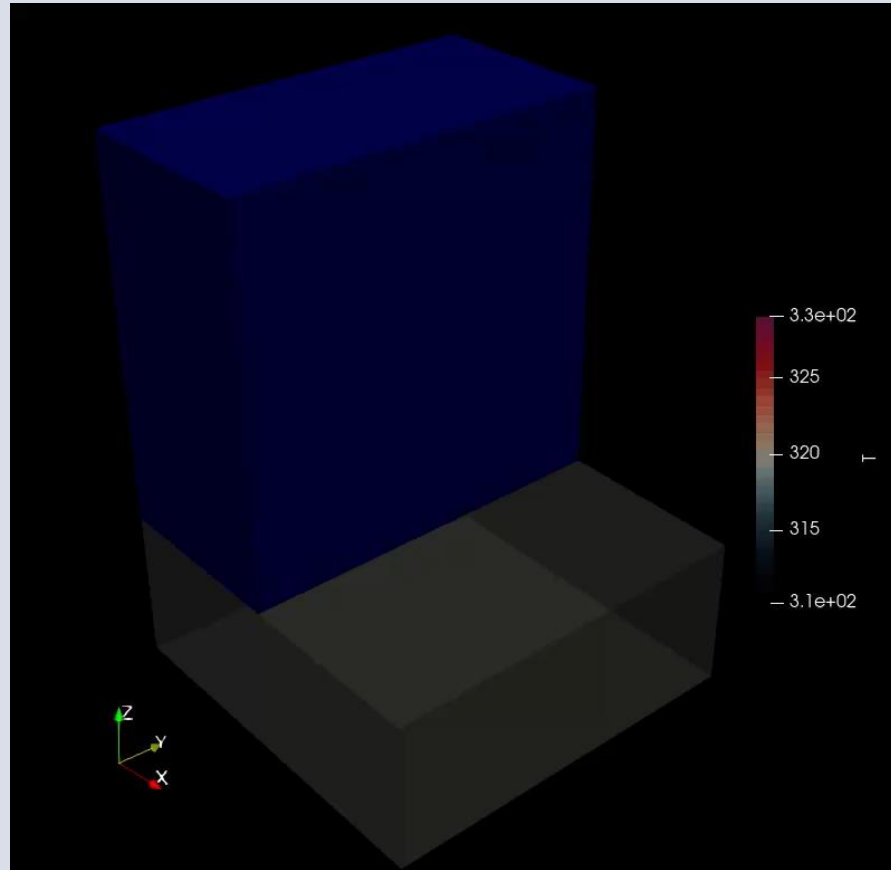
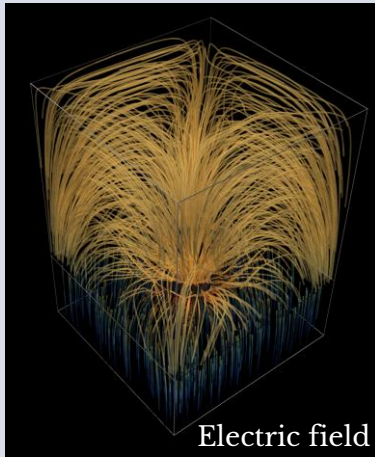
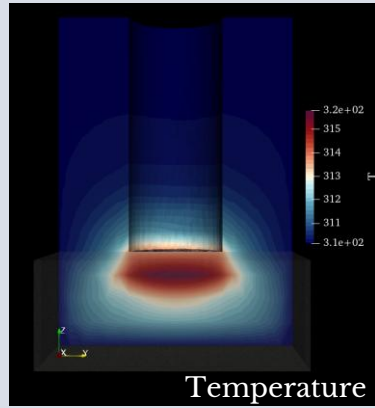
Electric potential



Temperature



(pseudo) RFA simulation



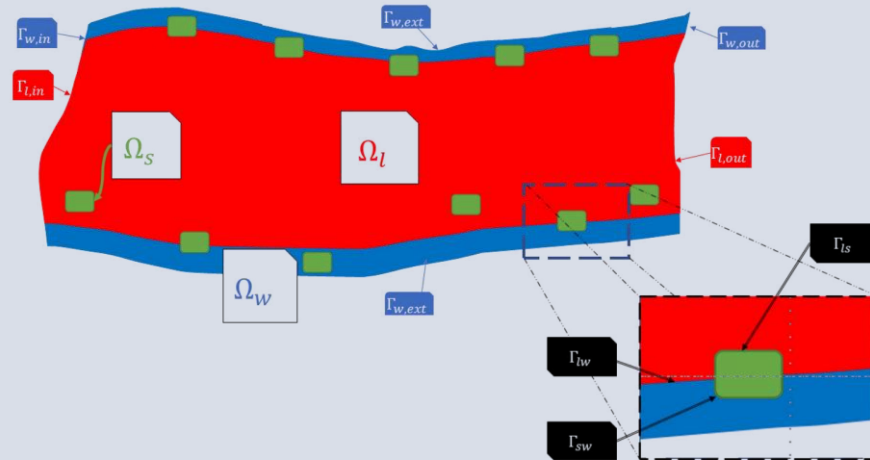


05.

Future work

Other applications

- ⬡ Different ablation sources: MWA, HIFU, Laser
- ⬡ Different ablated tissues: Liver, Intestine, Prostate, Bone
- ⬡ Erosion of bioresorbable stents



Future plan

⬡ Multiphysics model

- **1.1 Fluid dynamics:**

- 💡 Strongly consistent stabilization (SUPG, GLS, VMS), and turbulence model (EFR)

- **1.2 Heat transfer:**

- 💡 Extend tissue bioheat equation to **Non-Fourier models** (*SPL, DPL*) $\rho c \frac{\partial^2 T}{\partial t^2}, \quad \rho c \nabla \cdot (k \frac{\partial \nabla T}{\partial t})$

- **1.3 Mechanics:**

- 💡 Develop quasi-incompressible, hyperelastic, anisotropic **model for tissue mechanics**
 - 💡 Develop **electro-tissue contact model** (leverage MFEM-Tribol interface)

$$\sum_{k \in \mathcal{T}} (\mathcal{R}_\phi, \tau \mathcal{L}_{stab}(w)),$$

$$\mathcal{L}_{stab} = \begin{cases} a \cdot \nabla & SUPG \\ -a \cdot \nabla + \nu \Delta & GLS \\ -a \cdot \nabla - \nu \Delta & VMS \end{cases}$$

⬡ Improved computational efficiency

- 💡 **High fidelity simulations:** Extend PA to all solvers and enable GPU acceleration
 - 💡 **Clinically-suitable simulations:** investigate use of **Reduced Order Modeling (ROM)** techniques (**libROM**)

⬡ Ultimate goal: Surgical planning/therapy optimization

- 💡 Identification of optimality criteria
 - 💡 ROM techniques to accelerate identification of optimal solution
 - 💡 Ex-vivo and Pre-clinical Validation

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Scan me!

Thank you!