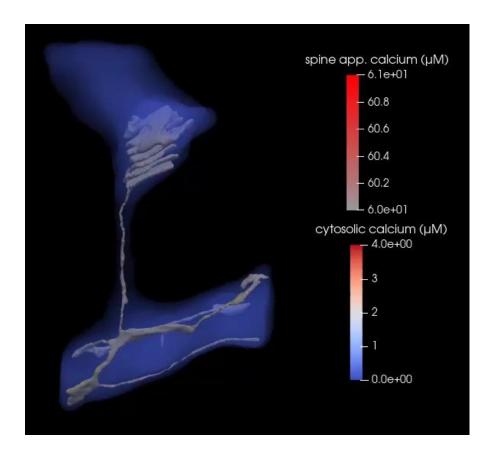
# Simulating cell signaling networks in realistic geometries, from dendritic spines to whole neurons

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### Difficulties of solving reaction-transport equations in cells

- 1. The equations are large, mixed dimensional systems of PDEs coupled across many cellular sub-compartments (plasma membrane, cytosol, organelle membranes, organelle interiors)
- 2. Many reactions depend nonlinearly on chemical constituents
- 3. Reaction and transport may involve many different physical mechanisms (diffusion, convection, electrodynamics, coupling to mechanics)
- 4. Cell geometries are complicated!

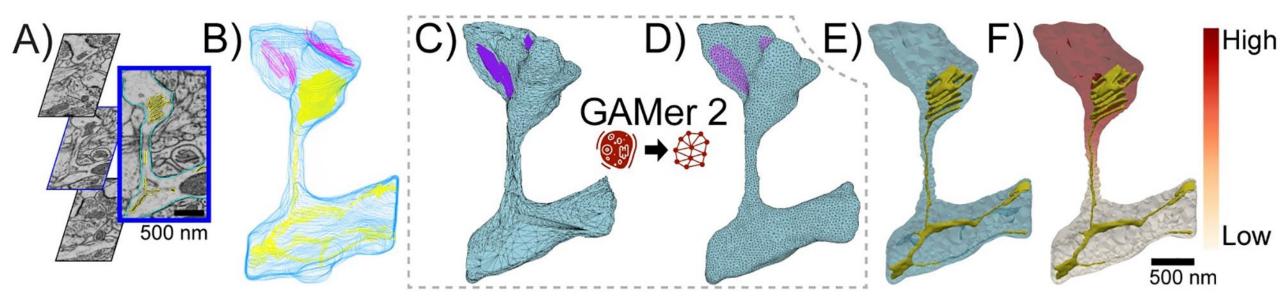
Example: ER-mitochondria geometry and dynamics (Guo et al 2018, *Cell*)

#### Part I

### The dynamic interactions between the ER and mitochondria

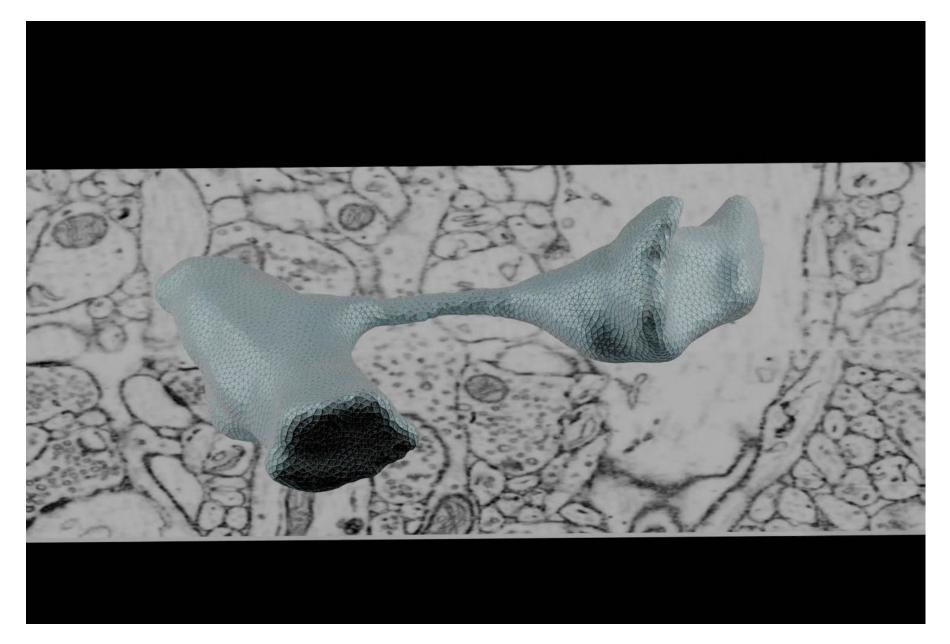
### Creation of meshes for complicated cell geometries – GAMer2

- Example below: meshes for a dendritic spine (bulbous protrusion from the dendrite of a neuron)
- GAMer2 allows for preservation of geometrical features while providing a smoother, better conditioned mesh for FEA



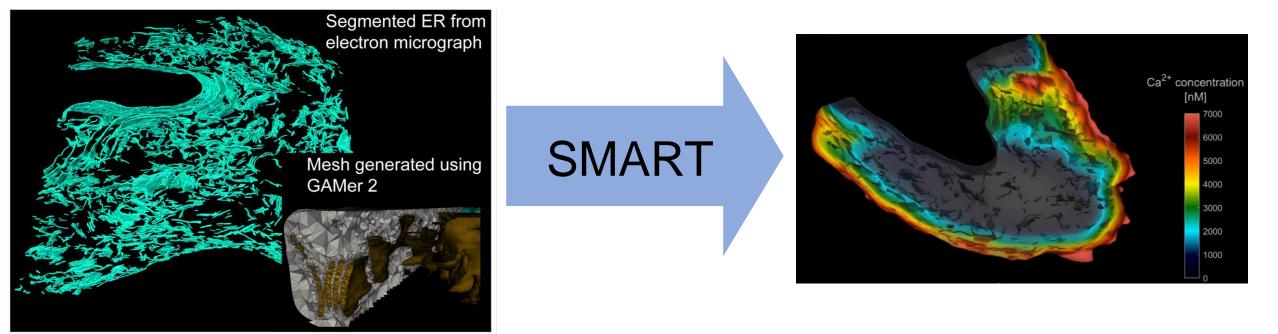
Lee et al 2020, PLOS Comp Bio

### Creation of meshes for complicated cell geometries – GAMer2



Lee et al 2020, *PLOS Comp Bio*4

## Spatial algorithms for reaction and transport (SMART) – modeling signaling networks in realistic cell geometries



Outline for today's talk:

- 1. Brief overview of formulation of equations and solution techniques in SMART.
- 2. Calcium dynamics in dendritic spines
- 3. Calcium dynamics in Purkinje neuron soma

### Formulation of mixed dimensional reaction-transport systems

Consider a volumetric compartment,  $\Omega^m$ , with boundary  $\Gamma^q$  and normal  $\mathbf{n}^m$ , and adjacent to other volumetric compartments  $\Omega^n$ 

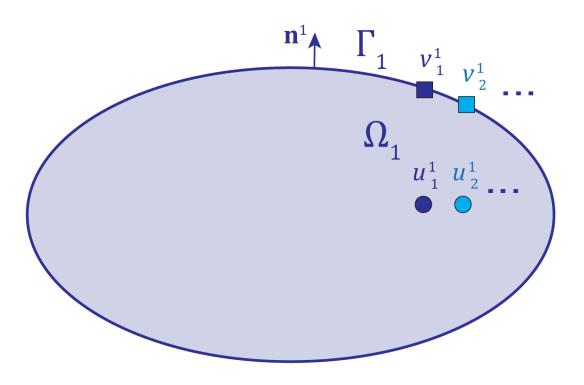
For each species in this compartment, with concentration  $u_i^m$ ,

Diffusion Volume reactions  $\partial_t u_i^m - \nabla \cdot (D_i \nabla u_i^m) - f_i^m (u^m) = 0$  in  $\Omega^m$ 

Diffusive flux	Surface reactions	
$D_i \nabla u_i^m \cdot \mathbf{n}^m -$	$-R_i^q(u^m, u^n, v^q) = 0$	on $\Gamma^q$

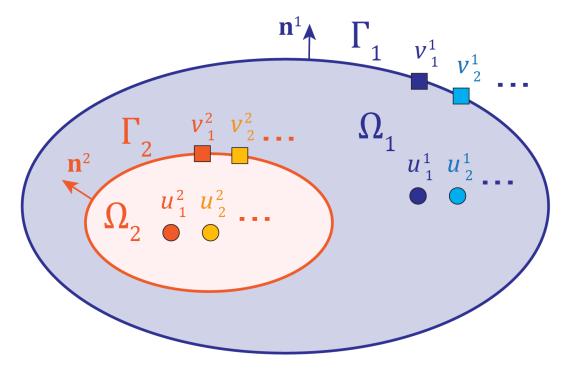
For each species on the boundary of this compartment, with concentration  $v_i^q$ ,

Surface diffusion Surface reactions  $\partial_t v_j^q - \nabla \cdot (D_j \nabla v_j^q) - g_j^q (u^m, u^n, v^q) = 0$  on  $\Gamma^q$ 



### Formulation of mixed dimensional reaction-transport systems

 $\partial_{t} u_{i}^{m} - \nabla \cdot (D_{i} \nabla u_{i}^{m}) - f_{i}^{m} (u^{m}) = 0 \quad \text{in } \Omega^{m}$   $D_{i} \nabla u_{i}^{m} \cdot \mathbf{n}^{m} - R_{i}^{q} (u^{m}, u^{n}, v^{q}) = 0 \quad \text{on } \Gamma^{q}$   $\rightarrow \text{Variational form } F_{i}^{m}$   $\partial_{t} v_{j}^{q} - \nabla \cdot (D_{j} \nabla v_{j}^{q}) - g_{j}^{q} (u^{m}, u^{n}, v^{q}) = 0 \quad \text{on } \Gamma^{q}$   $\rightarrow \text{Variational form } G_{j}^{q}$ 



Monolithic formulation – consider the sum of all variational forms for each subproblem:

 $F(u,v;\phi) + G(u,v;\psi) = 0,$ 

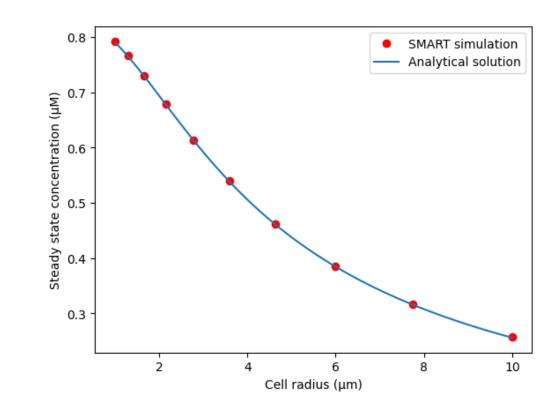
where both forms F and G are composed of sums over domains or surfaces and species:

$$F(u,v;\phi) = \sum_{m \in \mathcal{M}} \sum_{i \in \mathcal{I}^m} F_i^m(u,v;\phi_i^m), \qquad G(u,v;\psi) = \sum_{q \in Q} \sum_{i \in \mathcal{I}^q} G_i^q(u,v;\psi_i^q).$$

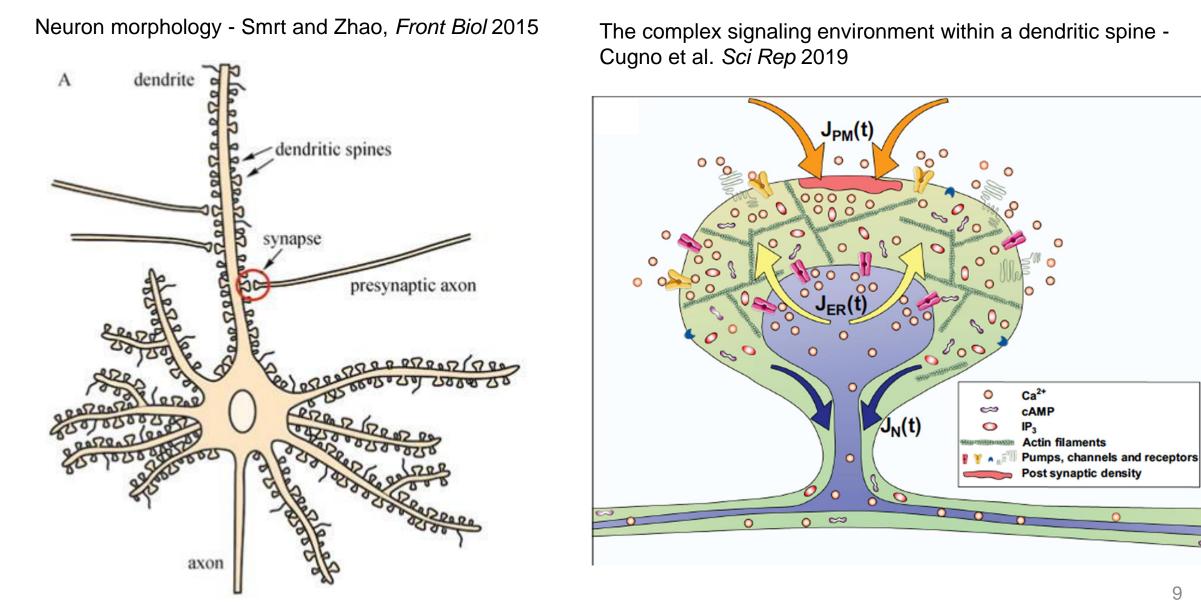
### Numerical methods and validation in SMART

- Use backward Euler for time discretization
- Assemble nonlinear finite element system using FEniCS
- Solve this system using Newton-Raphson iteration in PETSc

Validated SMART by testing problems with known analytical solutions



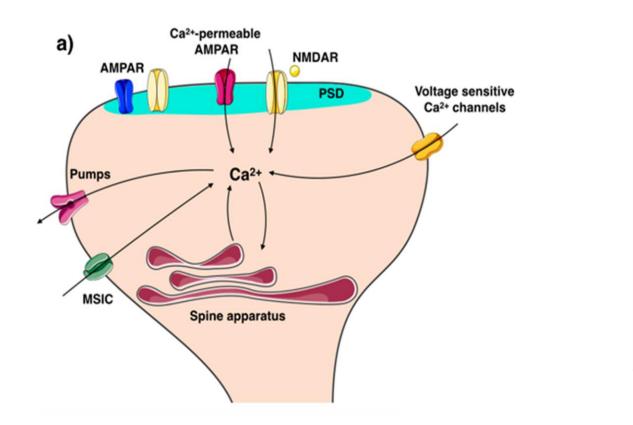
### Overview of Ca<sup>2+</sup> signaling networks in neurons



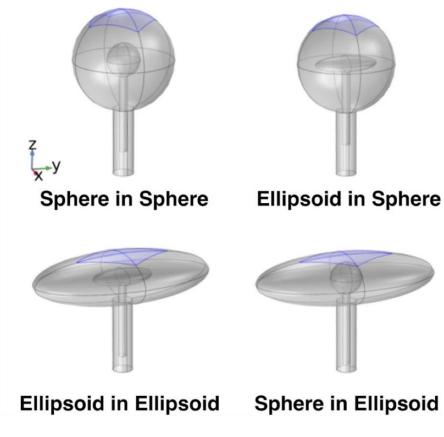
#### Testing Ca<sup>2+</sup> dynamics in idealized geometries for dendritic spines

Model from Bell et al 2019:

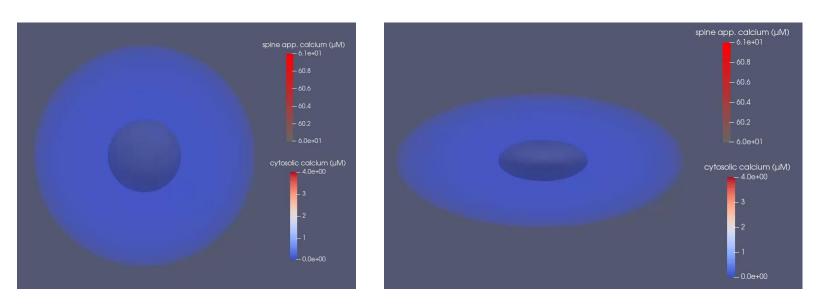
- Ca<sup>2+</sup> influx through VSCCs and NMDARs
- Ca<sup>2+</sup> exits the cytosol through pumps into the spine apparatus (SA) or out of the PM



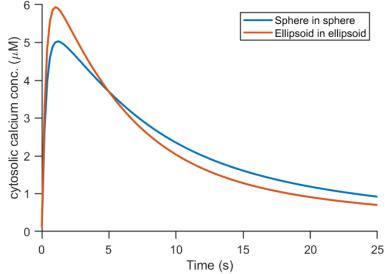
• 4 different geometries of spines with spine apparatus; volumes equal in all cases



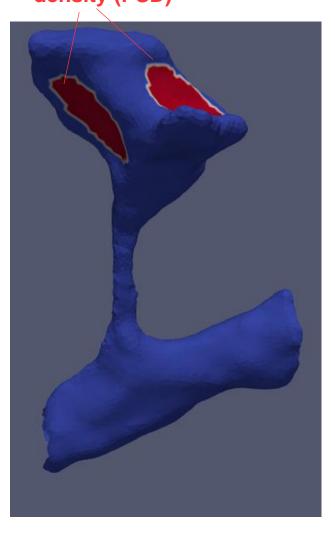
## Increased spine apparatus surface area allows for faster pumping of Ca<sup>2+</sup> out of the cytosol

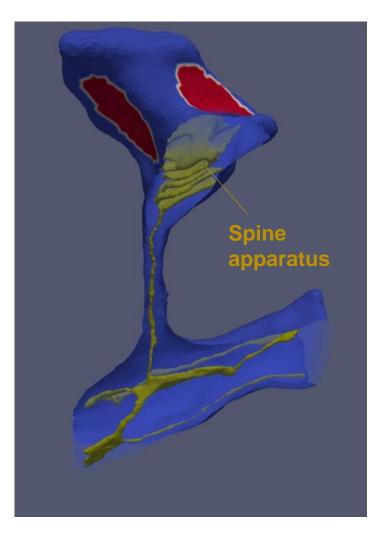


Ca<sup>2+</sup> decreases fastest when surface area to volume ratios are maximized

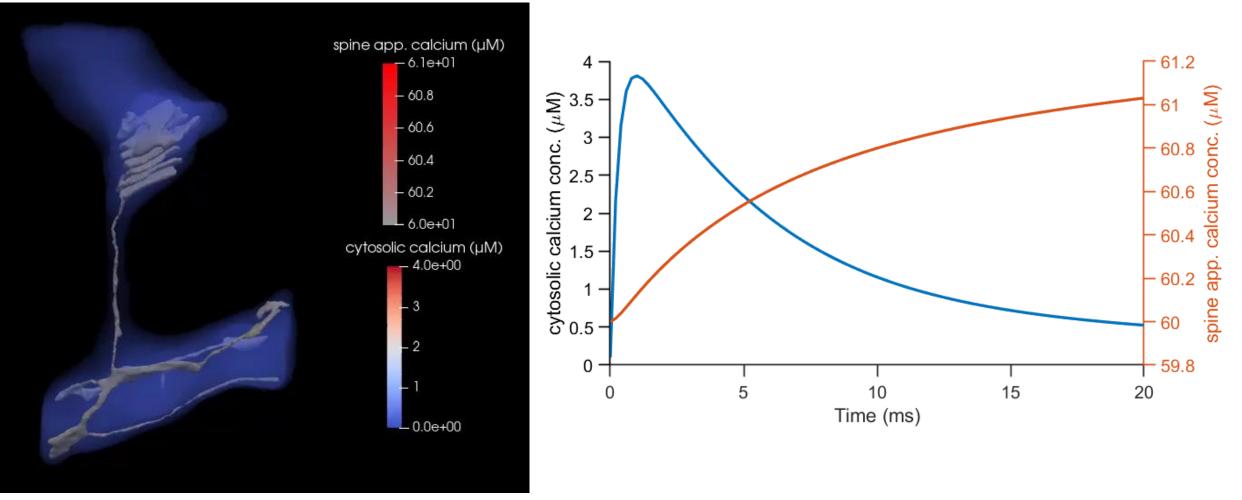


#### Ca<sup>2+</sup> dynamics in a realistic spine geometry Postsynaptic density (PSD)

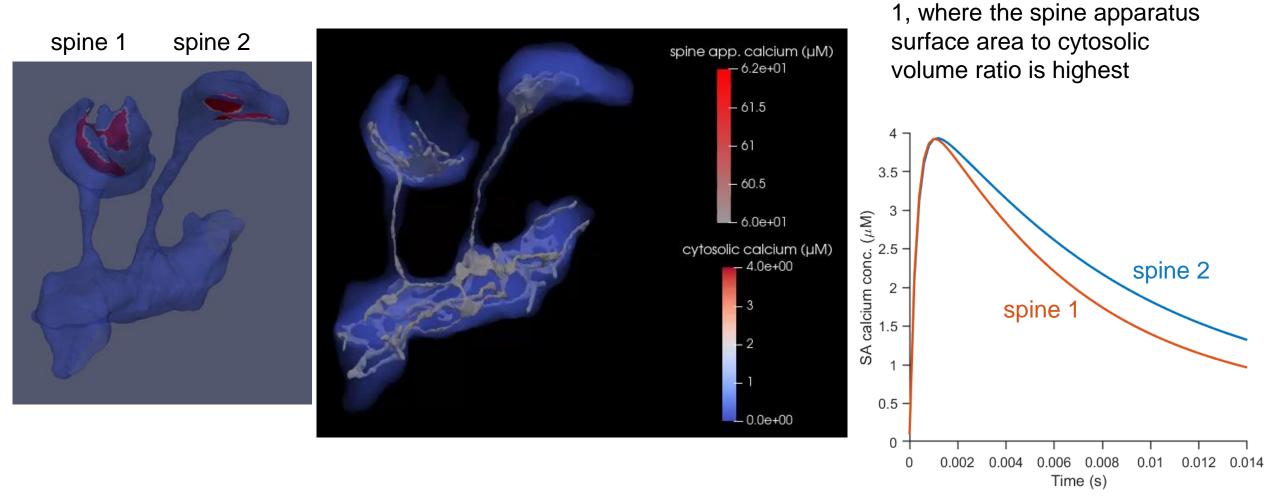




### Calcium dynamics in a realistic spine geometry



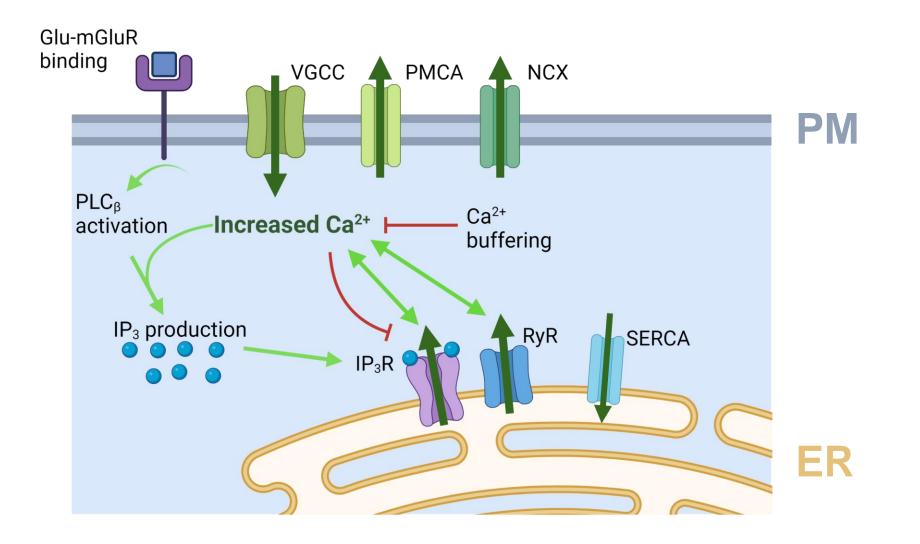
### Calcium dynamics in two realistic dendritic spines



Francis and Laughlin, in preparation

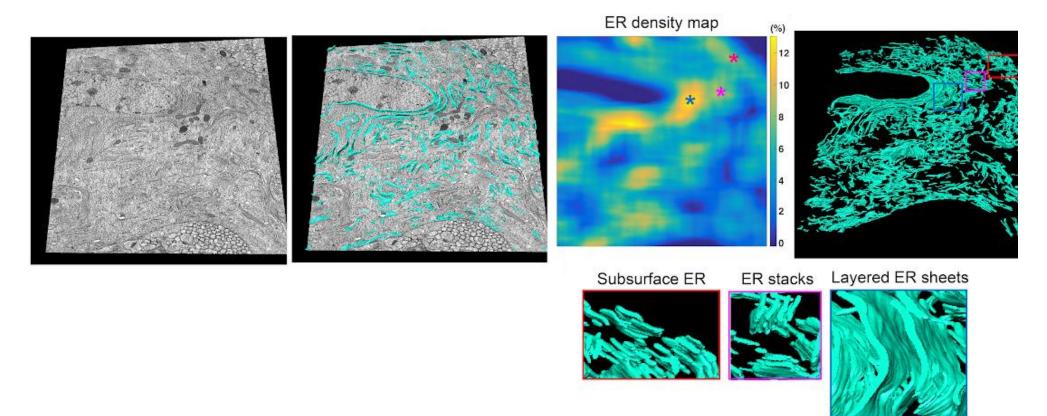
Ca<sup>2+</sup> decreases faster in spine

### Including Ca<sup>2+</sup> release through IP<sub>3</sub>Rs and RyRs



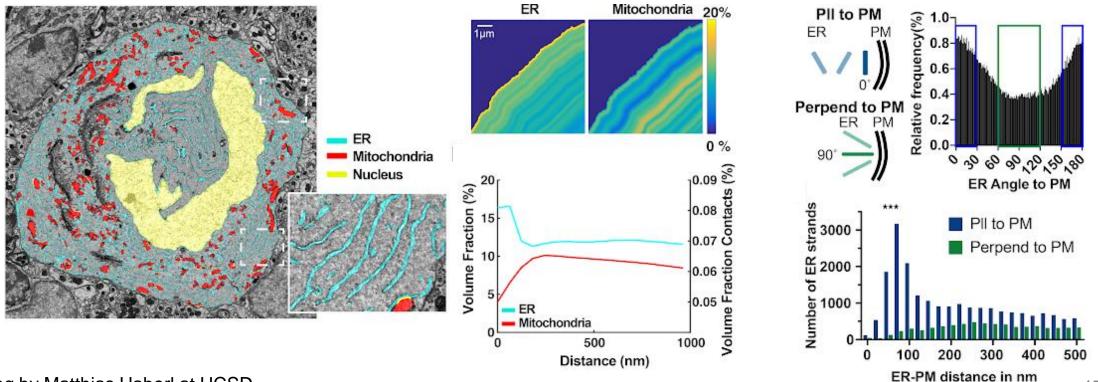
## Electron microscopy provides detailed characterization of the ER in the soma of Purkinje neurons

• Segmentation of electron micrographs identifies the ER within the main body (soma) of Purkinje neurons (collaboration with the Ellisman Lab)



## Electron microscopy provides detailed characterization of the ER in the soma of Purkinje neurons

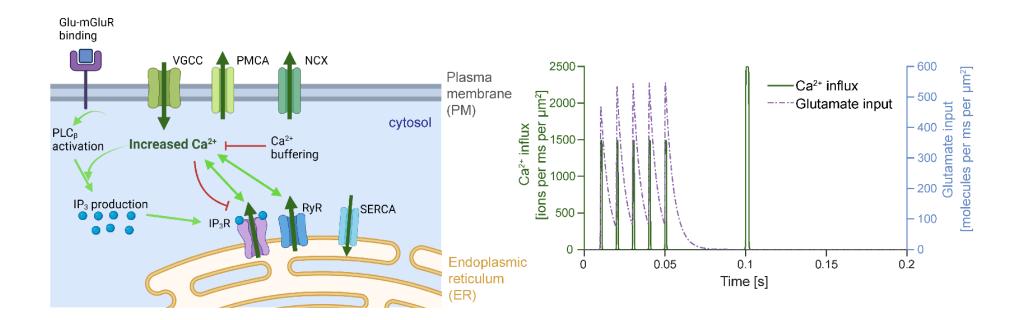
- ER is localized close to the PM (within 100nm)
- Membrane-adjacent ER is preferentially oriented parallel to the PM



Imaging by Matthias Haberl at UCSD

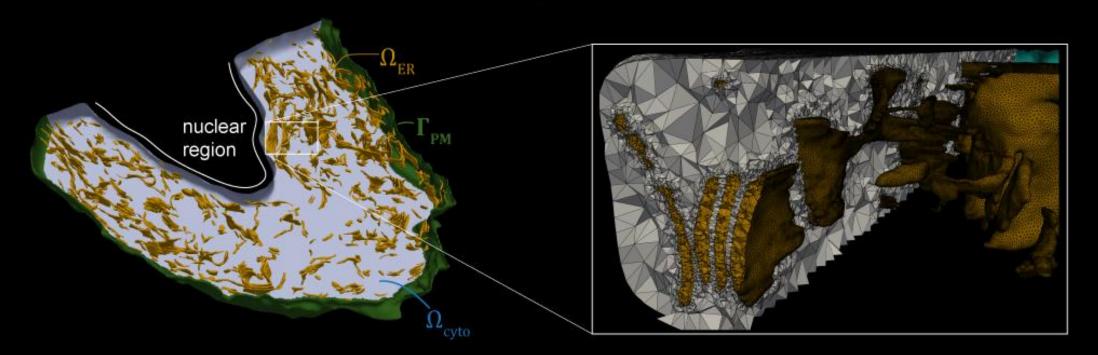
### Model of calcium dynamics in the Purkinje neuron

- Detailed signaling model adapted from previous well-mixed model (Doi et al 2005, *J Neurosci*), converted to a spatial model 26 species overall (21 surface, 5 volume)
- · Inputs chosen to match physiological stimulus used in Doi et al

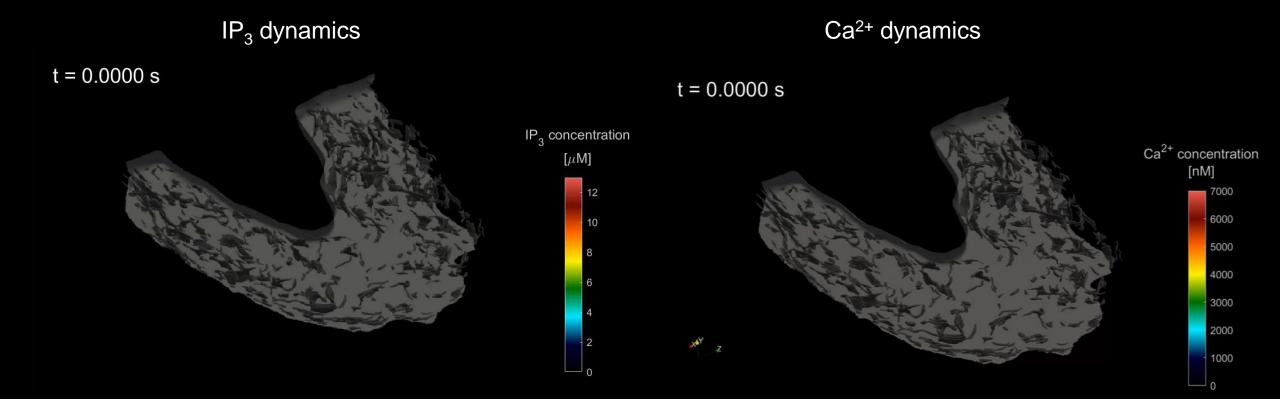


### Examining Ca<sup>2+</sup> dynamics in realistic soma geometries

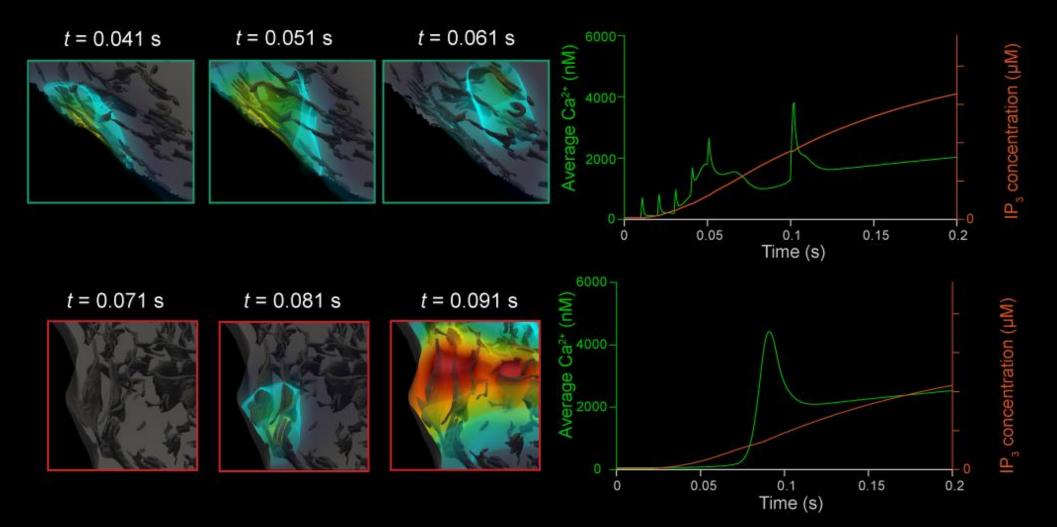
- Region of Purkinje soma from electron microscopy was segmented into ER and cytosol
- Resulting mesh was conditioned using GAMer2
- Mesh statistics:
  - ~62 million tetrahedra total (25m in the cytosol, 37m in the ER)
  - ~11.6 million surface triangles (11.5m in ER membrane, .1m in the PM)
  - ~10 million points total in the whole geometry



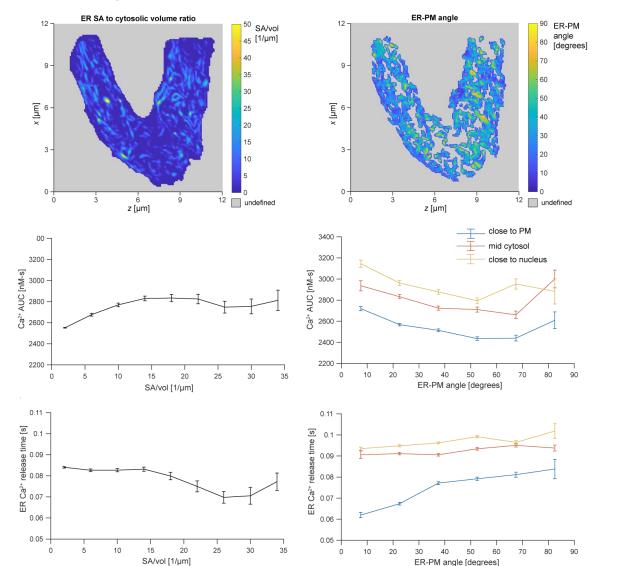
### IP<sub>3</sub> and Ca<sup>2+</sup> dynamics in a realistic Purkinje soma



### IP<sub>3</sub> and Ca<sup>2+</sup> dynamics in a realistic Purkinje soma

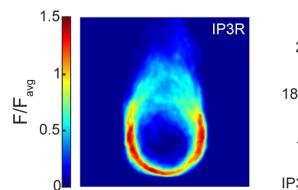


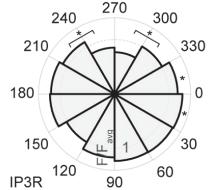
# ER spacing and orientation controls the timing and magnitude of calcium release

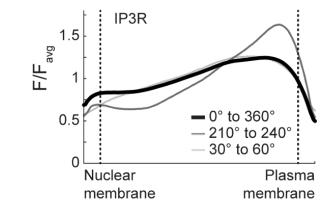


### Experimental measurements of receptor distributions





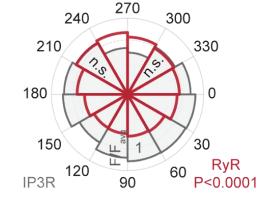


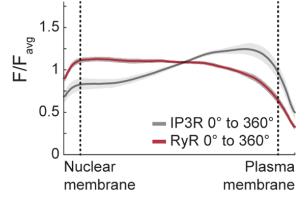


Imaging by the

**Bloodgood Lab** 

at UCSD

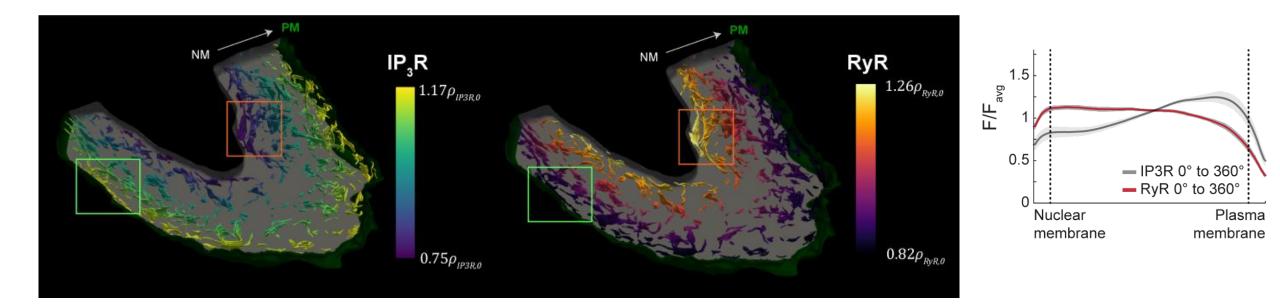




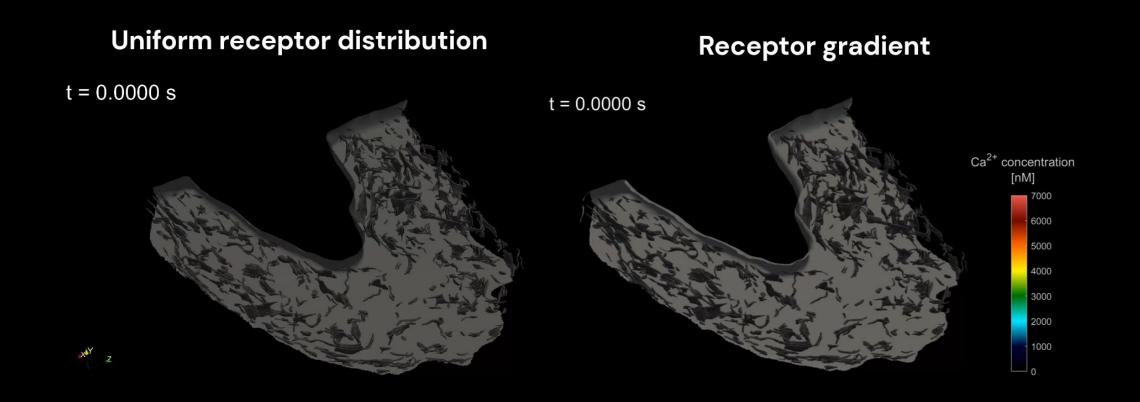
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### Whole-cell geometry with receptor gradients

• Gradients in realistic geometry chosen to match those observed experimentally (linear gradient as a function of distance from the PM)

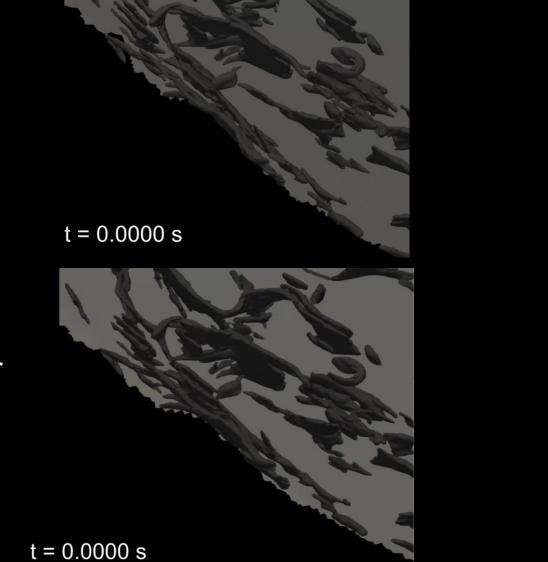


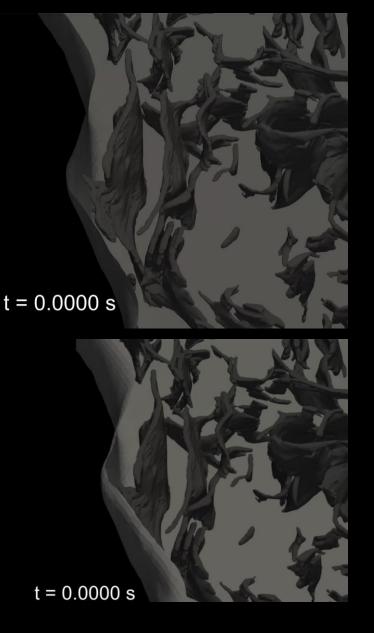
Ca<sup>2+</sup> dynamics for uniform receptors vs. realistic receptor gradient



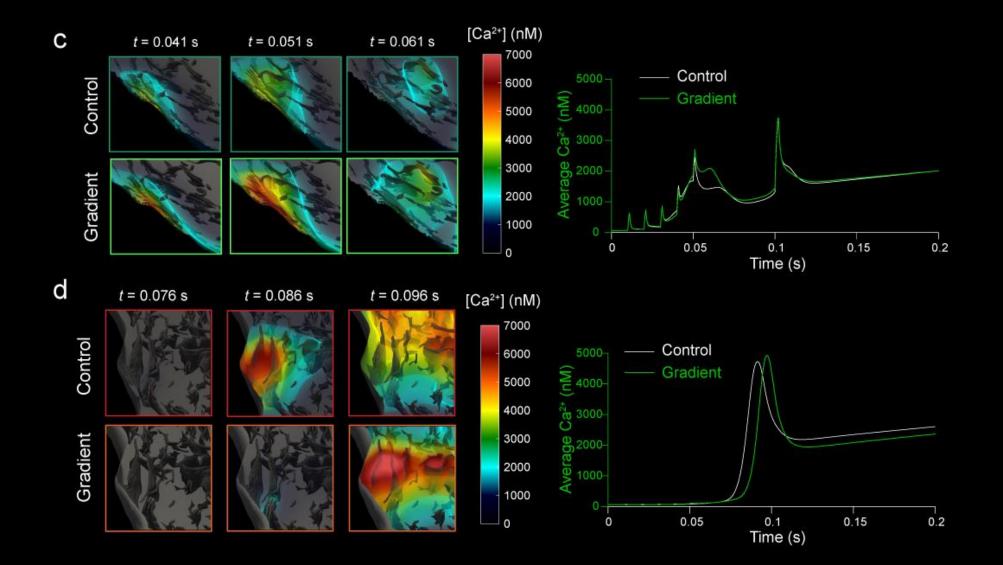
### Ca<sup>2+</sup> dynamics for uniform receptors vs. realistic receptor gradient

Control





### Ca<sup>2+</sup> dynamics for uniform receptors vs. realistic receptor gradient

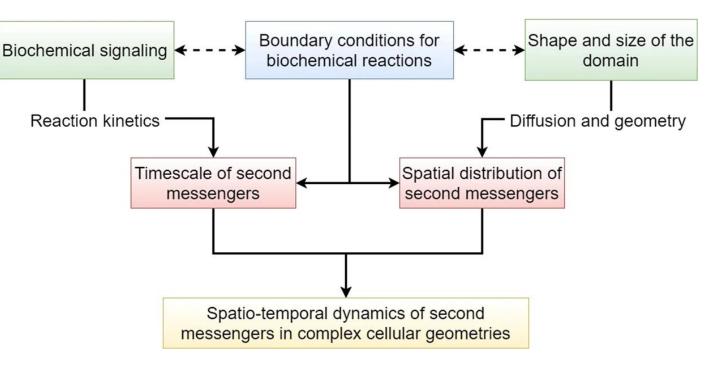


### **Conclusions / Summary**

- Together with GAMer2, SMART offers a platform to specify biological signaling networks in realistic cell geometries
- Realistic representation of cell signaling requires consideration of nonlinear reaction kinetics, surface-volume coupling, and detailed cell geometries

Lessons from spine and Purkinje simulations:

- Surface to volume ratios are important determinants of calcium influx, calcium release, and repackaging of calcium into the ER.
- 2. The orientation and spacing of ER with respect to the PM modulates the rate of calcium release.
- 3. Changes in receptor distribution influence calcium release dynamics.



#### Cugno et al 2019, Sci Reports

### UC San Diego

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Simula Research

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