

# Scalable parallel solvers for finite elements and isogeometric discretizations in computational cardiology

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This talk will be on the **use of PETSc in constructing and testing scalable domain decomposition solvers** for cardiac reaction-diffusion models, from a user (not developer) perspective. **Goal/hope** is to show some of PETSc's greatest features applied to very challenging biomechanical applications:

- **Modularity:** PETSc helps greatly in experimenting with complex choices of proper submodels/solvers in order to balance biophysical accuracy vs. computational costs
- **Depth of algorithmic options,** in particular for KSP, PC, SNES
- **Portability:** thanks to PETSc our codes are not bound to machine turnover (HP Superdome, Compaq Alpha, IBM SP3, 4, 5, BG/Q, Cray XC40, + about 10 Linux Clusters ...)
- **User support and documentation:** always there and helpful

**PETSc has also been essential** in my teaching parallel computing classes and in advising PhD students

# Introduction

We will focus on the first two of the three main cardiac functions:

- *bioelectrical*: ionic currents through ionic channels of the cell membrane, excitation front generation and propagation of cardiac action potential in cardiac tissue
- *mechanical*: contraction/relaxation of cardiac muscle
- *haemodynamical*: double pump → blood flow

Several large groups involved in cardiac modeling, e.g.:

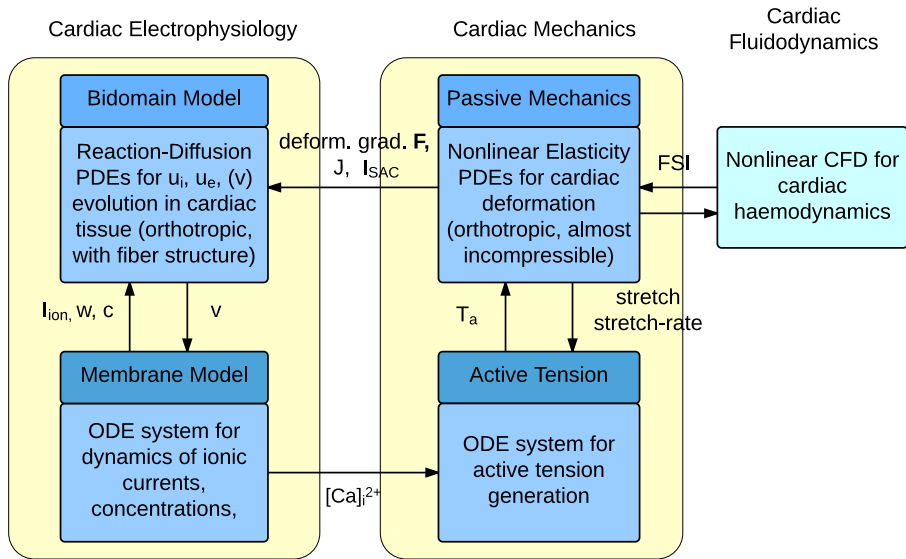
U Auckland (IUPS Physiome Project), Johns Hopkins (CCBM), INRIA (euHeart, CardioSense3D), Oxford (Chaste, COR), Simula, EPFL/Mox/Emory (LifeV, MathCard),...

Large European effort in [VPH Initiative \(Virtual Physiological Human\)](#), FP7 ICT 2007.5.3, ~ 72 M euro, [www.vph-noe.eu/home](http://www.vph-noe.eu/home)

(1 NoE, 3 IPs, 9 STREPs, 2CAs) → VPH Institute 2011 → Horizon 2020?

P. Hunter, A. Quarteroni, ... *A vision and strategy for the virtual physiological human: a 2012 update*, *Interface Focus* 3(2), 2013

# Coupled multiphysics in cardiac modeling



# 1.1 Cardiac bioelectrical model: the Bidomain system

Reaction-Diffusion system of degenerate parabolic PDEs:

- Given  $I_{app}^{i,e}$  (applied current),  $v_0, w_0$  (initial conditions)
- find  $u_i, u_e$  =intra and extracellular potentials,  
(and  $v = u_i - u_e$  = transmembrane potential),  
 $w$  =gating variables and  $c$  =ion concentrations such that:

Bidomain system (P-P formulation):

$$\chi C_m \frac{\partial v}{\partial t} - \operatorname{div}(D_i \nabla u_i) + \chi I_{ion}(v, w, c) = -I_{app}^i \quad \text{in } \Omega \times (0, T)$$

$$-\chi C_m \frac{\partial v}{\partial t} - \operatorname{div}(D_e \nabla u_e) - \chi I_{ion}(v, w, c) = I_{app}^e \quad \text{in } \Omega \times (0, T)$$

$$\frac{\partial w}{\partial t} = R(v, w), \quad \frac{\partial c}{\partial t} = S(v, w, c) \quad \text{in } \Omega \times (0, T)$$

with 0 Neumann b.c. for  $u_i, u_e$ , initial conditions for  $v, w, c$

$\chi$  = ratio membrane area/tissue volume,  $C_m$  = surface capacitance

## Conductivity tensors:

$$D_{i,e}(\mathbf{x}) = \sigma_l^{i,e} \mathbf{a}_l \mathbf{a}_l^T + \sigma_n^{i,e} \mathbf{a}_n \mathbf{a}_n^T + \sigma_t^{i,e} \mathbf{a}_t \mathbf{a}_t^T$$

$\sigma_l^{i,e}$ ,  $\sigma_n^{i,e}$ ,  $\sigma_t^{i,e}$  = conductivity coefficients along directions  $\mathbf{a}_l$ : along fiber,  $\mathbf{a}_n$ : normal to lamina,  $\mathbf{a}_t$  = tangent to lamina  
 $\Rightarrow$  electrical conductivity depends on fiber and laminar structure

## 1.2 Ionic model (Hodgkin - Huxley formalism, Nobel '63):

Ionic current  $I_{ion}$  and functions  $R, S$  in ODE systems are given by the chosen ionic membrane model:

- LR1, LRd00, LRd07, ... (ventricular, guinea pig)
- Shannon04, Mahajan07, ... (ventricular, rabbit)
- Ten Tusscher04, O'Hara-Rudy11, ... (ventricular, human)

Colli Franzone, LFP, Scacchi, *Mathematical Cardiac Electrophysiology*, Springer, 2014

# 1.3. Mechanical models of the cardiac tissue

Cardiac tissue modeled as a **nonlinear elastic material**. Notations:

- $\mathbf{X} = (X_1, X_2, X_3)^T \in \widehat{\Omega}$  undeformed cardiac domain
- $\mathbf{x} = (x_1, x_2, x_3)^T \in \Omega$  deformed cardiac domain
- $\mathbf{F}(\mathbf{X}, t) = \{F_{ij} = \frac{\partial x_i}{\partial X_j} \quad i, j = 1, 2, 3\}$  **deformation gradient tensor** and  $J(\mathbf{X}, t) = \det(\mathbf{F}(\mathbf{X}, t))$
- $\mathbf{C} = \mathbf{F}^T \mathbf{F}$  **Cauchy-Green deformation tensor**
- $\mathbf{E} = \frac{1}{2}(\mathbf{C} - \mathbf{I})$  **Lagrange-Green strain tensor** ( $\mathbf{I}$  identity)
- Div, div (Grad,  $\nabla$ ) the material, spatial divergence (gradient)

## Equilibrium equations

deformed body	undeformed body
$\operatorname{div} \boldsymbol{\sigma} = 0, \quad \mathbf{x} \in \Omega,$	$\operatorname{Div}(\mathbf{F}\mathbf{S}) = 0 \quad \mathbf{X} \in \widehat{\Omega},$

with  $\mathbf{S} = \{S_{ij}\} = J\mathbf{F}^{-1}\boldsymbol{\sigma}\mathbf{F}^{-T} =$  **2nd Piola-Kirchhoff stress tensor**

# The stress tensor: passive and active components

a) **Active stress** assumption:  $\mathbf{S}$  is the **sum** of

- an active biochemically generated component  $\mathbf{S}^{act}$ ,
- a passive elastic component  $\mathbf{S}^{pas}$ ,
- a volume component  $\mathbf{S}^{vol}$ ,

$$\mathbf{S} = \mathbf{S}^{act} + \mathbf{S}^{pas} + \mathbf{S}^{vol}$$

Most used in the literature: Nash and Hunter 2000, Vetter and McCulloch 2000; Kerckhoffs et al. 2003; Nash and Panfilov 2004; Sainte-Marie 2006; Pathmanathan and Whiteley 2009; Gotkepe and Kuhl 2010; Jie, Gurev and Trayanova 2010; Niederer, Nash, Hunter, Smith 2011; ...

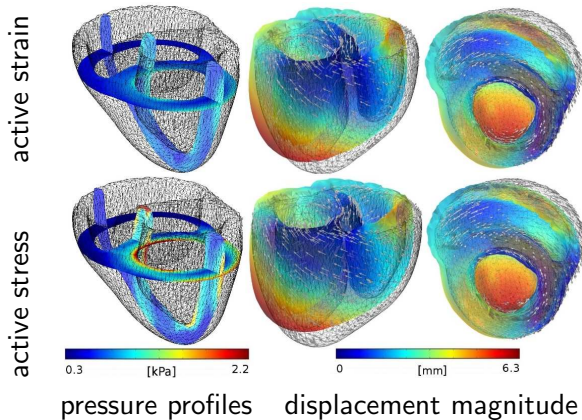
b) **Active strain** alternative assumption: **multiplicative strategy for combining the passive  $\mathbf{S}^{pas}$  and active  $\mathbf{S}^{act}$  components**,

Cherubini et al 2008, then used by Ambrosi et al. 2011, Nobile, Quarteroni and Ruiz-Baier 2012, Rossi et al. 2012...



# active stress vs. active strain

canine biventricular geometry from Ayache et al. 2007,  
orthotropic constitutive law from Holzapfel & Ogden 2009



Rossi, Ruiz-Baier, LFP, Quarteroni, *IJNMBE* 28, 2012

Passive component  $\mathbf{S}^{pas}$  given by suitable strain energy function  $W$

$$S_{ij}^{pas} = \frac{1}{2} \left( \frac{\partial W}{\partial E_{ij}} + \frac{\partial W}{\partial E_{ji}} \right) \quad i, j = 1, 2, 3.$$

We choose to model the myocardium as an **orthotropic hyperelastic material**, with the **exponential strain energy function** (see Eriksson et al. 2013)

$$W = \frac{a}{2b} \left( e^{b(\mathbf{I}_1 - 3)} - 1 \right) + \sum_{i=l,t} \frac{a_i}{2b_i} \left( e^{b_i(\mathbf{I}_{4i} - 1)^2} - 1 \right) + \frac{a_{lt}}{2b_{lt}} \left( e^{b_{lt}\mathbf{I}_{8lt}^2} - 1 \right),$$

$a$ ,  $b$ ,  $a_{(l,t)}$ ,  $b_{(l,t)}$  are positive material parameters,

$$\mathbf{I}_1 = \text{tr}(\mathbf{C}), \quad \mathbf{I}_{4l} = \hat{\mathbf{a}}_l^T \mathbf{C} \hat{\mathbf{a}}_l, \quad \mathbf{I}_{4t} = \hat{\mathbf{a}}_t^T \mathbf{C} \hat{\mathbf{a}}_t, \quad \mathbf{I}_{8lt} = \hat{\mathbf{a}}_l^T \mathbf{C} \hat{\mathbf{a}}_t,$$

$\hat{\mathbf{a}}_{l,t}$  are the directions along and across fiber

**Almost-incompressibility** of myocardium enforced by adding to the strain energy a bulk modulus  $K$  times a volume change penalization term

$$W^{vol} = K \left( \sqrt{\det(\mathbf{C})} - 1 \right)^2.$$

We assume the **active component**  $\mathbf{S}^{act}$ , **acting only in the direction of the fiber** (Pathmanathan et al. 2009, Whiteley 2007, Goktepek et al. 2010)

$$\mathbf{S}^{act} = T_a \frac{\hat{\mathbf{a}}_l \hat{\mathbf{a}}_l^T}{\hat{\mathbf{a}}_l^T \mathbf{C} \hat{\mathbf{a}}_l},$$

where  $\hat{\mathbf{a}}_l$  is the local fiber direction and  $T_a$  is the active tension.

## 1.4 Models of active tension

a)  $T_a = T_a(t, Ca_i)$  depends only on  $Ca_i$

$$\frac{dT_a}{dt} = \epsilon(Ca_i) [\eta([Ca_i - Ca_i^{rest}) - T_a]$$

where  $\epsilon(Ca_i) = \epsilon_0 + (\epsilon_\infty - \epsilon_0) \exp(-\exp(-\xi(Ca_i - Ca_i^{rest})))$   
(Kuhl et al., PBMB 2012, smooth variant of Nash, Panfilov, IJNMBE 2004)

b)  $T_a = T_a(Ca_i, \lambda)$  depends on  $Ca_i$  and fiber stretch  $\lambda = \sqrt{\hat{\mathbf{a}}_i^T \hat{\mathbf{C}} \mathbf{a}_i}$

$$T_a = \frac{Ca_i^n}{Ca_i^n + C_{50}^n} T_a^{max} (1 + \eta(\lambda - 1)) \quad (\text{Hunter et al. 1997})$$

c)  $T_a = T_a(Ca_i, \lambda, \frac{d\lambda}{dt})$  depends on  $Ca_i$ , stretch and stretch-rate  
system of 4 ODEs (Land et al., J. Physiol, 2012)

## P-P formulation

$$\begin{cases} \chi (C_m \frac{\partial v}{\partial t} + I_{ion}^{me}) - \frac{1}{j} \text{Div} (J \mathbf{F}^{-1} D_i \mathbf{F}^{-T} \text{Grad } u_i) = 0 \\ -\chi (C_m \frac{\partial v}{\partial t} + I_{ion}^{me}) - \frac{1}{j} \text{Div} (J \mathbf{F}^{-1} D_e \mathbf{F}^{-T} \text{Grad } u_e) = I_{app}^e \\ \frac{\partial w}{\partial t} = R(v, w), \quad \frac{\partial c}{\partial t} = S(v, w, c), \end{cases}$$

Mechano-electric feedback: deformation affects bioelectric phenomena, mostly during the repolarization phase.

- Conductivity coefficients modified with deformation gradient tensor  $\mathbf{F}(\mathbf{X}, t)$  and  $J(\mathbf{X}, t)$
- ionic term  $I_{ion}^{me}(v, w, c, \lambda) = I_{ion} + I_{SAC}$  augmented with the stretch-activated current  $I_{SAC}(v, \lambda)$
- possible presence of a convective term dependent on the velocity field  $\mathbf{V} = \frac{\partial \mathbf{x}(\mathbf{X}, t)}{\partial t}$  of the deformation field.

## 2. Numerical methods in space and time

$Q^1$  isoparametric FEM in space,

structured meshes based on PETSc DMDA objects

Splitting and IMEX method in time: given  $v^n, w^n, c^n, \mathbf{x}^n, \mathbf{F}^n$ ,

a. Solve the membrane model with a first order IMEX method to compute the new  $w^{n+1}, c^{n+1}$ , in particular  $Ca_i^{n+1}$

b. Solve the coupled active tension and mechanical models to compute new deformed coordinates  $\mathbf{x}^{n+1}$ , providing the new deformation gradient tensor  $\mathbf{F}^{n+1}$  and active tension  $T_a^{n+1}$ ,

c. Solve the Bidomain system. Given  $w^{n+1}, c^{n+1}, \mathbf{x}^{n+1}, \mathbf{F}^{n+1}$ , compute the new electric potentials  $u_i^{n+1}, u_e^{n+1}$ ,  
 $v^{n+1} = u_i^{n+1} - u_e^{n+1}$

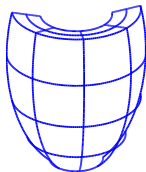
# Parallel Mechanical/Bidomain solvers at each time step

- **mechanical nonlinear system:**
  - outer iteration: Newton method
  - inner iteration (Jacobian system): GMRES, preconditioned by
  - preconditioner: Algebraic Multigrid (BoomerAMG from HYPRE) or BDDC (PCBDDC from PETSc)
- **Bidomain system (linear because of IMEX decoupling):**
  - Preconditioned Conjugate Gradient (PCG) method
  - preconditioner: Multilevel Hybrid Schwarz or PCBDDC
- **Parallel libraries:** mostly PETSc, + some MPI and HYPRE
- **Computational platforms (thanks to PETSc portability):**
  - local Linux clusters at Univ. of Milan/Pavia (small runs with  $O(10^2)$  cores)
  - MIRA BG/Q at ANL, Fermi BG/Q at CINECA, Shaheen2 Cray XC40 at Kaust, Piz Dora Cray XC40 at CSCS

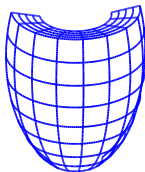
# Multilevel Additive Schwarz (MAS) preconditioners

For DD overview see: A. Toselli and O. Widlund, *Domain Decomposition Methods: Theory and Algorithms*, Springer, 2005

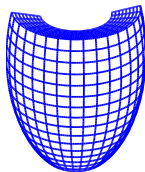
- $\mathcal{T}_k$ ,  $k = 0, \dots, L - 1$ : nested triangulations of  $\Omega$ ,  $\mathcal{T}_{L-1} = \mathcal{T}_h$



$$\mathcal{T}_0 = 4 \cdot 4 \cdot 2$$

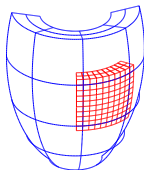
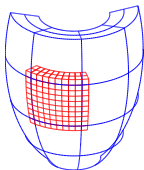
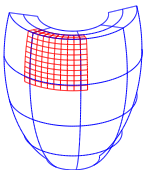


$$\mathcal{T}_1 = 2\mathcal{T}_0$$



$$\mathcal{T}_2 = 2\mathcal{T}_1 \dots$$

- $\mathcal{T}_k = \{\Omega_{km}\}_{m=1}^{N_k}$ , subdomains with overlap  $\delta_k$  and diameter  $H_k$





## Matrix form of MAS(L) preconditioner $\mathcal{P}_{MAS}^{-1}$ :

$$\mathcal{P}_{MAS}^{-1} = R_0^T B_0^{-1} R_0 + \sum_{k=1}^{L-1} \sum_{m=1}^{N_k} R_{km}^T B_{km}^{-1} R_{km}$$

- $B_{km}$  = local bidomain matrix on  $\Omega_{km}$  (level  $k$ , subdomain  $m$ )
- $R_{km}$  = restriction matrix to nodes in  $\Omega_{km}$
- $B_0$  = coarse bidomain matrix on  $\mathcal{T}_0$
- $R_0$  = restriction matrix to nodes in coarse mesh  $\mathcal{T}_0$

Structured FEM and DD data management greatly simplified by using PETSc DMDA objects  
(DMDACreate3d, DMCreateInterpolation, DMDAGetCorners,...)

## Theorem: MAS(L) convergence rate bounds

The condition number of the Multilevel Additive Schwarz operator  $\mathcal{P}_{MAS}^{-1}\mathcal{B}$  for the Bidomain system is bounded by

$$\kappa_2(\mathcal{P}_{MAS}^{-1}\mathcal{B}) \leq C \max_{k=1,\dots,L-1} \left( 1 + \frac{H_k}{\delta_k} \right)$$

with  $C$  constant independent of:

$L$  = number of levels,  $\delta_k$  = overlap at level  $k$ ,  
 $h_k$  = mesh size on level  $k$ ,  $H_k (= h_{k-1})$  subdomain diam. at level  $k$ .

Proof + numerical results in

*LFP, S. Scacchi, SIAM J. Sci. Comp., 31 (1), 2008*

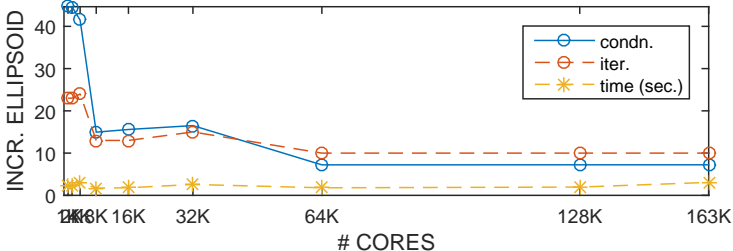
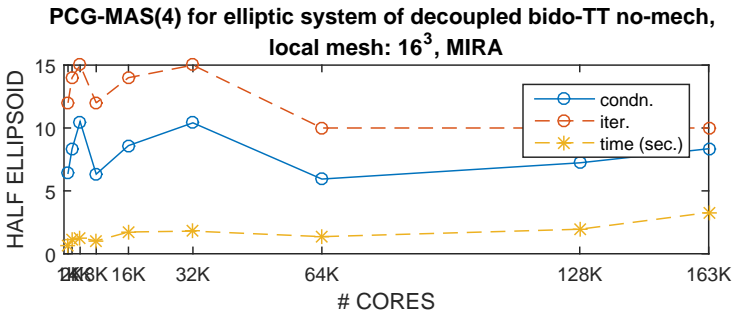
Analogous scalability bound holds for decoupled NKS MAS(2)

*M. Munteanu, LFP, S. Scacchi, SIAM J. Sci. Comp., 31 (5), 2009*

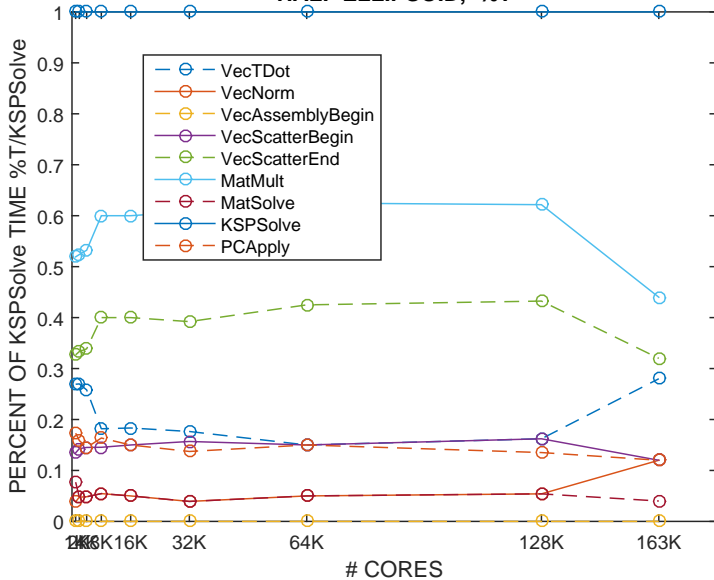
and for PE formulation with/without block-preconditioners

*LFP, S. Scacchi, SIAM J. Sci. Comp., 33 (4), 2011*

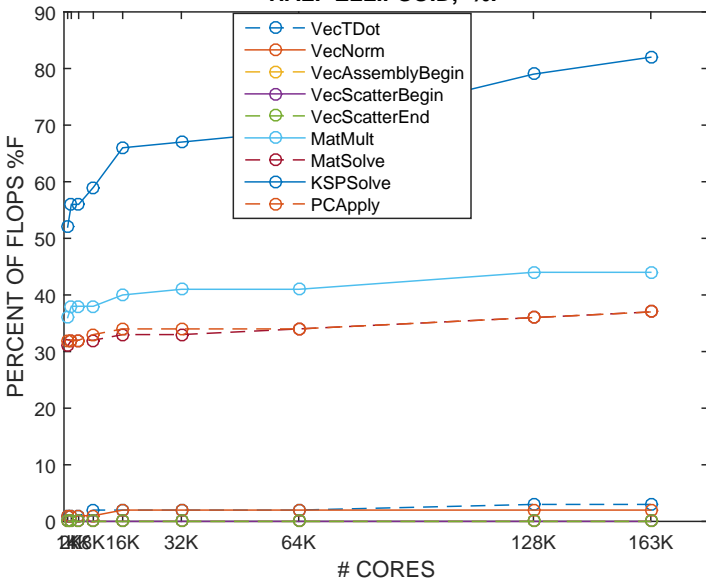
# 4.1 Bidomain parallel results (decoupled IMEX)

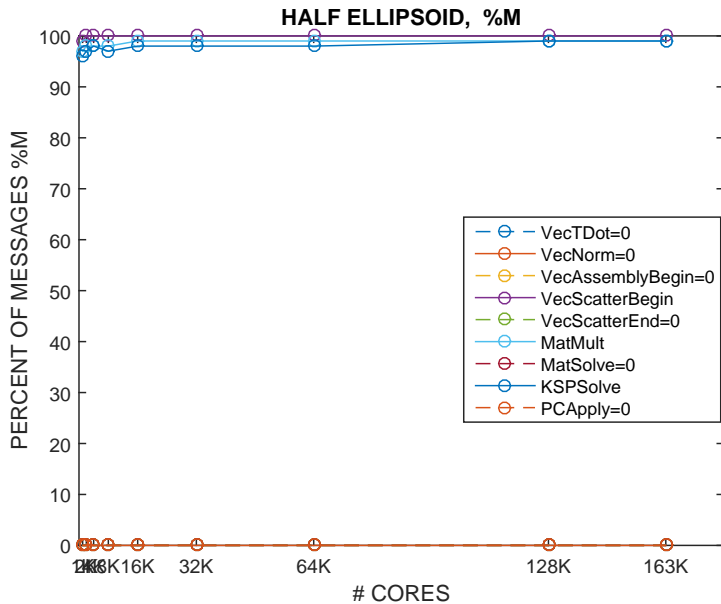


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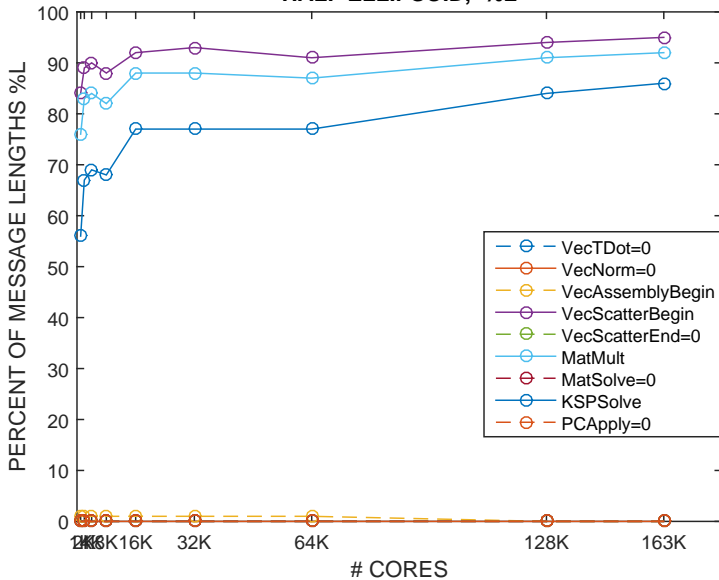


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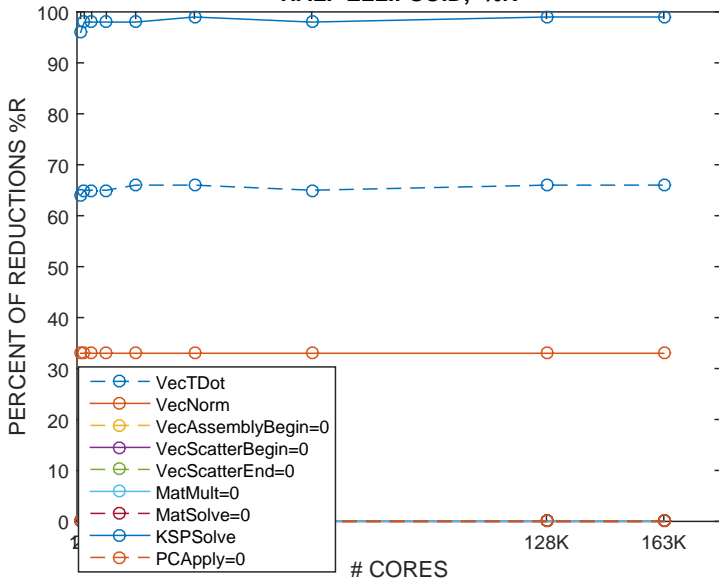




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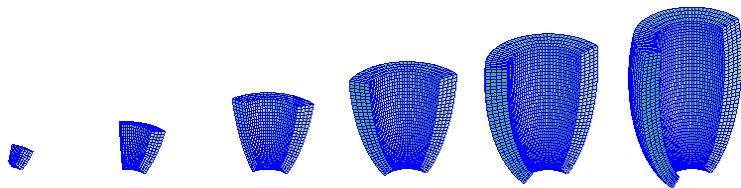




Recent extension of 2-level MAS to **Isogeometric Analysis (IGA)**  
 (L. Charawi PhD thesis, Univ of Milano, 2014 + DD22 Proc., to appear)

NURBS basis functions with  $p = 3$ ,  $k = 2$ ,  $H/h = 4$

Bidomain weak scalability test:



N	Unprec.		1-level MAS		2-level MAS	
	it.	$k_2$	it.	$\kappa_2 = \lambda_{max}/\lambda_{min}$	it.	$\kappa_2 = \lambda_{max}/\lambda_{min}$
$2 \times 2 \times 1$	765	$2.8e4$	14	$10.3=4.0/3.9e-1$	11	$5.7=4.9/8.7e-1$
$4 \times 4 \times 1$	1236	$4.9e4$	27	$58.6=4.8/8.2e-2$	10	$6.6=5.1/7.6e-1$
$6 \times 6 \times 1$	1539	$7.3e4$	35	$1.4e2=4.9/3.4e-2$	9	$6.3=5.0/8.0e-1$
$8 \times 8 \times 1$	1949	$1.0e5$	47	$2.7e2=4.9/1.8e-2$	8	$5.5=4.9/8.9e-1$
$10 \times 10 \times 1$	2180	$1.1e5$	55	$4.5e2=8.0/1.1e-2$	8	$5.5=4.9/9.0e-1$
$12 \times 12 \times 1$	2307	$1.2e5$	63	$6.7e2=4.9/7.3e-3$	8	$5.5=4.9/9.0e-1$

## 4.2 Nonoverlapping DD, decoupled IMEX (S. Zampini)

### A) BNN (Balancing Neumann-Neumann) scaled speedup on slabs

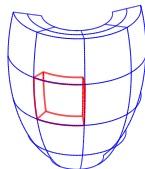
$$S_{BNN}^{-1} = S_0 + (I - S_0 S) \left( \sum_{i=1}^N D_i^T S^{i\dagger} D_i \right) (I - S S_0)$$

$S = B_{\Gamma\Gamma} - B_{\Gamma\Omega}^T B_{\Omega\Omega}^{-1} B_{\Omega\Gamma}$  (Bidomain Schur compl.)

$$B = \begin{bmatrix} B_{\Omega\Omega} & B_{\Omega\Gamma} \\ B_{\Gamma\Omega}^T & B_{\Gamma\Gamma} \end{bmatrix}.$$

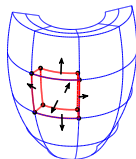
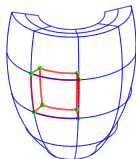
$S^{i\dagger}$  = pseudoinverse of local Schur complement  
(Neumann solver on  $\Omega_i$ )

$S_0$  = coarse solver,  $D_i$  = scaled restriction op.



procs		$\Omega$ dofs	$\Gamma$ dofs	it.	$\lambda_{min}$	$\lambda_{max}$	$\kappa_2$
16	(4x4x1)	235'225	14'325	14	1.00	5.08	5.08
64	(8x8x1)	931'225	66'325	15	1.00	5.75	5.75
144	(12x12x1)	2'088'025	155'925	16	1.00	5.91	5.91
256	(16x16x1)	3'705'625	283'125	16	1.00	5.84	5.84
400	(20x20x1)	5'784'025	447'925	16	1.00	5.93	5.93
576	(24x24x1)	8'323'225	650'325	16	1.00	5.99	5.99

## B) BDDC (Balancing Domain Decomposition with Constraints) scaled speedup on slabs



Coarse space  $\widehat{W}_\Pi$ :

a: vertex +  
edge averages

b: vertex +  
edge aver + face aver.

procs		a: $\widehat{W}_\Pi$ dofs	iter	$\lambda_{min}$	$\lambda_{max}$	$\kappa_2$
16	(4x4x1)	222	12	1.00	3.33	3.33
64	(8x8x1)	910	13	1.00	3.51	3.51
144	(12x12x1)	2046	14	1.00	3.56	3.56
256	(16x16x1)	3630	14	1.00	3.59	3.59
400	(20x20x1)	5662	14	1.00	3.59	3.59
576	(24x24x1)	8141	14	1.00	3.60	3.60

BCX/5120 Cineca cluster ([Zampini, Numer. Math. 2013](#) + [M3AS 2014](#))

more recent results on FERMI BG/Q

weak scaling with local size  $H/h = 20^3$

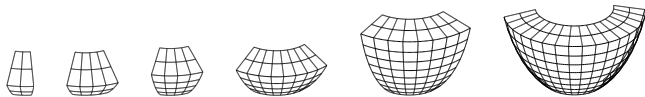
coarse problem solved by MUMPS Cholesky factorization

Uses Zampini's PETSc PCType PCBDDC and  
PCBDDCCreateFETIDPOperators

procs				BDDC		FETI-DP		
	$ \widehat{W} $	$ \widehat{W}_\Pi $	$ \widehat{W}_\Gamma $	iter	time	$ \Lambda $	iter	time
$5^3$	1.7M	1.3K	0.2M	21	0.28	0.2M	20	0.26
$10^3$	13M	9.6K	1.8M	22	0.30	2.0M	21	0.27
$15^3$	46.7M	30.8K	6.5M	22	0.36	7.1M	22	0.32
$20^3$	110.6M	70.9K	15.7M	23	0.40	17.1M	22	0.36
$25^3$	215.7M	135.9K	31.0M	23	0.53	33.8M	23	0.50
$30^3$	372.3M	231.8K	53.8M	23	0.81	58.8M	23	0.74

## 4.3 Mechanical solver: AMG weak scalability

- Simulation of 1.5 *ms* (30 time steps of  $\tau = 0.05$  *ms*) during the plateau phase on truncated ellipsoidal domains.
- Fixed local mechanical dof per subdomain: 13 476



Mechanical solver - AMG preconditioner				
procs	dof	outer iter.	inner iter.	CPU time
		Newton	GMRES	
8	107 811	2	42	12.06
27	352 947	2	42	16.70
64	823 875	2	39	23.45
125	1 594 323	2	39	30.66
216	2 738 019	2	40	49.12
512	6 440 067	2	40	75.09

# Mechanical solver: AMG strong scalability

Fixed global mechanical dof: 823 872

Mechanical solver - AMG preconditioner					
procs	local dof	nit	lit	time	speedup
8	102 984	2	41	110.84	-
16	51 492	2	40	63.61	1.74 (2)
32	25 746	2	41	34.64	3.20 (4)
64	12 873	2	39	23.26	4.76 (8)
128	6 436	2	40	16.08	6.89 (16)
256	3 218	2	40	15.50	7.15 (32)
512	1 609	2	41	16.97	6.53 (64)

nit = Newton iterations

it = CG iteration counts

time = CPU time in sec. to solve mechanical pb.

## 4.4 Bidomain - Multilevel Hybrid Schwarz weak scalability

Fixed local Bidomain dof per subdomain: 68 656

Bidomain solver - MHS(4) preconditioner							
procs	dof	non-deforming ( $\mathbf{C} = \mathbf{I}$ )			deforming		
		$\kappa_2$	it	time	$\kappa_2$	it	time
8	549 250	1.11	3	1.05	1.11	3	1.31
27	1 825 346	1.11	3	1.19	1.12	3	1.17
64	4 293 378	1.12	3	1.23	1.13	3	1.21
125	8 346 562	1.13	3	1.31	1.18	4	1.49
216	14 378 114	1.18	4	1.55	1.20	4	1.55
343	22 781 250	1.15	4	1.62	1.17	4	1.66
512	33 949 186	1.14	4	1.96	1.17	4	1.67

$\kappa_2$  = average condition number per time step

it = average CG iteration counts per time step

time = average CPU time in seconds to solve one Bidomain linear system

# Bidomain - Multilevel Hybrid Schwarz strong scalability

Fixed global Bidomain dof: 4 293 376

Bidomain solver - MHS(4) preconditioner					
procs	local dof	$\kappa_2$	it	time	speedup
8	536 672	1.13	3	9.18	-
16	268 336	1.13	3	5.16	1.78 (2)
32	134 168	1.14	3	2.62	3.50 (4)
64	67 084	1.15	3	1.30	7.06 (8)
128	33 542	1.16	4	0.72	12.75 (16)
256	16 771	1.19	4	0.48	19.12 (32)
512	8 385	1.20	4	0.26	35.31 (64)



## 4.5 Whole heartbeat: ventricular wedge deformation + v

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endo

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endo

# Whole heartbeat

Simulation of 500 ms on a truncated half ellipsoidal domain modeling half left ventricle. Number of processors = 24

Mechanical Solver: dof = 32967, time step = 0.25 ms

Prec.	<i>Newton nit</i>	<i>Total nit</i>	<i>GMRES it</i>	<i>Total it</i>	<i>time cpu</i>	<i>Tot cpu</i>
AMG	3	7031	796	6.5 ML	28.42s	15h 47m

Bidomain solver: dof = 9 655490, time step = 0.05 ms

Prec.	$\kappa_2$	$CG_{it}$	$Total_{it}$	$time_{cpu}$	$Total_{cpu}$
MAS(4)	6.18	8	81 178	1.54s	4h 16m

$\kappa_2$ : average condition number per time step

P. Colli Franzone, LFP, S. Scacchi, *Appl. Numer. Math.*, 2015

## 4.6 Better mechanical solvers: BDDC strong scalability

Land - Niederer et al. active tension model

Fixed global mesh on ellipsoidal domain:  $769 \times 769 \times 97$

3-level BDDC code by S. Zampini

procs	V			VE			VEF		
	nit	lit	time	nit	lit	time	nit	lit	time
512	3	353	31.1	3	144	21.4	3	143	22.1
1024	3	317	12.8	3	132	9.4	3	127	9.9
2048	3	464	9.5	3	196	8.1	3	189	10.2
4096	3	341	5.6	3	163	5.5	3	158	7.4
8192	3	279	6.3	3	141	6.9	3	166	8.1

Shaheen 2 (KAUST)

/medskip

LFP, S. Scacchi, S. Zampini, submitted, 2015

# Mechanical solver: 2-level BDDC weak scalability

Shaheen2 (KAUST)

Passive model by Vetter & M, active model by Goktepe et al.

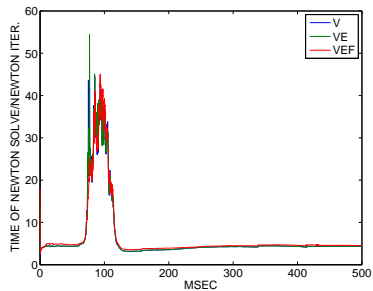
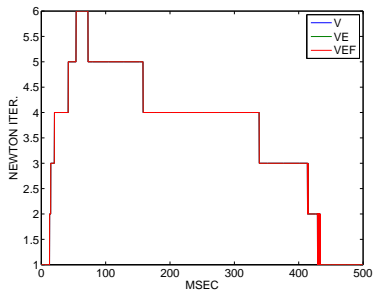
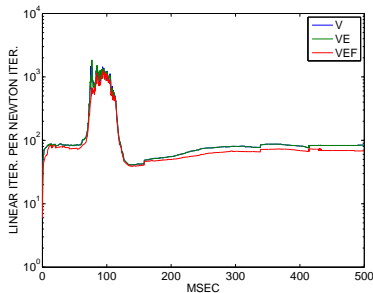
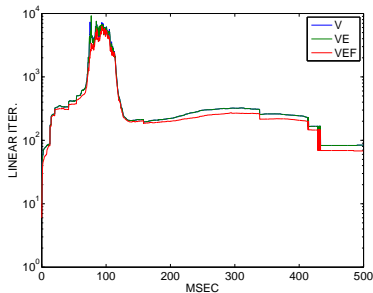
local mesh  $20^3$  (electrical),  $5^3$  (mechanical), global size increases with core count,

nit  $\approx 4 - 40$  (not reported)

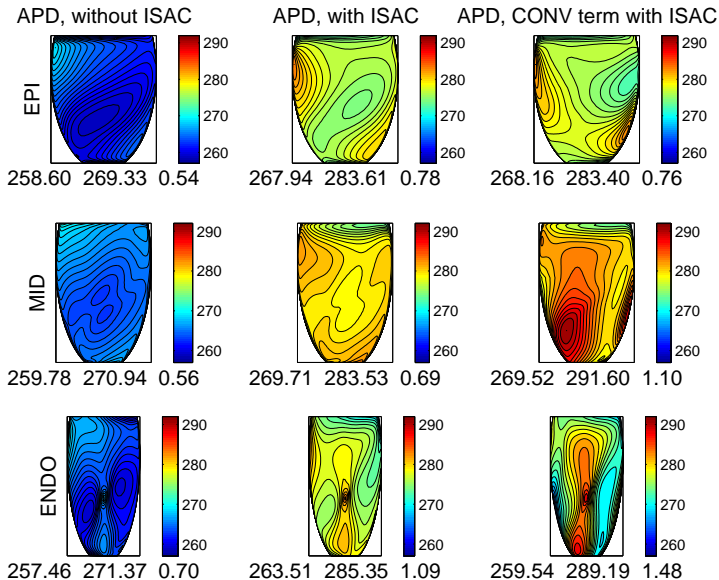
procs	Slab domains						Ellipsoidal domains					
	V		VE		VEF		V		VE		VEF	
	lit	time	lit	time	lit	time	lit	time	lit	time	lit	time
256	96	0.3	41	0.4	37	0.4	473	0.7	165	0.9	150	0.8
512	93	0.4	40	0.4	37	0.5	566	1.5	185	1.0	163	1.1
1024	90	0.5	38	0.6	34	0.6	598	2.1	177	1.6	153	1.7
2048	96	0.7	38	0.7	34	0.9	644	3.1	180	2.1	158	2.6
4096	88	0.9	38	1.1	29	1.3	589	5.0	181	3.4	154	4.2
8192	89	1.5	36	1.9	29	2.7	626	28.3	192	6.5	167	7.9
16384	88	2.4	33	3.4	26	5.4	796	33.2	189	11.2	148	14.1

V = Vertex, E = Edges averages, F = Faces averages

# Whole beat: mechanical BDDC with Land-Niederer $T_a$



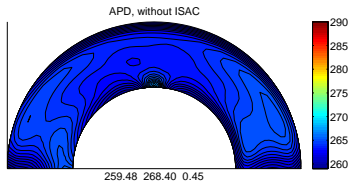
# 5. Applications: epicardial APD distributions



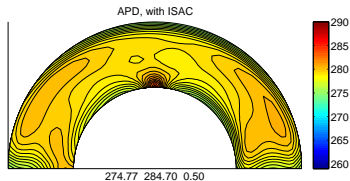


# Transmural APD distributions

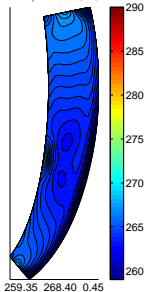
without ISAC



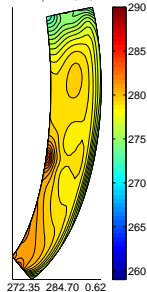
with ISAC



APD, without ISAC



APD, with ISAC



# Conclusions

- **Large-scale 3D cardiac simulations** now possible due to advances in parallel architectures, libraries (PETSc), solvers
- **Scalable and efficient DD solvers** (multilevel Schwarz, BDDC,...) available for both:
  - Bidomain model (IMEX linearization) and
  - cardiac electromechanical models with orthotropic strain energy functions.Mechanical solvers (Newton-Krylov) using AMG not quite scalable, better performance with DD preconditioners

**PETSc modularity** helps greatly in experimenting with complex choices of proper submodels/solvers to balance biophysical accuracy vs. computational costs: ionic model, calcium dynamics, Bidomain/Monodomain, active tension, mechanical constitutive law.

**PETSc has been essential** in my professional work, in advising PhD students, in teaching parallel computing classes

# Current/future work

- extend multilevel DD solvers to **anatomic cardiac geometries**:
  - unstructured meshes (go from DMDA to IS PETSc objects),
  - isogeometric analysis (via PetIGA library)
- explore other **nonlinear solvers** (nonlinear Schwarz, FAS, ...)
- explore better **coupling/decoupling strategy** of submodels in order to increase numerical stability/efficiency
- continue **advanced Bidomain simulations**: reentry genesis/termination, virtual electrode polarization, shock waveform (mono/biphasic) and energy optimization,...
- add coupling with **haemodynamical models**

Many thanks to Barry and the PETSc Team for this great library:  
Happy 20th Birthday and many more!

THANK YOU



Volume 13

## Mathematical Cardiac Electrophysiology

Piero Colli Franzone, Luca F. Pavarino, Simone Scacchi

This book covers the main mathematical and numerical models in computational electrocardiology, ranging from microscopic membrane models of cardiac ionic channels to macroscopic bidomain, monodomain, eikonal models and cardiac source representations. These advanced multiscale and nonlinear models describe the cardiac bioelectrical activity from the cell level to the body surface and are employed in both the direct and inverse problems of electrocardiology.

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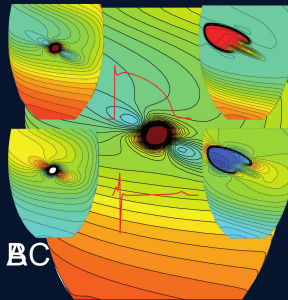
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