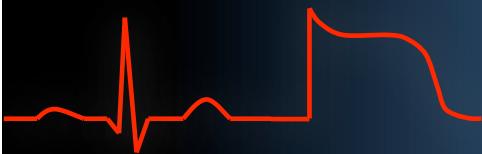


Cardiovascular Dynamics *in-silico*

Pharmacological Targeting of Long QT 3 Syndrome: Proof of Concept for an *in-silico* Drug Development Platform



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Outline

- Biophysics | Clinical Medicine | Cardiovascular Pharmacology | Mathematics | Engineering | HPC
- Ion-channel mutations (long QT)
 - Biophysics, genotype, phenotype and clinical characteristics
- Current treatment strategies
- *In-silico* predictive modeling and computational biomedicine
 - Formulation of Markov models of drug-receptor interactions
 - Channel | cell | cable | tissue | 3D heart
 - Numerical techniques
 - High performance computing

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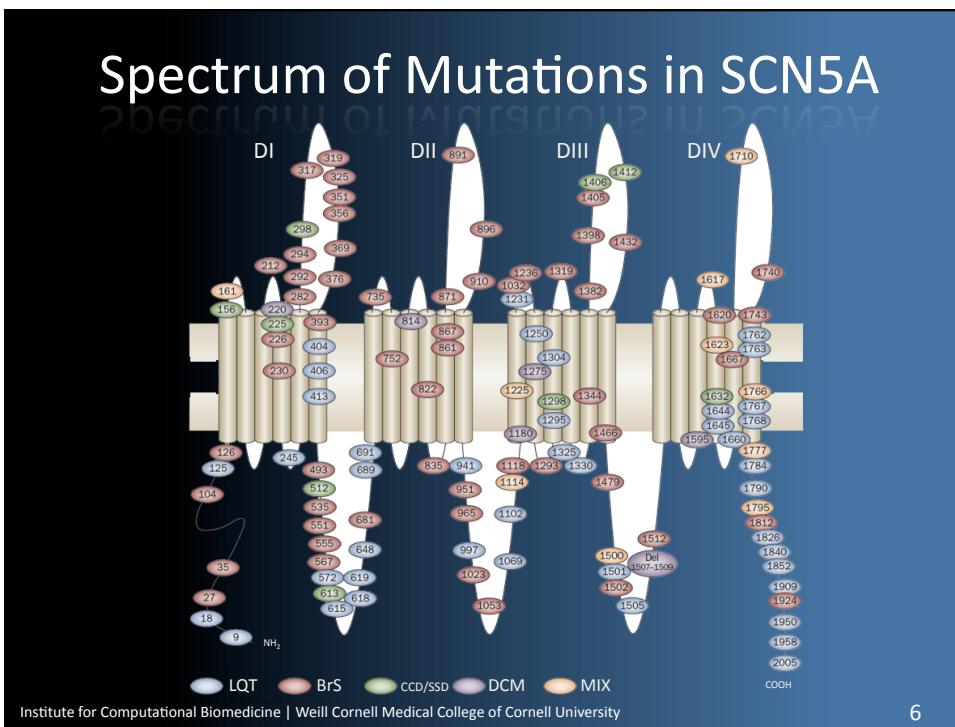
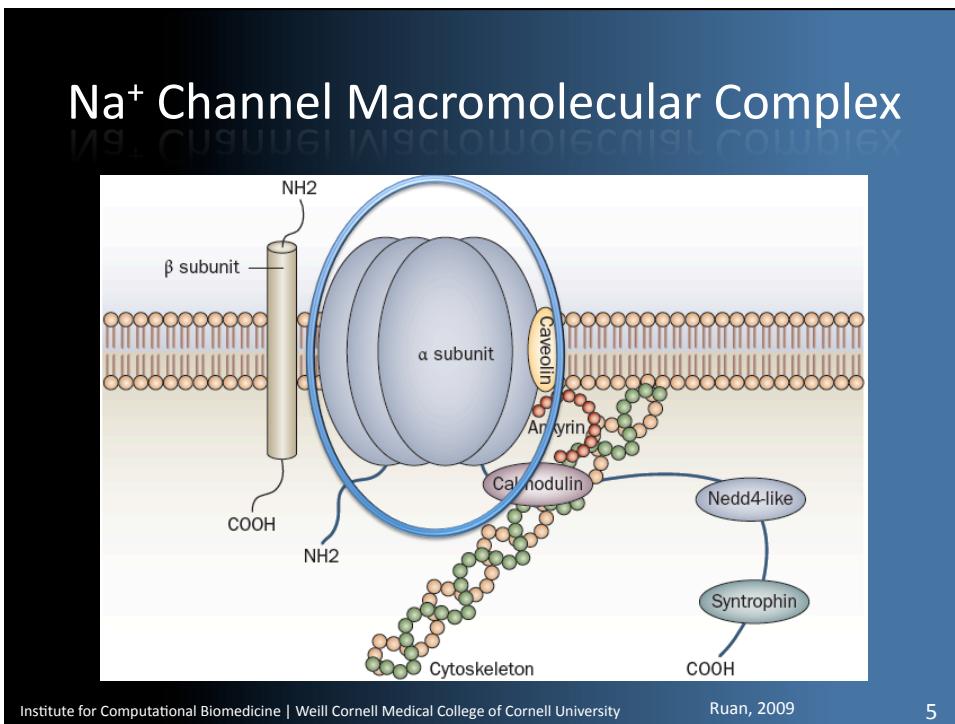
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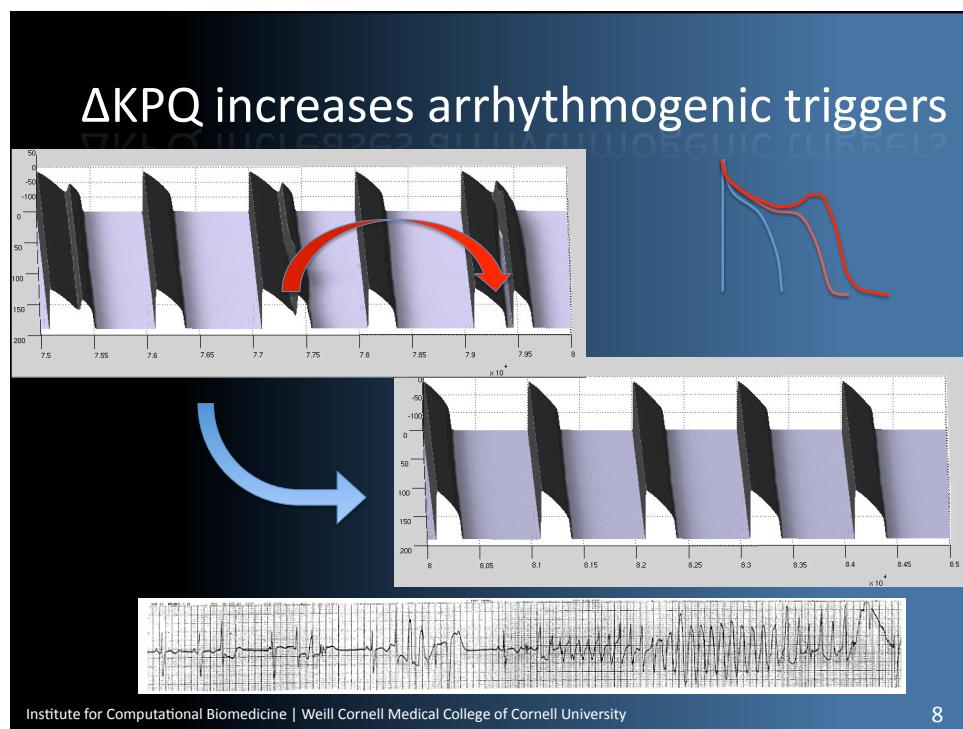
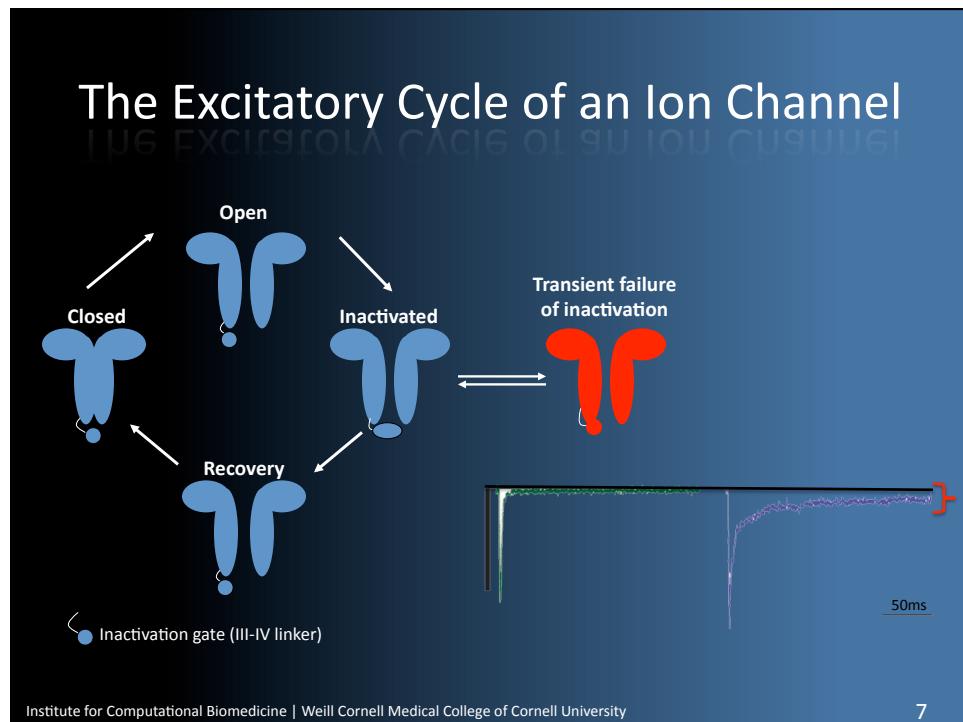
Ion-channel mutations and their arrhythmogenicity

- **Channelopathy:** Mutations in genes that encode ion channels can lead to abnormal channel function leading to perturbation of the AP to cause arrhythmias. Specific Na^+ channel mutations:
 - Gain of function during AP plateau (LQT)
 - Overall loss of channel function (Brugada, ICCD, SSS)
- Hereditary Long QT (LQT) syndrome
 - Prolongation of QT interval on ECG can lead to life-threatening arrhythmias and sudden cardiac death
 - **LQT3:** Usually bradycardic; occurring during sleep or relaxation

The Long QT (3) Syndrome

- Heterogeneous group of mutations in cardiac sodium channel α subunit
 - Genetically distinct, clinical presentation similar with subtle differences
 - Overlapping syndromes with clinical characteristics coexisting in a single patient
- ΔKPQ identified in 1995
 - Transient failure of inactivation → persistent I_{Na}
- D1790G: C terminal mutation, LQT3 clinical phenotype, possible distinct mechanism of action





Hodgkin Huxley Formulation

- HH formulation computes conductance for each current as a function of the P(O) of a series of hypothetical gates
 - Conductance is $f(t, V)$ via gates
 - 1st order transitions from C → O and O → C that are *independent* of the other gates
 - Ions can only pass through the open state of the gate
- Based on experimental data, Na⁺ activation can be modeled by 3 identical activation gates (m^3)
- Inactivation shortly after activation (h)
- LRd also includes slow inactivation gate (j)

$$I_{Na} = G_{Na} * m^3 * h * j * (V - E_{Na})$$

Markov-based Models

Motivation

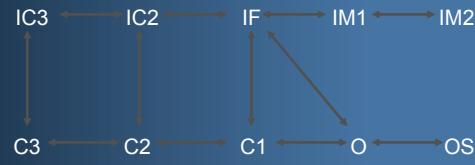
- Need for models with explicit representation of single ion-channel states
- HH models are limited in their ability to describe specific aspects of single channel behavior
 - Inactivation of the Na⁺ channel greater when in state O
- Assumption of independent gating ($m^3 \cdot h \cdot j$) fails
- MM can represent the dependence of a given transition on the occupancy of different states of the channel
 - Assume that transitions between channel states depend on the present conformation of the channel, but not on previous behavior

Formulating the model

- Wild Type: Use as many states as necessary to recapitulate the kinetics of channels

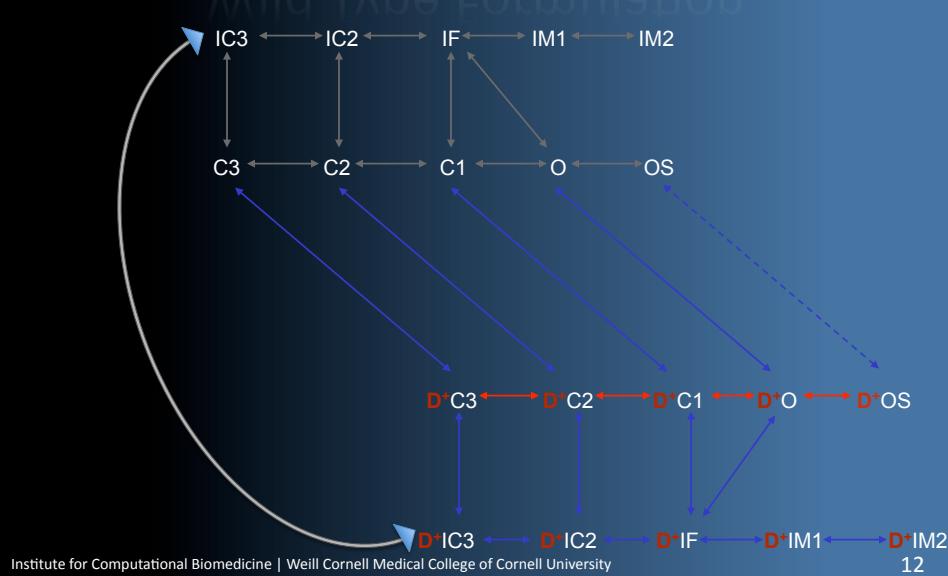
- Experimental data:

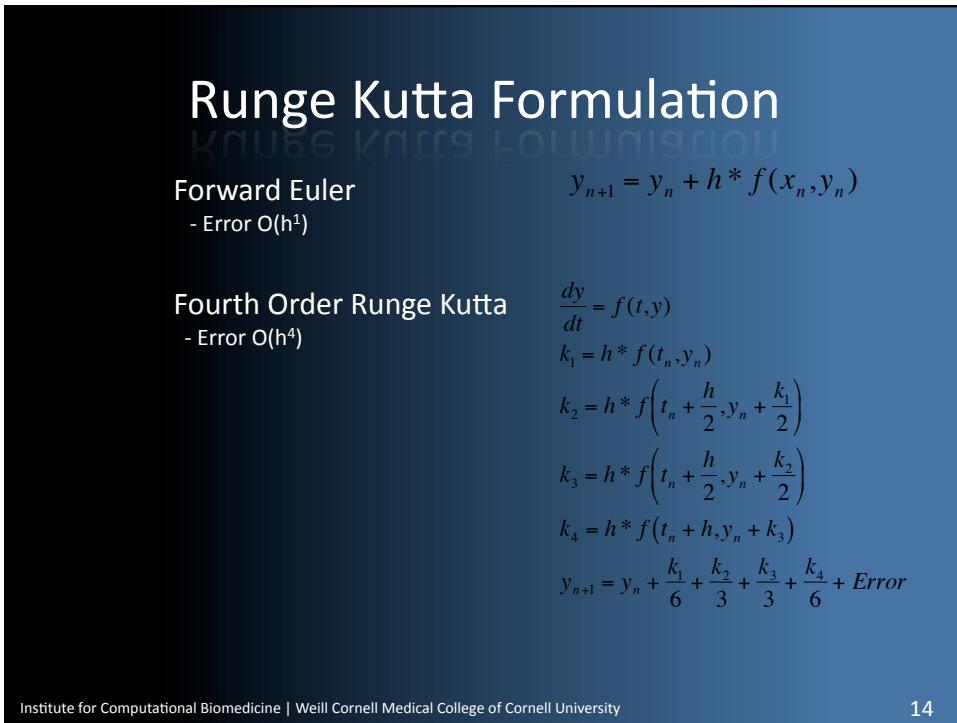
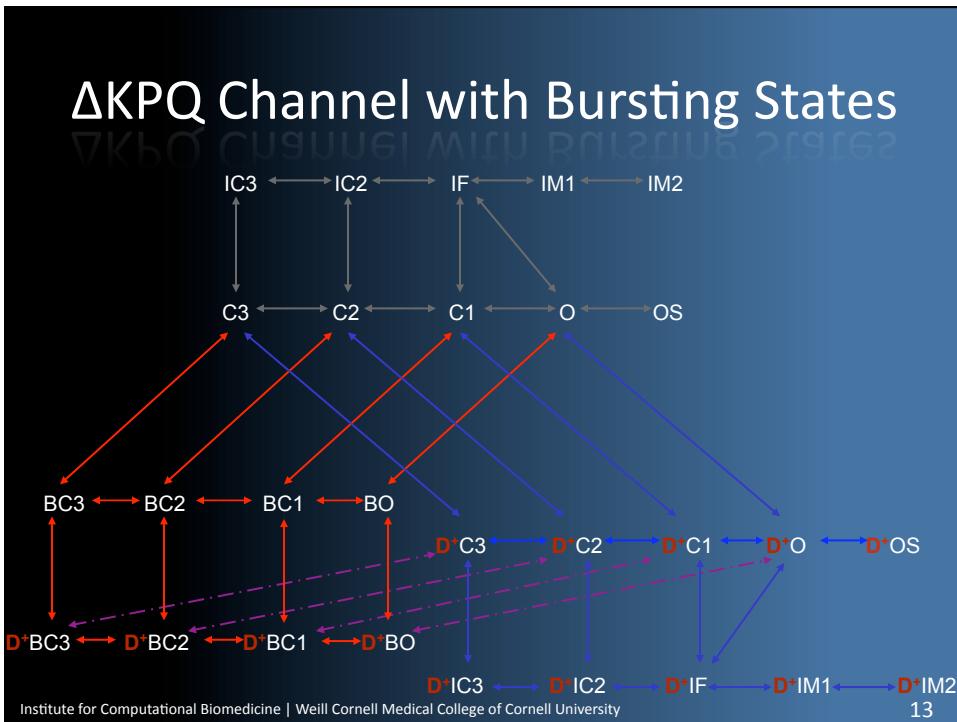
- k_{on} , k_{off} rates
- pKa – for drug partitioning
- Steady State Availability
- Tonic Block - k_d^{closed}
- Use Dependent Block - k_d^{open}
- Frequency dependence of UDB
- Recovery from UDB



- Models adhere to 2nd law of thermodynamics (microscopic reversibility)
- Start playing!

Wild Type Formulation





...some code

```

O_1K = Cell<ptr>_C1 + K1_OZ;
C1_K1 = Cell<ptr>_MC1 + K1_C1Z;
C1_C2 = Cell<ptr>_MC1 + K1_C2Z;
C3_K1 = Cell<ptr>_MC3 + K1_C3Z;
C3_K2 = Cell<ptr>_MC3 + K1_C2Z;
C3_K3 = Cell<ptr>_MC3 + K1_C3Z;
C3_K4 = Cell<ptr>_MC3 + K1_C4Z;
C3_K5 = Cell<ptr>_MC3 + K1_C5Z;
C3_K6 = Cell<ptr>_MC3 + K1_C6Z;
IM1_K1 = Cell<ptr>_IM1 + K1_IM1Z;
IM1_K2 = Cell<ptr>_IM1 + K1_IM2Z;
IM1_K3 = Cell<ptr>_IM1 + K1_IM3Z;
IM1_K4 = Cell<ptr>_IM1 + K1_IM4Z;
IM1_K5 = Cell<ptr>_IM1 + K1_IM5Z;
IM1_K6 = Cell<ptr>_IM1 + K1_IM6Z;
OS_1K = Cell<ptr>_MOS + K1_OSZ;
D0_1K = Cell<ptr>_D0 + K1_D0Z;
DC1_K1 = Cell<ptr>_DC1 + K1_DC1Z;
DC1_C2 = Cell<ptr>_DC1 + K1_DC2Z;
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DOS_1K = Cell<ptr>_DOS + K1_DOSZ;
DC3_K4 = Cell<ptr>_DC3 + K1_DC4Z;
DC3_K5 = Cell<ptr>_DC3 + K1_DC5Z;
DC3_K6 = Cell<ptr>_DC3 + K1_DC6Z;
DIF_1K = Cell<ptr>_DIF + K1_DIFZ;
DIM1_K1 = Cell<ptr>_DIM1 + K1_DIM1Z;
DIM2_K1 = Cell<ptr>_DIM1 + K1_DIM2Z;
DIM2_K2 = Cell<ptr>_DIM1 + K1_DIM3Z;
DIM2_K3 = Cell<ptr>_DIM1 + K1_DIM4Z;
DRCB_1K = Cell<ptr>_DRCB + K1_DRCBZ;
DBC2_1K = Cell<ptr>_DBC2 + K1_DBC2Z;
DBC3_1K = Cell<ptr>_DBC3 + K1_DBC3Z;

D_0_K1 = Cell<ptr>_D0 + K1_D0Z;
D_C1_K1 = Cell<ptr>_D_C1 + K1_D_C1Z;
D_C2_K1 = Cell<ptr>_D_C2 + K1_D_C1Z;
D_C3_K1 = Cell<ptr>_D_C3 + K1_D_C1Z;
D_C3_K2 = Cell<ptr>_D_C3 + K1_D_C2Z;
DOS_0_K1 = Cell<ptr>_DOS + K1_D_OSZ;
D_C3_K3 = Cell<ptr>_D_C3 + K1_D_C3Z;
D_C3_K4 = Cell<ptr>_D_C3 + K1_D_C4Z;
D_C3_K5 = Cell<ptr>_D_C3 + K1_D_C5Z;
D_C3_K6 = Cell<ptr>_D_C3 + K1_D_C6Z;
F4_K1 = Cell<ptr>_F4 + K1_F4Z;
F4_K2 = Cell<ptr>_F4 + K1_F4Z;
F4_K3 = Cell<ptr>_F4 + K1_F4Z;
F4_K4 = Cell<ptr>_F4 + K1_F4Z;
F4_K5 = Cell<ptr>_F4 + K1_F4Z;
F4_K6 = Cell<ptr>_F4 + K1_F4Z;

```

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The failure of antiarrhythmics

- Sudden Cardiac Death is the #1 cause of death in the US, CVD in general will be global cause of death by 2020
 - CAST, CAST II trials:
 - an incomplete understanding and inability to predict interaction of intrinsically complex drug pharmacology with complexity of cardiac tissue
 - Drug receptor dynamics
 - Anatomical and electrical heterogeneity
 - Mutant channel interactions
 - Off target effects of noncardiovascular therapeutics
 - 40% of all new pharmaceuticals affect repolarization process
 - Vast majority of drugs pulled off market (Vioxx)
 - Side effects may occur <1% of patient population
 - Predict vulnerable populations?
 - Discard compounds earlier in the product development pipeline?
 - 50:1 failure to success rate

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Cardiovascular Dynamics *in-silico*

- Single channel → single cell → 1D fiber → 2,3D tissues
 - *Spatiotemporal trends that emerge in higher dimensions that allow for propagation of arrhythmia are missed at smaller spatial scales*
- Develop a computational framework for an *in-silico* drug screen of cardiovascular medications
 - Predictive simulations of pharmacodynamics in an engineered virtual cardiac tissue
 - Identification of key parameters for drug-receptor dynamics

Biophysics and
experimental data

Computational and
engineering methods



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Model “wish list”

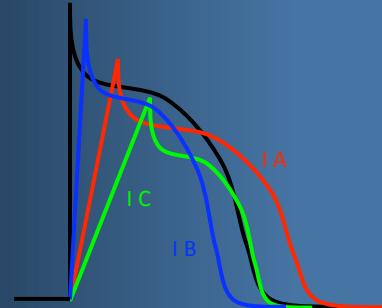
- Transport phenomena: Diffusion
- Kinetics of drug receptor interactions
- Macroscopic wave propagation
- Scalability: 0D → 3D
- Optimize with high performance computing
 - Numerical optimizations
 - Parallel computations
 - GPGPU

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Sodium Channel Block Subclassification

- Class IA (e.g., **quinidine**)
 - Moderate Na^+ channel blockade
 - \uparrow ERP
- Class IB (e.g., **lidocaine**)
 - Weak Na^+ channel blockade
 - \downarrow ERP
- Class IC (e.g., **flecainide**)
 - Strong Na^+ channel blockade
 - \rightarrow ERP



Antiarrhythmics

Flecainide

- Class 1C antiarrhythmic
- $pK_a \sim 9.3$
- 99% charged at physiologic pH

Lidocaine

- Class 1B antiarrhythmic
- $pK_a \sim 7.6$
- 50% charged at physiologic pH

