

# Computational Models of Cardiac Arrhythmia

PATRICK BOYLE

Biomedical Modeling Using Imaging Data  
 AAPM Annual Meeting, Tuesday, July 16, 2019

**W** BIOENGINEERING UNIVERSITY of WASHINGTON pmjboyle@uw.edu

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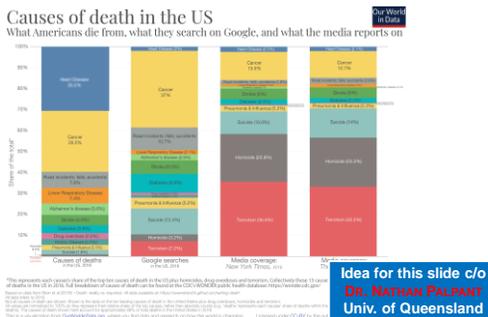
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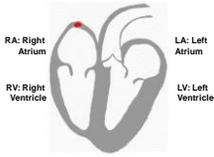
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## What is cardiac arrhythmia?



Normal heart rhythm, Wikimedia Commons

- **Cardiac excitation** precedes contraction, delivery of blood to body
- **Electrocardiogram** has principal electrical events
- No coordinated excitation = no O<sub>2</sub> delivered to body = **death within minutes**

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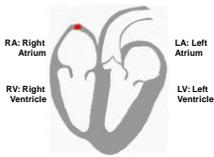
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## What is cardiac arrhythmia?



Normal heart rhythm, Wikimedia Commons

- **Cardiac Arrhythmia** Any disruption of heart's intrinsic rhythm regulation
- Some arrhythmias are less lethal than others
- e.g., **Atrial fibrillation** poses no acute threat but increases risk of stroke

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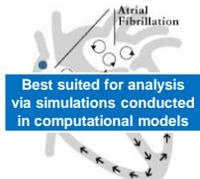
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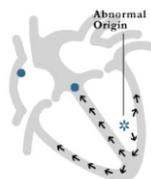
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## Major arrhythmia categories

### Reentrant self-perpetuation



### Focal/triggered ectopy



Brigham and Women's Hospital (2015)

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## Methodological overview

1. What are we trying to model?
2. How do these models work?
3. How can these models be used?

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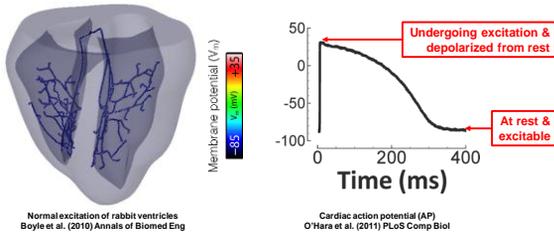
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## Excitation at cell and organ scales




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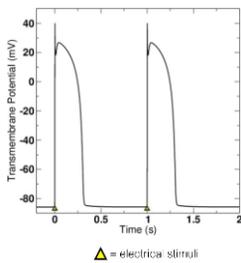
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## Refractoriness



- **Generally**, two stimuli = two action potentials
- What if we shorten the **coupling interval (CI)**?
- At its extreme, this can lead to complete action potential suppression

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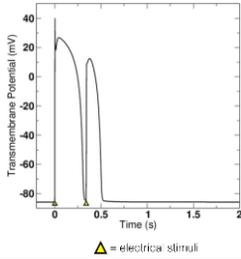
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## Refractoriness



- **Generally**, two stimuli = two action potentials
- What if we shorten the **coupling interval (CI)**?
- At its extreme, this can lead to complete action potential suppression
- **Major consequences at the tissue scale!**

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## Normal AP Propagation in Tissue




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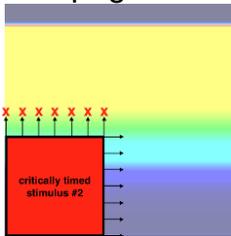
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## Normal AP Propagation in Tissue



Self-perpetuating excitation = **BAD**

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## Why do we care so much about reentry?

- Two “ingredients” for reentrant arrhythmia:
  - **Substrate:** state of tissue after stimulus #1 (i.e., *spatially heterogeneous refractoriness*)
  - **Trigger:** critically-timed stimulus #2
- Many pathophysiological factors can give rise to these types of pro-arrhythmic substrates and triggers!
- Reentrant arrhythmias are complex; simulations can help explain how they work and reveal new treatments

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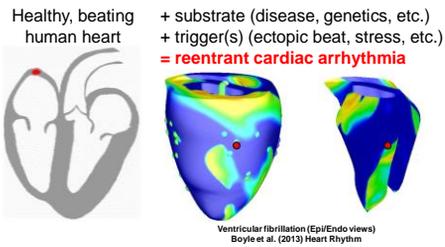
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## Conceptual basis of reentrant arrhythmia




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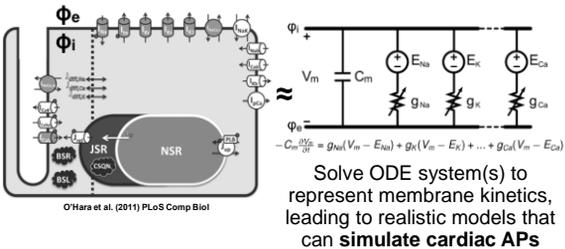
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## Cell-scale building blocks




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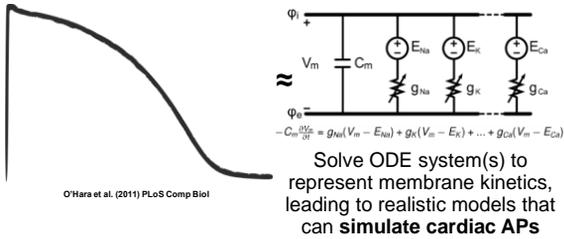
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### Cell-scale building blocks



O'Hara et al. (2011) PLoS Comp Biol

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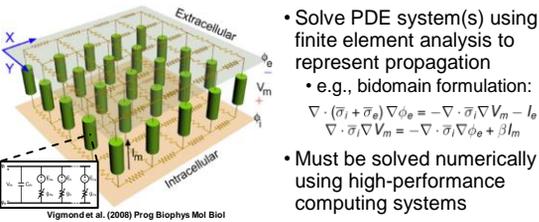
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### Tissue-scale building blocks



Vigmond et al. (2008) Prog Biophys Mol Biol

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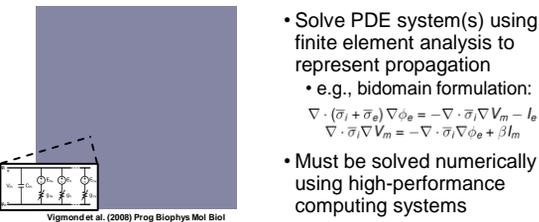
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### Tissue-scale building blocks



Vigmond et al. (2008) Prog Biophys Mol Biol

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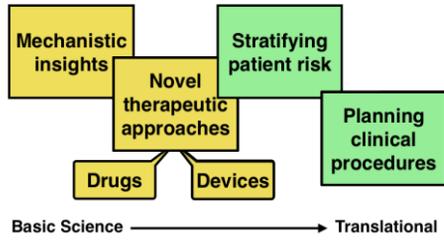
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### Potential applications



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## Case study #1

*Fast-turnaround patient-specific modeling of atrial fibrillation (AFib)*

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### The AFib epidemic

- AFib is an enormous health problem worldwide
- Most prevalent sustained arrhythmia (1-2%)
- Prevalence on the rise (~2.5x increase by 2050)
- Many patients transition to **persistent AFib** from less severe forms of disease



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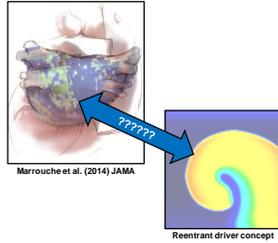
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## Persistent AFib is very hard to treat

- Long-term ablation success rates are low
- Relationship between **fibrotic remodeling** and **reentrant drivers** of AFib is not well characterized
- **Computational models can be a huge help!**




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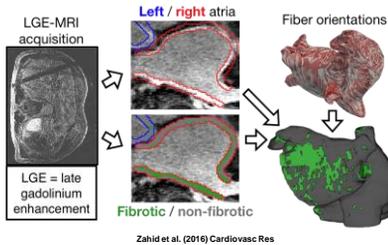
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## Model reconstruction from LGE-MRI



Zahid et al. (2016) Cardiovasc Res

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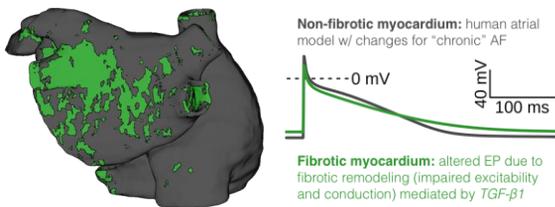
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## Model parameterization



**Non-fibrotic myocardium:** human atrial model w/ changes for "chronic" AF

**Fibrotic myocardium:** altered EP due to fibrotic remodeling (impaired excitability and conduction) mediated by *TGF-β1*

Zahid et al. (2016) Cardiovasc Res

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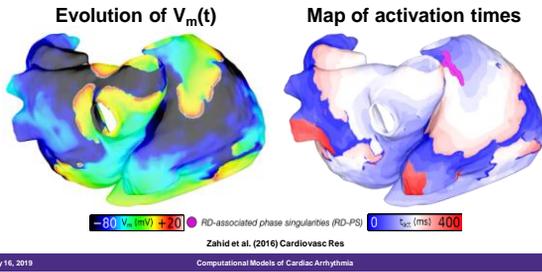
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## Dynamics of simulated AFib




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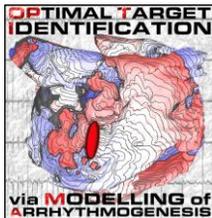
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## Potential value of modeling is high

- Better understand underlying mechanisms of AFib
- Identify patients at risk of developing AFib in the future (e.g., cryptogenic stroke)
- **Patient-specific persistent AFib treatment planning (OPTIMA trial at JHU)**



Boyle et al. (2019) Nature Biomedical Engineering (IN PRESS)

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## OPTIMA Workflow

OPTimal  
Target  
Identification via  
Modeling of  
Arrhythmogenesis

Boyle et al. (2019) Nature Biomedical Engineering (IN PRESS)

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## OPTIMA outcomes are good, but...

ID	Rhythm at end of FU	FU duration (days)	Atrial ablation during FU
1	SR	543	No
2	SR	347	No
3	PAF/AfI	417	PAF/AfI
5	SR	399	No
6	SR	302	No
7	SR	315	No
8	SR	264	No
9	SR	239	No
10	SR	197	No

- No recurrence of persistent AF: 10/10
- 4/10 patients had paroxysmal AF/flutter during follow-up, including one repeat ablation

Boyle et al. (2019) *Nature Biomedical Engineering* (IN PRESS)

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## ...uptake will be hindered by complexity!

### • Staggering computational complexity

- Model reconstruction: ~0.5-2 days
- "Substrate stress test": ~2-4 days
- 15-20 total days of CPU time

### • Difficult to carry out simple parameter sensitivity analyses

- Deng et al. (2017) *Chaos*: 60k CPU hours to test  $\pm 10\%$  APD/CV in 12 atrial models

### • The JAWS philosophy does not apply



"We're gonna need more than just a bigger super computer."  
-Pat (2019)

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## Long-Term Objective: Streamlined framework for atrial model reconstruction and realistic simulation of AFib

- Automate! Automate! Automate!
- Prioritize Validation, Verification, and Uncertainty Quantification (VVUQ)
  - Produce clinically actionable results within a suitable timeframe

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**Issue:** Atrial image segmentation can be time consuming and labor intensive

- **Solution?** 3<sup>rd</sup> party image segmentation service
  - Every AFib ablation patient at UW LGE-MRI pre-procedure
  - Every scan gets processed off-site (Merisight, Utah)
- **Benefits:** no time spent on image segmentation or processing, consistency, ease of clinical integration
- **Drawbacks:** loss of precise control, current workflow produces single LA surface w/ normalized LGE values

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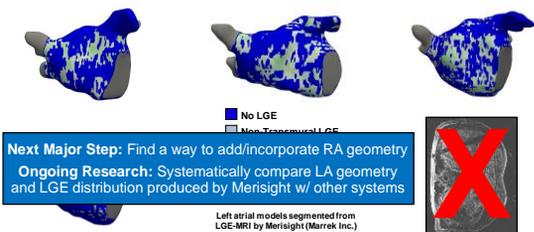
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3<sup>rd</sup> party image segmentation




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**Issue:** Mesh generation and refinement are time-consuming and unreliable

- **Meshes produced by Merisight** are too coarse for running cardiac EP simulations, have weird elements
- **Commercial software options** (Simpleware ScanIP, Materialise Mimics) are \$\$\$, produce meshes suitable for different types of finite element analysis and nearly always involve a GUI (i.e., death to automation)
- **Solution?** Explore Open Source Software (OSS) for potential solutions that can be run from command line

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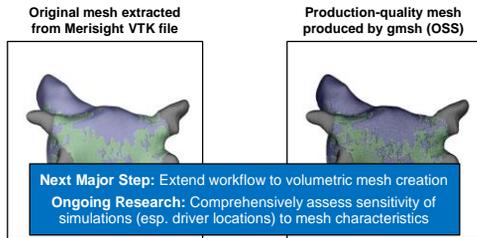
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## Fully automated (re)-meshing



Geuzaine & Remacle (2009) Int J Num Meth Eng

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## Issue: Simulation run-times are long

- **Computational complexity** impairs rapid turnaround times (clinical applications) and makes it challenging to carry out VVUQ (e.g., sensitivity analysis, etc.)
- **Solution?** Surface-only instead of volumetric models
- **Benefits:** huge performance boost, simpler meshing
- **Drawbacks:** information on transmuralilty abandoned

Using surface-only meshes is not a new idea!  
 e.g., Labarthe et al. (2014) Europace; Roney et al. (2016) Europace; Hwang et al. (2018) PLoS One; [many others](#)

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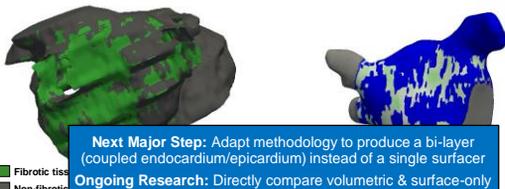
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## Simplified geometry (no transmuralilty)

Volumetric bi-atrial mesh

Surface-only mesh of LA



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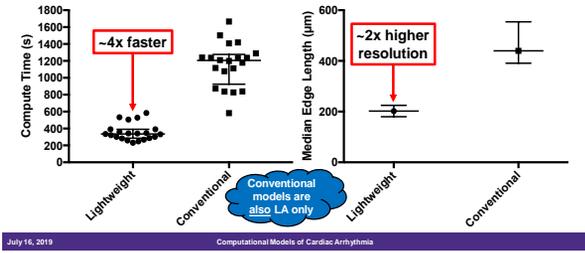
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### Performance of new models is great!



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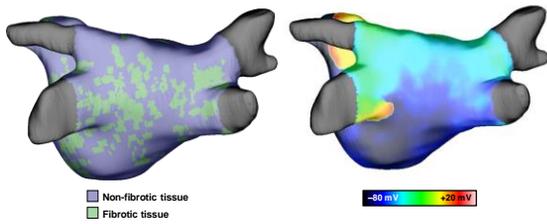
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### Simulated AFib in new models



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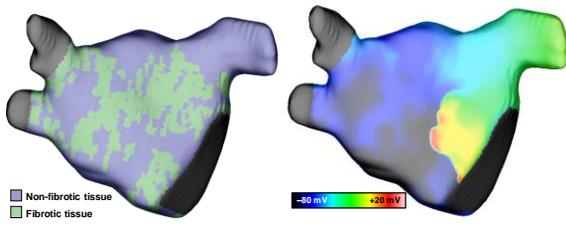
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### Simulated AFib in new models



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## Simulated AFib in new models

- Reentrant drivers are **qualitatively similar** to those seen in conventional models
- Dynamic localization to **fibrotic tissue boundaries**
- High spatiotemporal **stability** (i.e., little meandering)
  
- **Key distinction:** Lack of chaotic activity and/or transient reentry in areas peripheral to driver
  - Consequence of surface only vs. volumetric models?

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## Next steps

Model Framework Features	Lightweight	Conventional*
Completely automated	Yes	No
Rapid compute time	Yes	No
High mesh resolution	Yes	No
Reentrant arrhythmia inducible	Yes	Yes
Realistic fiber orientations	No	Yes
Atrial thickness, transmural fibrosis	No	Yes

1. Map fibers from atlas geometry via new **UAC method**
2. Approximate transmural fibers via **bi-layer surface models**

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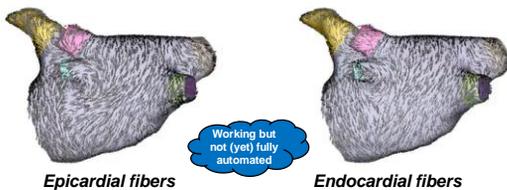
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## Preliminary data: UAC fiber mapping



UAC = Universal Atrial Coordinates  
Roney et al. (2019) Med Imag Anal

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## Conclusions

**A streamlined simulation framework is within reach!**

If successful, this research promises to realize personalized computational models that could "follow the patient" in their electronic health records, **with near-real-time updates**

The new computational framework is not yet ready for showtime (validation is still needed), **but it will be soon!**

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## Case study #2

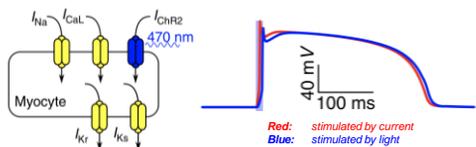
*Exploring anti-arrhythmia approaches based on cardiac optogenetics*

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**"Optogenetics":** genetic encoding of light-sensitive proteins in excitable cells

- Channelrhodopsin-2 (ChR2) light-gated NS cation channel
- Irradiation with blue light elicits a large depolarizing current



Simulation data from Boyle et al. (2013) Nature Communications

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## Noteworthy advantages of optogenetics

- Different opsins alter cell behavior in different ways
- Contact-free perturbation
- Patterned illumination = exquisite spatial and temporal precision
- Selective expression in specific cells/tissues = selective stimulation



Deisseroth lab experiment, Stanford University  
<http://www.youtube.com/watch?v=7uRFV98PBU>

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## Optogenetic stimulation of the heart



Bruegmann et al. (unpublished, 2010)

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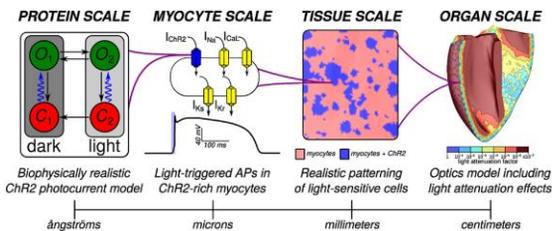
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## Comprehensive modeling framework



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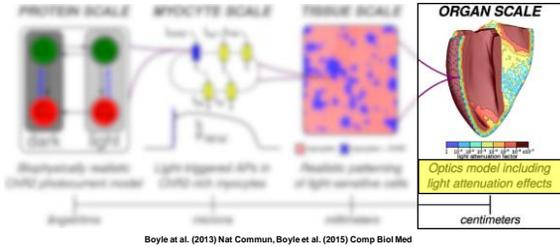
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## Comprehensive modeling framework



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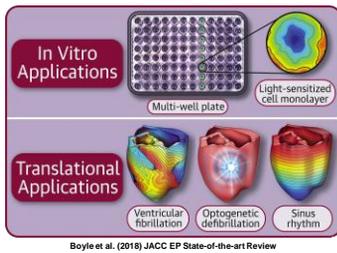
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## A novel approach to arrhythmia therapy?



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## A novel approach to arrhythmia therapy?

### As a clinical treatment?

- Optogenetic defibrillation, pacing, APD modulation
- Two key prerequisites:
  - Light sensitization via gene/cell therapy
  - Illumination

### As an experimental tool?

- New assays to quantify EP properties (contact-free, high-throughput)
- Patterned illumination can impose contrived scenarios (e.g., rotors)
- Light-based modulation of non-EP properties

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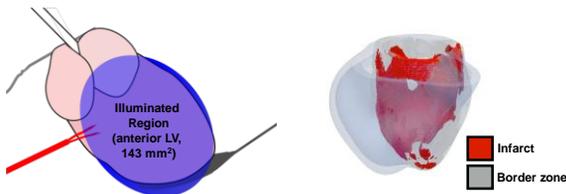
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## Optogenetic defibrillation

In perfused mouse hearts      In a simulated human heart



Bruegmann, Boyle, et al. (2016) J Clin Invest

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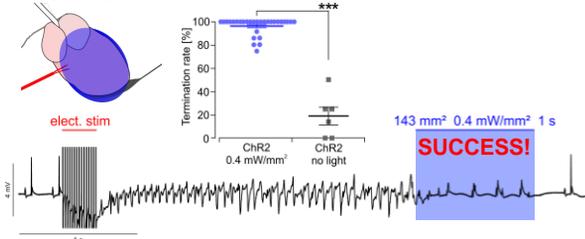
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## Optogenetic defibrillation



Bruegmann, Boyle, et al. (2016) J Clin Invest

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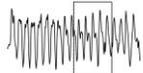
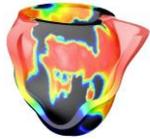
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### Optogenetic defibrillation



-90 mV +40 mV Bruegmann, Boyle, et al. (2016) J Clin Invest

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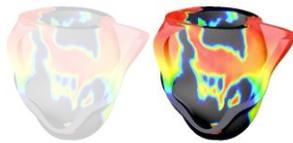
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### Optogenetic defibrillation



-90 mV +40 mV Bruegmann, Boyle, et al. (2016) J Clin Invest

**Blue light** failed to defibrillate the heart!  
Likely cause: attenuation from photon scattering and energy absorption  
Possible solution: **RED light**

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### Optogenetic defibrillation



**Conclusion:** Optogenetic defibrillation is feasible in principle and may even work in the light-sensitized human ventricles, but extra steps are needed to overcome light attenuation.

-90 mV +40 mV Bruegmann, Boyle, et al. (2016) J Clin Invest

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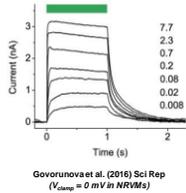
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### Optogenetic defibrillation/cardioversion via anion channelrhodopsins (ACRs)



- **GtACR1:** Green Light Sensitive Anion Channel
- Conducts anions and elicits strong currents in cardiomyocytes
- GtACR1 reversal potential near that of chloride ions (approx.  $-40$  mV)

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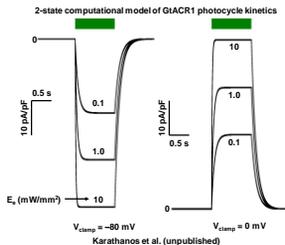
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- **GtACR1:** Green Light Sensitive Anion Channel
- Conducts anions and elicits strong currents in cardiomyocytes
- GtACR1 reversal potential near that of chloride ions (approx.  $-40$  mV)

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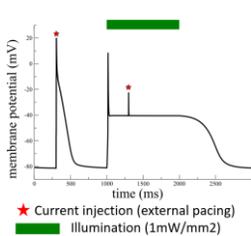
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### Optogenetic defibrillation/cardioversion via anion channelrhodopsins (ACRs)



Proof-of-principle termination of AFib driven by a stable reentrant source located in the left atrium

**Preliminary Conclusion:** Arrhythmia termination via GtACR1 is feasible

Karathanos et al. (unpublished)

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## Conclusions

~10 years on, cardiac optogenetics has emerged as a **vibrant and rapidly-growing field**

Development of a **comprehensive framework for computational modeling** has helped drive progress

**New opsins and illumination strategies** have created opportunities for optogenetics-based innovation, both experimentally and as a potential new treatment approach

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### Cardiac Systems Simulation Lab UW Department of Bioengineering

**Clinical EP team at UW Medical Center**



AHA Funding  
SDG16-30440006  
(PI: Boyle)

**Nazem Akoum**  
Arun Sridhar  
Zeinab Birjandian  
Sakher Sarairah



NIH Funding  
DP1-HL123271  
(PI: Trayanova)

**Optogenetics**



Collaborators:  
Johns Hopkins  
N Trayanova

**Emilia Entcheva**  
Cookie Yu  
Thomas Karathanos



Collaborators:  
CARP consortium  
G Plank, E Vigmond

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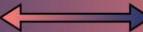
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## OPTIMA Team (JHU)

Simulations



Clinical EP

Natalia Trayanova  
Patrick Boyle  
Sohail Zahid  
Dong Dong Deng  
Adityo Prakosa  
Michael Murphy  
William Franceschi

Hugh Calkins  
Saman Nazarian  
Joseph Marine  
Hiroshi Ashikaga  
David Spragg  
Ronald Berger  
Tarek Zghaib

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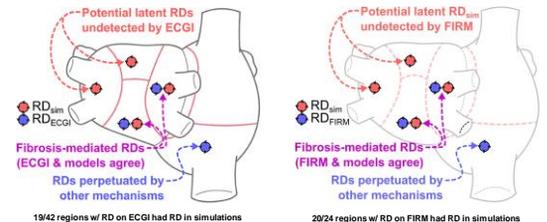
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### OPTIMA Patient Characteristics

ID	Age	Sex	BMI	LA % fibrosis	RA % fibrosis	LA V (ml/m <sup>2</sup> )	RA V (ml/m <sup>2</sup> )	CHA <sub>2</sub> DS <sub>2</sub> -VASc	# Prior Ablations	Reconnected Pvs (if any)		
1	69	M	31	35.8	38.65	73.3	62.5	2	1	No		
2	60	M	26	26.7	5.8	35.5	52.6	0	0			
3	73	M	27	46.9	34.4	87.8	74.4	5	3	No		
4	72	F	34	0.7	5.5	59.3	58.7	2	0			
5	68	M	25	32.9	34.9	71.0	46.7	2	1	RSPV, RIPV		
6	67	M	29	19.7	15.6	43.3	38.2	2	0			
7	49	M	35	25.4	7.2	42.4	43.7	1	1	No		
8	71	M	27	24.8	33.9	32.3	26.0	1	1	LSPV, LIPV, RIPV, LSPV		
9	72	M	23.8	11.9	3.1	28.5	38.2	2	2			
10	59	M	40.2	15.8	6.8	41.9	34.4	1	0			
				<b>6627</b>	<b>9(90%)</b>	<b>3015</b>	<b>24.1±12.4</b>	<b>18.6±14.2</b>	<b>51.5±20</b>	<b>47.5±14.6</b>	<b>2[1;2]</b>	<b>1[0;1.5]</b>

### Comparison to clinical data

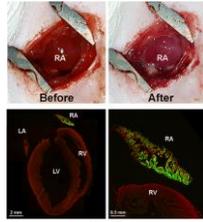
Boyle et al. (2018a) Front Physiol      Boyle et al. (2018b) Front Physiol





## Closed-chest, automated optogenetic termination of atrial fibrillation in mice

- Gene painting technique used to light-sensitize RA
- Selective expression of ReaChR (red-sensitive)
- System is fully automated and self-contained
- AFib termination "in vivo" and in perfused hearts



Nyns et al. (2019) Sci Transl Med

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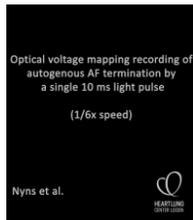
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Nyns et al. (2019) Sci Transl Med

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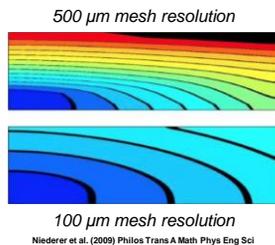
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## Computational complexity

- Ionic models have 20+ state variables
- $\mu$ s-order time steps required due to ODE system stiffness
- Millions of nodes due to stringent resolution limit (max: 200-300  $\mu$ m)



Niederer et al. (2009) Philos Trans A Math Phys Eng Sci

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## Computational complexity

- Even with the best available optimizations, clinically relevant simulations of cardiac electrophysiology require **HOURS OR DAYS OF RUNTIME ON DOZENS OF CPUs IN PARALLEL**
  - Millions of nodes due to stringent resolution limit (max: 200-300  $\mu\text{m}$ )
- 
- 100  $\mu\text{m}$  mesh resolution  
Niederer et al. (2009) Philos Trans A Math Phys Eng Sci

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