

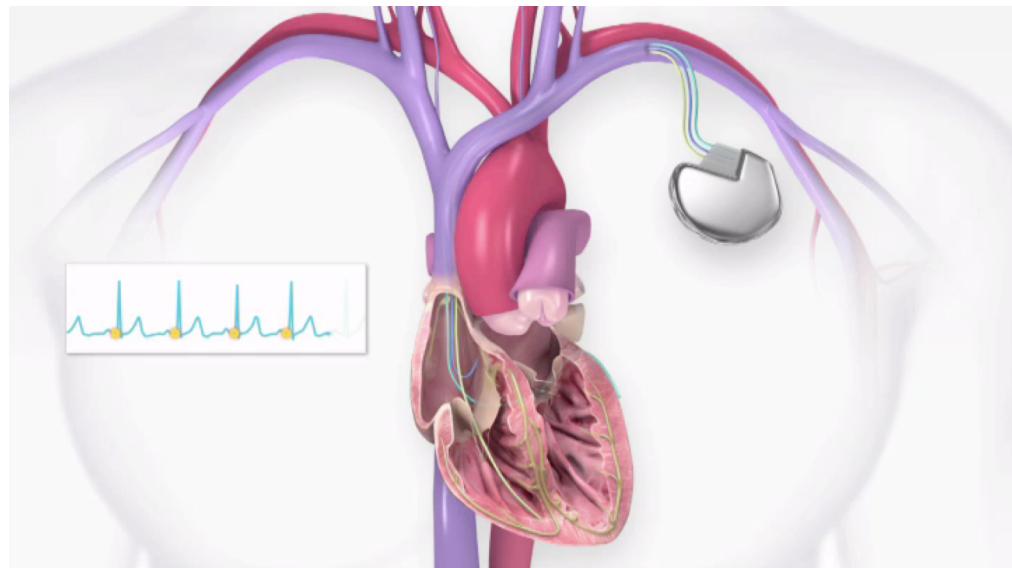
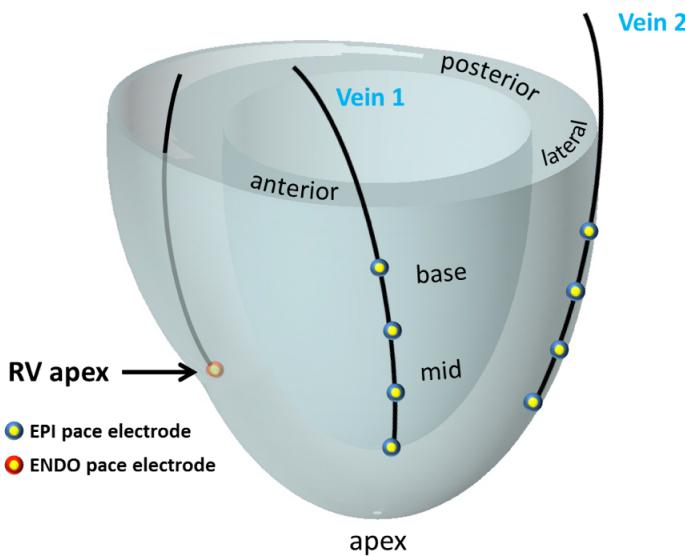
Heart diseases is the no. 1 cause of death in the Western World



- **31 % of all deaths**
- **5 % of people between 60 and 70 years live with heart failure**

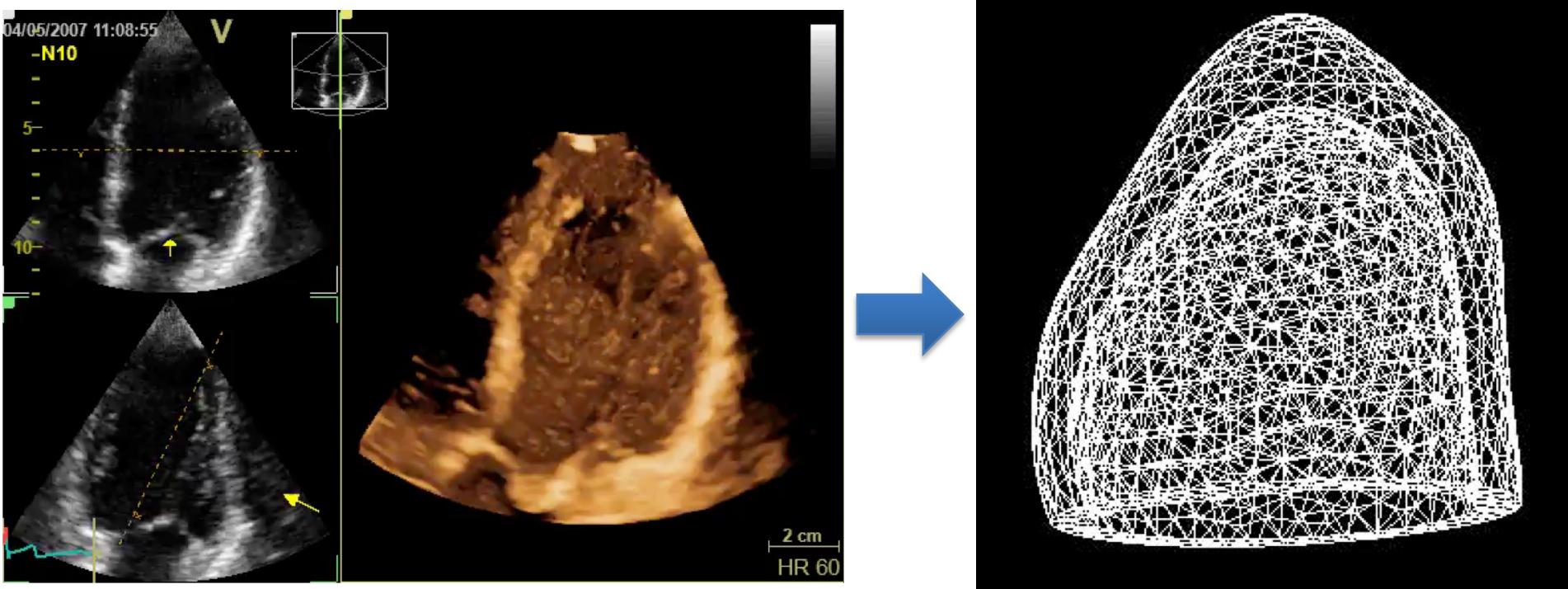
Understanding the mechanical events occurring during a cardiac cycle may improve understanding of optimal treatment

Electrode positions



30 – 40 % do not respond

Patient Specific modeling of the heart can provide us with insight and optimize patient treatment



FEniCS' 16

Optimization of a Spatially Varying Cardiac Contraction Parameter using the Adjoint Method

Henrik Finsberg*, Gabriel Balaban, Joakim Sundnes, Marie Rognes, Hans Henrik Odland, Stian Ross, Samuel Wall

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The problem is formulated as a PDE-Constrained Optimization problem

$$\begin{aligned} & \underset{\gamma}{\text{minimize}} && I(\mathbf{u}, \gamma) \\ & \text{subject to} && R(\mathbf{u}, \gamma) = 0 \\ & && \gamma \in [0, 1) \end{aligned}$$

I : Misfit functional between observed and simulated data

R : Force-balance equation

\mathbf{u} : displacement field

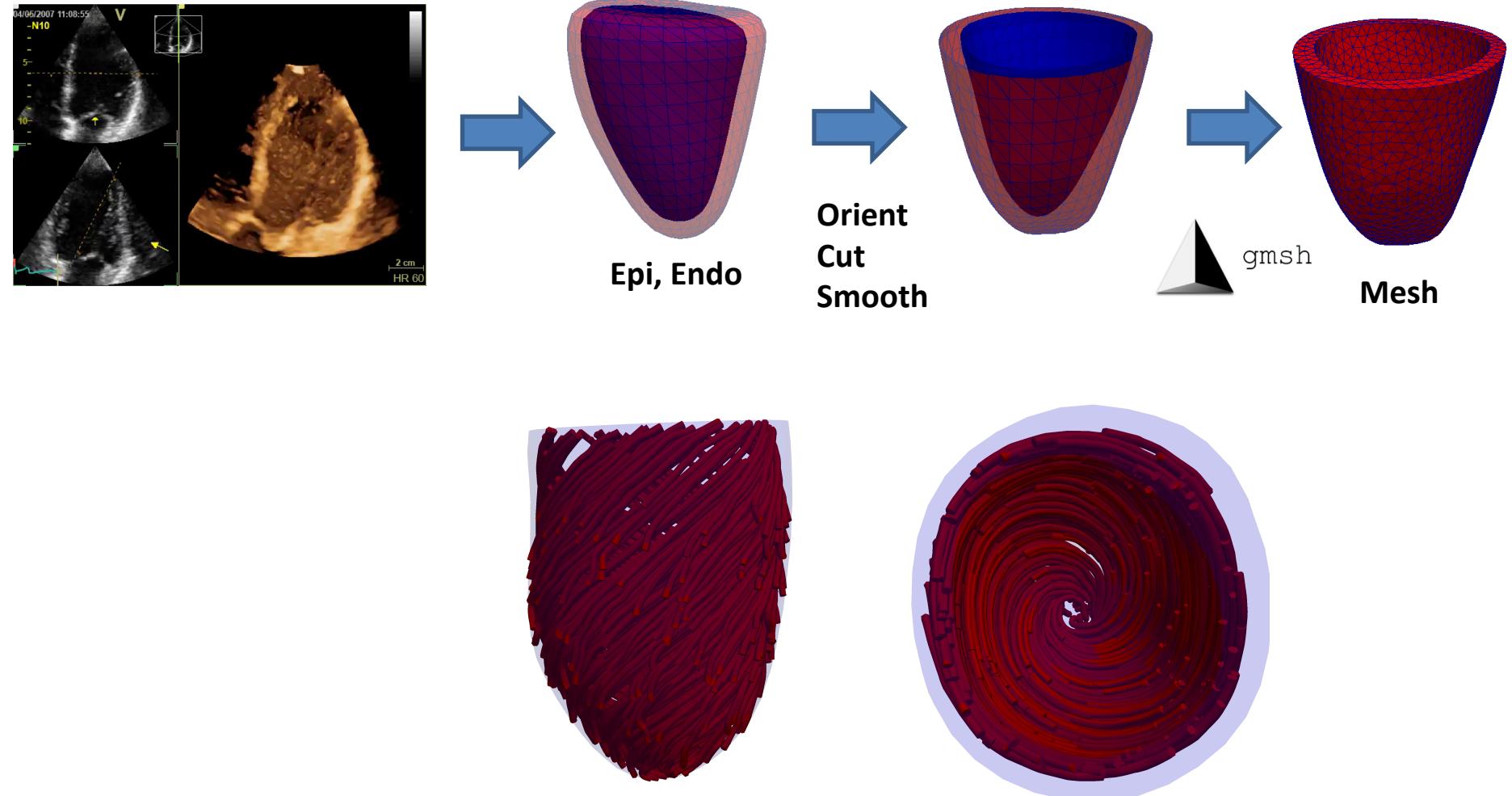
$\gamma = \gamma(\mathbf{X})$: contraction field



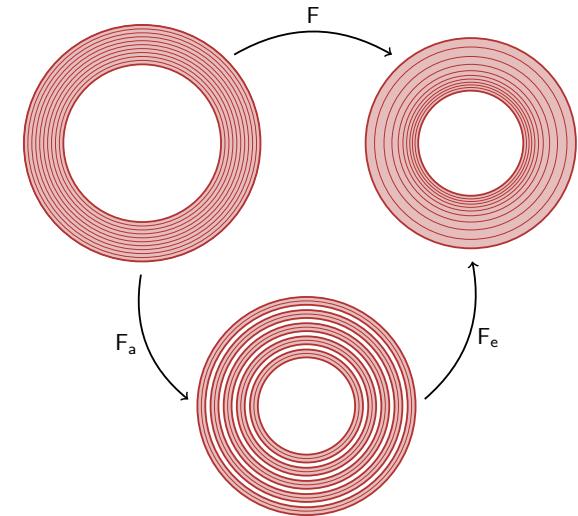
dolfin-adjoint



The geometry is extracted from 4D echocardiography,
fiber orientations are assigned using a rule-based algorithm



The mechanical model is based on the active strain framework



Active Strain Decomposition:

$$\mathbf{F} = \mathbf{F}_e \mathbf{F}_a$$

Passive Mechanics:

$$\mathcal{W}(\mathbf{C}_e) = \frac{a}{2b} \left(e^{b(I_1 - 3)} - 1 \right) + \frac{a_f}{2b_f} \left(e^{b_f(I_{4,\mathbf{f}_0} - 1)^2} - 1 \right)$$

$$\mathbf{C}_e = \mathbf{F}_e^T \mathbf{F}_e \quad I_{4,\mathbf{f}_0} = \mathbf{f}_0 \cdot (\mathbf{C}_e \mathbf{f}_0) \quad I_1 = \text{tr}(\mathbf{C}_e)$$

Active Mechanics:

$$\mathbf{F}_a = (1 - \gamma) \mathbf{f}_0 \otimes \mathbf{f}_0 + \frac{1}{\sqrt{1 - \gamma}} (\mathbf{I} - \mathbf{f}_0 \otimes \mathbf{f}_0)$$

parameters = # vertices

Boundary Conditions:

Endocardial pressure from clinical data :

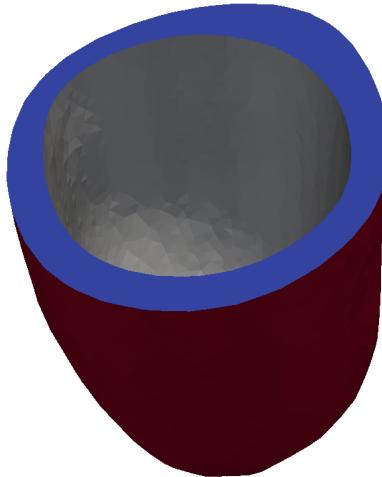
$$\mathbf{T} = J p_{LV} \mathbf{F}^{-T} \mathbf{N}, \mathbf{x} \in \partial\Omega_{endo}$$

Fix base in longitudinal direction. Allow some displacement in the base plane:

$$\mathbf{u} = (u_1, u_2, u_3)$$

$$u_1 = 0, \mathbf{x} \in \partial\Omega_{base}$$

$$\mathbf{T} = -k\mathbf{u}, \mathbf{x} \in \partial\Omega_{base}$$



$\partial\Omega_{epi}$
 $\partial\Omega_{endo}$
 $\partial\Omega_{base}$

The force-balance equation represents the minimum potential energy state

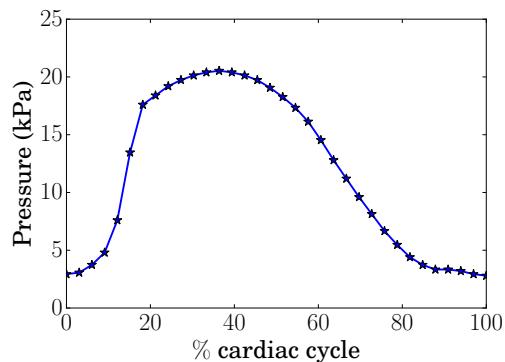
$$\Pi = \int_{\Omega} \mathcal{W}(\mathbf{C}_e) + p(J-1)dV$$

+ boundary conditions

$$R(\mathbf{u}, p) = \begin{pmatrix} D_{\delta\mathbf{u}}\Pi \\ D_{\delta p}\Pi \end{pmatrix} = \mathbf{0}$$

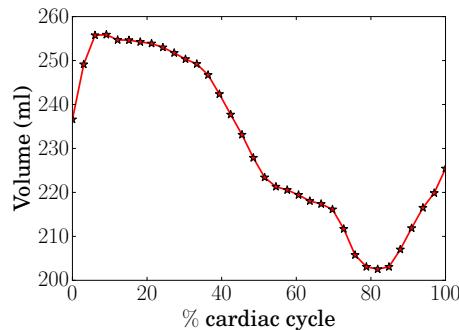
We use measured pressure, volume and regional strain to personalize the mechanics

LV Pressure:



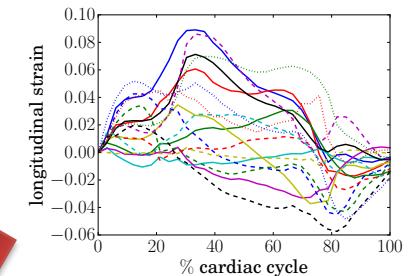
Boundary
Condition

LV Volume:

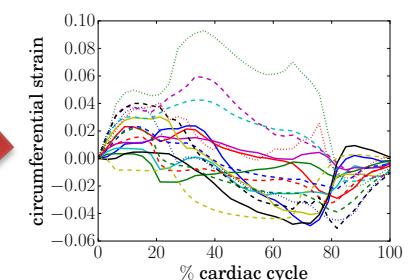


Objective
function

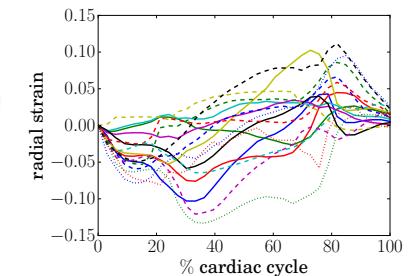
Longitudinal



Circumferential

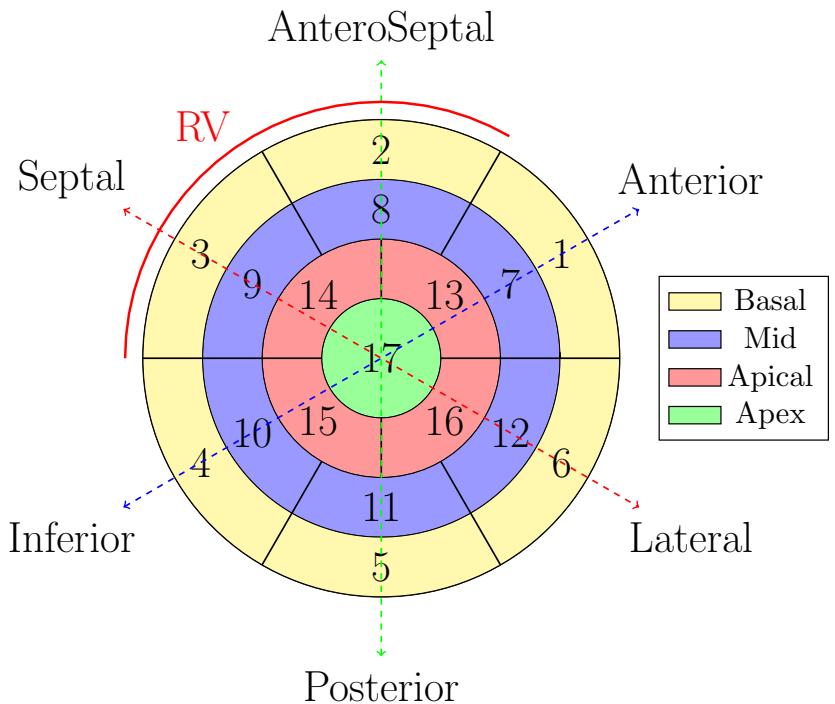
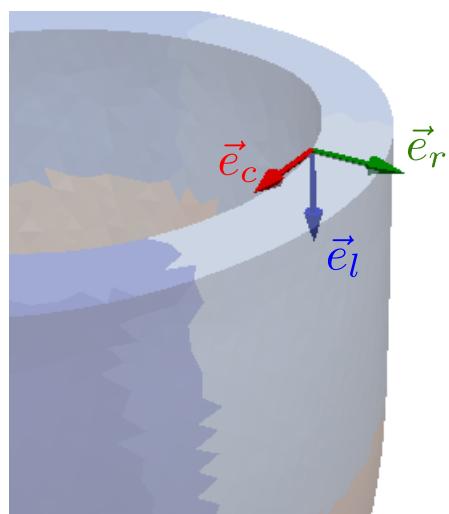
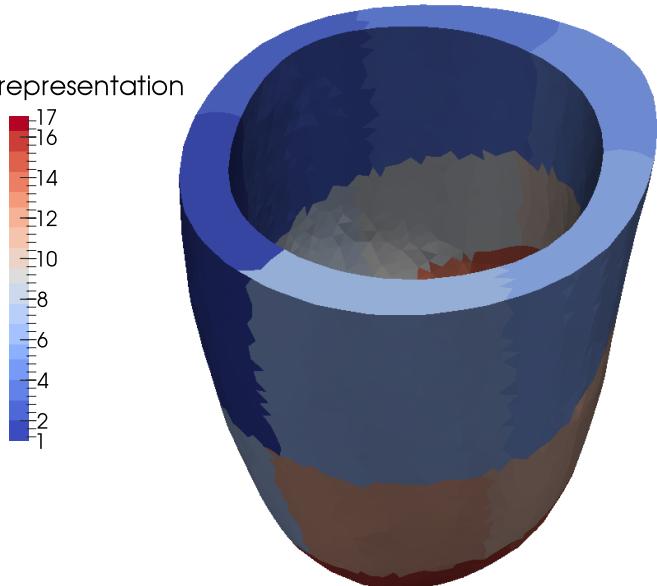


Radial



The average regional strain is computed in three directions in the left ventricle(LV)

AHA-zone representation



$$\Omega_{LV} = \bigcup_{j=1}^{17} \Omega^j$$

$$\tilde{\varepsilon}_{k,j} = \frac{1}{|\Omega^j|} \int_{\Omega^j} e_k^T \nabla \mathbf{u} \cdot e_k \, dx$$

We minimize the difference between measured and simulated strain and volume

Volume misfit functional:

$$I_{\text{vol}}^i = \left(\frac{V^i - \tilde{V}^i}{V^i} \right)^2, \quad \tilde{V}^i = -\frac{1}{3} \int_{\partial\Omega_{\text{endo}}} (\mathbf{X} + \mathbf{u}) \cdot J \mathbf{F}^{-T} \mathbf{N} \, dS,$$

Strain misfit functional:

$$I_{\text{strain}}^i = \sum_{j=1}^{17} \sum_{k \in \{c, r, l\}} (\varepsilon_{k,j}^i - \tilde{\varepsilon}_{k,j}^i)^2$$

Total misfit functional (weighted):

$$I_\alpha^i = \alpha I_{\text{vol}}^i + (1 - \alpha) I_{\text{strain}}^i$$

Parameters are estimated by solving a PDE-constrained optimization problem

Passive material parameters:

$$\text{minimize}_{\mathbf{m}} \quad \sum_{i=0}^{N_{\text{ED}}} I_{\alpha}^i$$

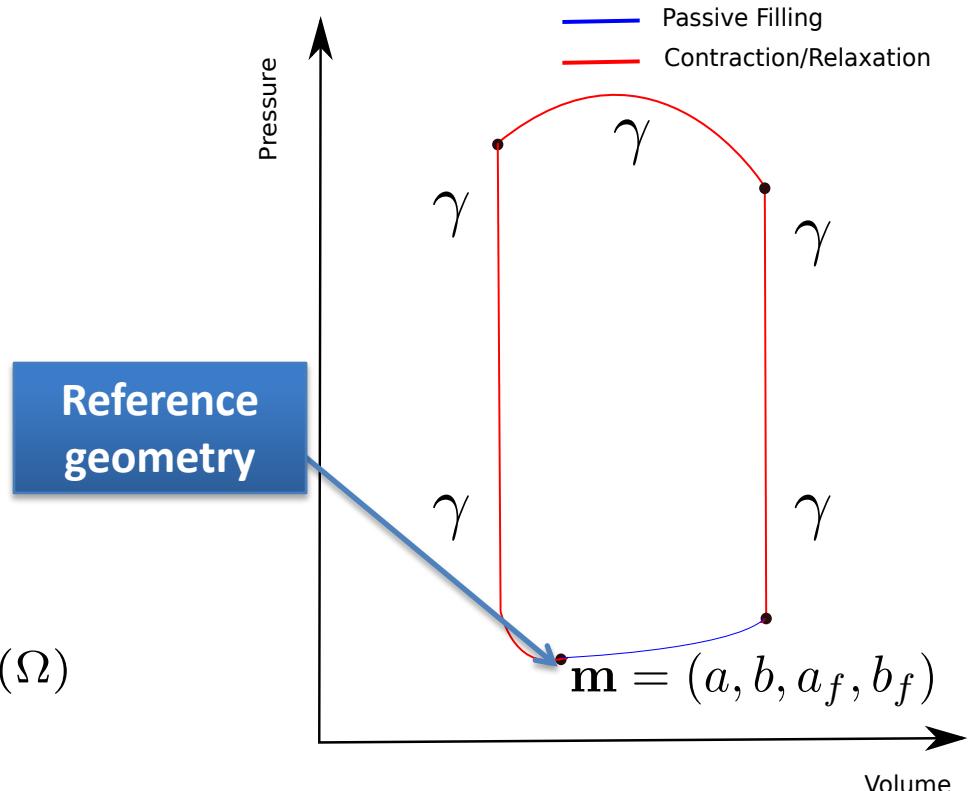
$$\text{subject to} \quad R(\mathbf{u}, p) = 0.$$

Active contraction parameter:

$$\text{minimize}_{\gamma(\mathbf{x}, i)} \quad I_{\alpha}^i + \lambda \|\nabla \gamma\|_{L^2(\Omega)}^2$$

$$\text{subject to} \quad R(\mathbf{u}, p) = 0,$$

$$\gamma(\mathbf{x}, i) \in [0, 1], \quad \mathbf{x} \in \Omega, \quad i = N_{\text{ED}} + 1, \dots, N.$$



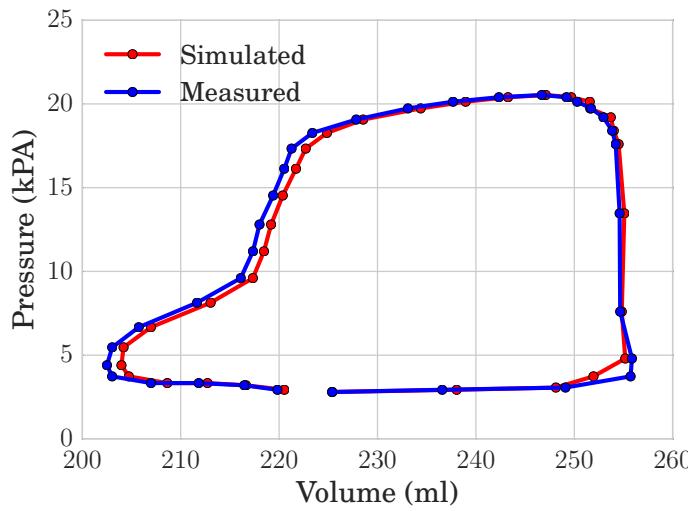
dolfin-adjoint

SciPy

FEniCS
project

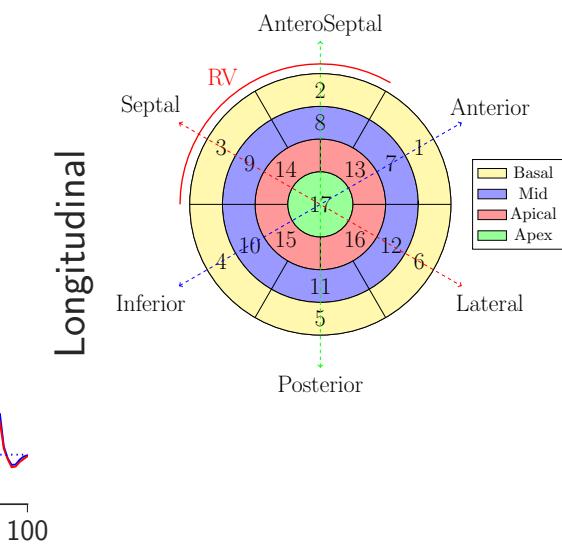
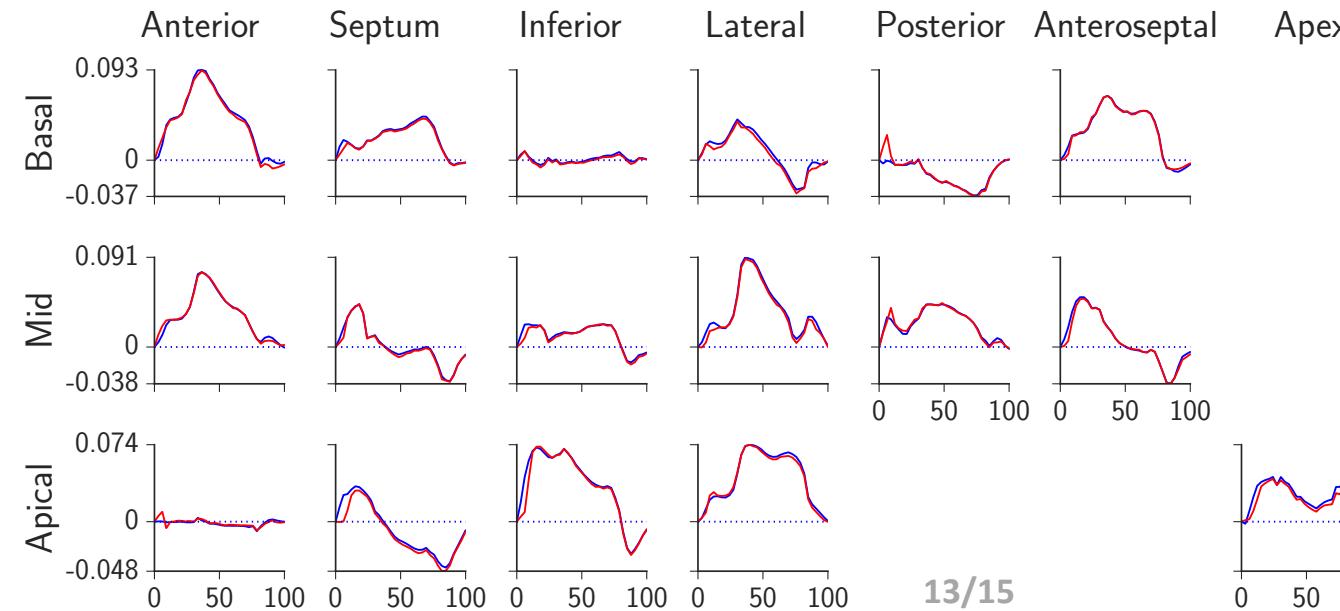


We get excellent match between simulated and measured data, at a low computational cost

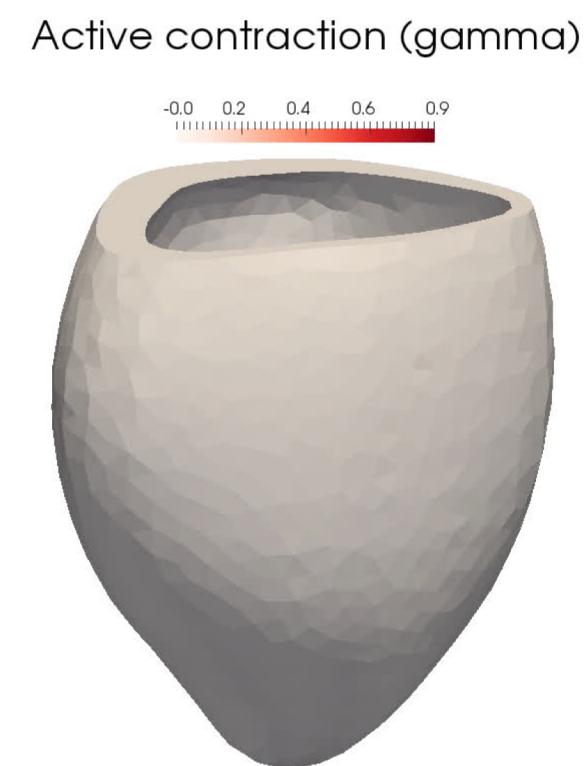
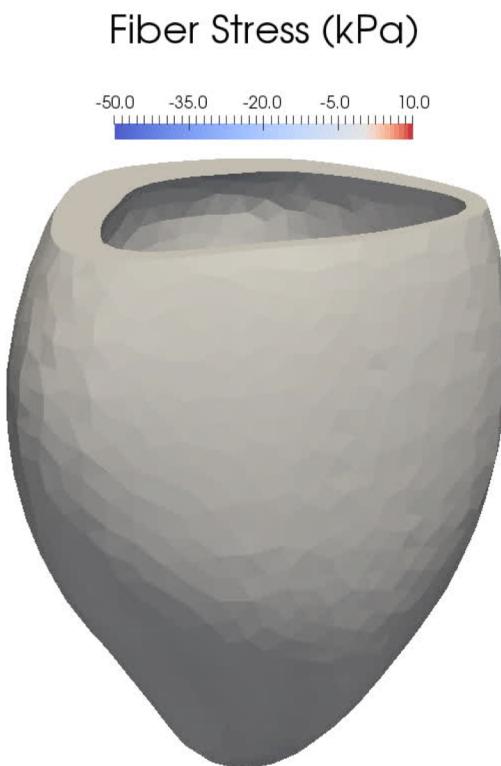


Timings running on 8 cores

Number of vertices	Functional evaluation (sec)	Derivative Evaluation (sec)
1457	3.6	13.0
2661	8.4	13.5
4812	19.3	16.5



We can use the simulation to visualize mechanical features which are impossible to measure



In summary, we are able to fit a model to clinical measurements, a key step towards bringing modeling into the clinic.

Fiber Stress (kPa)



-50.0 -35.0 -20.0 -5.0 10.0



Acknowledgements

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- Stian Ross
- Samuel Wall

Thank you!

