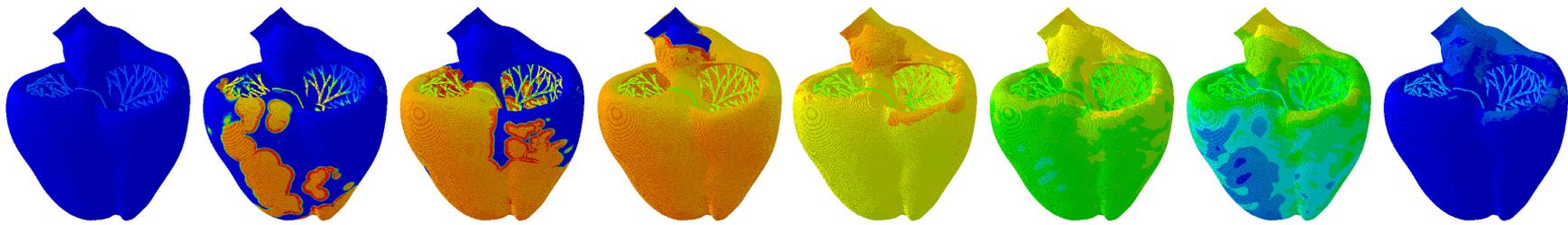
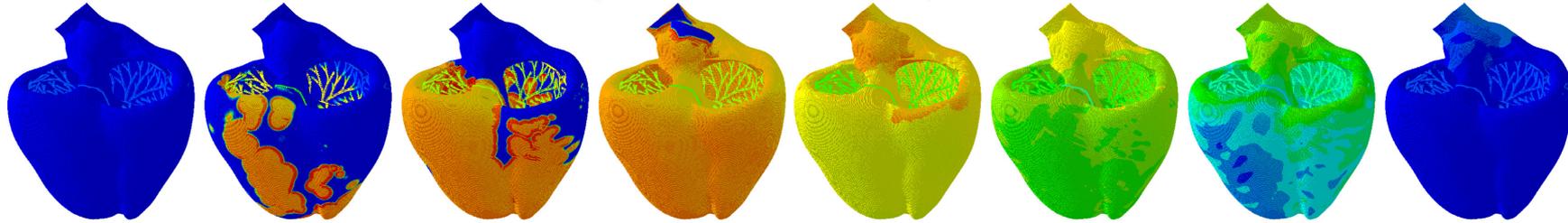


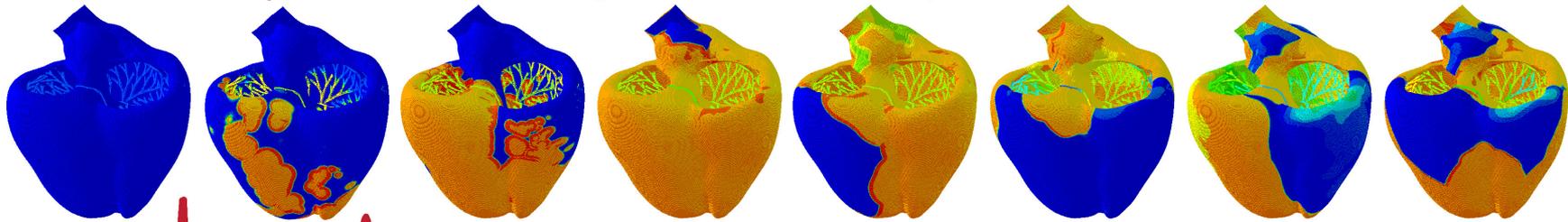
baseline



ranolazine



quinidine



# machine learning in drug development

FRANCISCO SAHLI COSTABAL • JIANG YAO • KINYA  
SEO • PARIS PERDIKARIS • ANNA SHER • EUAN  
ASHLEY • ELLEN KUHL

multiscale modeling consortium meeting • march 6-7, 2019



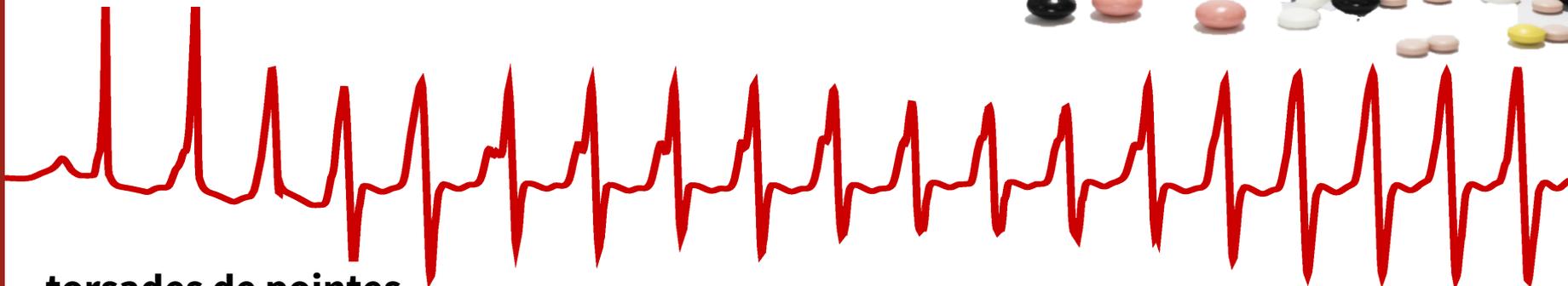
advania



# drug development



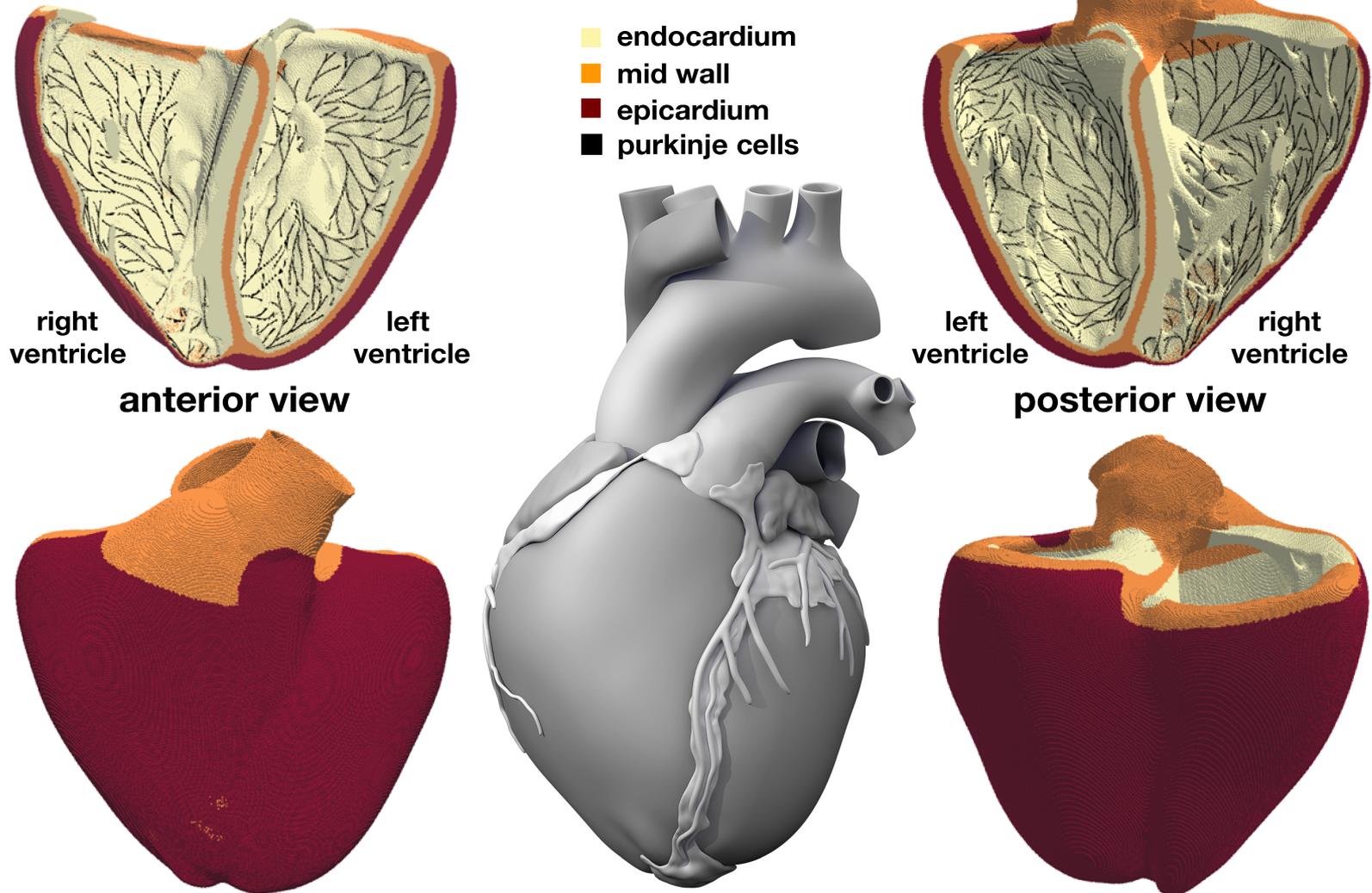
- numerous drugs have **serious side effects on the heart**
- gold standard safety test **action potential** and **QT interval** lengths
- criteria are **non-specific** / useful drugs are **falsely screened out**
- new drug - average cost **\$2.5 billion** / average time **> 10 years**
- CiPA initiative by FDA – **new paradigm for drug safety evaluation**



## torsades de pointes

colatsky et al. [2016], crumb et al. [2016], gintant et al. [2016], johannesen et al. [2014], mirams et al. [2011], sager et al. [2014], stockbridge et al. [2013], wang et al. [2017], vincente et al. [2016,2018]

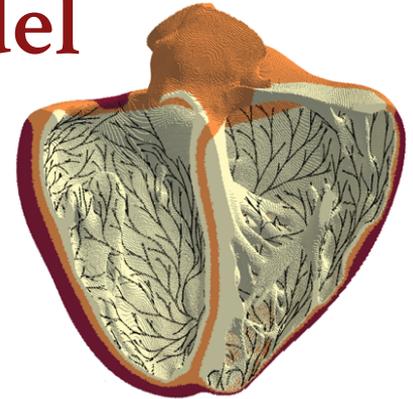
# organ model - the living heart



**spatial discretization: 0.3mm, 7M linear hexahedral elements, 8M nodes, 250M internal variables; temporal discretization: 0.005ms, 1M steps, 5 beats.**

sahli costabal, hurtado, kuhl [2016], <https://github.com/fsahli/fractal-tree>

# tissue model - monodomain model



- monodomain model - **action potential**

$$\dot{\phi} = \text{div}(\mathbf{D} \cdot \nabla \phi) + f^{\phi}$$

- flux term - second order **conductivity tensor**

$$\mathbf{D} = D_{\text{iso}} \mathbf{I} + D_{\text{ani}} \mathbf{f} \otimes \mathbf{f}$$

- source term - **ionic currents**

$$f^{\phi} = -I_{\text{ion}}/C_m \quad \text{with} \quad I_{\text{ion}} = I_{\text{ion}}(\phi, \mathbf{q}(\phi); t)$$

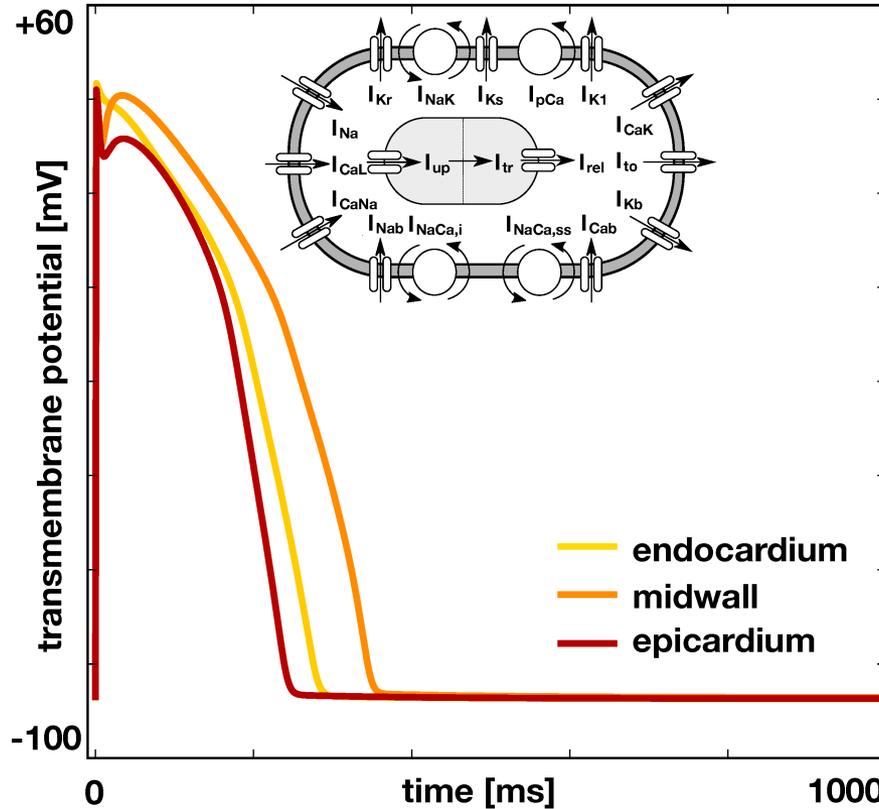
- ordinary differential equations for **state variables**

$$\dot{\mathbf{q}} = \mathbf{g}(\phi, \mathbf{q}(\phi); t)$$

- ventricular cells - **o'hara rudy model** - 15 currents / 39 state variables
- purkinje cells - **stewart model** - 14 currents / 20 state variables

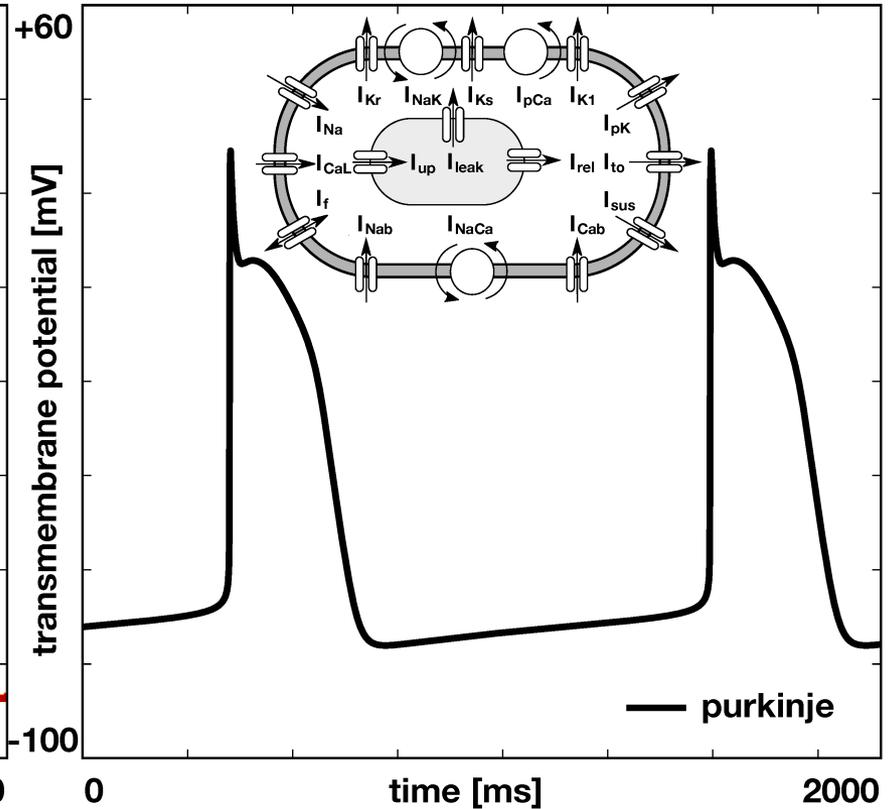
# cell model - ventricular and purkinje cells

## ventricular cells



$$\begin{aligned}
 I_{ion} = & I_{CaL} + I_{Na} + I_{Cab} + I_{Nab} \\
 & + I_{K_r} + I_{K_s} + I_{K_1} + I_{to} \\
 & + I_f + I_{sus} \\
 & + I_{NaK} + I_{pCa} + I_{pK} + I_{NaCa}
 \end{aligned}$$

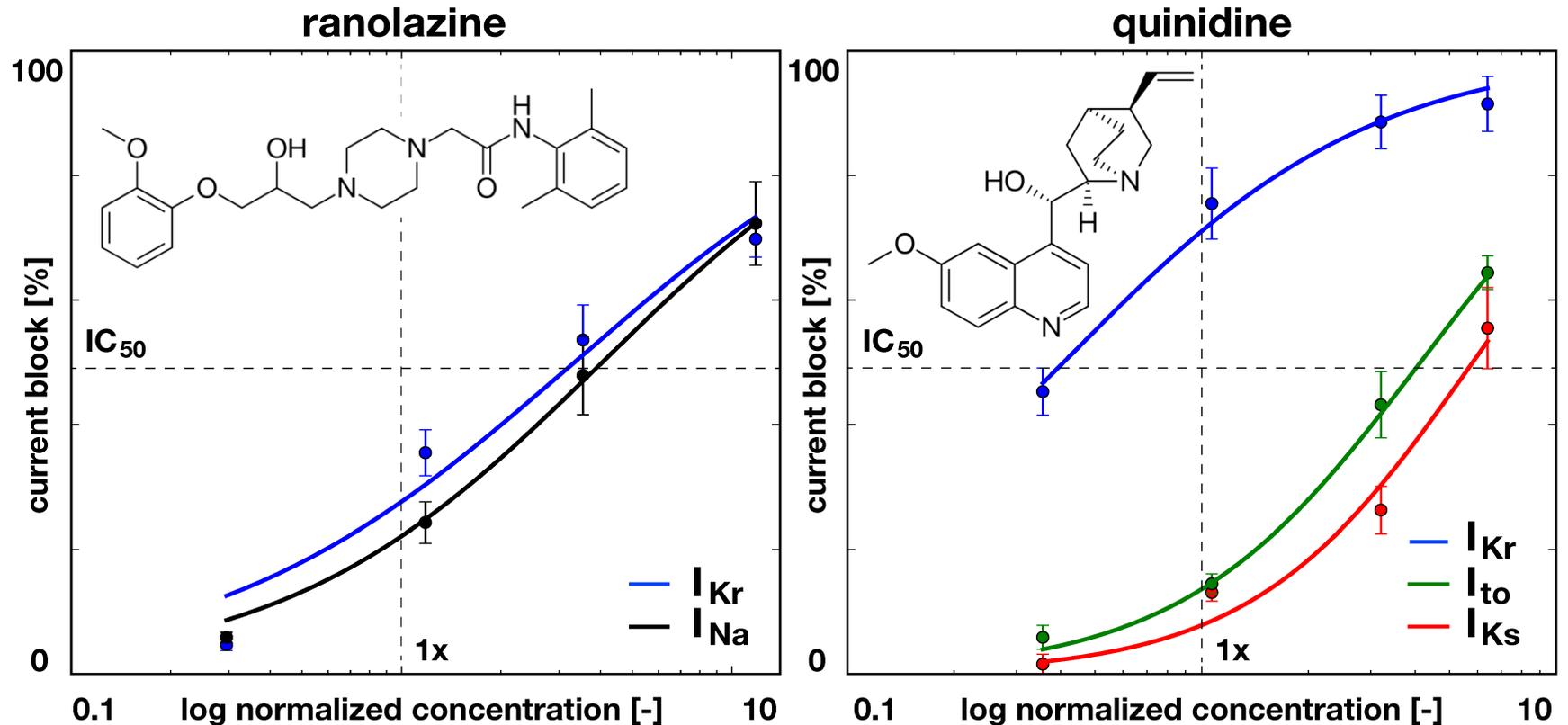
## purkinje cells



$$\begin{aligned}
 I_{ion} = & I_{CaL} + I_{Na} + I_{CaNa} + I_{CaK} \\
 & + I_{Cab} + I_{Nab} + I_{K_b} \\
 & + I_{K_r} + I_{K_s} + I_{K_1} + I_{to} \\
 & + I_{NaK} + I_{pCa} + I_{NaCa,i} + I_{NaCa,ss}
 \end{aligned}$$

o'hara, virag, varro, rudy [2011], stewart, aslanidi, noble, noble, boyett, zhang [2009]

# drug model - ranolazine and quinidine



- calculate **ionic current**
- two-parameter **Hill-type model**
- calculate drug-specific block

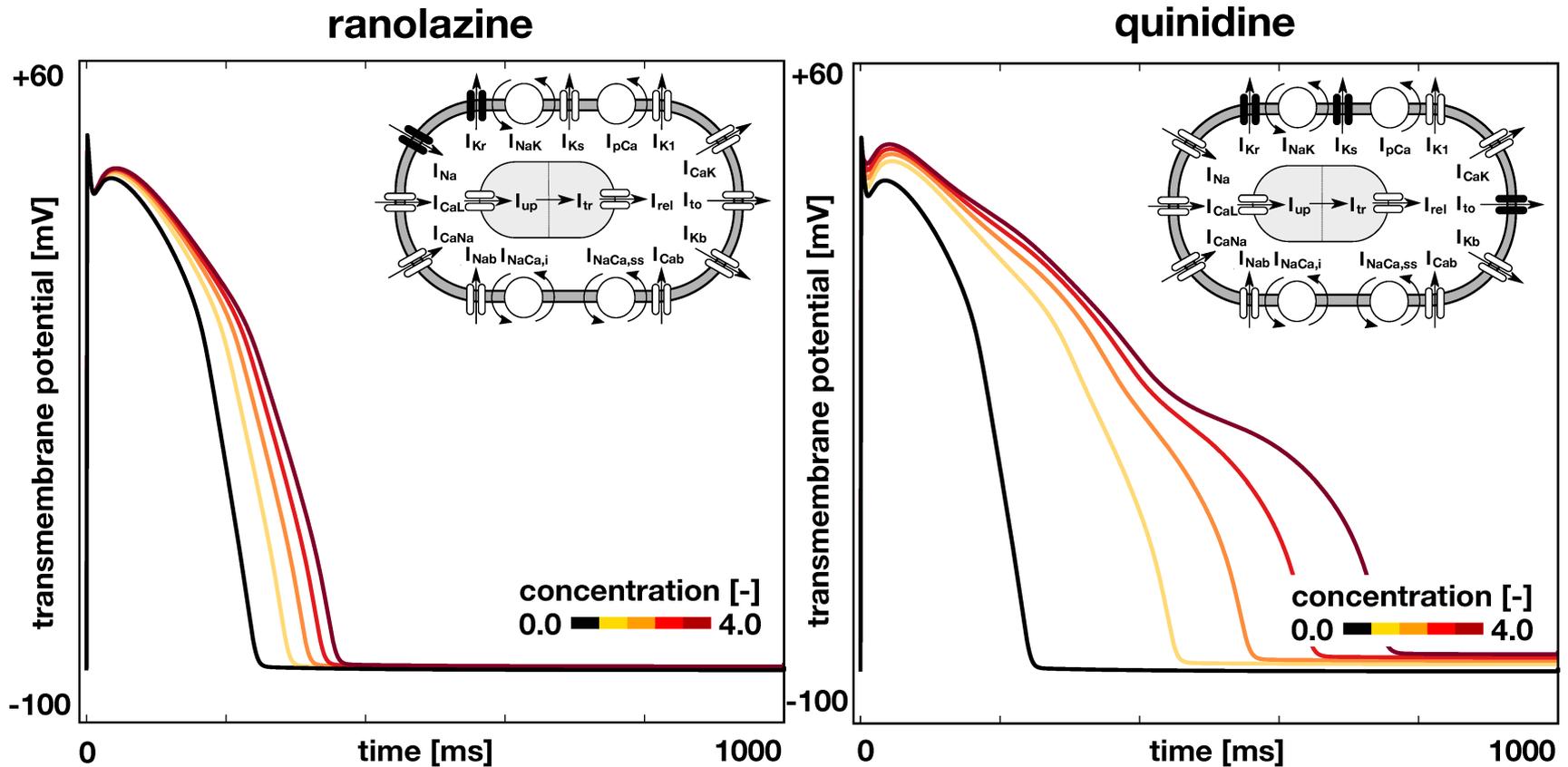
$$I_{ion} = I_{ion}(\phi, q(\phi); t)$$

$$\beta = C^h / [IC_{50}^h + C^h]$$

$$I_{drug} = [1 - \beta] I_{ion}$$

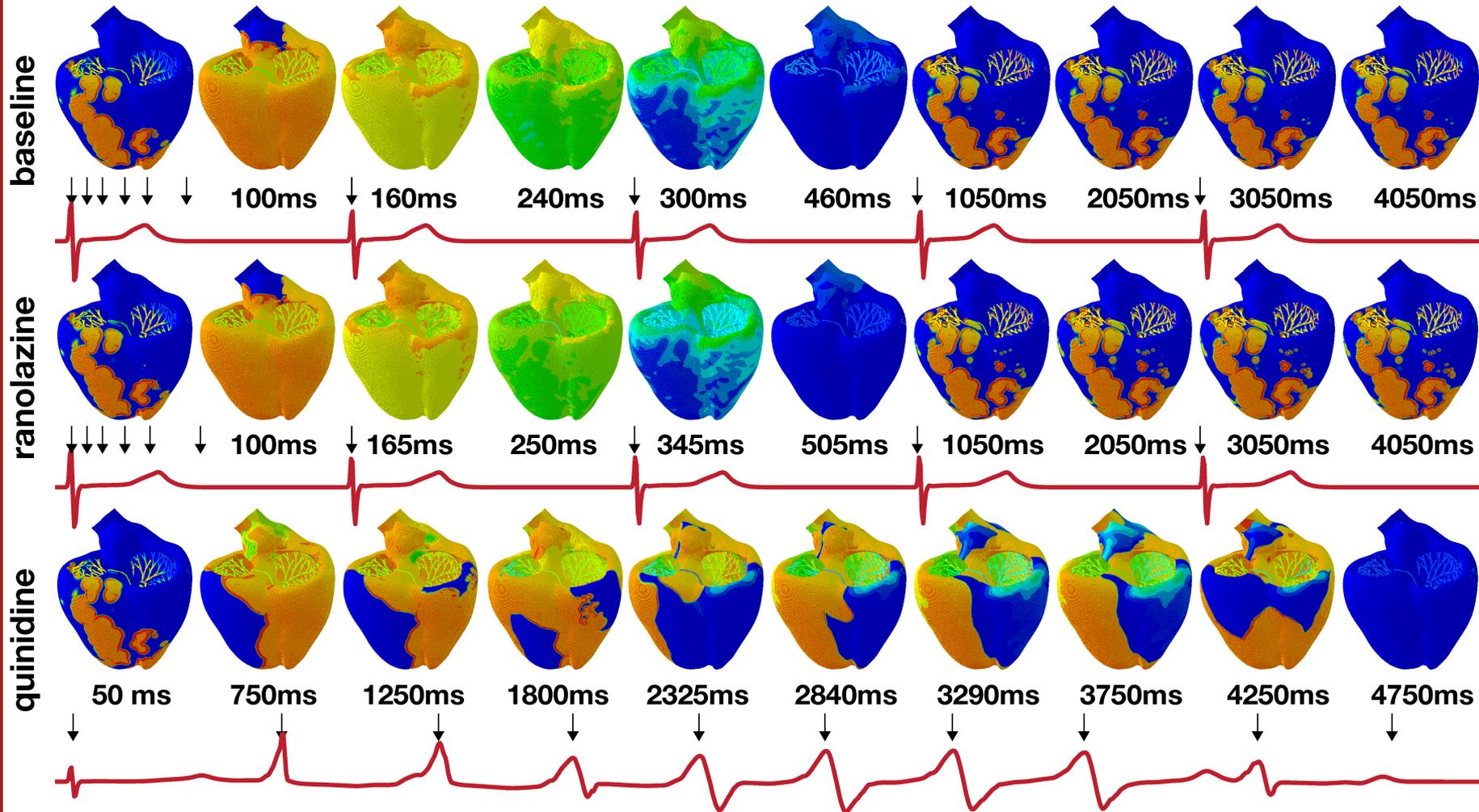
mirams, cui, sher, fink, cooper, heath, mc mahon, gavaghan, noble [2011], colatsky, fermini, gintant, pierson, sager, sekino, strauss, stockbridge[2016], crumb, vicente, johannesen, strauss[2016]

# drug model - effects on the cell level

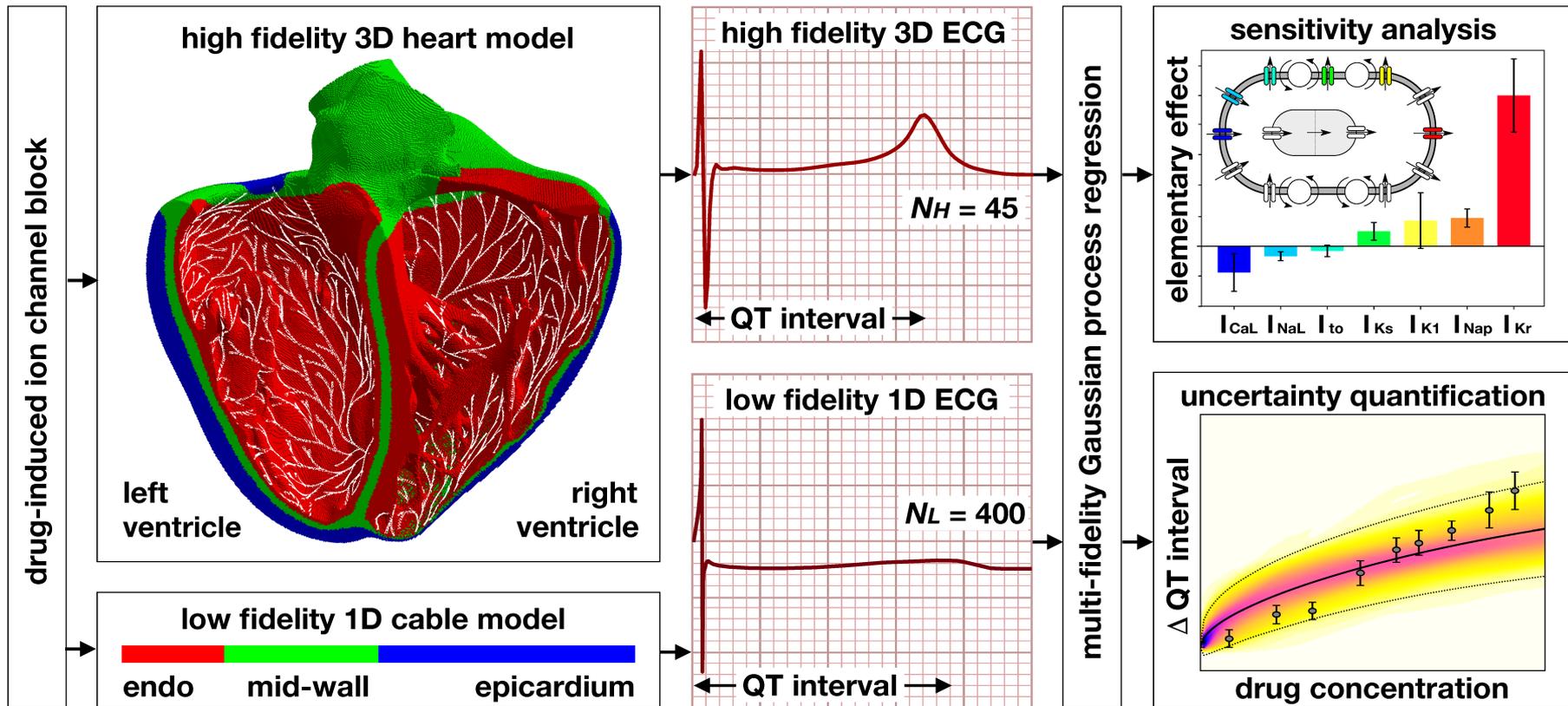


- **ranolazine** chronic angina drug
- blocks  $I_{Kr}$  and  $I_{Na}$
- **mildly prolongs** APD and QT
- **low** torsades de pointes **risk**
- **quinidine** antiarrhythmic agent
- blocks  $I_{Kr}$  and  $I_{Ks}$  and  $I_{to}$
- **severely prolongs** APD and QT
- **high** torsades de pointes **risk**

# drug model - effects on the organ level

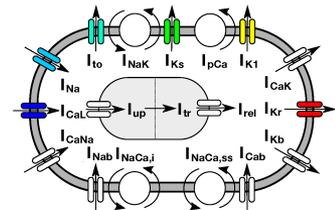
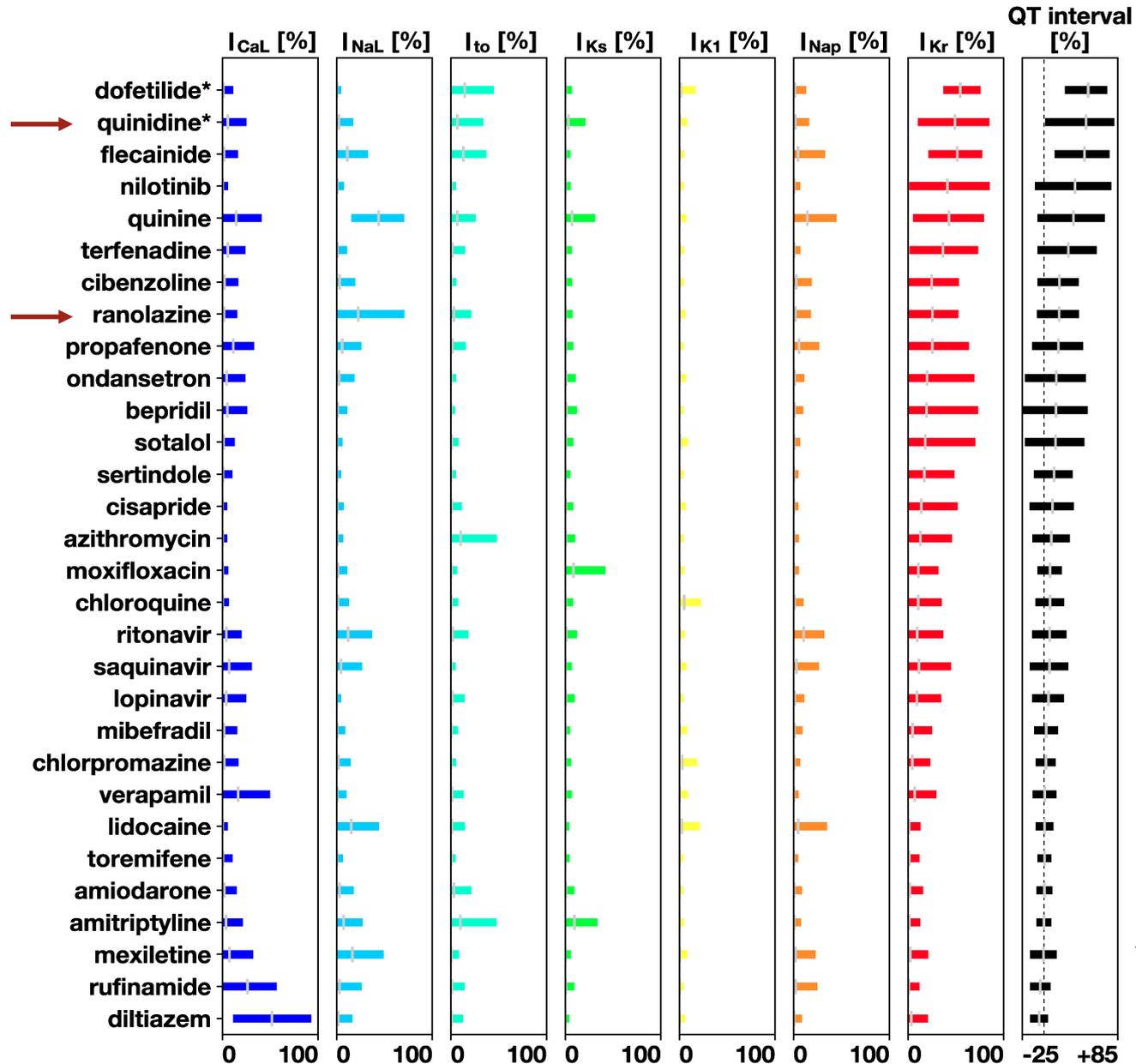


# using machine learning in drug development

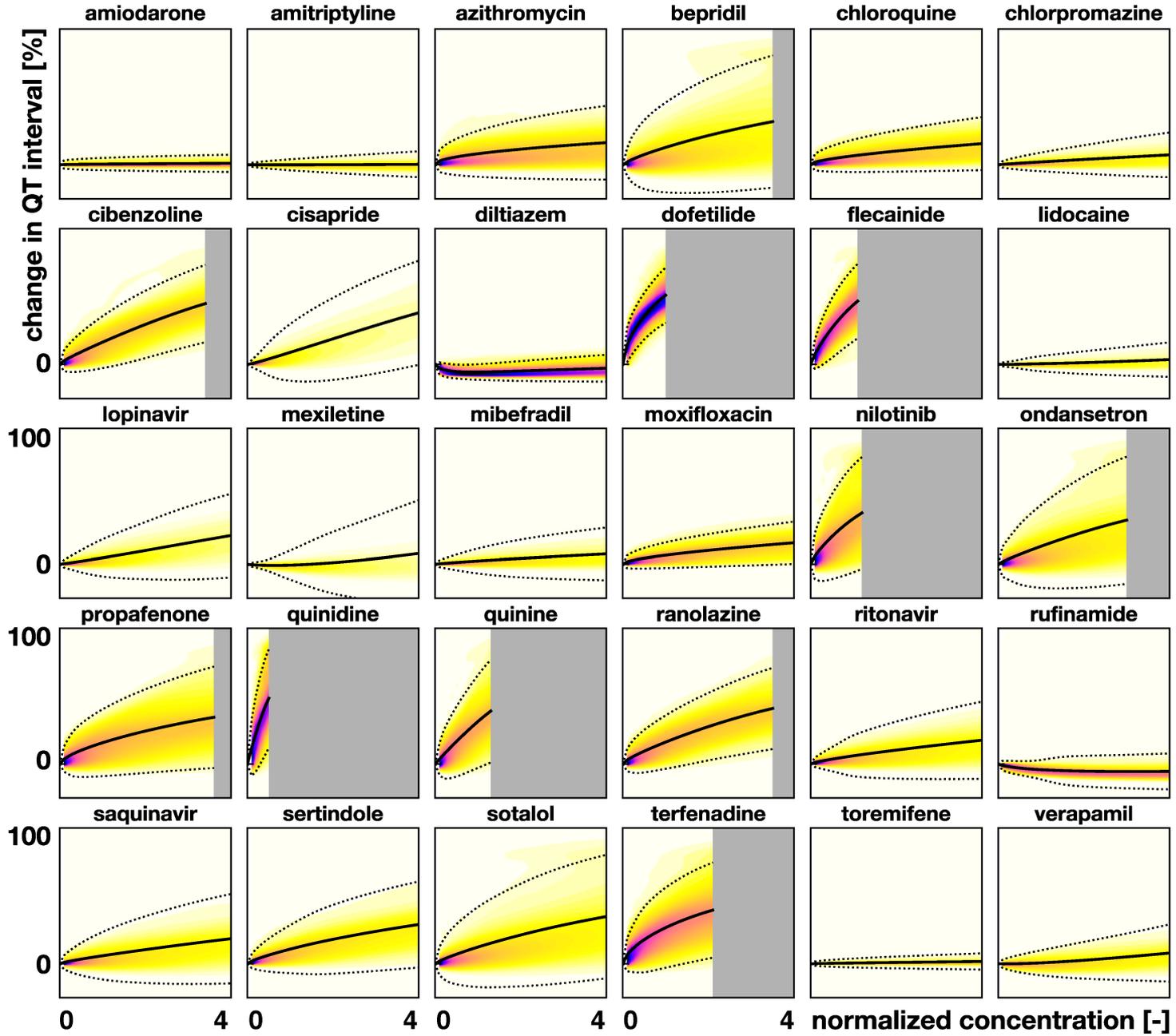


characterizing effect of 30 drugs on the QT interval using gaussian process regression, surrogate model for sensitivity analysis and uncertainty quantification

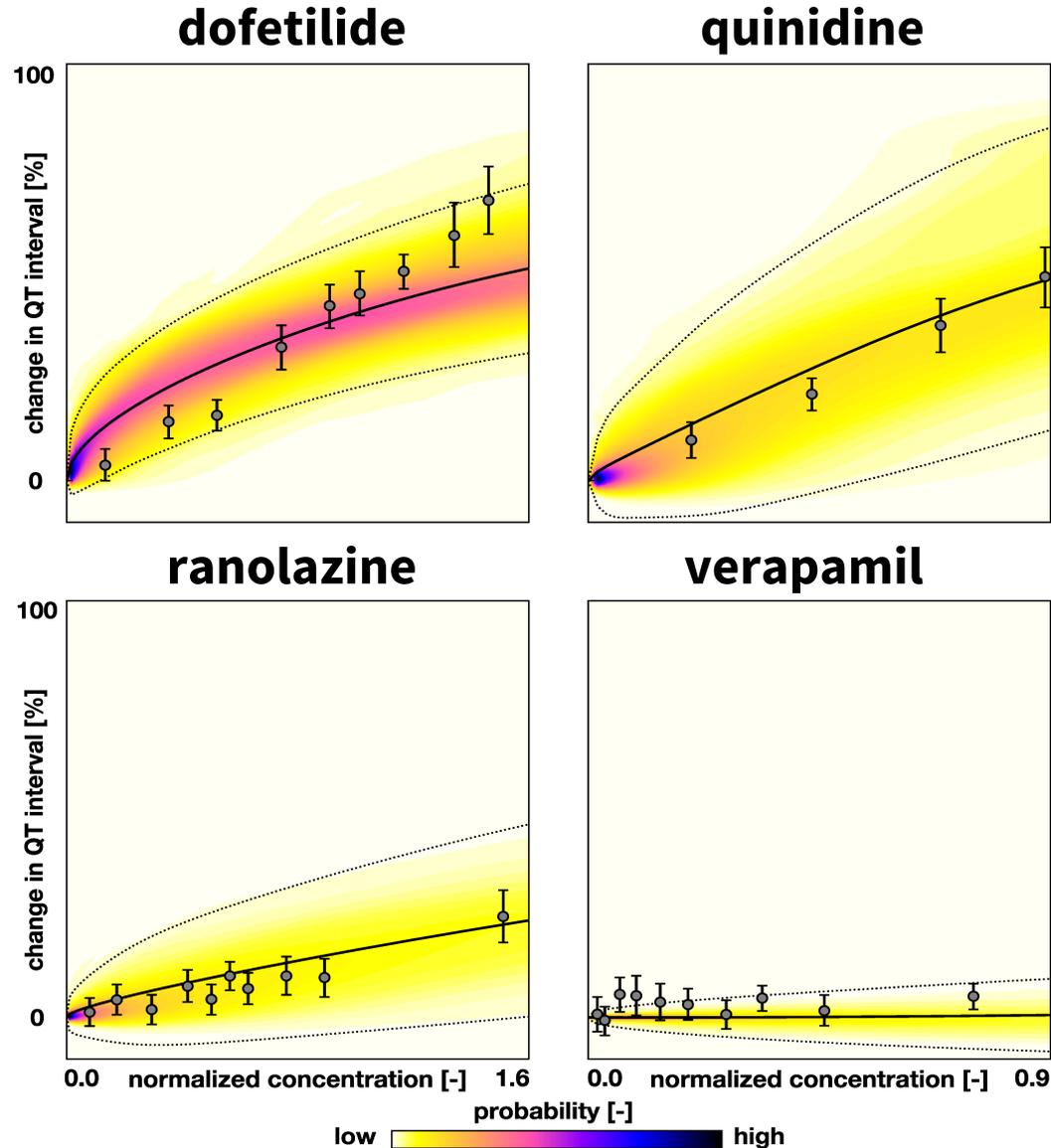
# uncertainty quantification for 30 drugs



# uncertainty quantification for 30 drugs

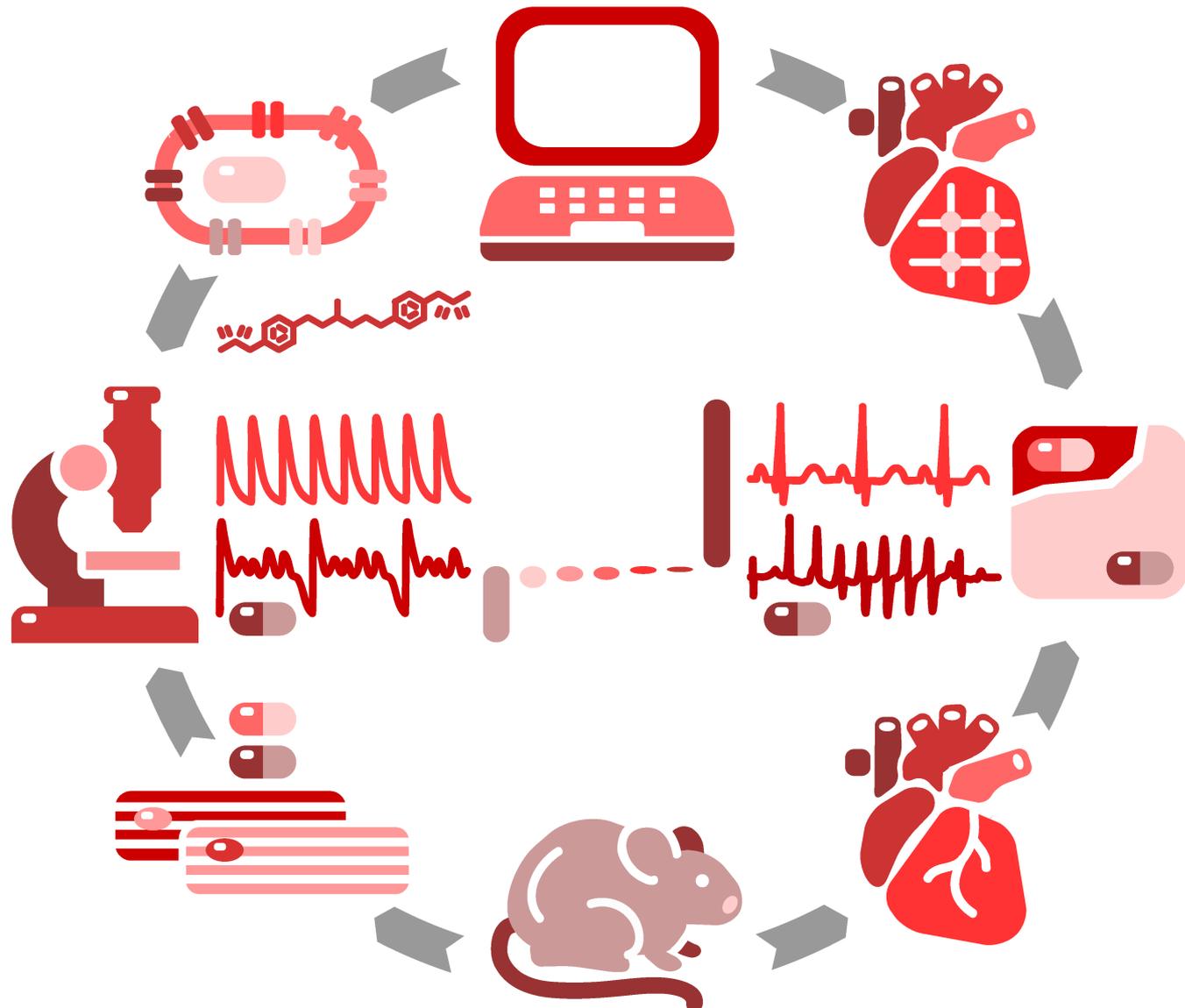


# validation of uncertainty quantification



validation of QT interval change for drugs dofetilide, quinidine, ranolazine, and verapamil data from randomized clinical trial, error bars 95% confidence; johannesen et al. [2014]

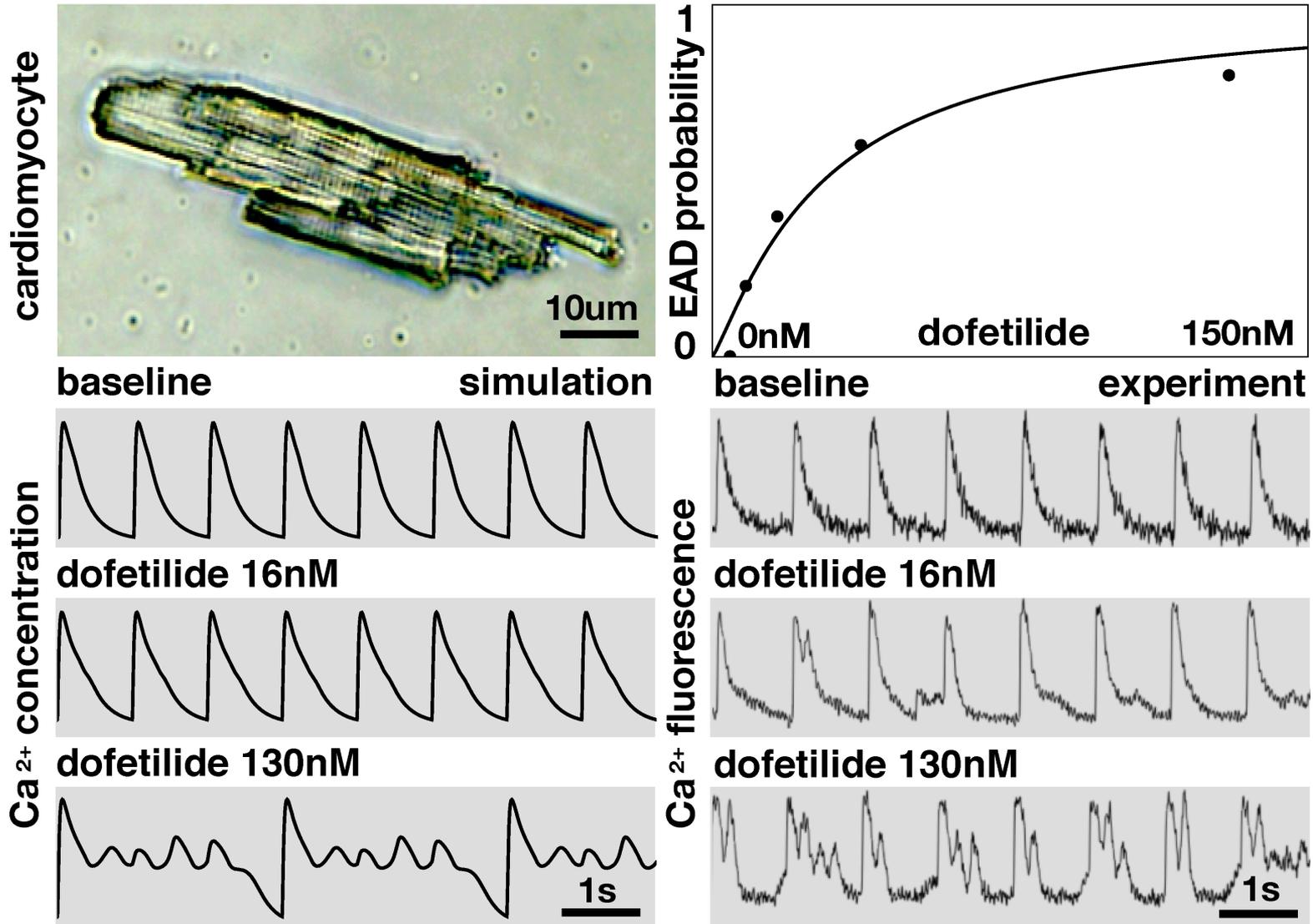
# validation on cell and organ levels



sahli costabal, seo, ashley, kuhl [2019]

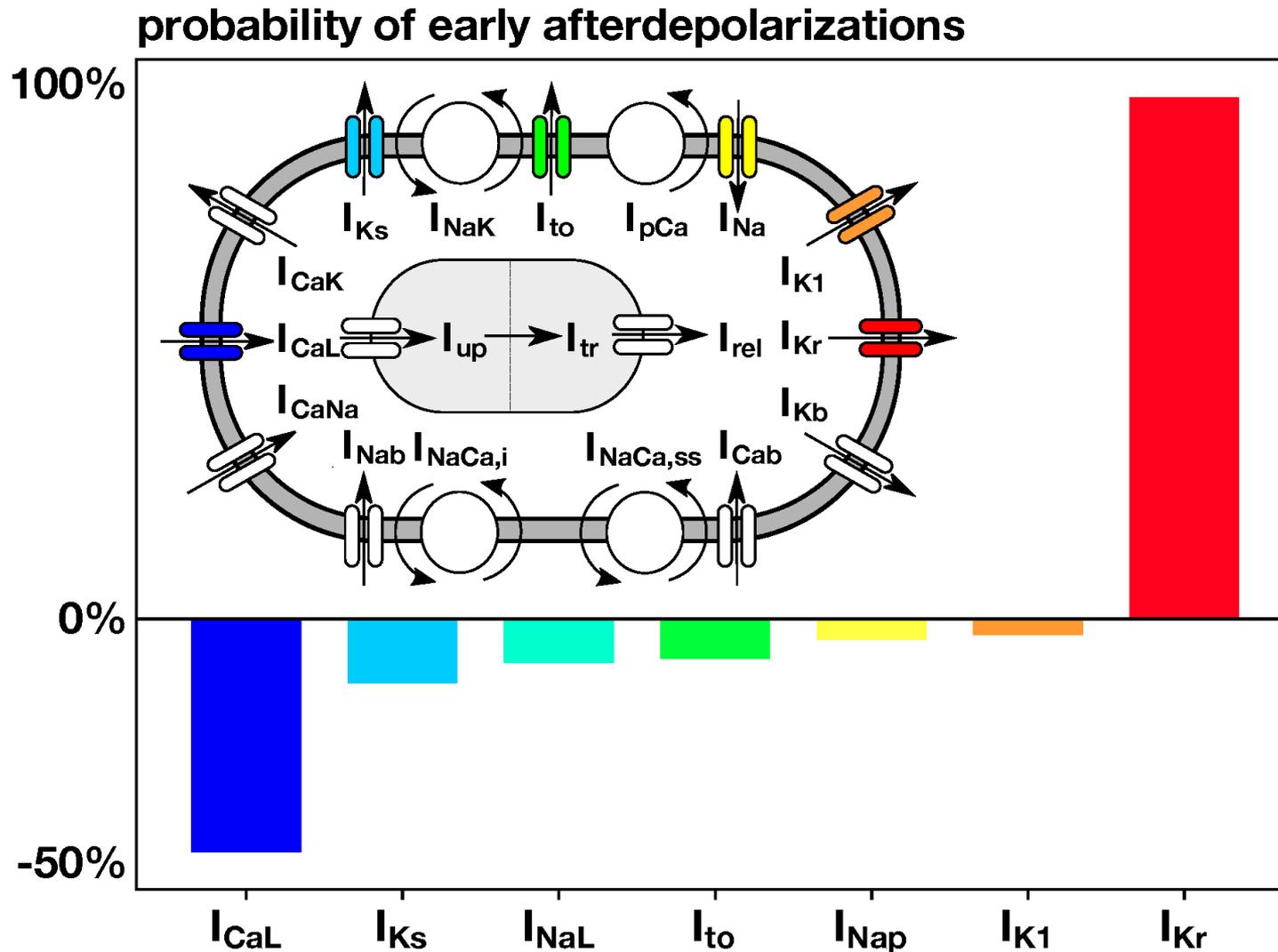
Stanford University

# cell level validation - early afterdepolarizations



early afterdepolarizations. simulation and isolated rat cardiomyocytes at dofetilide concentrations of 4nM, 8nM, 16nM, 38nM, 130nM (n=6 cells each).

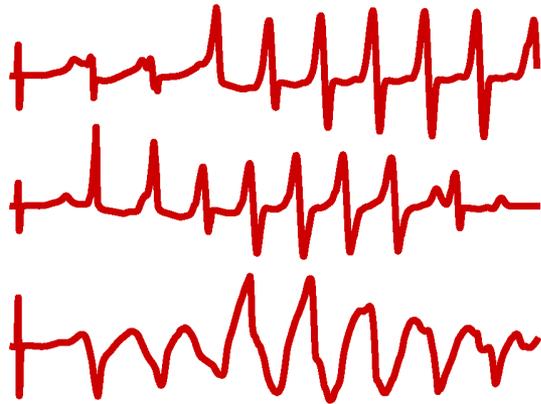
# cell level sensitivity analysis - ion channels



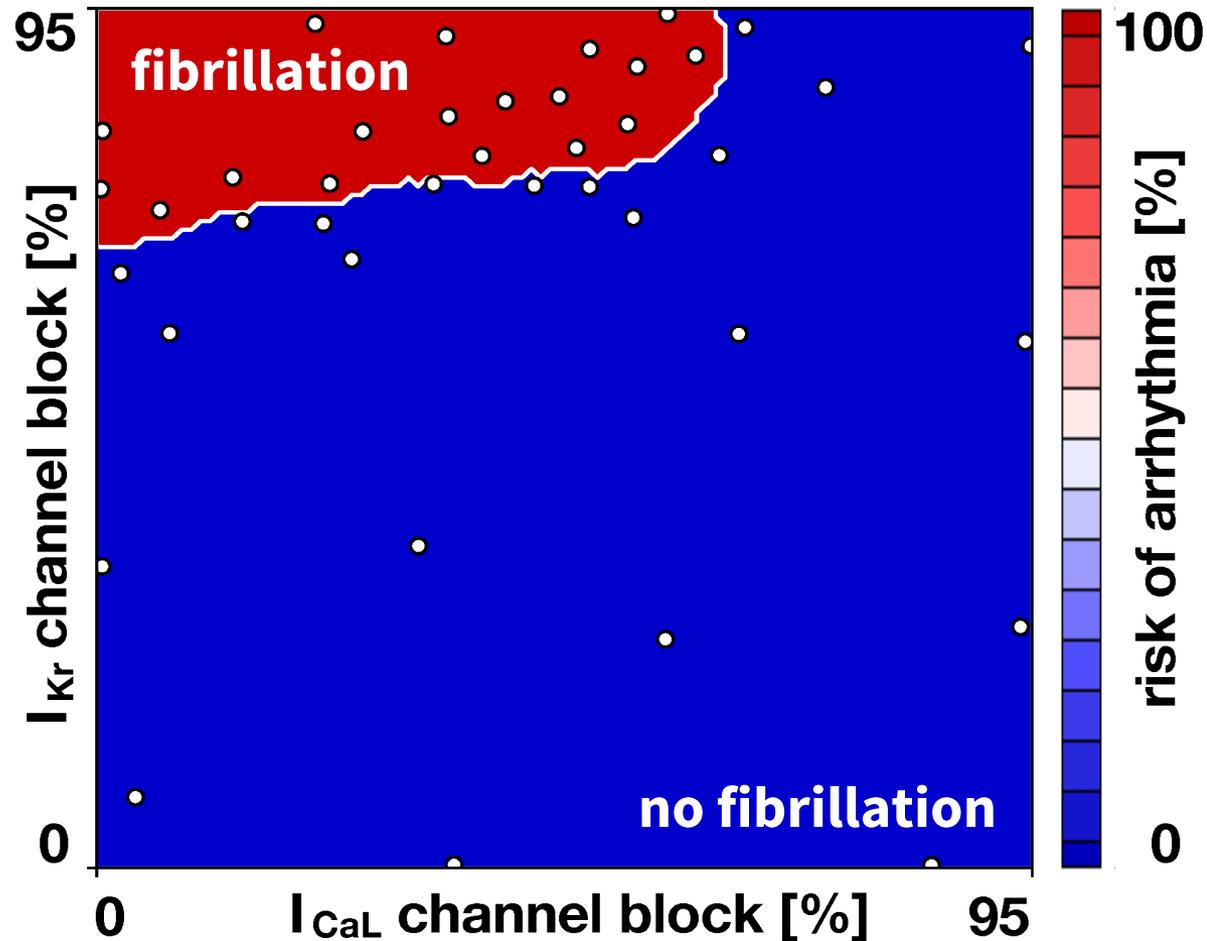
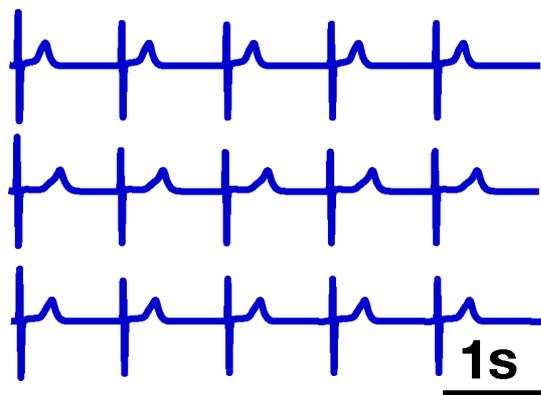
n = 500 single cell simulations > EAD > logistic regression > marginal effects.  
 blocking  $I_{Kr}$  and  $I_{CaL}$  increases and reduces risk of early afterdepolarizations

# new pro-arrhythmic risk classifier

high risk

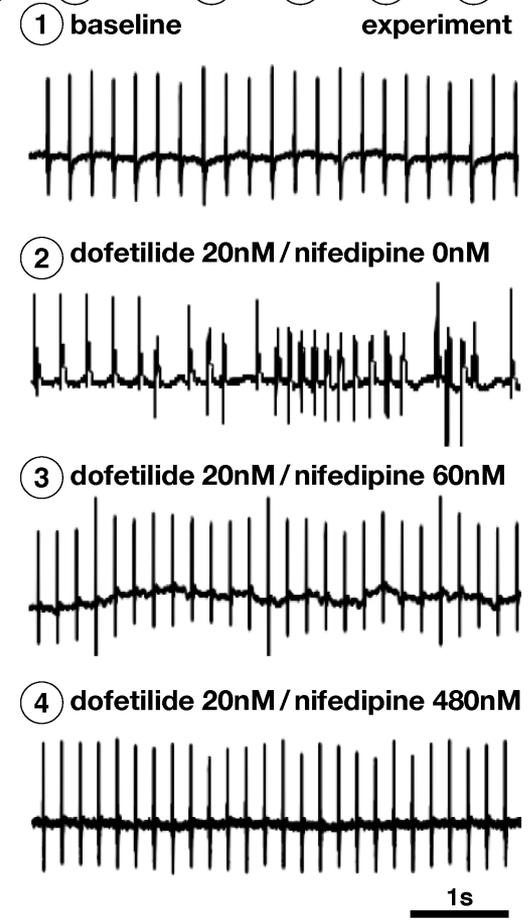
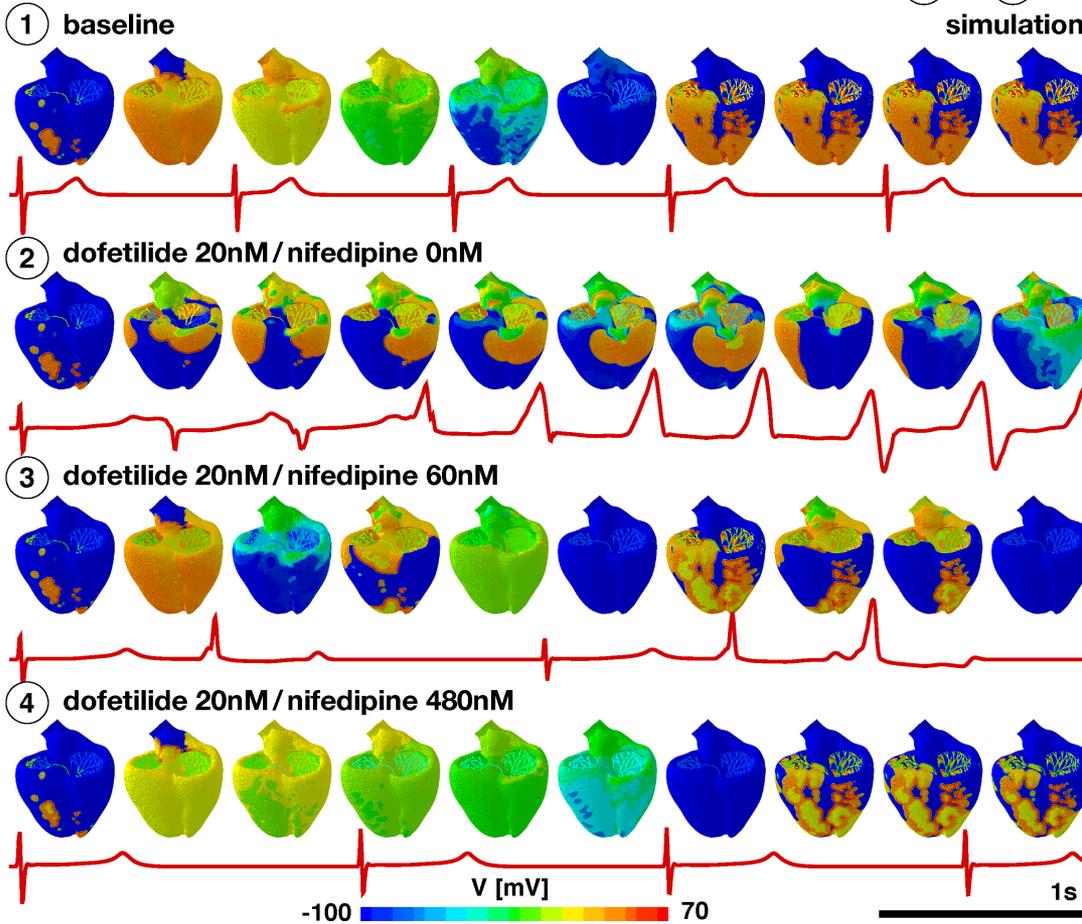
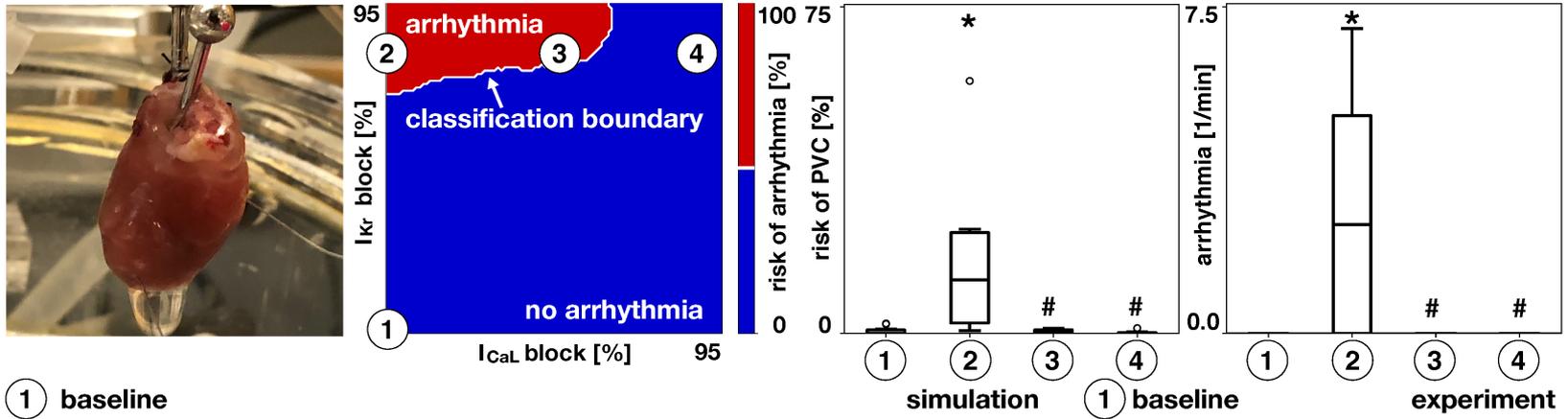


low risk

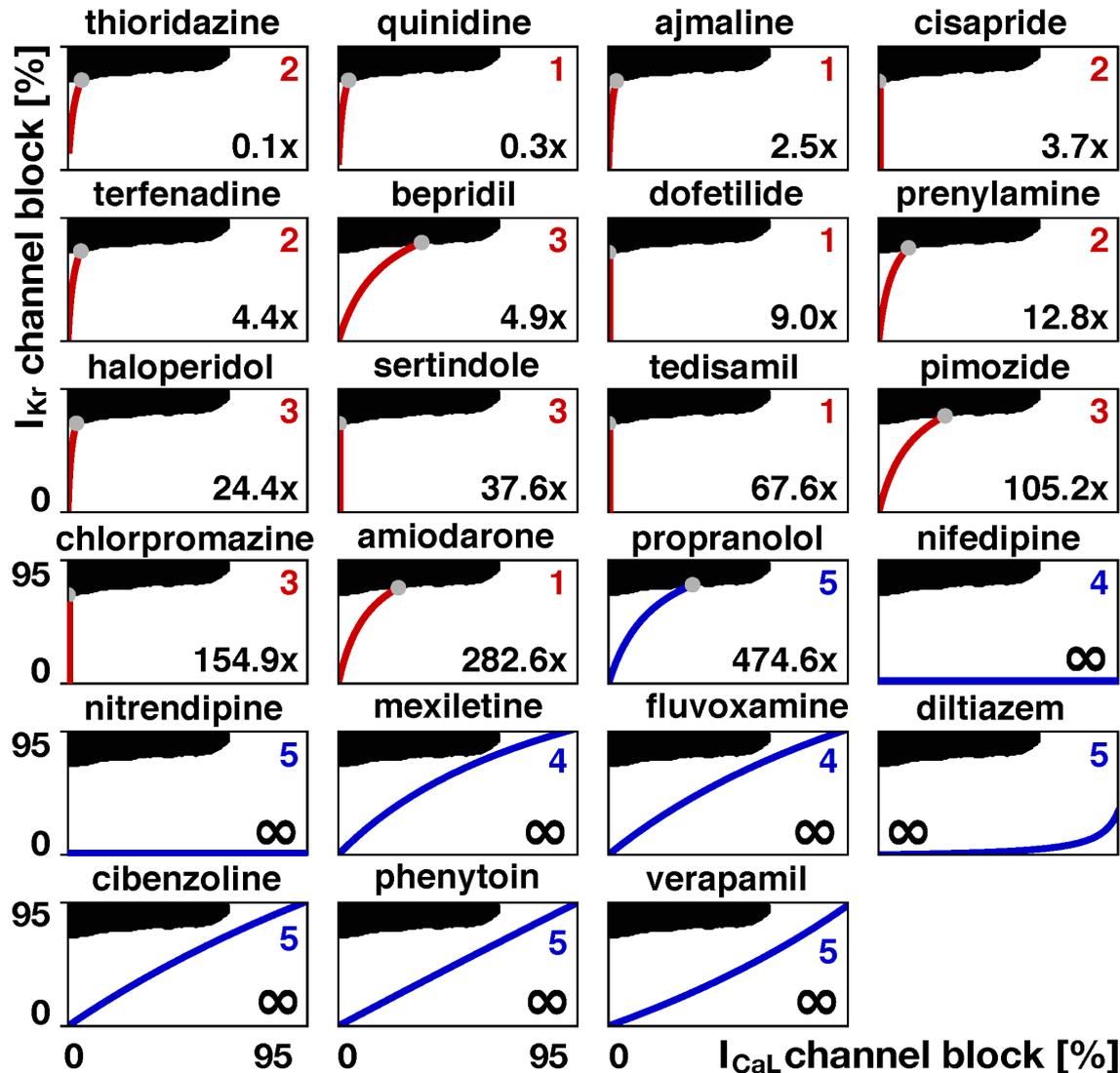


particle learning method to sample classification boundary within  $I_{Kr}$  /  $I_{CaL}$  space, gaussian process classifier, adaptively sample of point of maximum entropy, create  $n = 10$  samples from latin hypercube design, sample  $n = 30$  samples adaptively

# organ level validation - arrhythmogenic risk

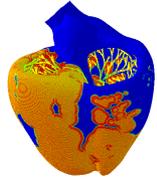


# new paradigm for drug safety evaluation?

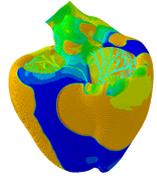


risk stratification of 23 drugs using our pro-arrhythmic risk classifier. numbers x indicate critical concentration; 1-5 risk category; red = torsadogenic, blue = safe.

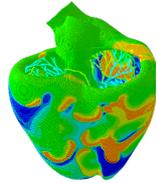
# machine learning in drug development



multi fidelity **gaussian process regression** - sensitivities  $I_{CaL} / I_{Kr}$



**uncertainty quantification** – effect of variations on **QT interval**



gaussian process **classification** – risk classifier in **polypharmacy**

FRANCISCO SAHLI COSTABAL • PARIS PERDIKARIS • JIANG YAO • ANNA SHER • KINYA SEO • EUAN ASHLEY



NSF CAREER award the virtual heart  
BioX interdisciplinary seed grant 2018



NIH U01HL119578 multi-scale laws of myocardial growth and remodeling

Stanford University