

Questions and Challenges in Tissue and Whole Organ Modeling

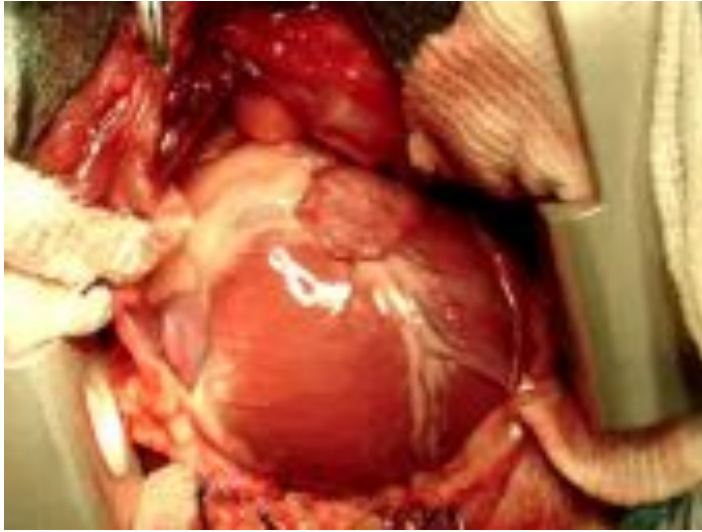
Radu Grosu

Vienna University of Technology

NSF-Expeditions Project CMACS

NSF-Frontiers Project CyberHeart

The Grand Challenge



Error-Free Heart

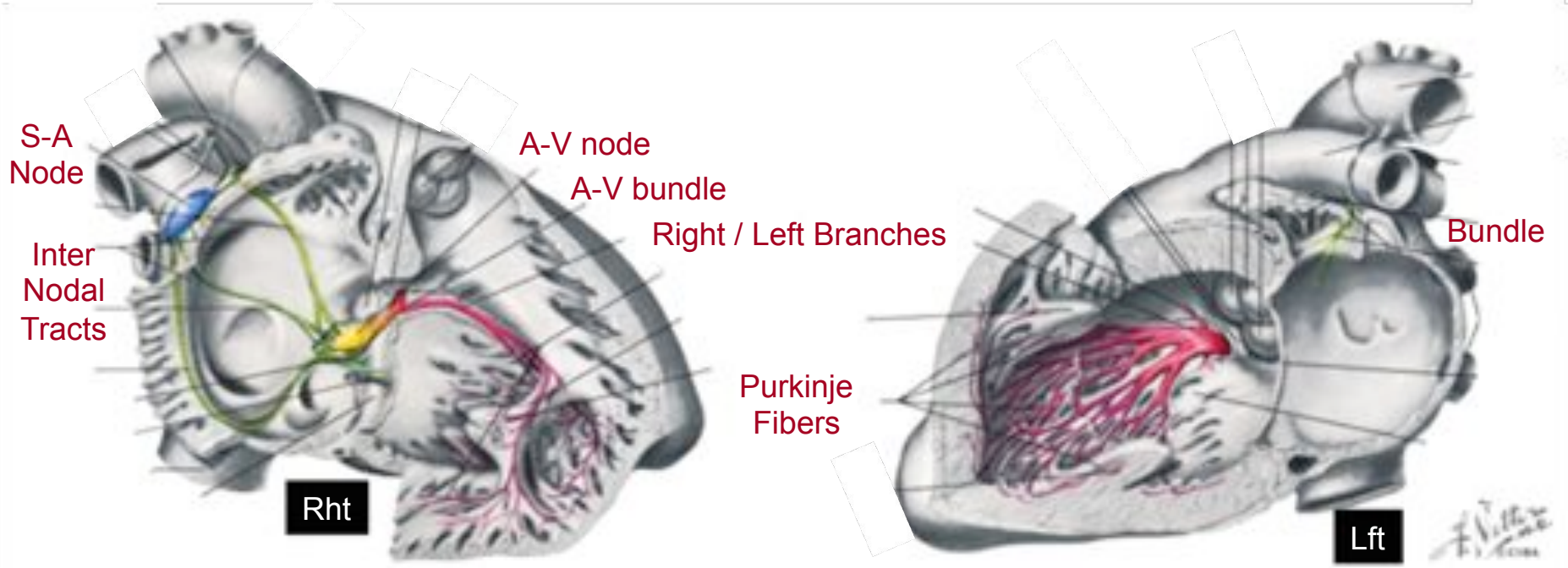


Error-Prone Heart

Can we predict and control abnormal behavior?

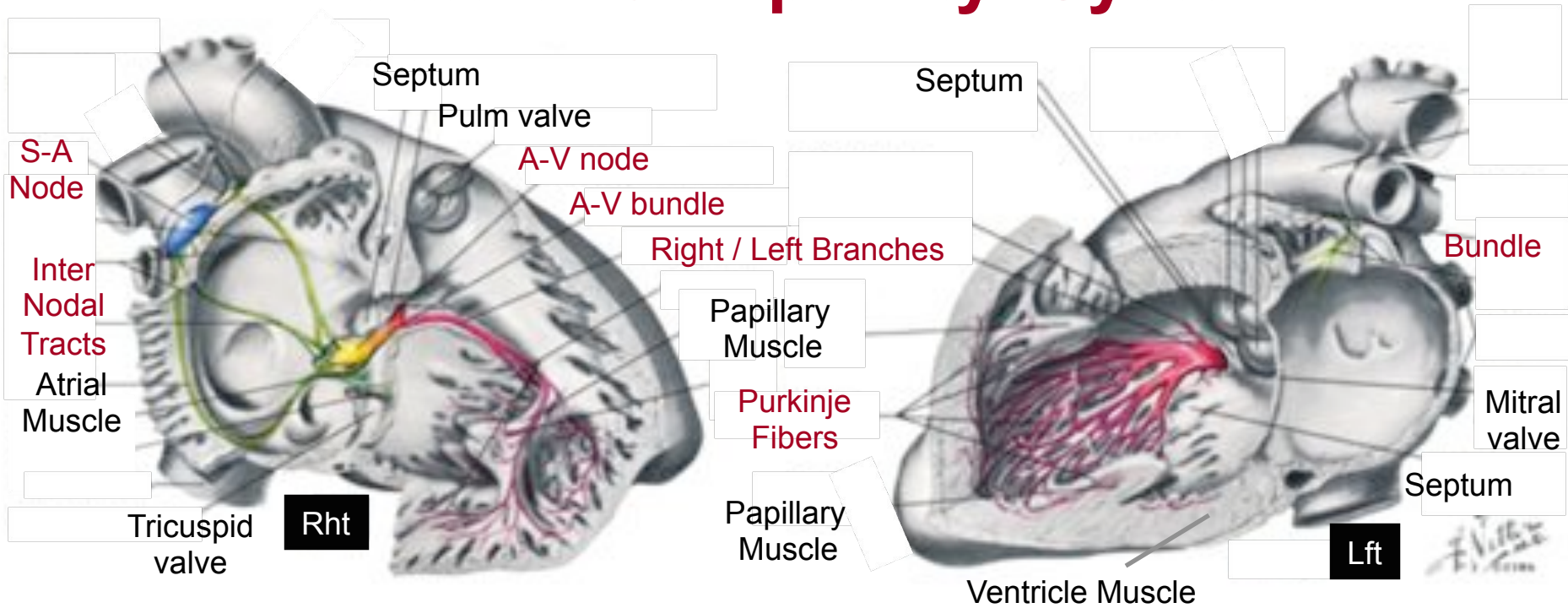
- **Abstraction, approximation and composition**

A Taste of Complexity: Systems



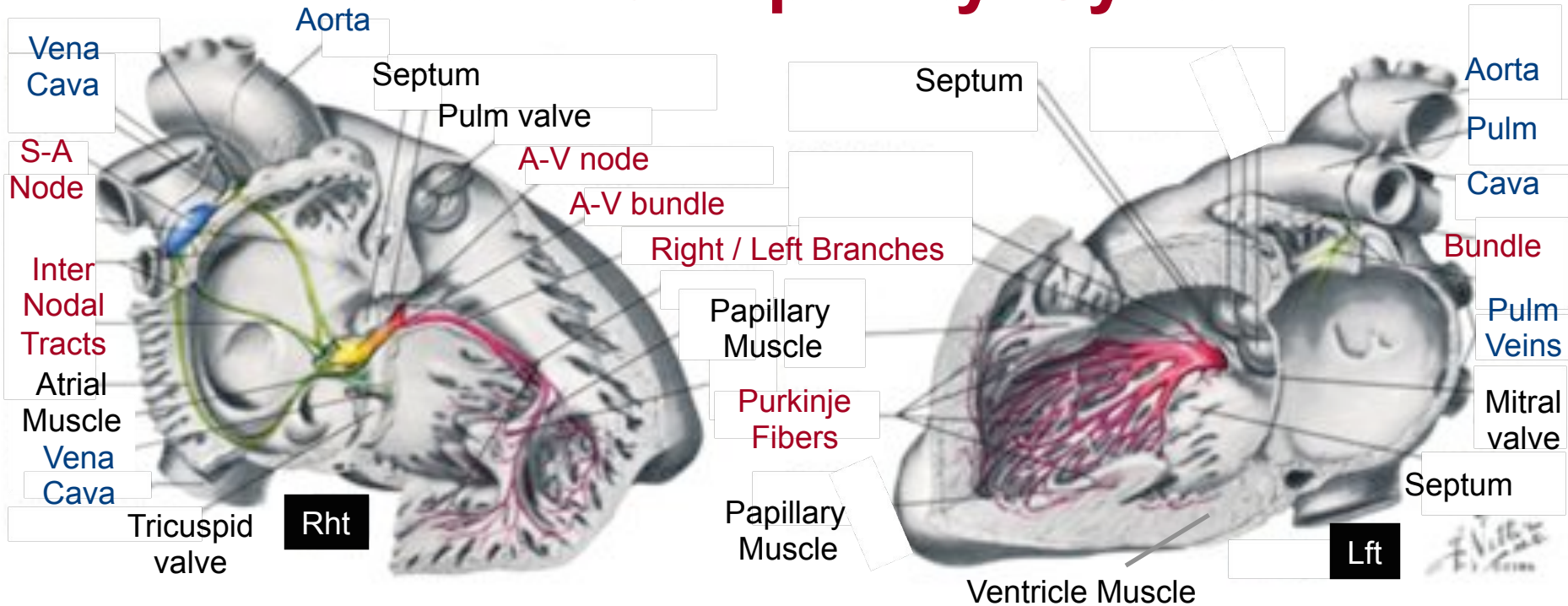
Electrical system: Responsible for synchroniztion

A Taste of Complexity: Systems



Mechanical system: responsible for contraction

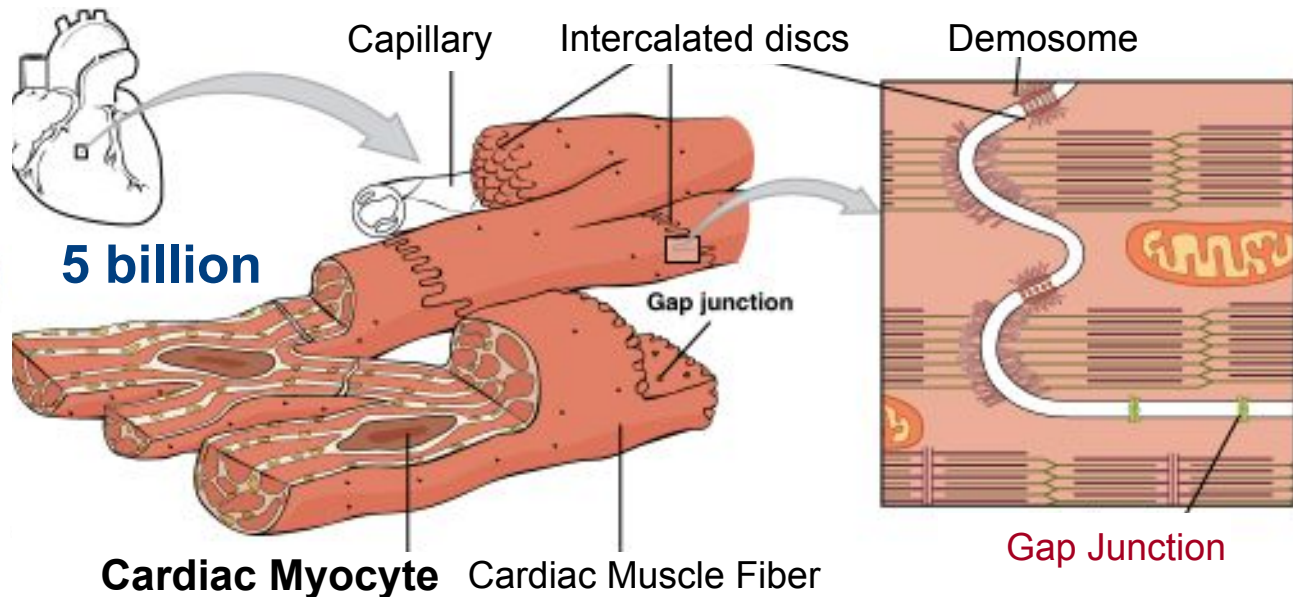
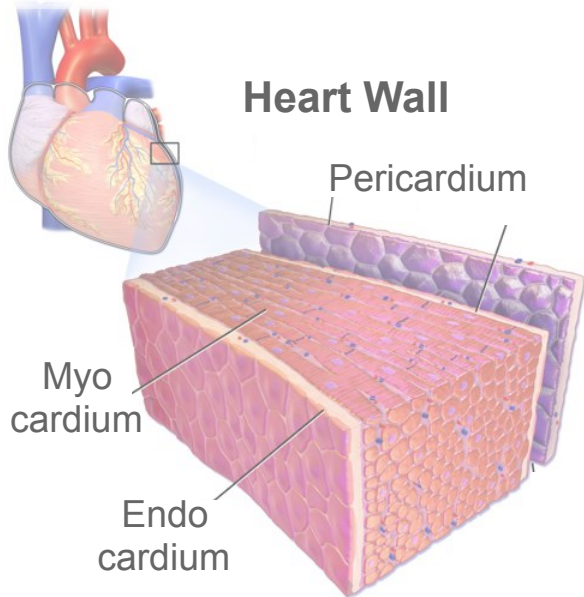
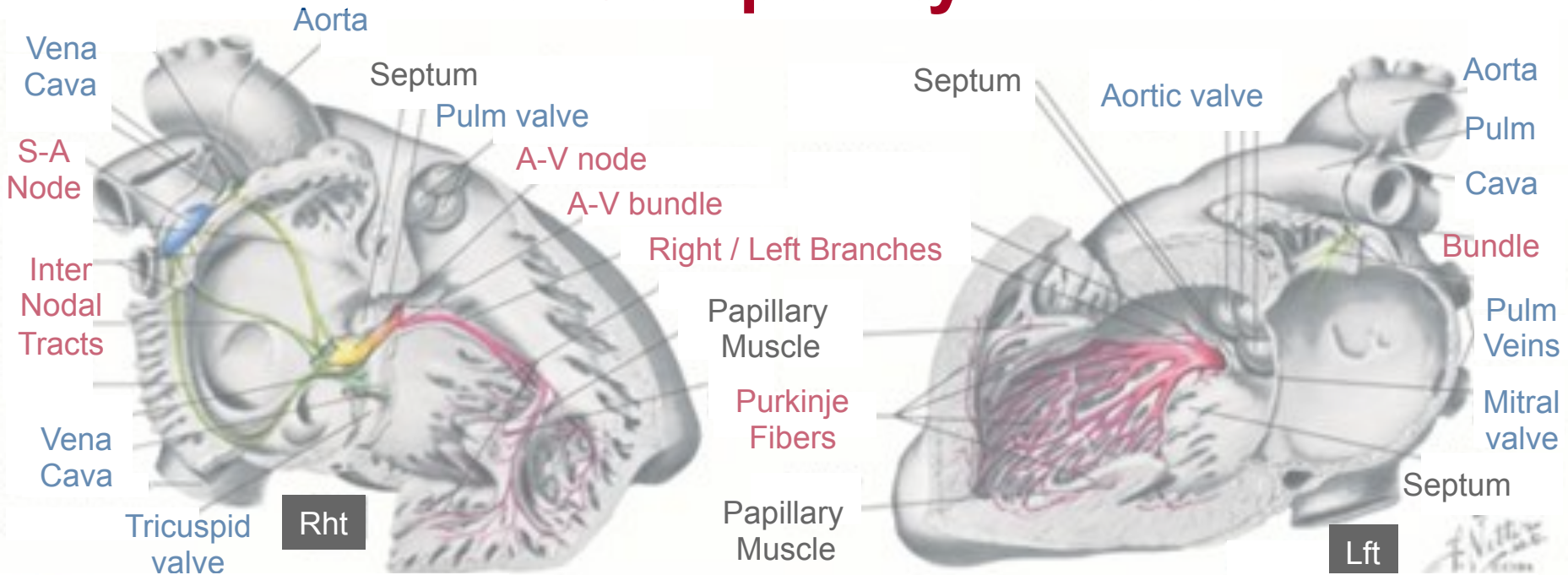
A Taste of Complexity: Systems



Vascular system: responsible for transportation

System abstraction: electrical, mechanical, vascular

A Taste of Complexity: Multiscale

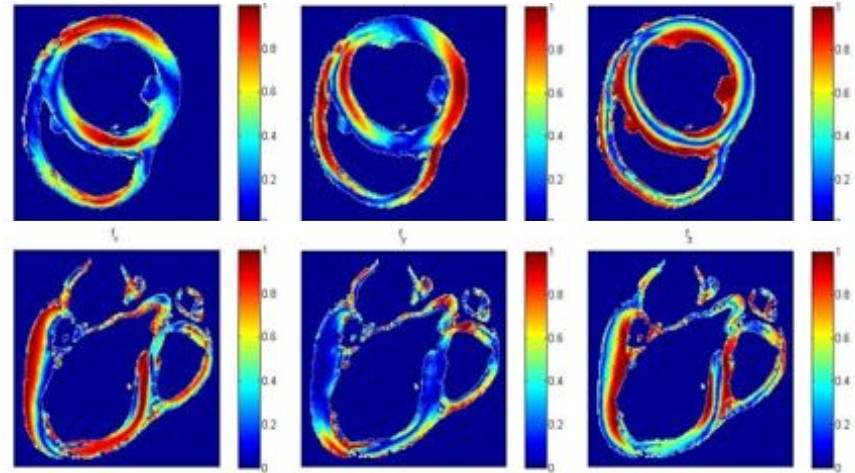


A Taste of Complexity: Tools

Complicated structure: Atrium

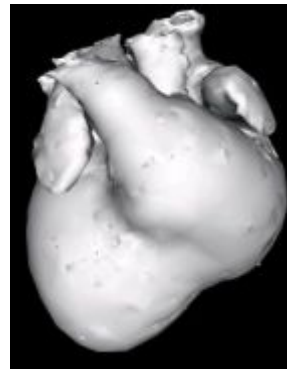


Canine heart: slices
(DTMRI @ 250 microns resolution)



Ventricle

Matrix

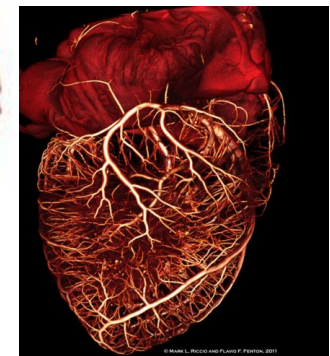


Pittsburgh NMR Center

Fibers



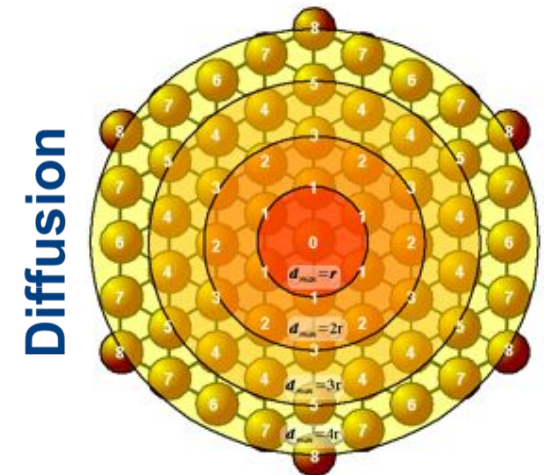
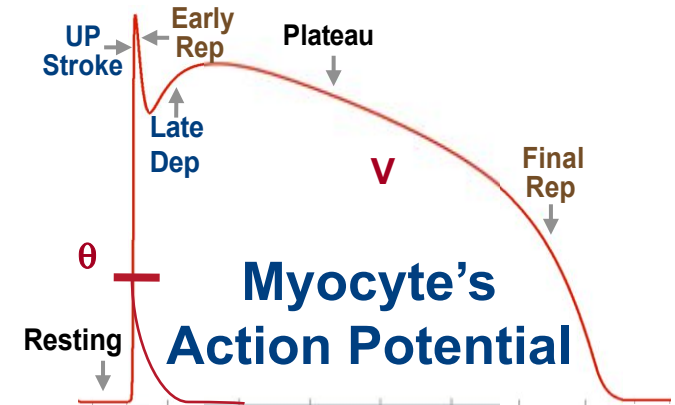
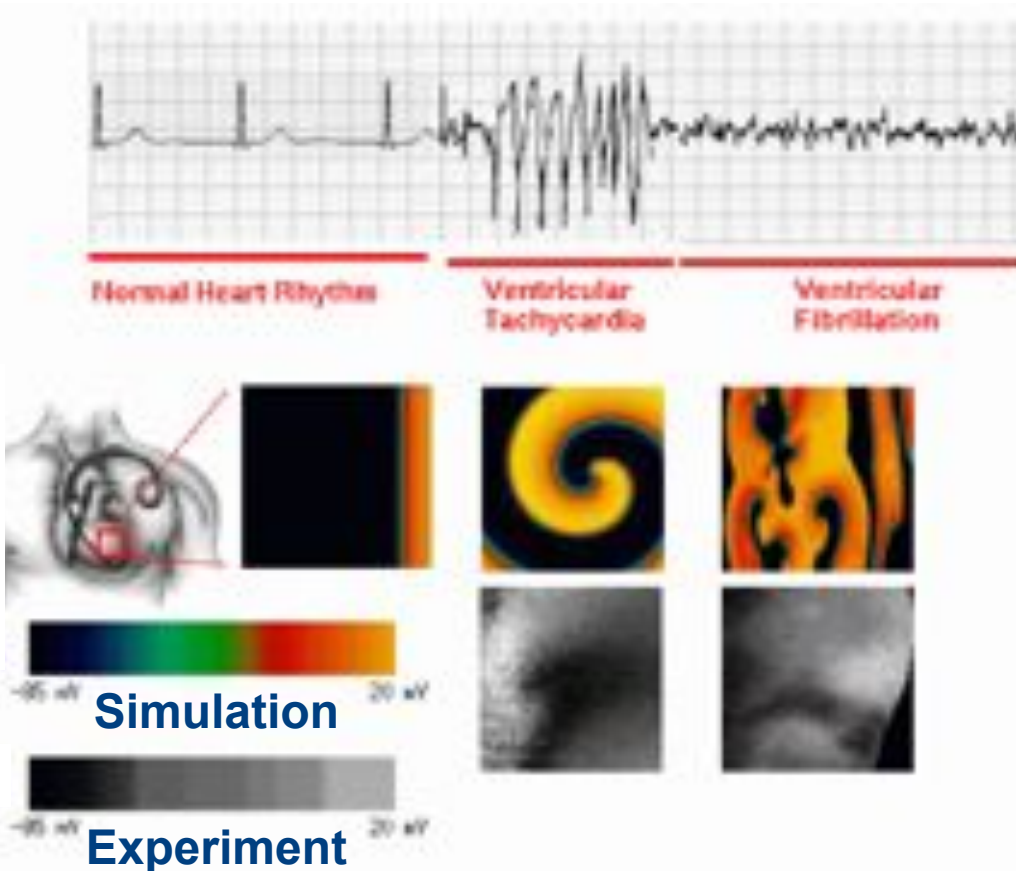
Vessels



MicroCT Cornell

5 billion cells synchronize their contraction to create a heart beat!

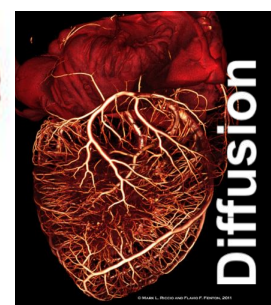
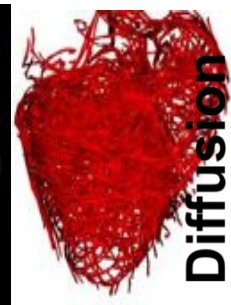
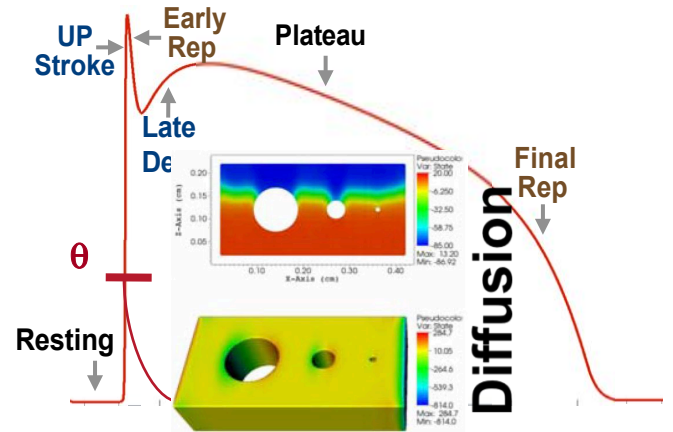
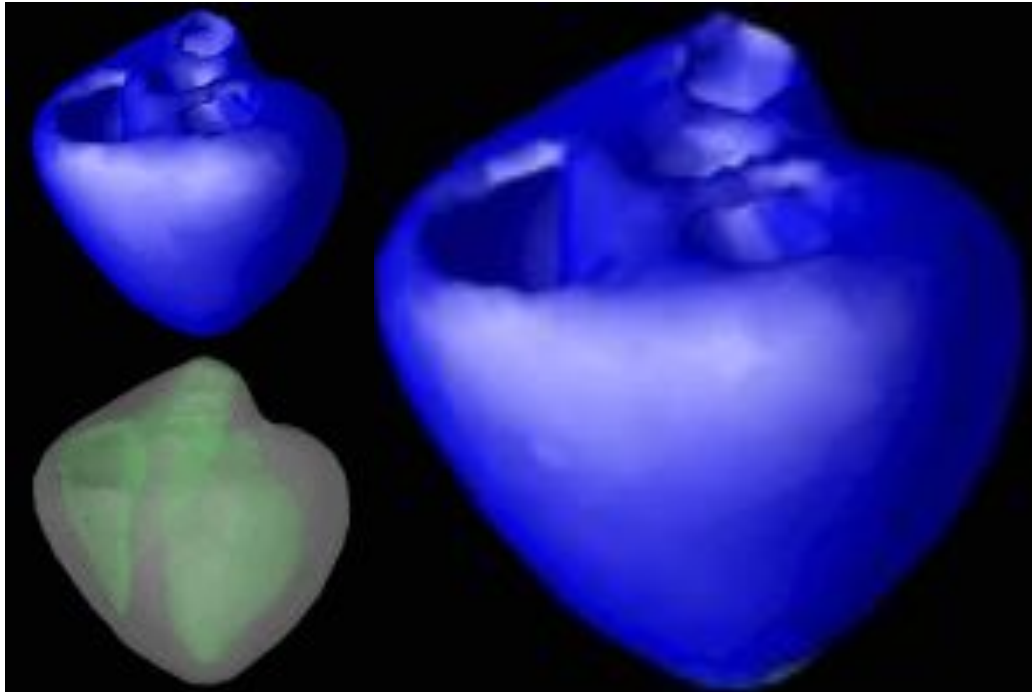
From Cell to EKG and Tissue



$$\frac{\partial V}{\partial t} = \nabla(D\nabla V) - I_{react}$$

From Cell to Whole Heart

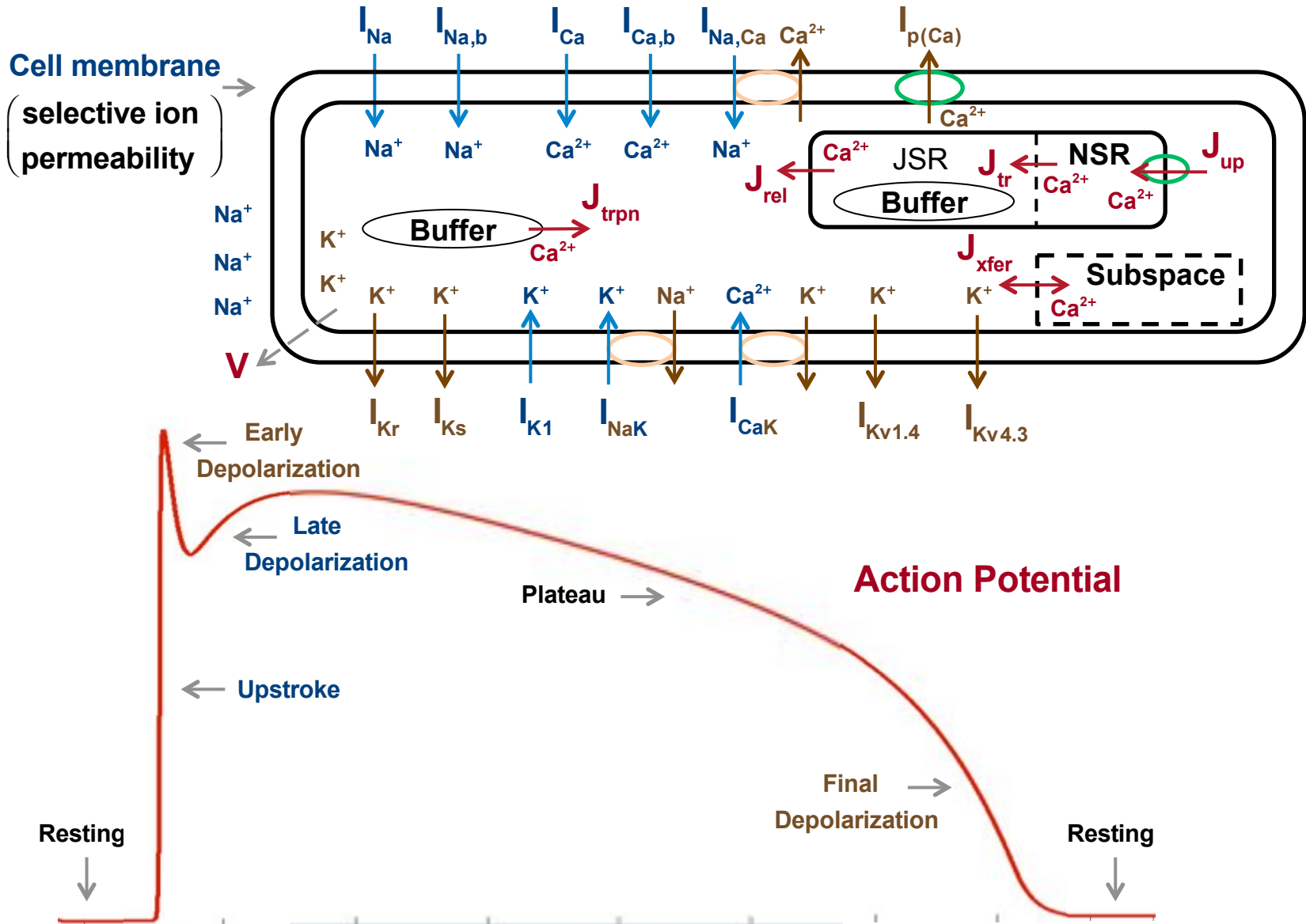
3D Model of a Pig Heart (FK 3V Model)



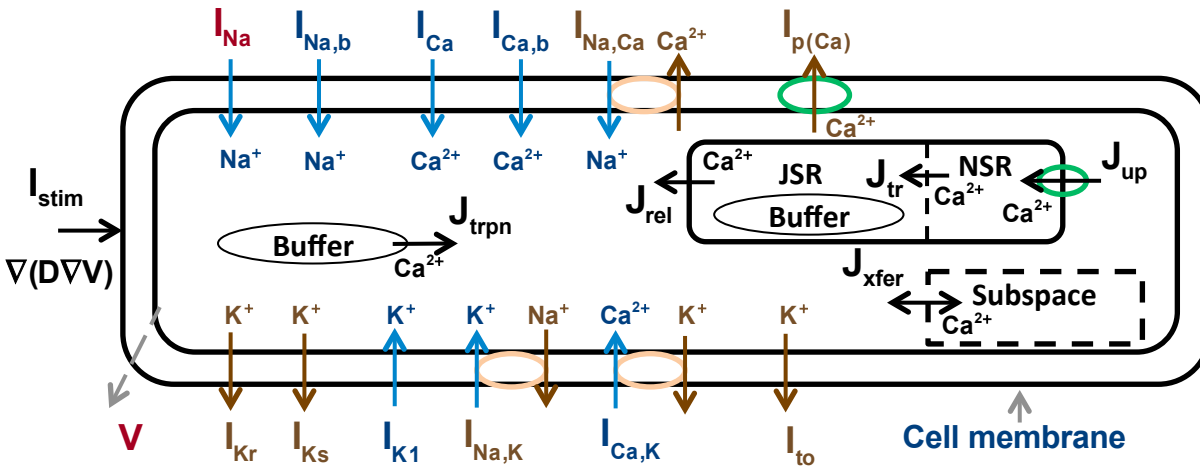
Topological approximations:

- 0D, 1D: EKG (no space), cable propagation
- 2D, 3D: tissue/organ, (an)isotropic, (no)veins, etc.

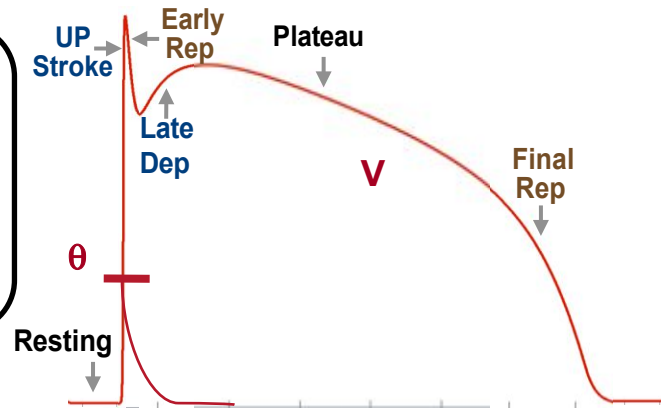
The Cell: Action-Potential



Cellular Approximation Challenge



The 67-variable IMW myocyte model



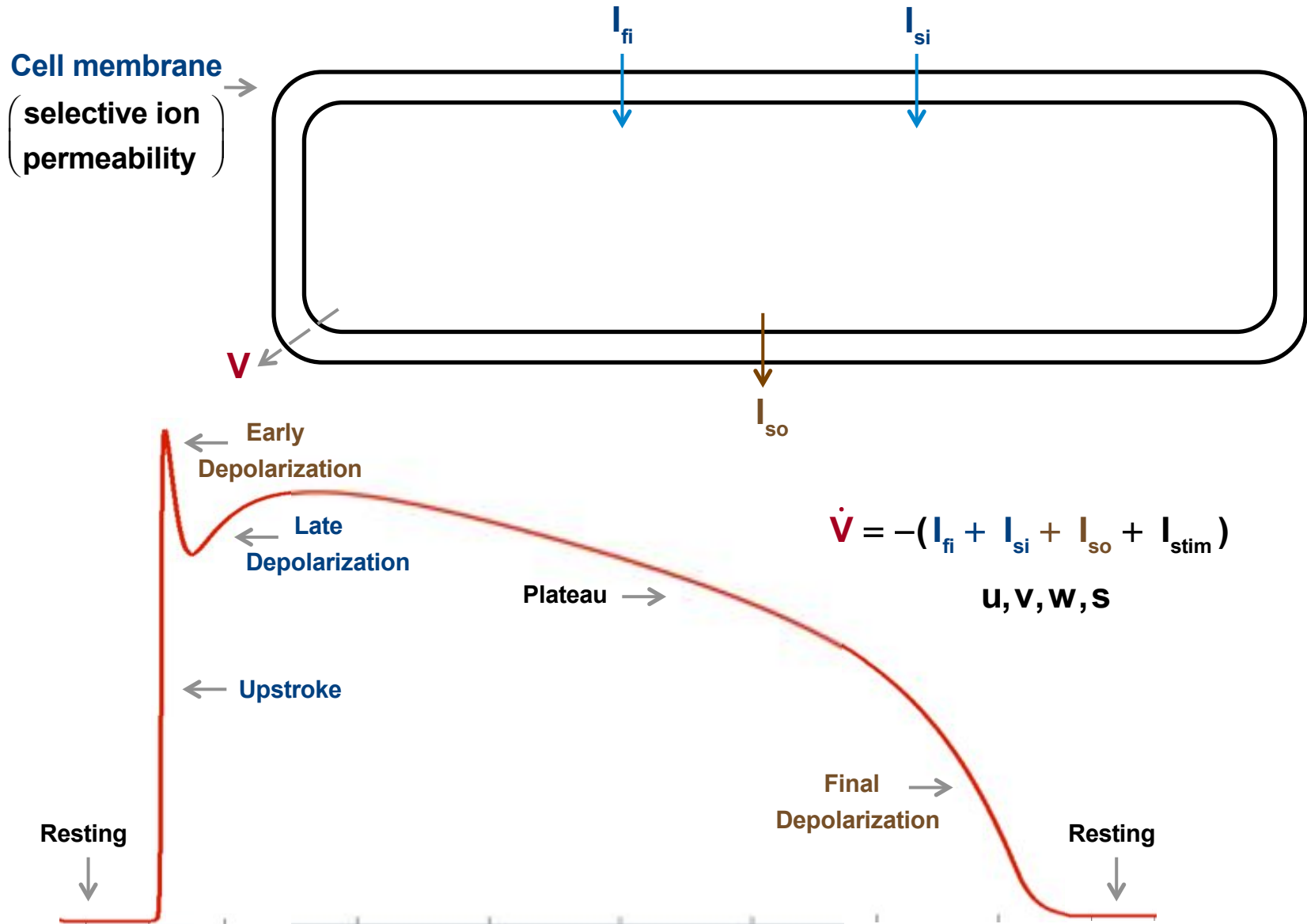
Myocyte Action Potential

$$\frac{\partial V}{\partial t} = -(\nabla(D\nabla V) + I_{Na} + I_{Ca} + I_{Ca,K} + I_{Kr} + I_{Ks} + I_{K1} + I_{Na,Ca} + I_{Na,K} + I_{to} + I_{p(Ca)} + I_{Cab} + I_{Nab})$$

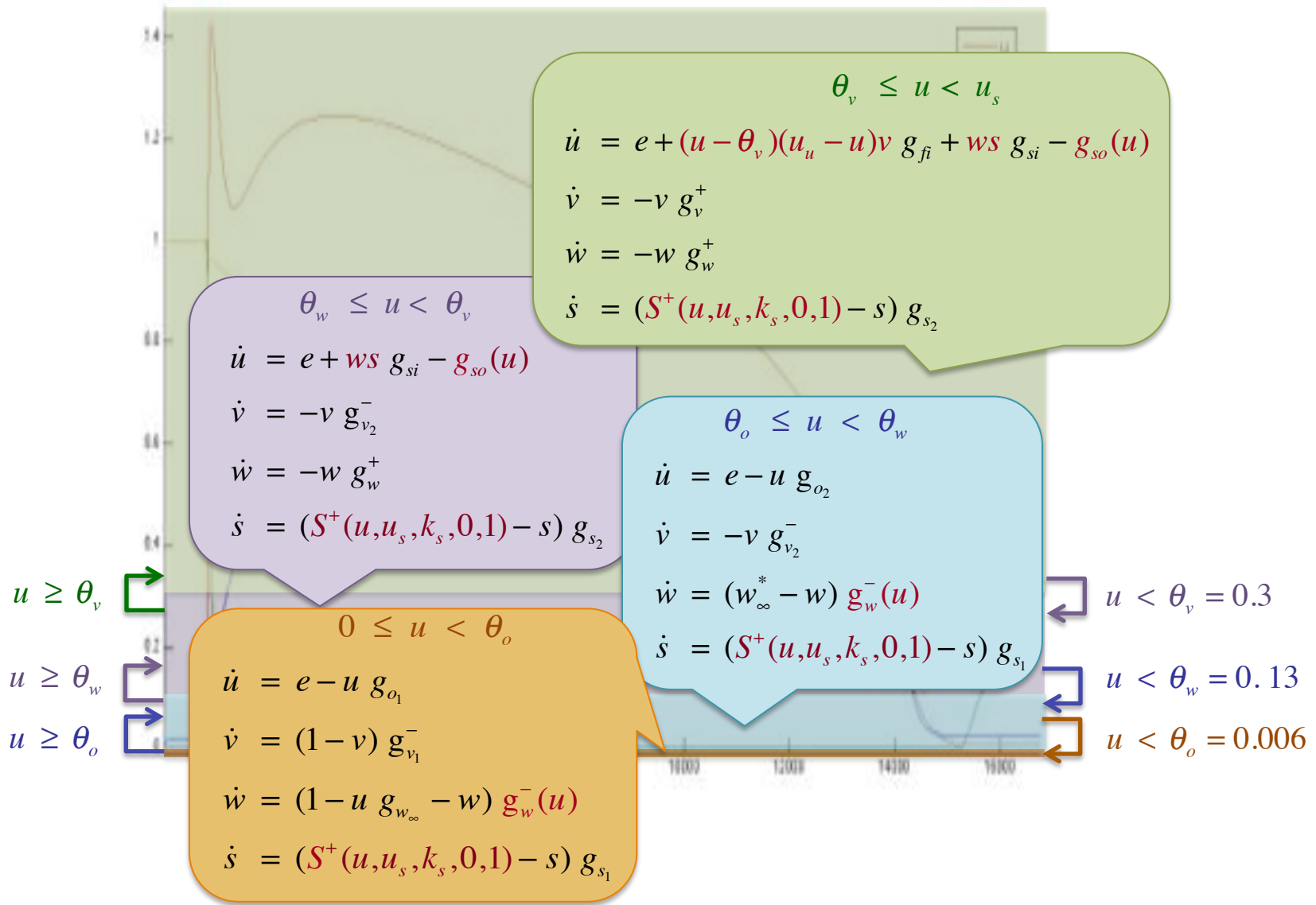
Assumption: V is the only variable of interest

Can we construct simpler models?

Cellular-Approx: Nonlinear HA-Model



Cellular-Approx: Nonlinear HA-Model



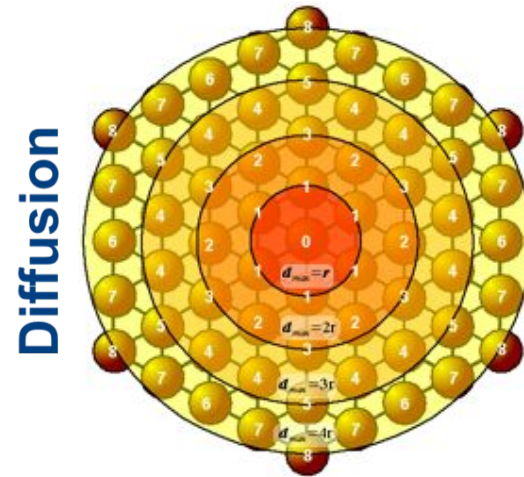
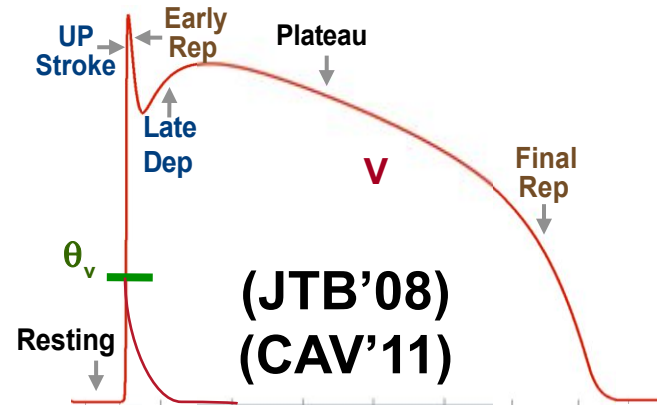
Cellular-Approx: Nonlinear HA-Model

$$\begin{aligned}
 & (\theta_v \leq u < u_u) \\
 \dot{u} &= e + (u - \theta_v)(u_u - u)v g_{fi} + ws g_{si} - g_{so}(u) \\
 \dot{v} &= -v g_v^+ \\
 \dot{w} &= -w g_w^+ \\
 \dot{s} &= S^+(u, k_s, u_s, 0, 1) g_{s_2} - s g_{s_2}
 \end{aligned} \quad (4)$$

$$\begin{aligned}
 & \theta_w \leq u < \theta_v \\
 \dot{u} &= e + ws g_{si} - g_{so}(u) \\
 \dot{v} &= -v g_{v_2}^- \\
 \dot{w} &= -w g_w^+ \\
 \dot{s} &= S^+(u, k_s, u_s, 0, 1) g_{s_2} - s g_{s_2}
 \end{aligned} \quad (3)$$

$$\begin{aligned}
 & \theta_o \leq u < \theta_w \\
 \dot{u} &= e - u g_{o_2} \\
 \dot{v} &= -v g_{v_2}^- \\
 \dot{w} &= (w_\infty^* - w) g_w^-(u) \\
 \dot{s} &= S^+(u, k_s, u_s, 0, 1) g_{s_1} - s g_{s_1}
 \end{aligned} \quad (2)$$

$$\begin{aligned}
 & 0 \leq u < \theta_o \\
 \dot{u} &= e - u g_{o_1} \\
 \dot{v} &= (1 - v) g_{v_1}^- \\
 \dot{w} &= (1 - u g_{w_\infty} - w) g_w^-(u) \\
 \dot{s} &= S^+(u, k_s, u_s, 0, 1) g_{s_1} - s g_{s_1}
 \end{aligned} \quad (1)$$



$$\frac{\partial u}{\partial t} = \nabla(D\nabla u) - (I_{fi} + I_{si} + I_{so})$$

PDEs are simulated as Finite Difference Equations

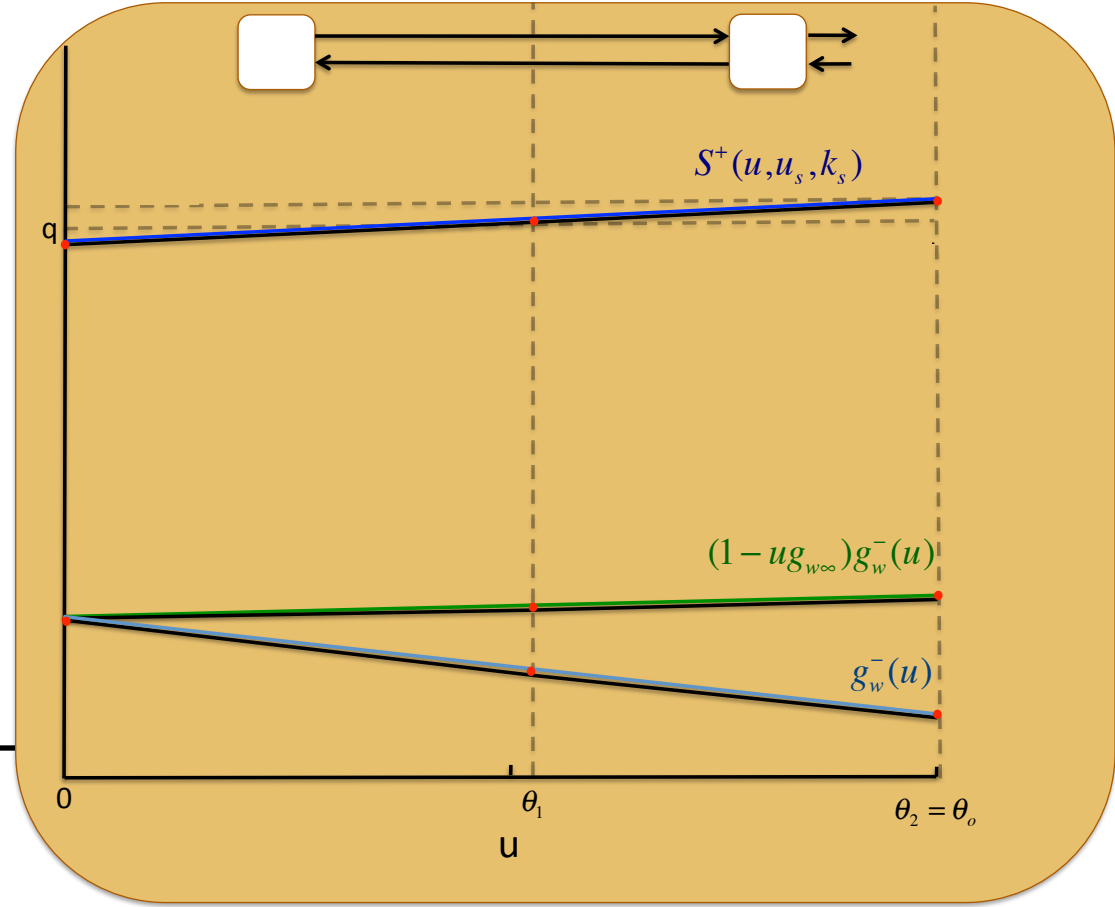
Cellular-Approx: Multi-Affine HA-Model

$$\begin{aligned}
 & (\theta_v \leq u < u_u) \\
 \dot{u} &= e + (u - \theta_v)(u_u - u)v g_{fi} + ws g_{si} - g_{so}(u) \\
 \dot{v} &= -v g_v^+ \\
 \dot{w} &= -w g_w^+ \\
 \dot{s} &= S^+(u, k_s, u_s, 0, 1) g_{s_2} - s g_{s_2}
 \end{aligned}$$

$$\begin{aligned}
 & \theta_w \leq u < \theta_v \\
 \dot{u} &= e + ws g_{si} - g_{so}(u) \\
 \dot{v} &= -v g_{v_2}^- \\
 \dot{w} &= -w g_w^+ \\
 \dot{s} &= S^+(u, k_s, u_s, 0, 1) g_{s_2} - s g_{s_2}
 \end{aligned}$$

$$\begin{aligned}
 & \theta_o \leq u < \theta_w \\
 \dot{u} &= e - u g_{o_2} \\
 \dot{v} &= -v g_{v_2}^- \\
 \dot{w} &= (w_\infty^* - w) g_w^-(u) \\
 \dot{s} &= S^+(u, k_s, u_s) g_{s_1} - s g_{s_1}
 \end{aligned}$$

$$\begin{aligned}
 & 0 \leq u < \theta_o \\
 \dot{u} &= e - u g_{o_1} \\
 \dot{v} &= (1 - v) g_{v_1}^- \\
 \dot{w} &= (1 - u g_{w_\infty} - w) g_w^-(u) \\
 \dot{s} &= S^+(u, k_s, u_s) g_{s_1} - s g_{s_1}
 \end{aligned}$$



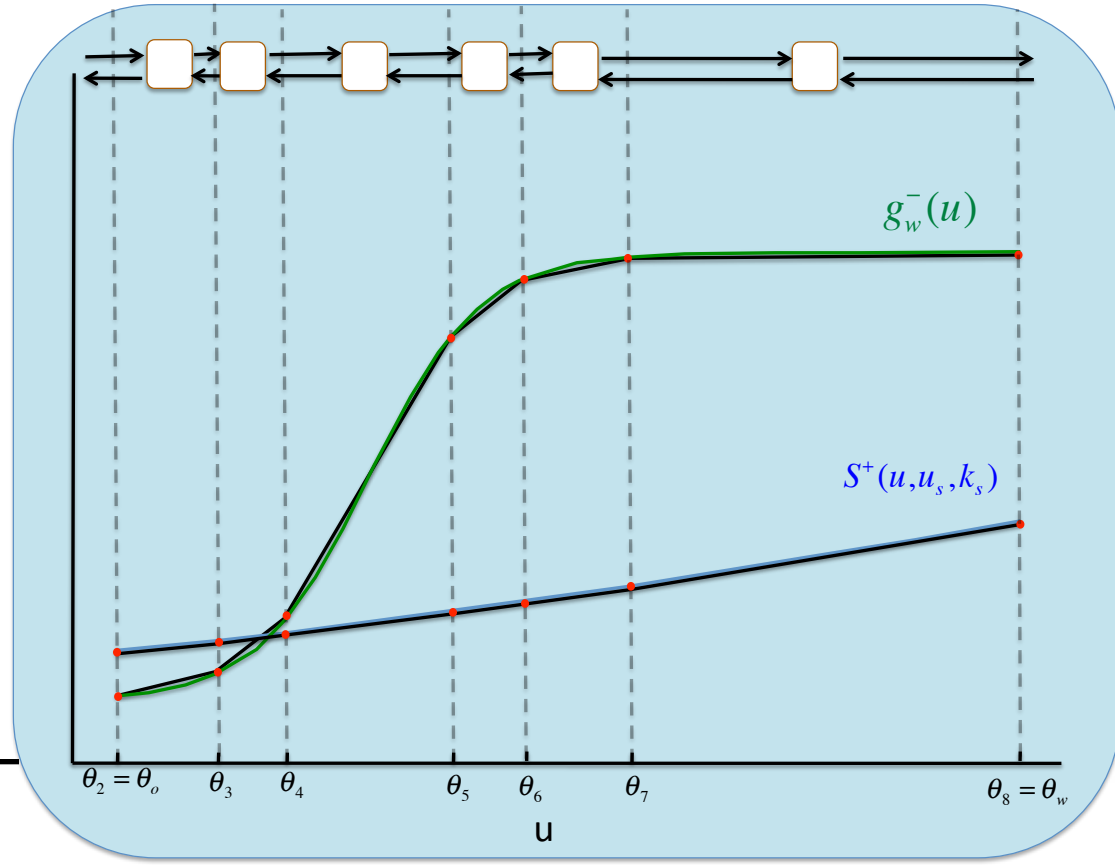
Cellular-Approx: Multi-Affine HA-Model

$$\begin{aligned}
 & (\theta_v \leq u < u_u) \\
 \dot{u} &= e + (u - \theta_v)(u_u - u)v g_{fi} + ws g_{si} - g_{so}(u) \\
 \dot{v} &= -v g_v^+ \\
 \dot{w} &= -w g_w^+ \\
 \dot{s} &= S^+(u, k_s, u_s, 0, 1) g_{s_2} - s g_{s_2}
 \end{aligned}$$

$$\begin{aligned}
 & \theta_w \leq u < \theta_v \\
 \dot{u} &= e + ws g_{si} - g_{so}(u) \\
 \dot{v} &= -v g_{v_2}^- \\
 \dot{w} &= -w g_w^+ \\
 \dot{s} &= S^+(u, k_s, u_s, 0, 1) g_{s_2} - s g_{s_2}
 \end{aligned}$$

$$\begin{aligned}
 & \theta_o \leq u < \theta_w \\
 \dot{u} &= e - u g_{o_2} \\
 \dot{v} &= -v g_{v_2}^- \\
 \dot{w} &= (w_\infty^* - w) g_w^-(u) \\
 \dot{s} &= S^+(u, k_s, u_s) g_{s_1} - s g_{s_1}
 \end{aligned}$$

$$\begin{aligned}
 & 0 \leq u < \theta_o \\
 \dot{u} &= e - u g_{o_1} \\
 \dot{v} &= (1 - v) g_{v_1}^- \\
 \dot{w} &= (1 - u g_{w_\infty} - w) g_w^-(u) \\
 \dot{s} &= S^+(u, k_s, u_s) g_{s_1} - s g_{s_1}
 \end{aligned}$$



Cellular-Approx: Multi-Affine HA-Model

$(\theta_v \leq u < u_u)$

$$\begin{aligned} \dot{u} &= e + (u - \theta_v)(u_u - u)v g_{fi} + ws g_{si} - g_{so}(u) \\ \dot{v} &= -v g_v^+ \\ \dot{w} &= -w g_w^+ \\ \dot{s} &= S^+(u, k_s, u_s) g_{s_2} - s g_{s_2} \end{aligned}$$

$\theta_w \leq u < \theta_v$

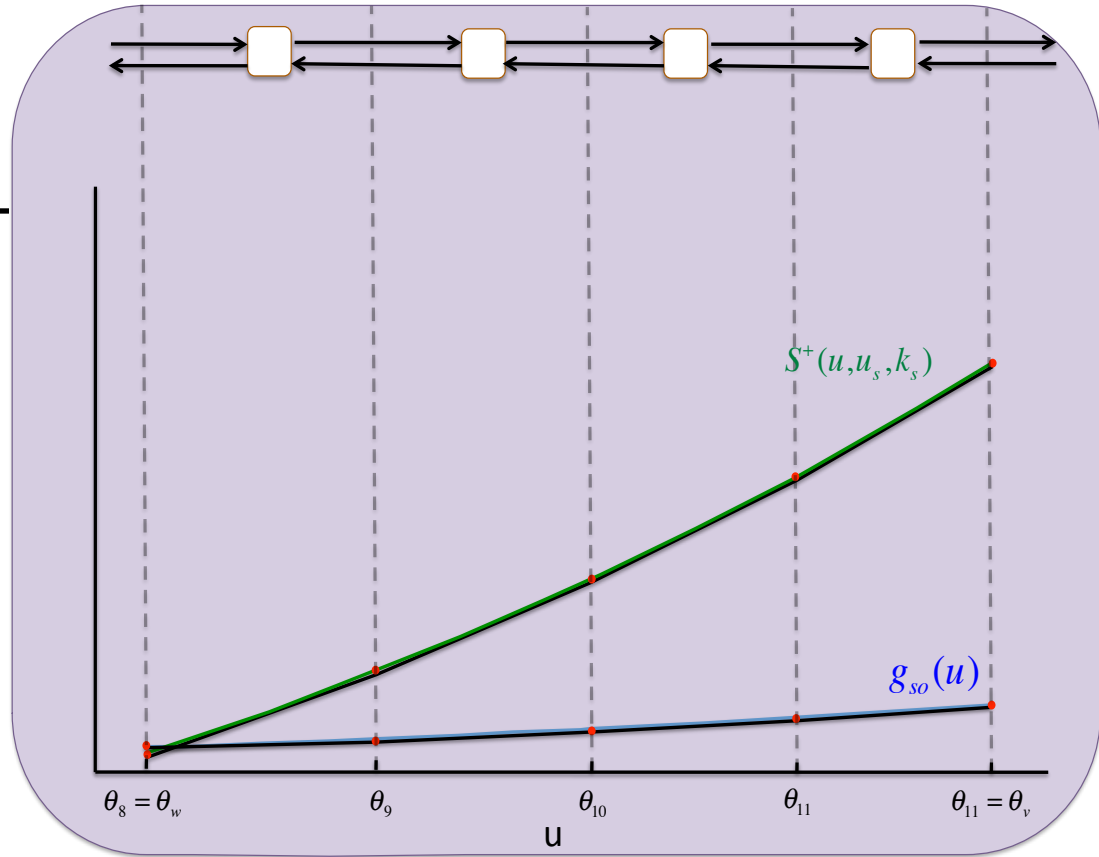
$$\begin{aligned} \dot{u} &= e + ws g_{si} - g_{so}(u) \\ \dot{v} &= -v g_{v_2}^- \\ \dot{w} &= -w g_w^+ \\ \dot{s} &= S^+(u, k_s, u_s) g_{s_2} - s g_{s_2} \end{aligned}$$

$\theta_o \leq u < \theta_w$

$$\begin{aligned} \dot{u} &= e - u g_{o_2} \\ \dot{v} &= -v g_{v_2}^- \\ \dot{w} &= (w_\infty^* - w) g_w^-(u) \\ \dot{s} &= S^+(u, k_s, u_s) g_{s_1} - s g_{s_1} \end{aligned}$$

$0 \leq u < \theta_o$

$$\begin{aligned} \dot{u} &= e - u g_{o_1} \\ \dot{v} &= (1 - v) g_{v_1}^- \\ \dot{w} &= (1 - u g_{w_\infty} - w) g_w^-(u) \\ \dot{s} &= S^+(u, k_s, u_s) g_{s_1} - s g_{s_1} \end{aligned}$$



Cellular-Approx: Multi-Affine HA-Model

$(\theta_v \leq u < u_u)$

$$\dot{u} = e + (u - \theta_v)(u_u - u)v g_{fi} + ws g_{si} - g_{so}(u)$$

$$\dot{v} = -v g_v^+$$

$$\dot{w} = -w g_w^+$$

$$\dot{s} = S^+(u, k_s, u_s) g_{s_2} - s g_{s_2}$$

$\theta_w \leq u < \theta_v$

$$\dot{u} = e + ws g_{si} - g_{so}(u)$$

$$\dot{v} = -v g_{v_2}^-$$

$$\dot{w} = -w g_w^+$$

$$\dot{s} = S^+(u, k_s, u_s) g_{s_2} - s g_{s_2}$$

$\theta_o \leq u < \theta_w$

$$\dot{u} = e - u g_{o_2}$$

$$\dot{v} = -v g_{v_2}^-$$

$$\dot{w} = (w_\infty^* - w) g_w^-(u)$$

$$\dot{s} = S^+(u, k_s, u_s) g_{s_1} - s g_{s_1}$$

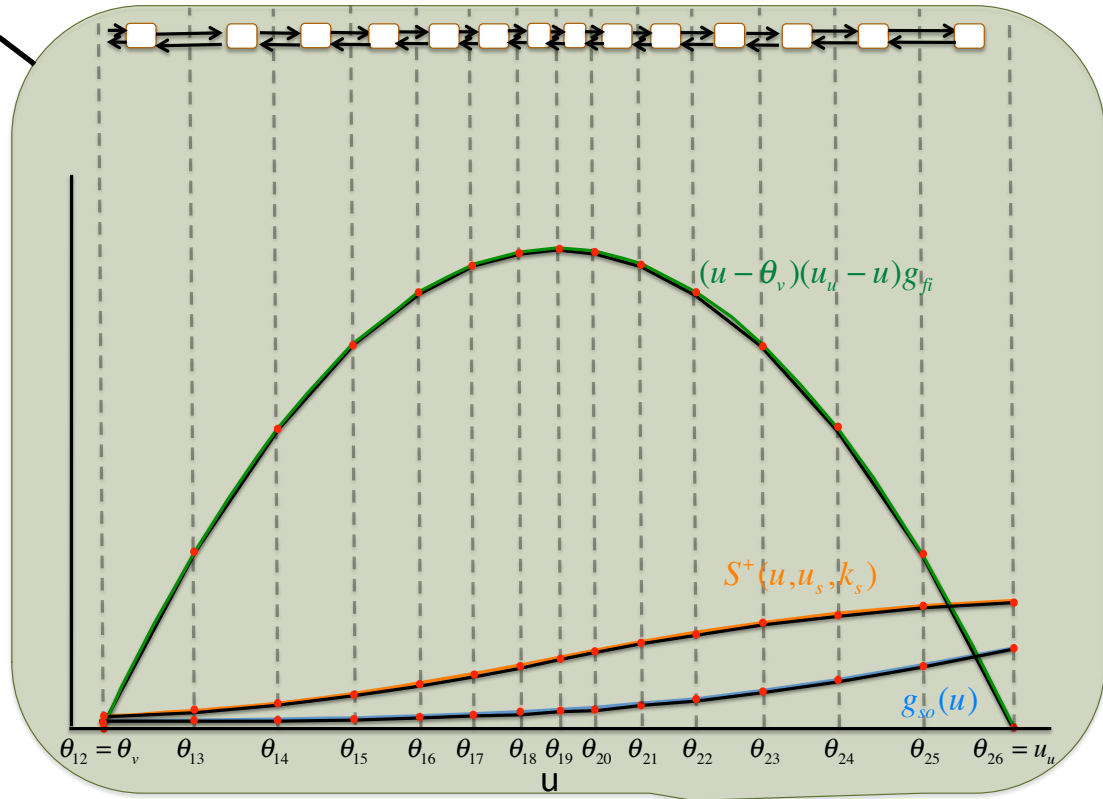
$0 \leq u < \theta_o$

$$\dot{u} = e - u g_{o_1}$$

$$\dot{v} = (1 - v) g_{v_1}^-$$

$$\dot{w} = (1 - u g_{w_\infty} - w) g_w^-(u)$$

$$\dot{s} = S^+(u, k_s, u_s) g_{s_1} - s g_{s_1}$$



Cellular-Approx: Multi-Affine HA-Model

(CAV'11)

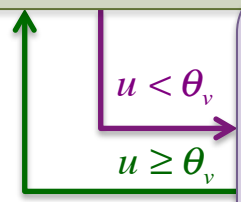
$$\theta_{12} = \theta_v < u \leq u_u = \theta_{26}$$

$$\dot{u} = e + \sum_{i=12}^{25} R(u, \theta_i, \theta_{i+1}, u_{f_i}, u_{f_{i+1}}) v g_{f_i} + ws g_{s_i} - \sum_{i=12}^{25} R(u, \theta_i, \theta_{i+1}, u_{s_{o_i}}, u_{s_{o_{i+1}}}) g_{s_{o_i}}$$

$$\dot{v} = -v g_v^+$$

$$\dot{w} = -w g_w^+$$

$$\dot{s} = \left(\sum_{i=12}^{25} R(u, \theta_i, \theta_{i+1}, u_{s_i}, u_{s_{i+1}}) - s \right) g_{s_2}$$



$$\theta_8 = \theta_w \leq u < \theta_v = \theta_{12}$$

$$\dot{u} = e + ws g_{s_i} - \sum_{i=8}^{11} R(u, \theta_i, \theta_{i+1}, u_{s_{o_i}}, u_{s_{o_{i+1}}}) g_{s_{o_i}}$$

$$\dot{v} = -v g_{v_2}^-$$

$$\dot{w} = -w g_w^+$$

$$\dot{s} = \left(\sum_{i=8}^{11} R(u, \theta_i, \theta_{i+1}, u_{s_i}, u_{s_{i+1}}) - s \right) g_{s_2}$$

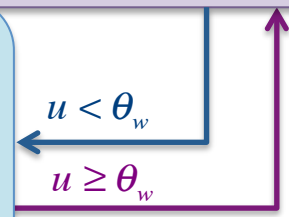
$$\theta_2 = \theta_o \leq u < \theta_w = \theta_8$$

$$\dot{u} = e - u g_{o_2}$$

$$\dot{v} = -v g_{v_2}^-$$

$$\dot{w} = (w_\infty^* - w) \sum_{i=2}^7 R(u, \theta_i, \theta_{i+1}, u_{w_i}, u_{w_{i+1}}) g_{w_b}$$

$$\dot{s} = \left(\sum_{i=2}^7 R(u, \theta_i, \theta_{i+1}, u_{s_i}, u_{s_{i+1}}) - s \right) g_{s_1}$$



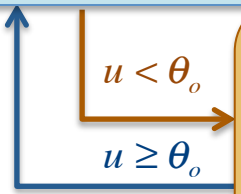
$$\theta_0 = 0 \leq u < \theta_o = \theta_2$$

$$\dot{u} = e - u g_{o_1}$$

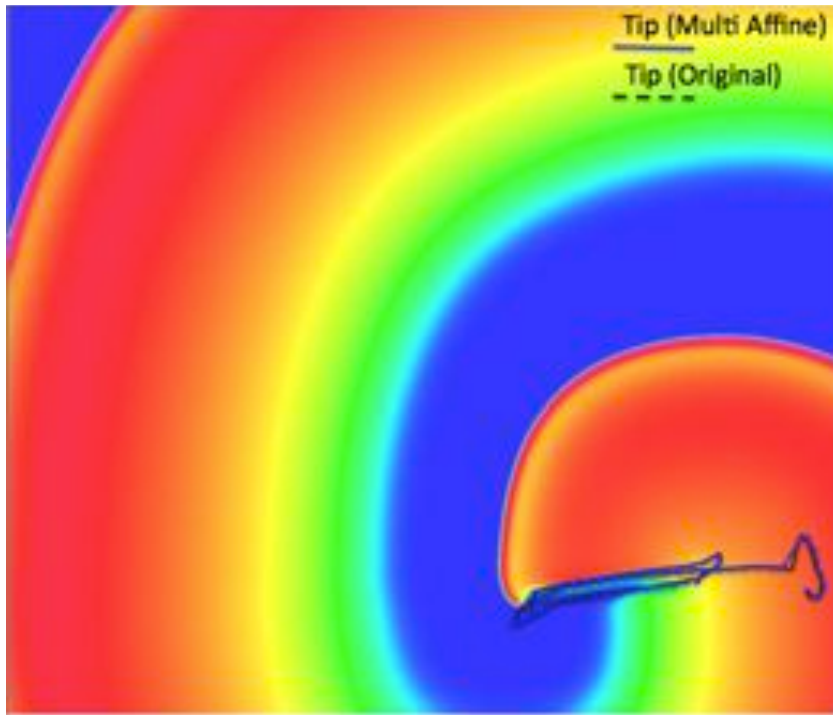
$$\dot{v} = (1 - v) g_{v_1}^-$$

$$\dot{w} = \left(\sum_{i=0}^1 (R(u, \theta_i, \theta_{i+1}, u_{w_i}^+, u_{w_{i+1}}^+) - w R(u, \theta_i, \theta_{i+1}, u_{w_i}^-, u_{w_{i+1}}^-)) \right) g_{w_a}$$

$$\dot{s} = \left(\sum_{i=0}^1 R(u, \theta_i, \theta_{i+1}, u_{s_i}, u_{s_{i+1}}) - s \right) g_{s_1}$$

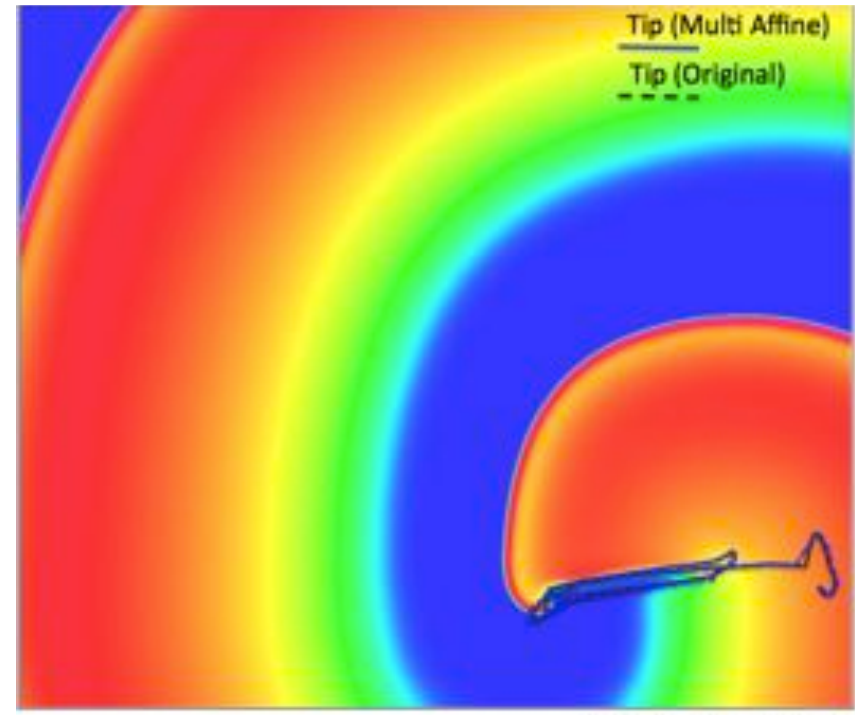


Nonlinear vs Multi-Affine HA Model



(a)

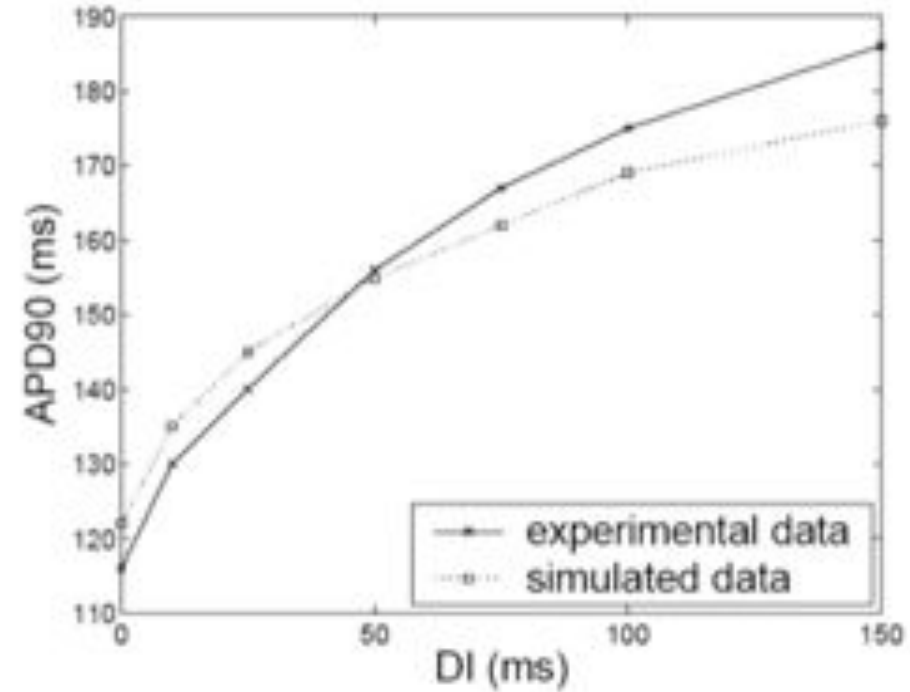
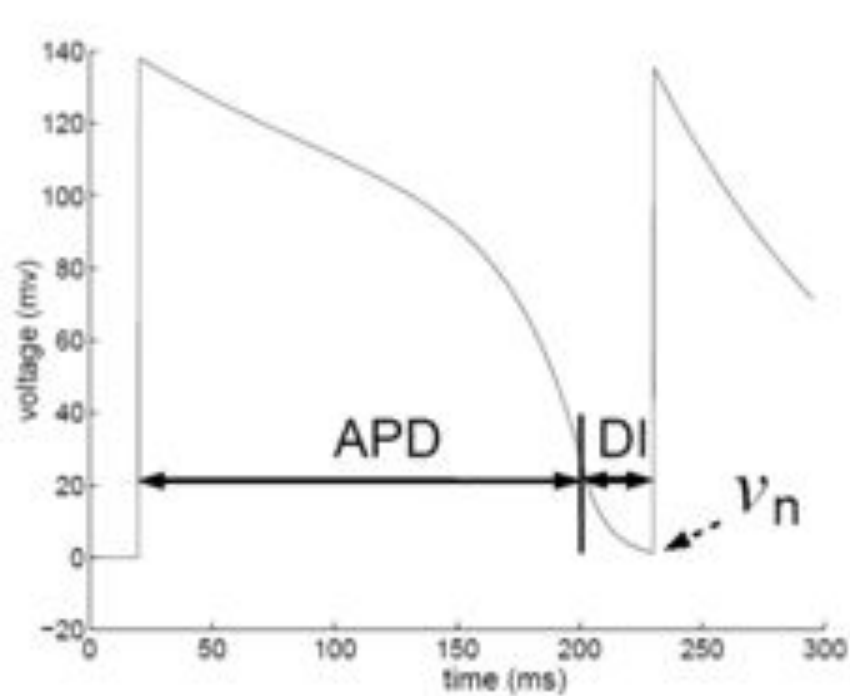
Original MRM



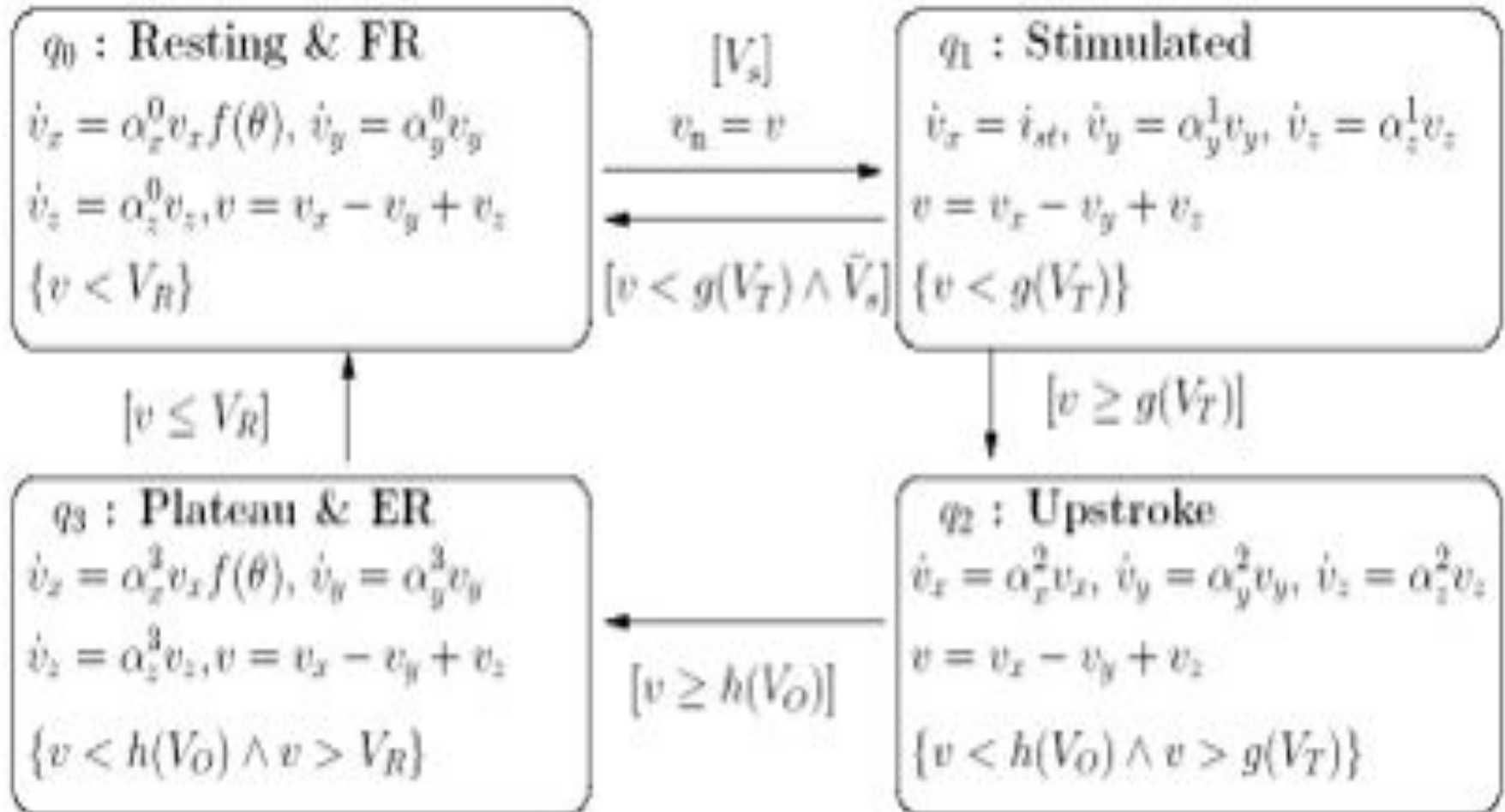
(b)

Multi Affine (26 ramps)

Cellular-Approx: Restitution

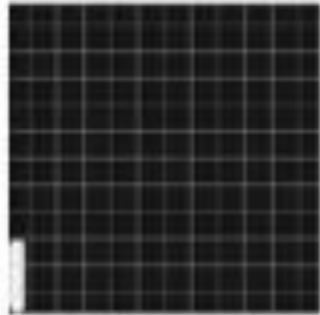


Cellular-Approx: Cycle-Linear HA-Model



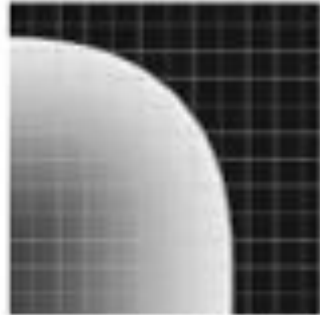
(CMSB'05)

Cellular-Approx: Cycle-Linear HA-Model

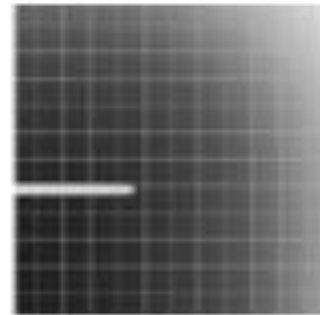


time=0s

1st stimulus occurs

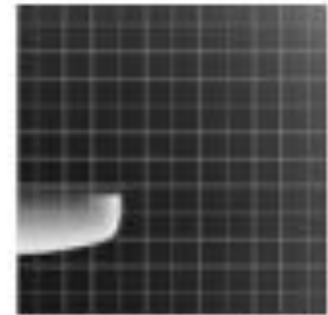


time=0.07s

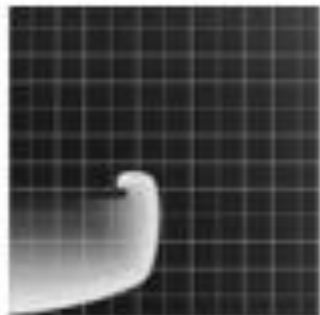


time=0.145s

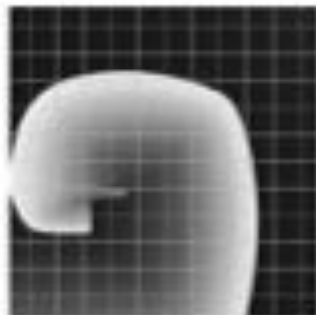
2nd stimulus occurs



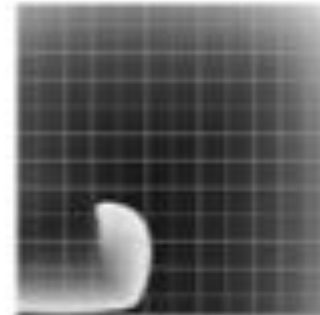
time=0.18s



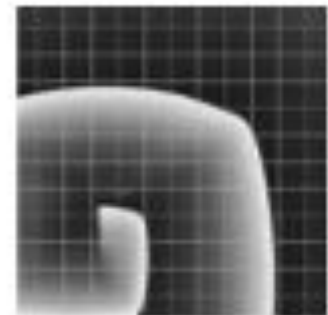
time=0.21s



time=0.27s



time=0.34s

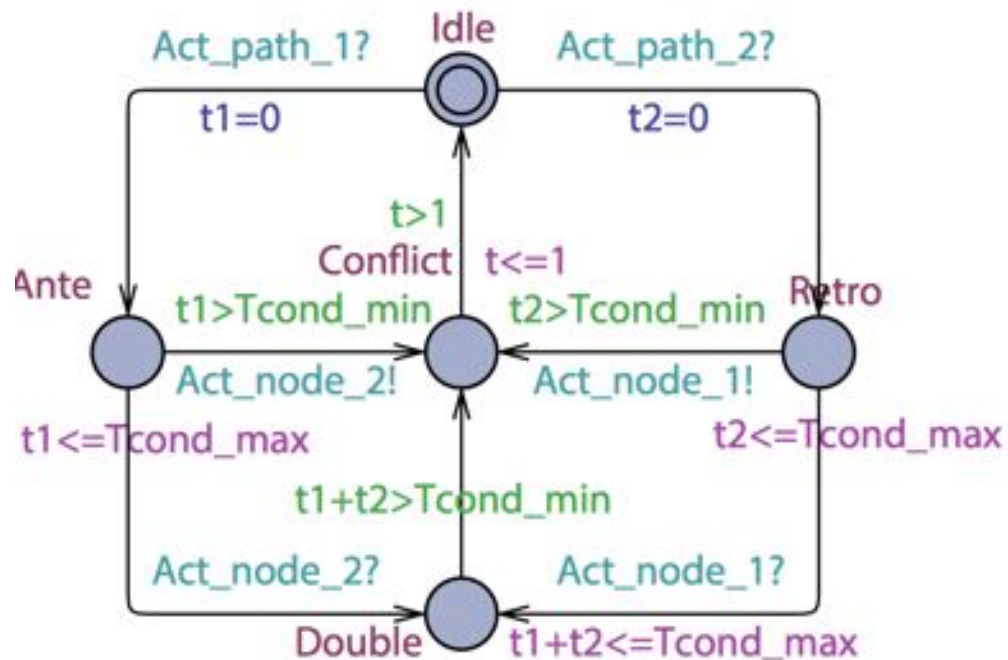
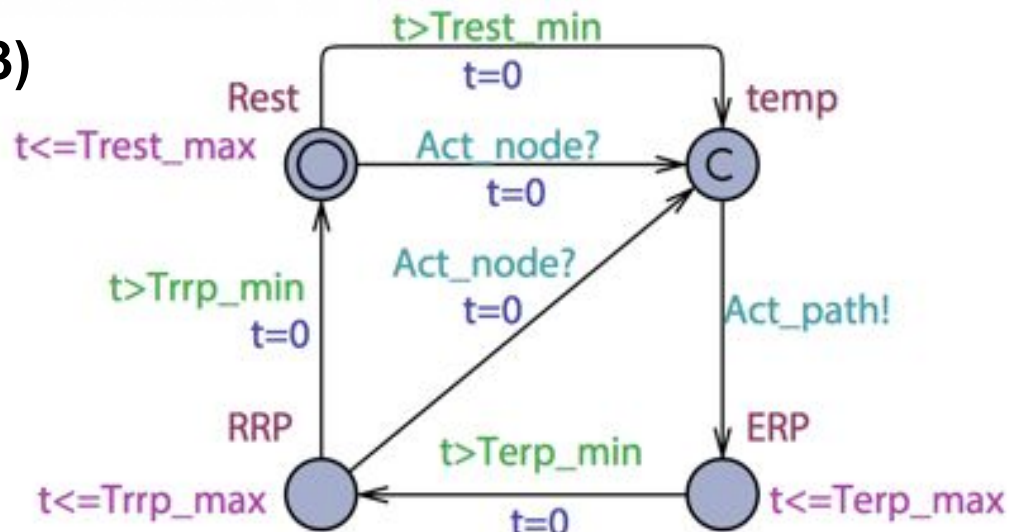
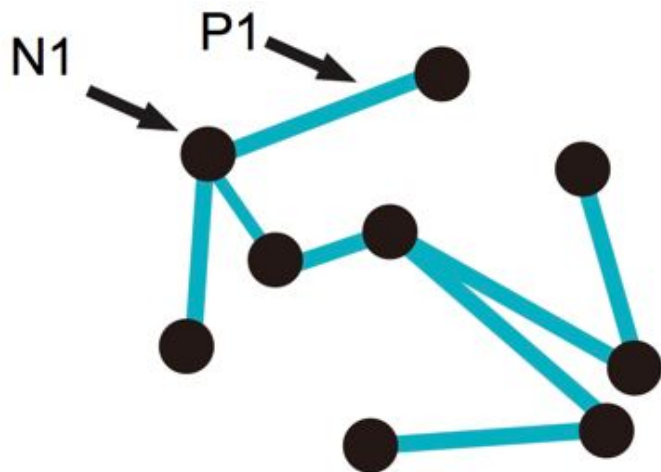
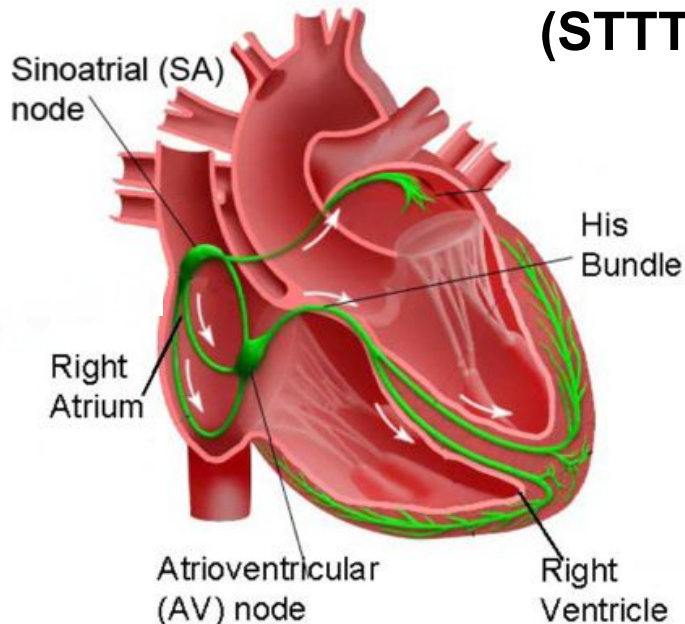


time=0.55s

(CMSB'05)

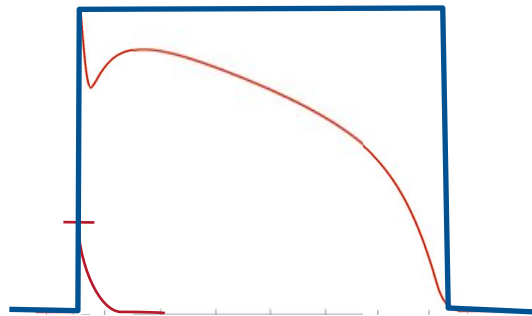
Cellular-Approx: Timed-Automata

(STTT'13)

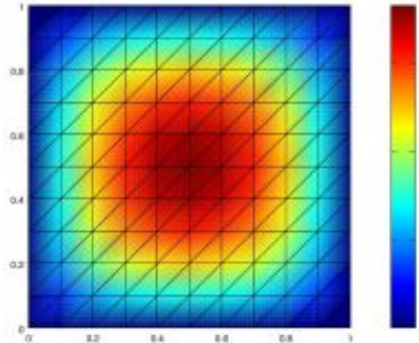


Cellular Approx: Cellular-Automata

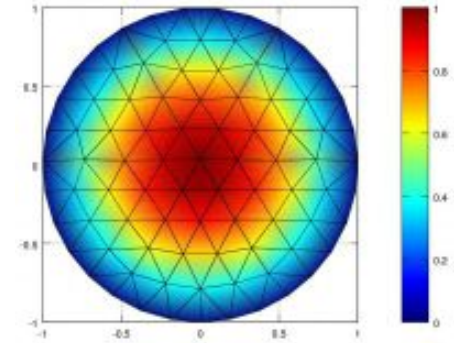
– Waves take the form of lattice! Curvature, velocity,...



Action Potential

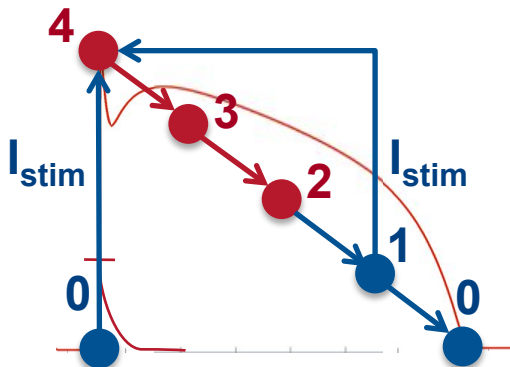


Square Wave

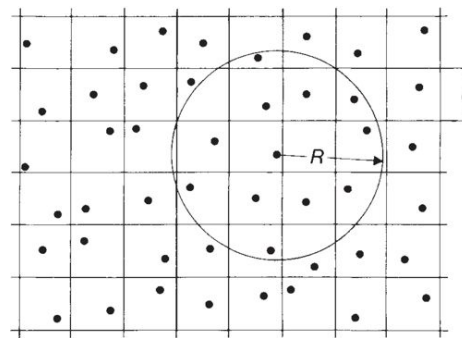


Hexagonal Wave

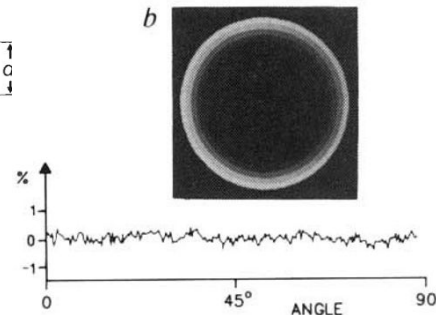
– Recent approach: Preserves most wave properties



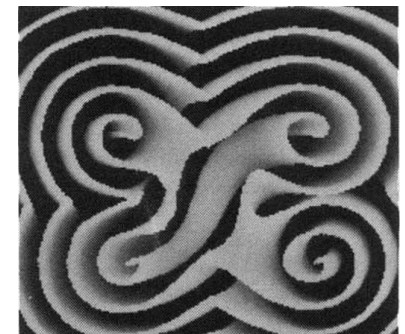
Action Potential



Random Grid

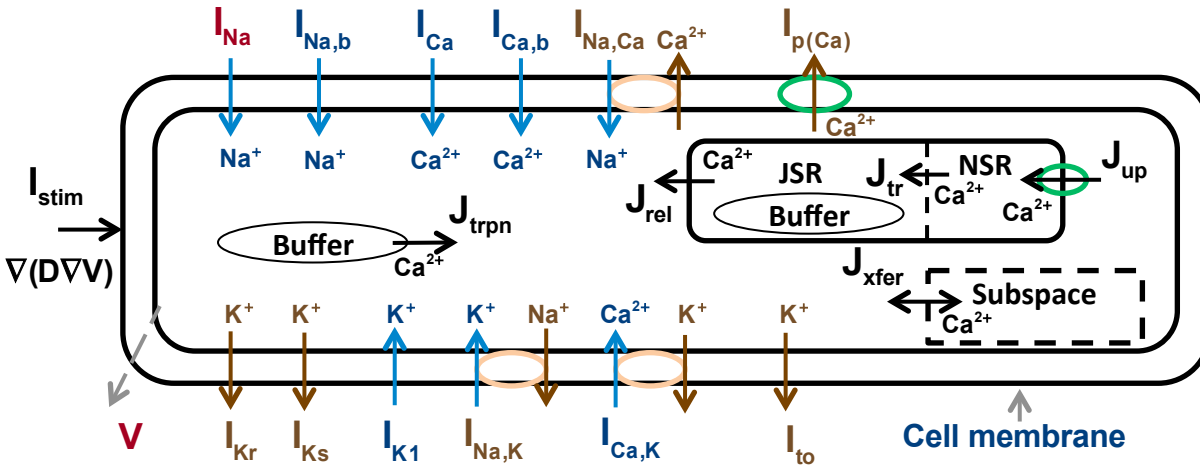


Isotropy

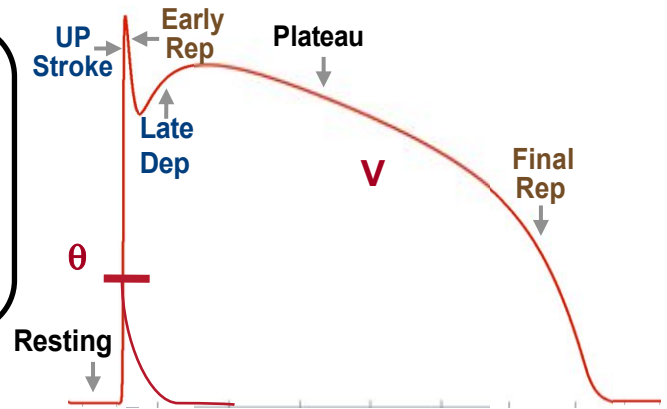


3D Wave

Cellular Approximation Challenge



The 67-variable IMW myocyte model



Myocyte Action Potential

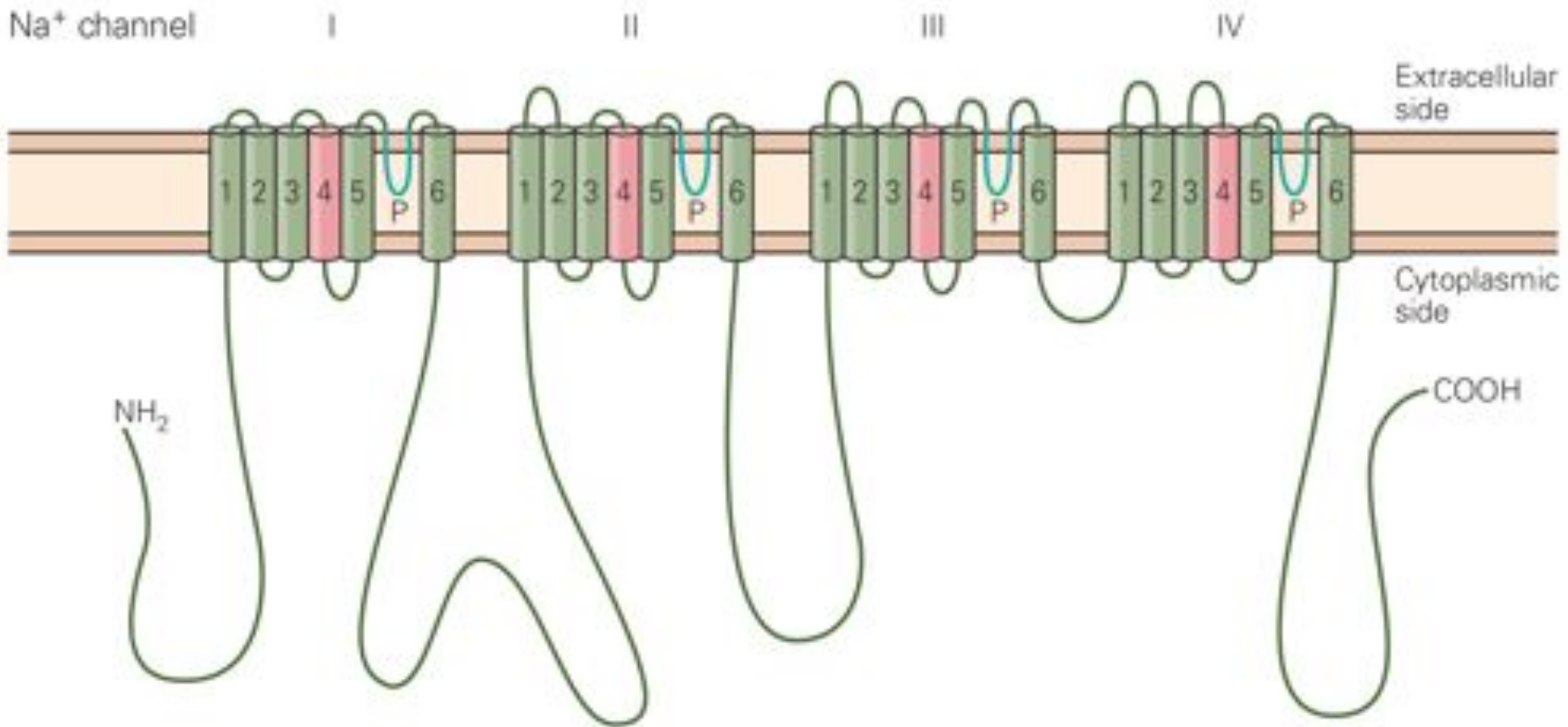
$$\frac{\partial V}{\partial t} = -(\nabla(D\nabla V) + I_{Na} + I_{Ca} + I_{Ca,K} + I_{Kr} + I_{Ks} + I_{K1} + I_{Na,Ca} + I_{Na,K} + I_{to} + I_{p(Ca)} + I_{Cab} + I_{Nab})$$

Assumption: V is the only variable of interest

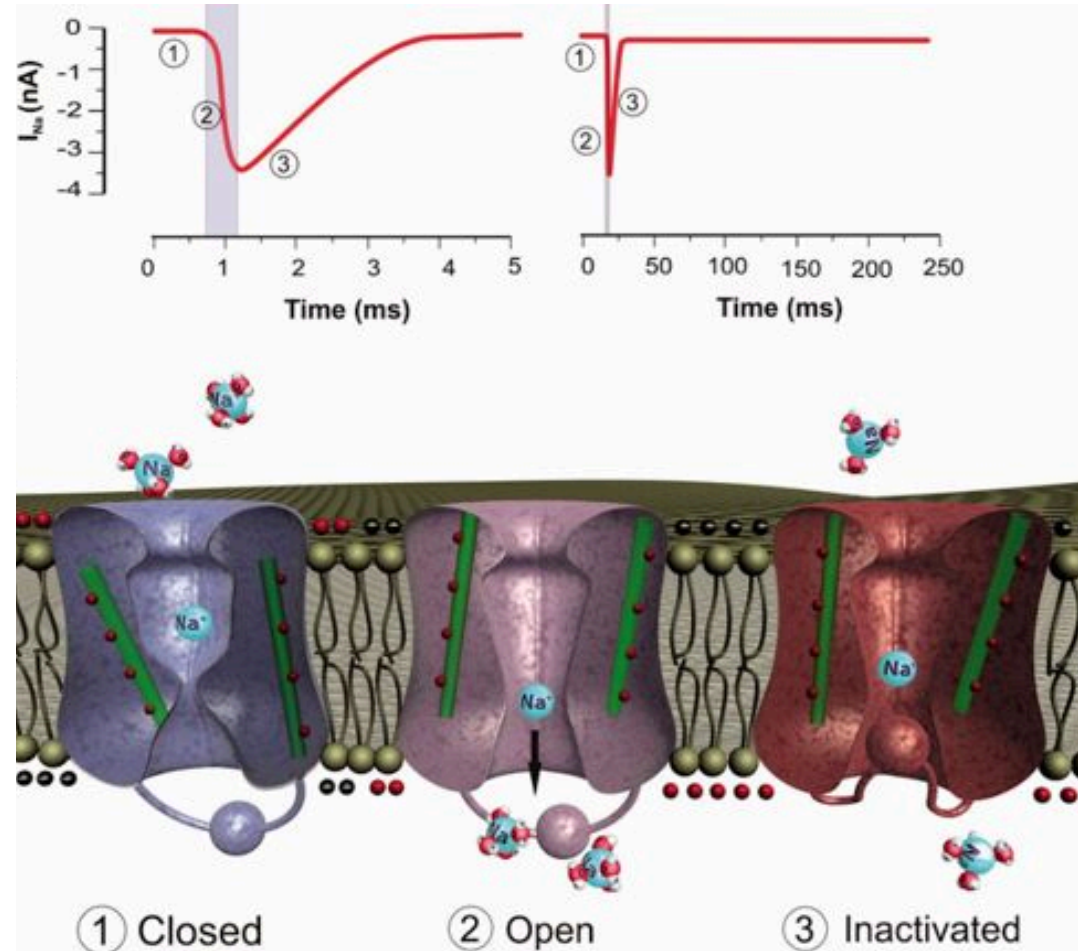
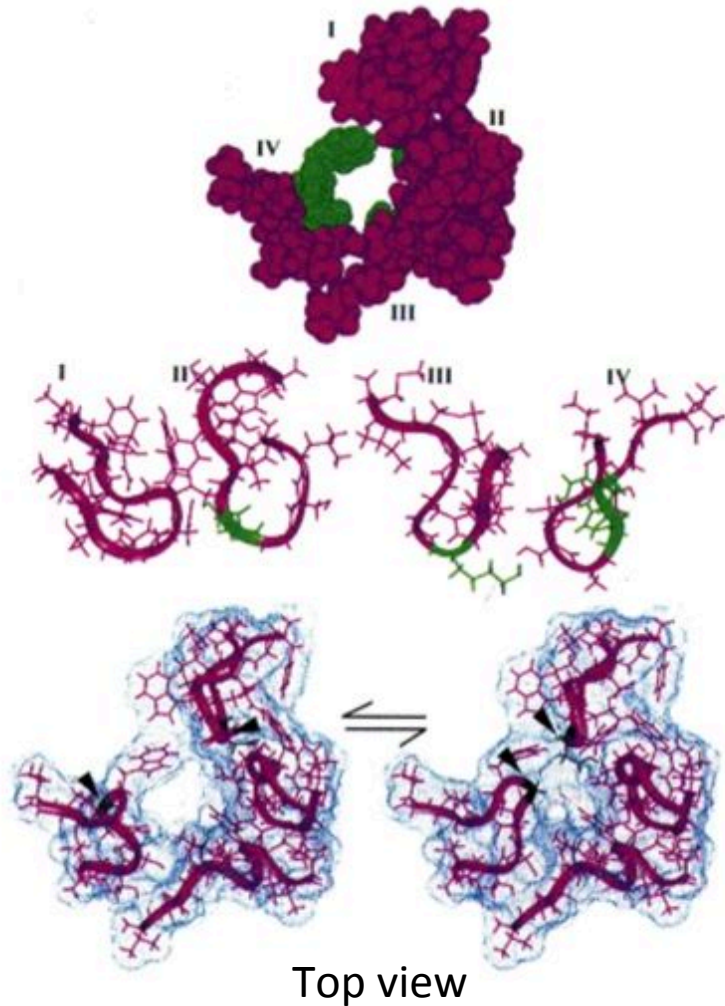
Question: Is there a systematic way to approximate the model?

Idea: Approximate each channel at a time

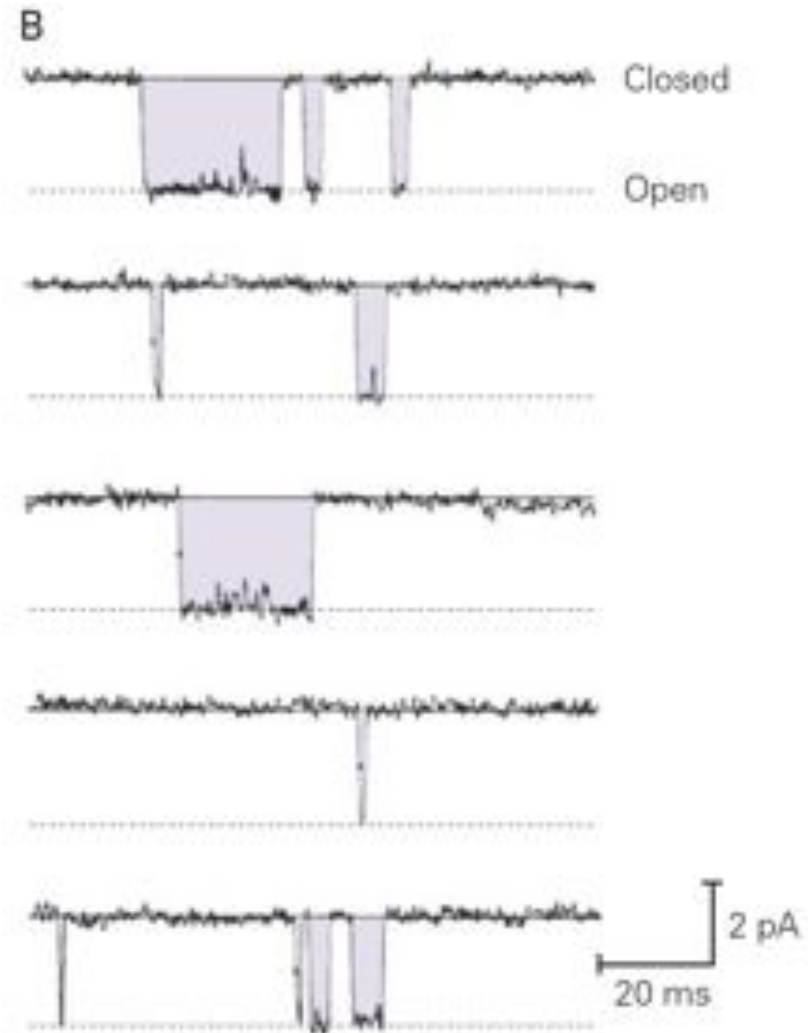
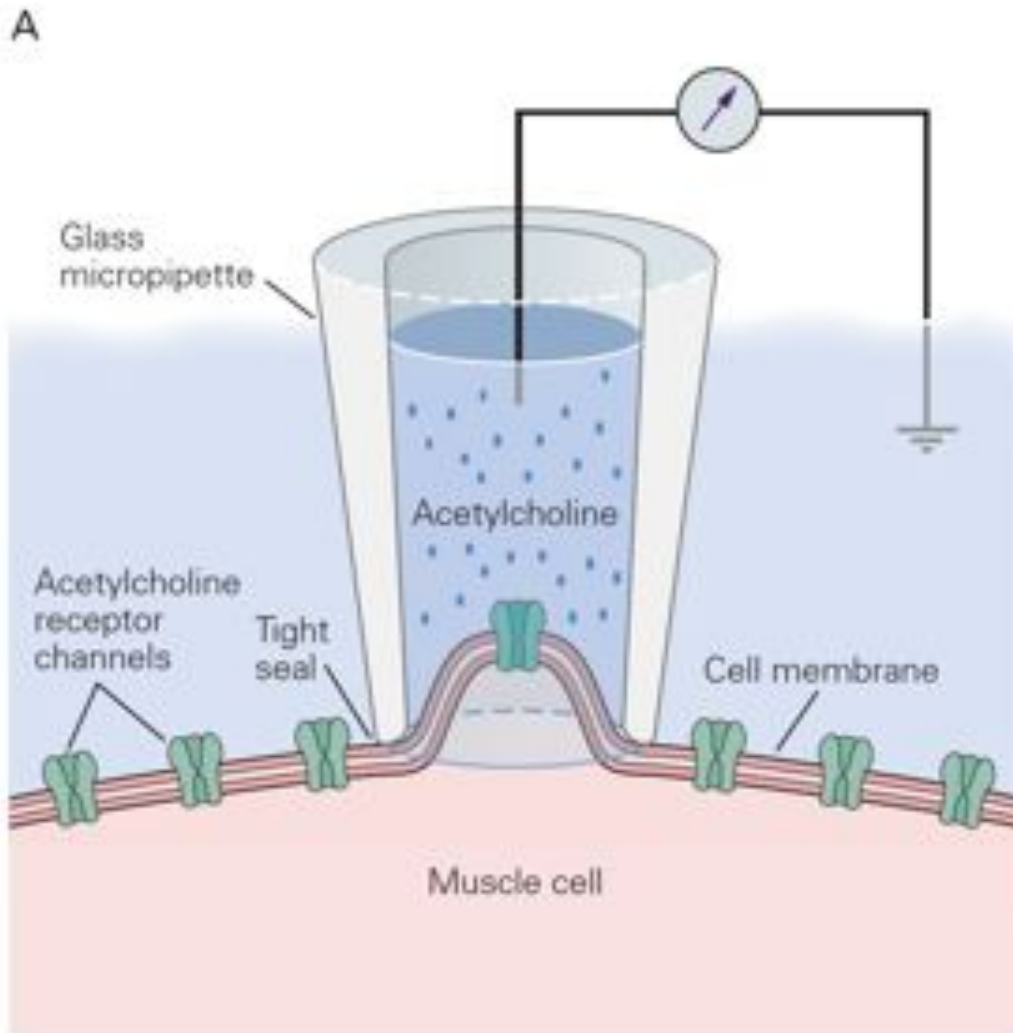
Channel Approximation: Na⁺ Channel



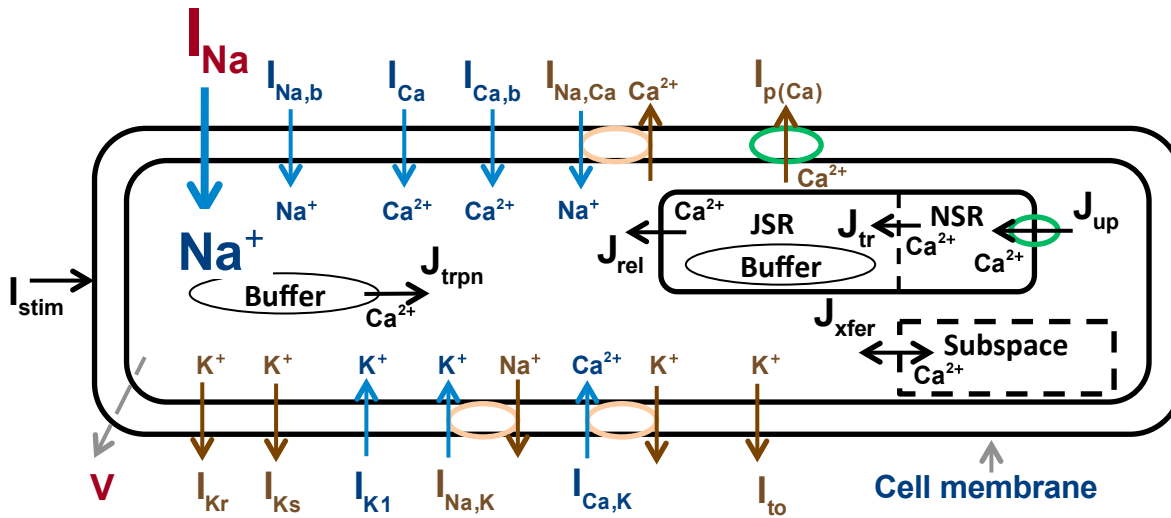
Channel Approximation: Na⁺ Channel



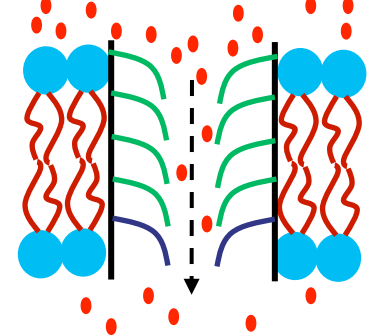
Channel Approximation: Patch-Clamp



Channel Approx: IMW Model



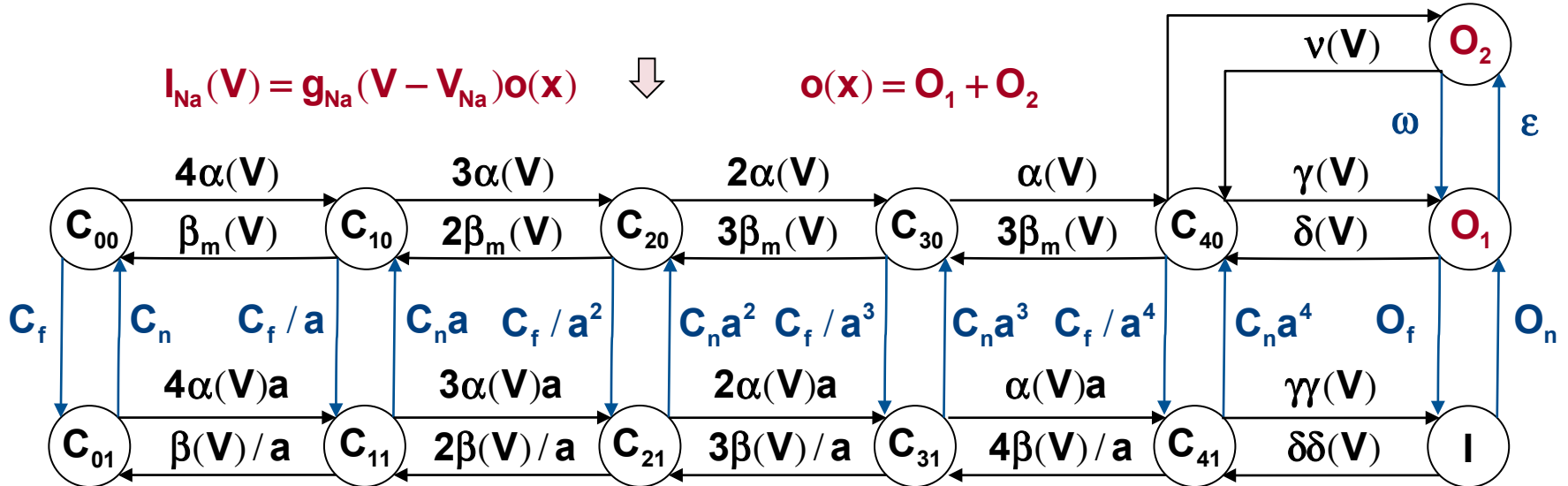
dependent gates



12-States CT-MDP

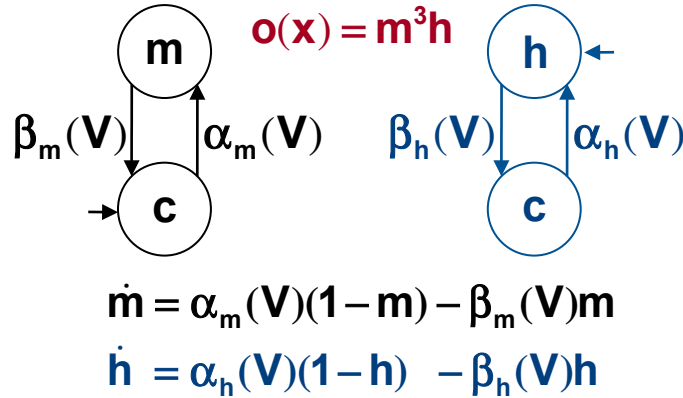
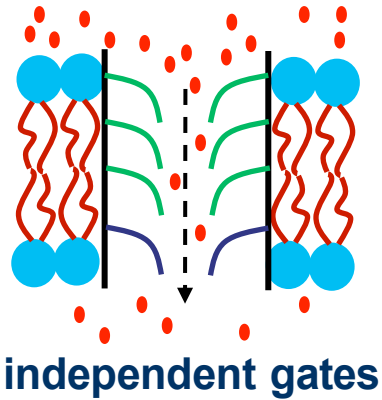
$$I_{Na}(V) = g_{Na}(V - V_{Na})o(x)$$

$$o(x) = O_1 + O_2$$

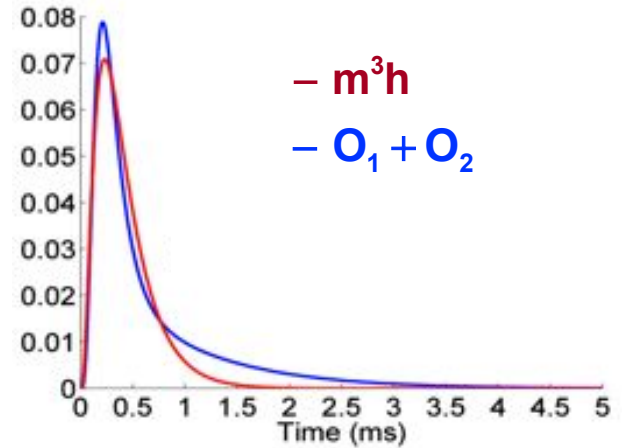


$$\dot{C}_{01} = C_f C_{00} + \beta(V) C_{11} / a - (C_n + 4\alpha(V)a) C_{01}$$

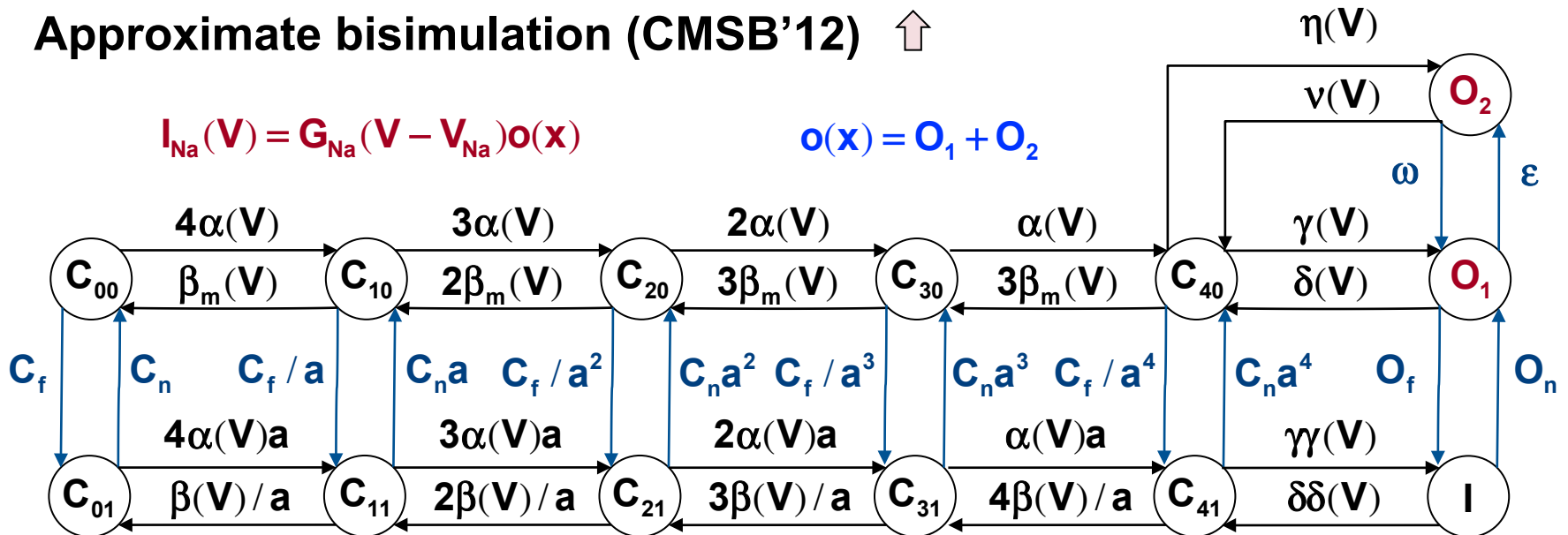
Channel Approx: HH-Na-Channel



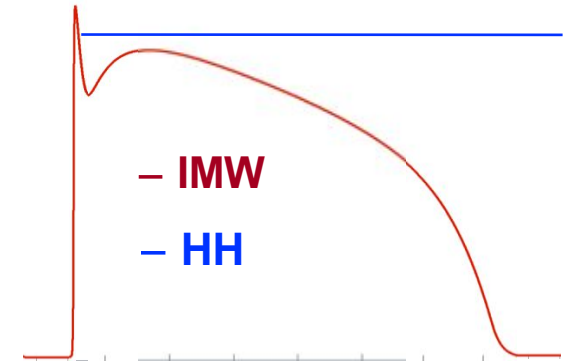
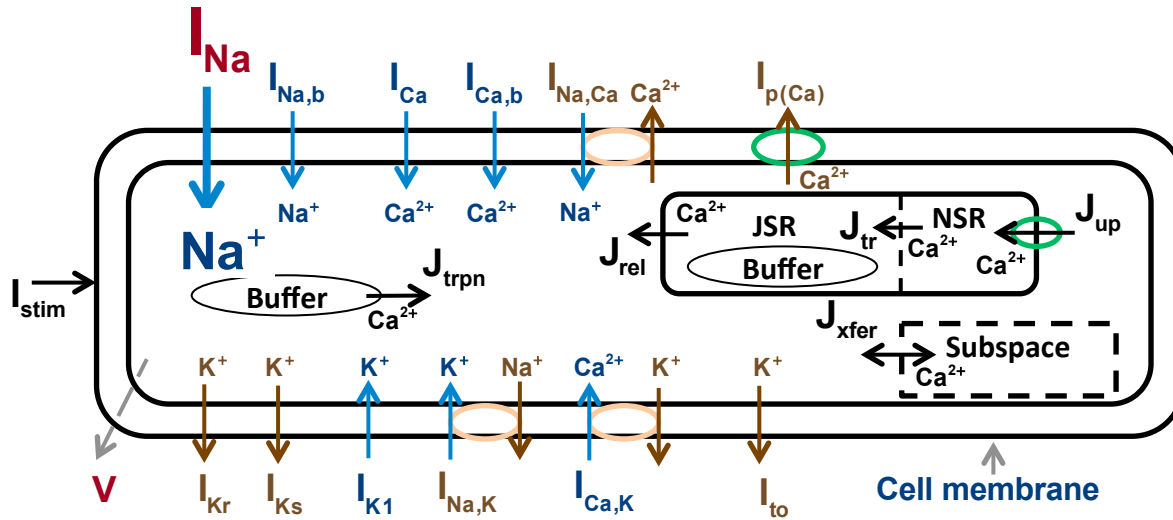
$$o(x) = m^3 h$$



Approximate bisimulation (CMSB'12) \uparrow



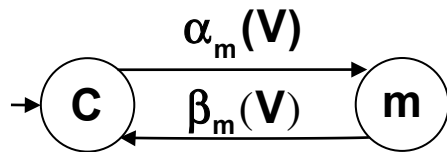
Compositionality Challenge: IMW with HH



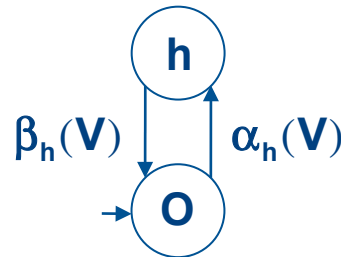
$$I_{Na}(V) = G_{Na}(V - V_{Na})o(V) \quad \Downarrow \quad o(x) = m^3h$$

2-States CTMC

Rates depend on V

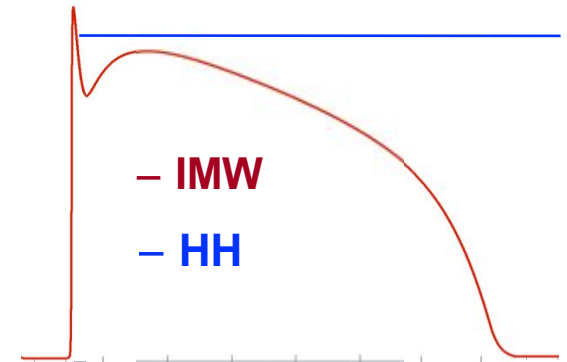
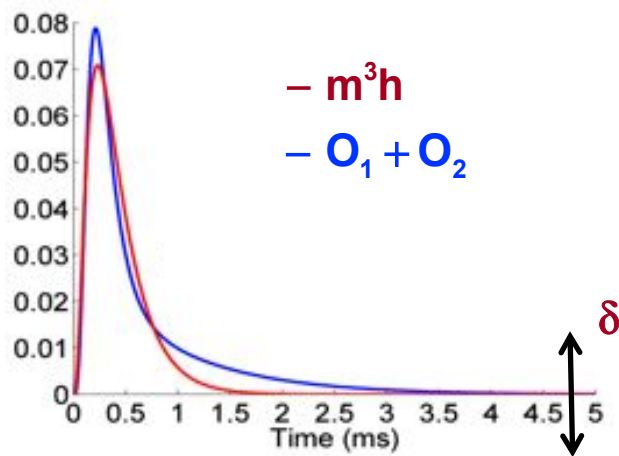


$$\dot{m} = \alpha_m(V)(1 - m) - \beta_m(V)m$$



$$\dot{h} = \alpha_h(V)(1 - h) - \beta_h(V)h$$

Compositionality Challenge: IMW with HH



Problem: HH-Channel was not completely closing

Learned a better δ -bisimulation
 A way to predict such problems?

Error δ was amplified
 δ -bisim not compositional

$$\begin{array}{c}
 \text{IMW}_{\text{Na}} \approx^{\delta} \text{HH}_{\text{Na}} \\
 \hline
 \text{R} \parallel \text{IMW}_{\text{Na}} \approx^{f(\delta)} \text{R} \parallel \text{HH}_{\text{Na}}
 \end{array}$$

\Downarrow

Bisimulation Functions with SGC

Small-gain condition (SGC) for bisimulation function S :

1. S bounds output difference :

$$\|o(x_1) - o(x_2)\| \leq S(x_1, x_2)$$

$$x_1 = [m, h]$$

$$x_2 = [C_{00}, \dots, C_{41}, I, O_1, O_2]$$

2. S decays along trajectories :

$$\forall u_1, u_2, x_1, x_2, \exists \lambda > 0, \gamma \geq 0 \text{ st.}$$

$$\dot{S} = \frac{\partial S}{\partial x_1} \underbrace{f_1(x_1, u_1)}_{\dot{x}_1} + \frac{\partial S}{\partial x_2} \underbrace{f_2(x_2, u_2)}_{\dot{x}_2} \leq -\lambda S(x_1, x_2) + \gamma \|u_1 - u_2\|$$

BFs compose linearly subject to SGC:

-BFs computed using SoS-TOOLS
(HSCC'14-15)

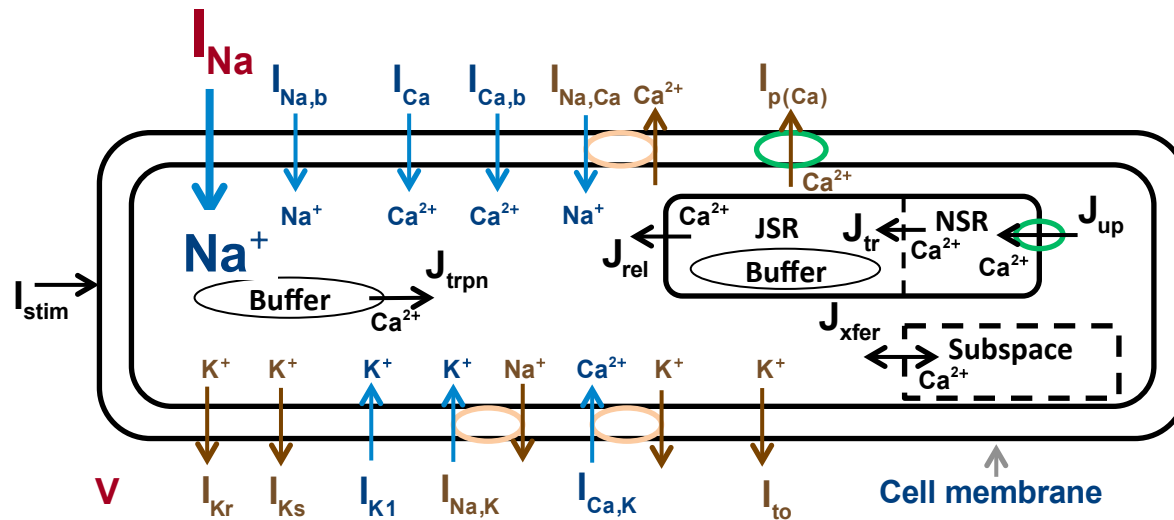
$$R \cong^S R \quad \mathbb{I}_{Na}^{IMW} \cong^T \mathbb{I}_{Na}^{HH}$$

$$R \mathbb{I} \mathbb{I}_{Na}^{IMW} \cong^{aS+bT} R \mathbb{I} \mathbb{I}_{Na}^{HH}$$

Conclusions

Models become increasingly sophisticated

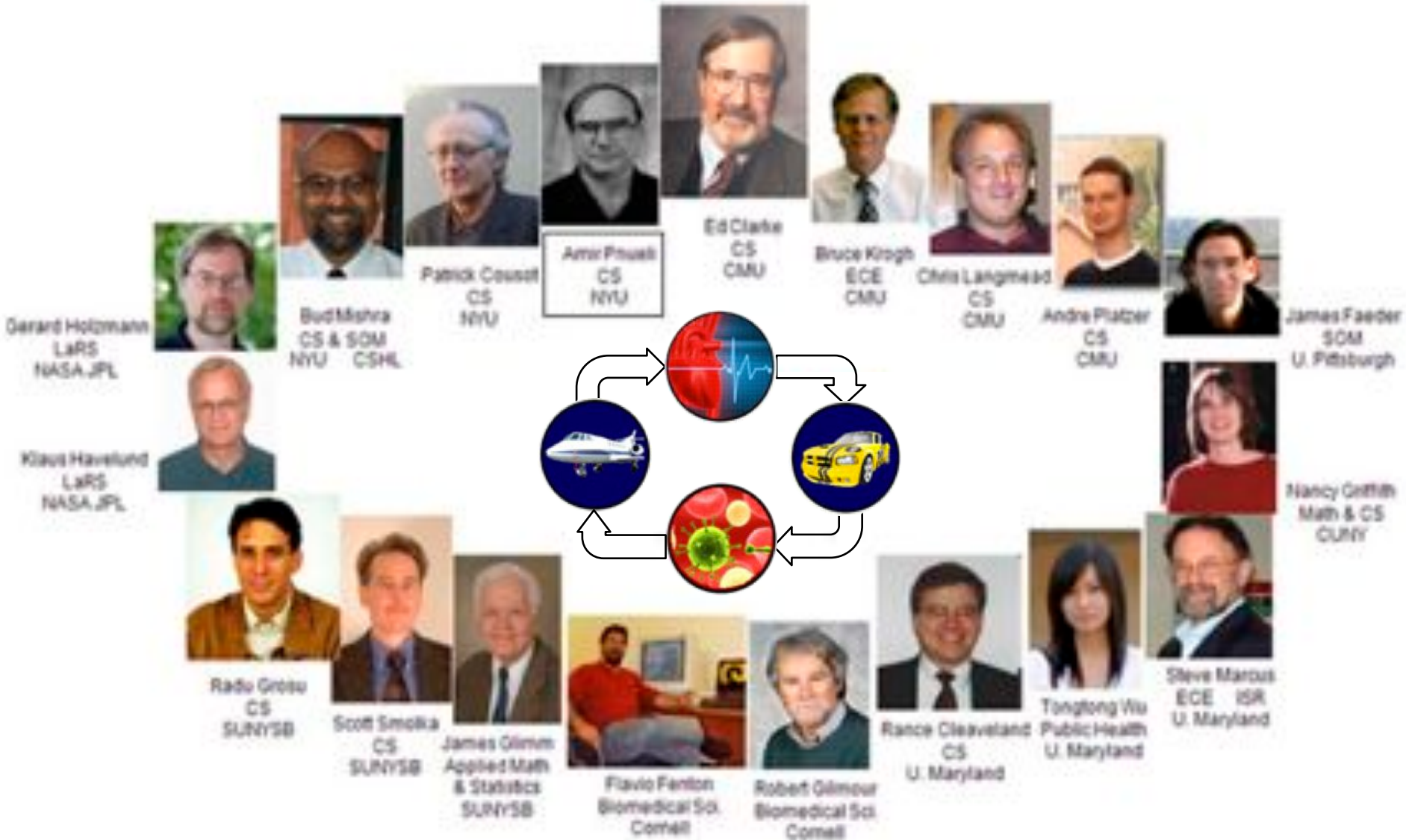
- Molecular processes better understood
- **Investigation of very specific treatments**



However, simulation of whole organ infeasible

- No method to automatically learn a channel's env
- **Compositionality is a big obstacle**

CMACS: Multi-Institutional, -Disciplinary Team



CyberHeart: Multi-Institutional, -Disciplinary Team

Scott Smolka
Stony Brook

Rance Cleaveland
UMD / Fraunhofer

Rick Gray
FDA



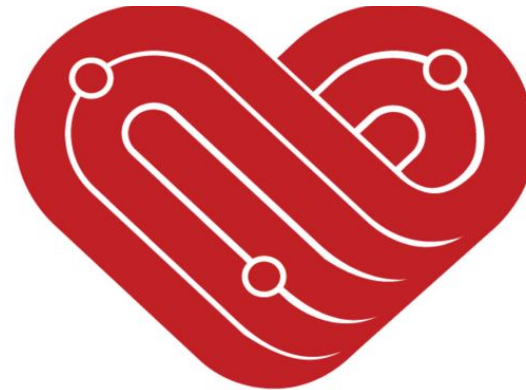
Elizabeth Cherry
RIT

Ed Clarke
CMU



James Glimm
Stony Brook

Sean Gao
CMU



Radu Grosu
Stony Brook /
Vienna

Arnab Ray
Fraunhofer



Rahul Mangharam
Penn



Flavio Fenton
Gatech



Sanjay Dixit
Director of Cardiac
Electrophysiology
Philadelphia VA Hospital