

Experimental Bifurcation Analysis In Neurons Using Control-Based Continuation

Mark Blyth



About me

- First year PhD student (started in September)
- Supervised by Lucia and Ludovic
- Studied EngMaths for my undergrad
- Research interests are in dynamical systems theory and applied nonlinear mathematics



₭ How do neurons work?



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- Why should mathematicians get excited by neurons?



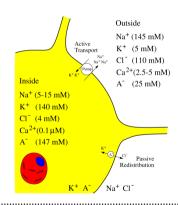
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- What is my research topic? Why am I doing what I'm doing?
- What challenges am I trying to solve, and how?

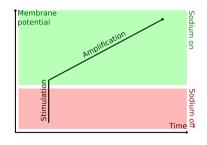


- Neurons are cells; they and their surrounding media contain charged ions
- Acive transport across the cell membrane means that, at rest, there's a voltage over the membrane
- At rest, this membrane potential is typically around -70 mV



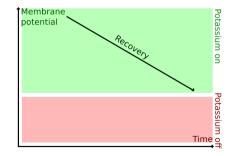


- Sodium current activates as membrane potential increases
- Simple model: current switches on when membrane potential exceeds a threshold
- It's an inward current, so it brings positive ions into the cell and increases membrane potential
- This causes positive feedback!





- Potassium currents activate as membrane potential increases
- Potassium forms an outward current positive ions flow out, returning the membrane potential to its rest value
- The potassium current turns on and off slower than the sodium current





The interplay of slow potassium and fast sodium currents causes neurons to spike, rather than settling to a steady state

A recipe for spiking

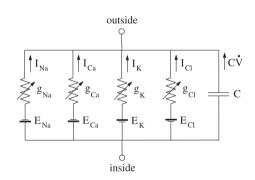
- Sodium currents switch on and off fast
- Potassium currents switch on and off slowly
- K Slow potassium activation allows the membrane potential to increase fast
- ✓ Once it activates, the potassium current pulls the membrane potential back down
- Potassium current takes a while to switch off again, so membrane potential gets pulled down to below the turn-on threshold for the two currents



Currents flow through different ion channels; let's consider each one separately. Using current laws,

$$C\dot{V} = I_{Na} + I_{Ca} + I_K + I_{Cl}$$
 (1)

The Hodgkin-Huxley model gives each ionic current as a function of membrane potential. This is exciting, as we now have a mathematical model of a neuron, to which we can apply a rigorous analysis.





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- Highly nonlinear
- High-dimensional
- Multi-timescale dynamics
- Stochastic behaviour



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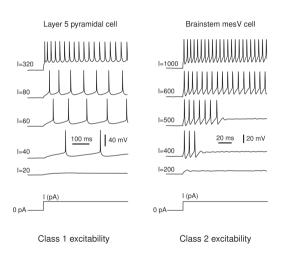
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- ✓ Dynamical neuroscience is also something of a new field, though, so there's still a big research gap in experimental bifurcation analysis



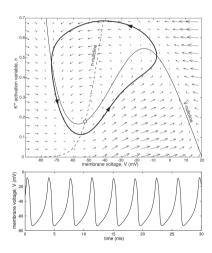
Project goal

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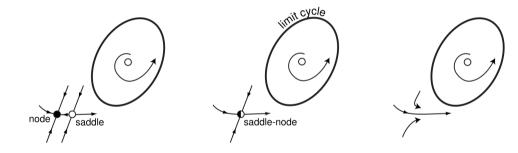
Spike trains and neural computation



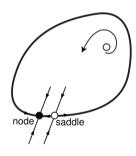
Phase diagrams

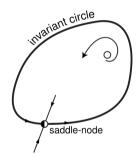


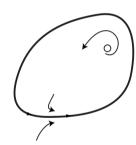




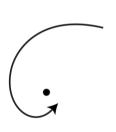


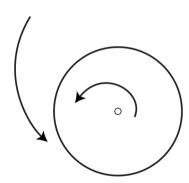




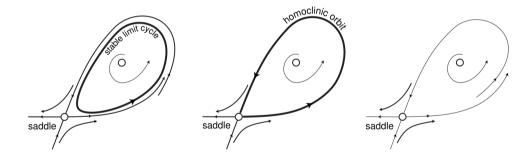














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George Box

All models are wrong, but some are useful



Numerical continuation

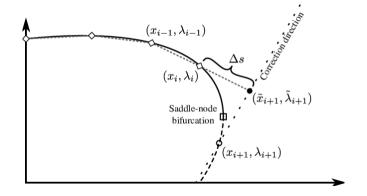
Consider $f(x, \lambda) = 0$. Numerical continuation seeks to track x, as λ varies. For ODEs of form

$$\dot{x} = f(x, \lambda) ,$$

this can be used to find bifurcations.



Numerical continuation





Control-based continuation

CBC allows us to apply continuation methods on black-box numerical or physical systems, no model needed.

- ✓ Use control theory to steer the system onto a (possibly unstable) natural invariant set

This tracking step can be a classical psuedo-arclength continuation, or something more problem-specific.



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- We The system has poor observability (eg. can't easily see population ion channel conductance); how do we control a system that we can't observe?
- We have limited control inputs; how can we use them to steer the dynamics effectively?