

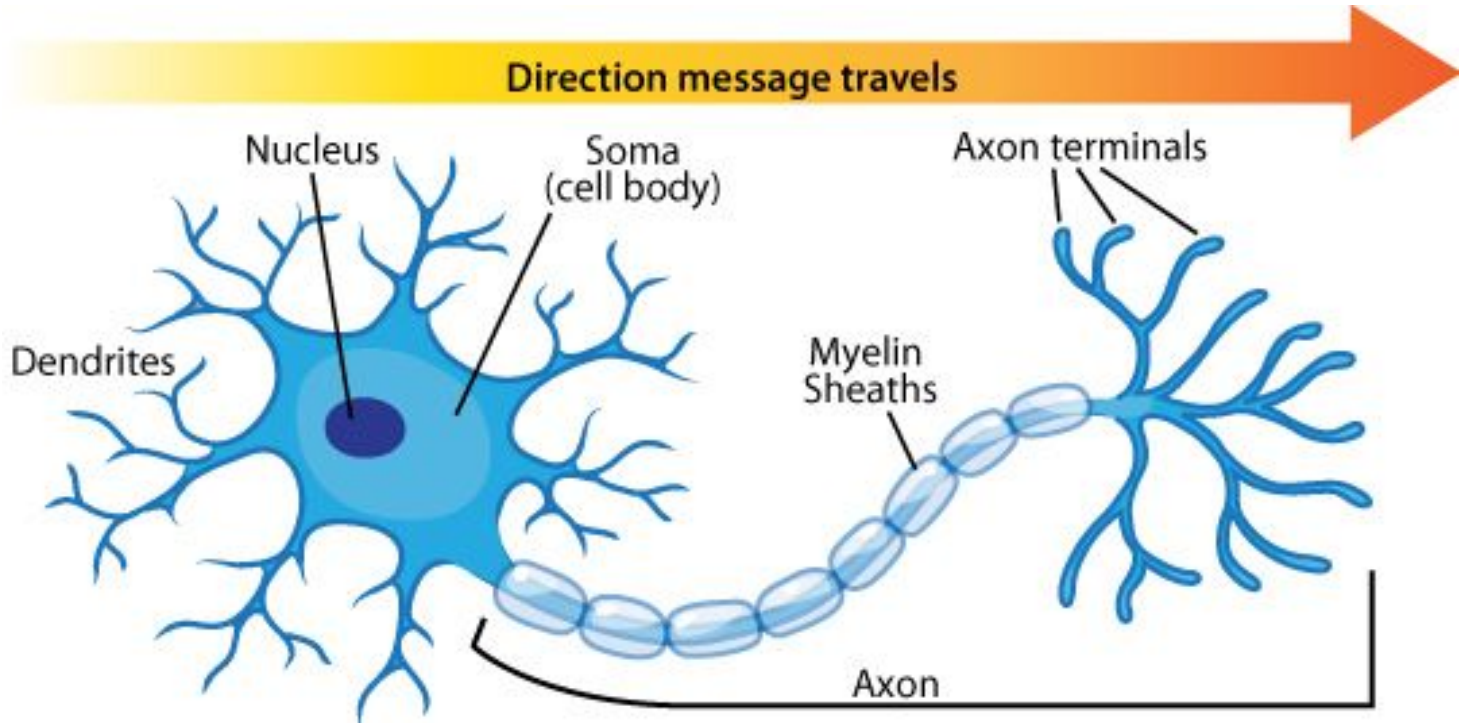
PDE of Neuron Models

Maire Keene & Michael Sheets
PDE Final Spring 2016

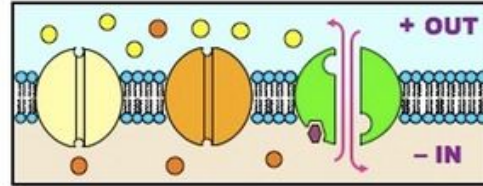
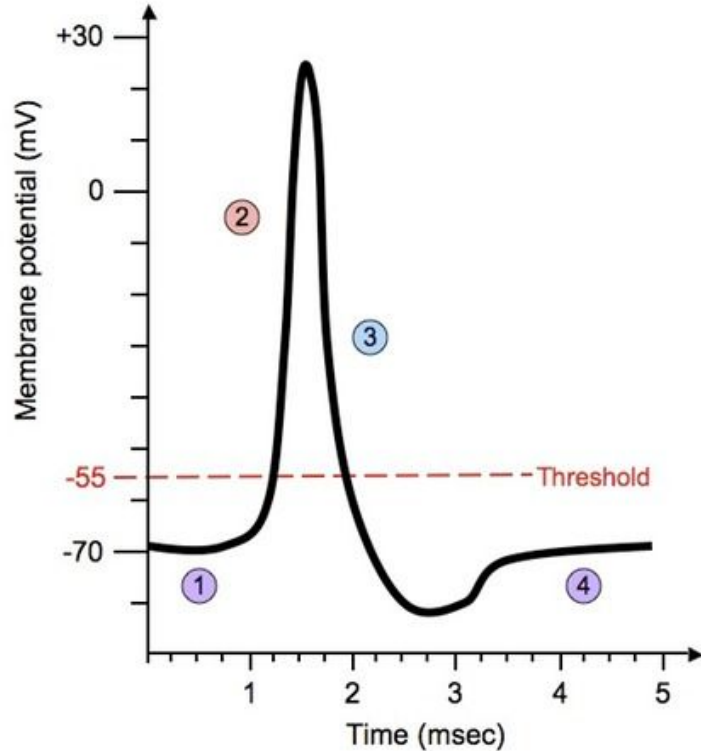
This is a 10 minute introduction to an introduction.

1. So what's a neuron, again?
2. Action Potential
3. Hodgkin-Huxley
4. FitzHugh-Nagumo
 - a. 'Fast' and 'Slow' Time
 - b. Phase Space
 - c. Space *and* Time: Travelling Waves

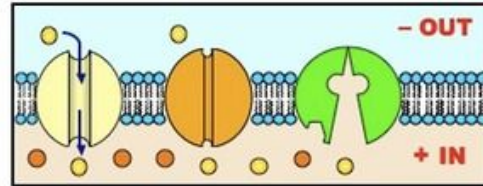
This is a neuron:



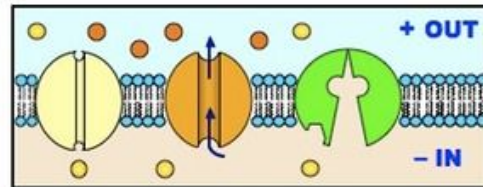
Action Potential



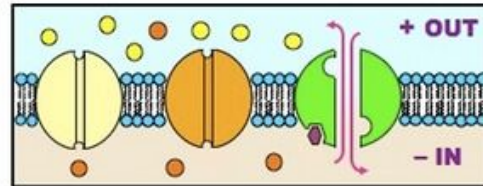
- ① **Resting Potential**
Na⁺/K⁺ pump



- ② **Depolarisation**
Voltage-gated Na⁺ channel

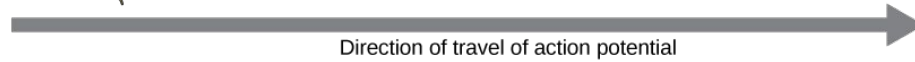
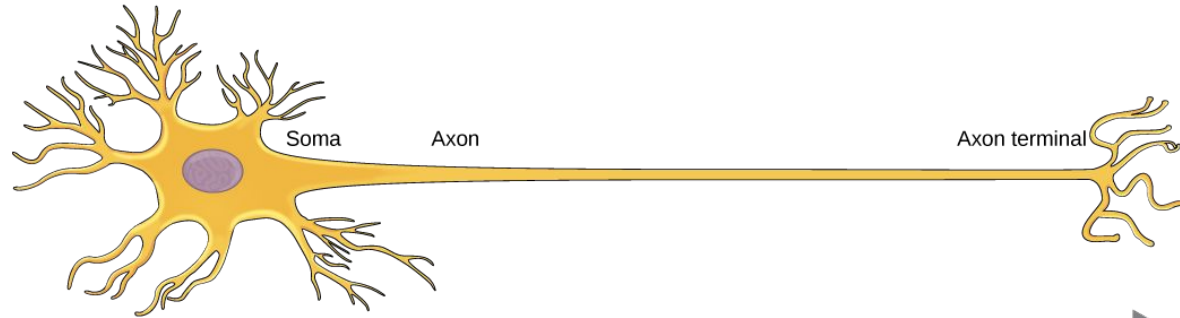


- ③ **Repolarisation**
Voltage-gated K⁺ channel



- ④ **Resting Potential**
Na⁺/K⁺ pump

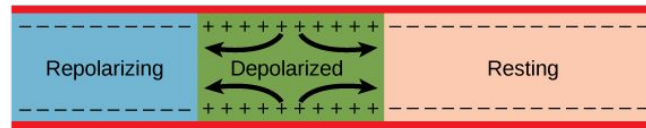
Action Potential



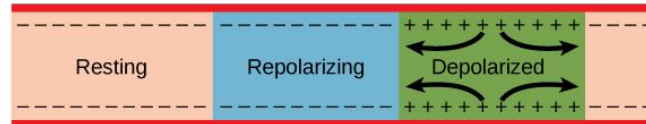
- a. In response to a signal, the soma end of the axon becomes depolarized.



- b. The depolarization spreads down the axon. Meanwhile, the first part of the membrane repolarizes. Because Na^+ channels are inactivated and additional K^+ channels have opened, the membrane cannot depolarize again.



- c. The action potential continues to travel down the axon.



Please Don't Panic: Hodgkin-Huxley

By applying $Q=CV$ to the Ion Cascades:

Current/Area = sum of currents

$$\left\{ \begin{array}{l} I \leftarrow C_m \frac{dV_m}{dt} + \bar{g}_K n^4 (V_m - V_K) + \bar{g}_{Na} m^3 h (V_m - V_{Na}) + \bar{g}_l (V_m - V_l) \\ \frac{dn}{dt} = \alpha_n(V_m)(1 - n) - \beta_n(V_m)n \\ \frac{dm}{dt} = \alpha_m(V_m)(1 - m) - \beta_m(V_m)m \\ \frac{dh}{dt} = \alpha_h(V_m)(1 - h) - \beta_h(V_m)h \end{array} \right.$$

Terms are
each the
constituent
currents/area

Hodgkin-Huxley

$$\begin{cases} I = C_m \frac{dV_m}{dt} + \bar{g}_K n^4 (V_m - V_K) + \bar{g}_{Na} m^3 h (V_m - V_{Na}) + \bar{g}_l (V_m - V_l) \\ \frac{dn}{dt} = \alpha_n(V_m)(1 - n) - \beta_n(V_m)n \\ \frac{dm}{dt} = \alpha_m(V_m)(1 - m) - \beta_m(V_m)m \\ \frac{dh}{dt} = \alpha_h(V_m)(1 - h) - \beta_h(V_m)h \end{cases}$$

Max Capacitance

Membrane Voltage

Variable [0,1] characterizing activation

Channel Voltage Limit

Hodgkin-Huxley

$$\left\{ \begin{array}{l} I = C_m \frac{dV_m}{dt} + \bar{g}_K n^4 (V_m - V_K) + \bar{g}_{Na} m^3 h (V_m - V_{Na}) + \bar{g}_l (V_m - V_l) \\ \frac{dn}{dt} = \alpha_n(V_m)(1 - n) - \beta_n(V_m)n \quad \leftarrow \text{Sodium Channel Activation} \\ \frac{dm}{dt} = \alpha_m(V_m)(1 - m) - \beta_m(V_m)m \quad \leftarrow \text{Potassium Channel Activation} \\ \frac{dh}{dt} = \alpha_h(V_m)(1 - h) - \beta_h(V_m)h \quad \leftarrow \text{Sodium Channel Deactivation} \end{array} \right.$$

FitzHugh-Nagumo

- Frustration → Simpler model based on the van der Pol oscillator:

Voltage variable

“Recovery” variable

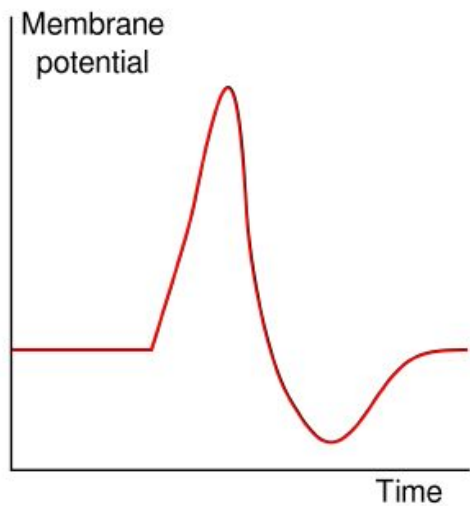
$$\begin{cases} \dot{V} &= V - \frac{V^3}{3} - W + I \\ \dot{W} &= 0.08(V + 0.7 - 0.8W) \end{cases}$$

Nonlinear competing term feedback

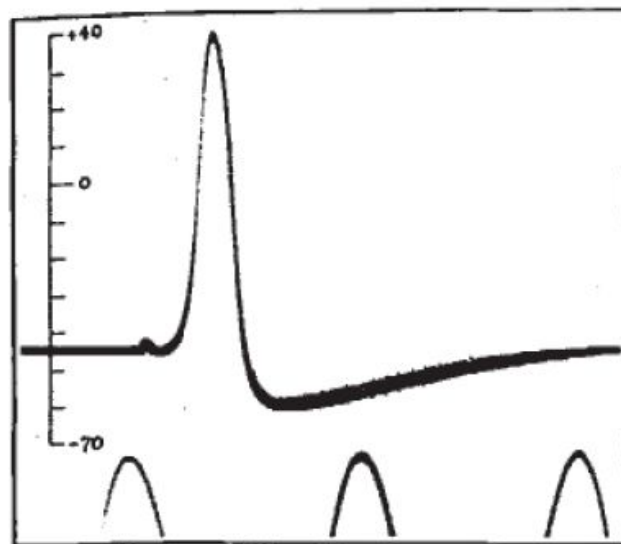
Linear, negative, and smaller magnitude feedback

- 2 state variables → much easier to visualize and analyze

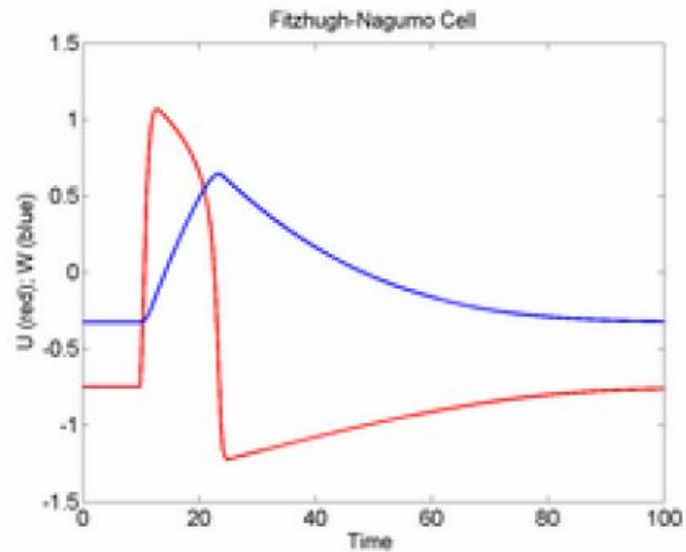
Outputs



Actual



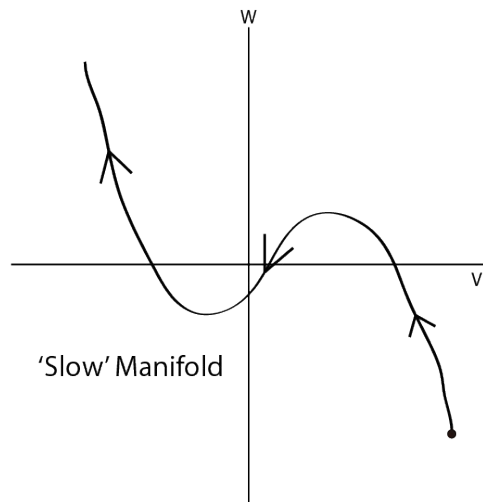
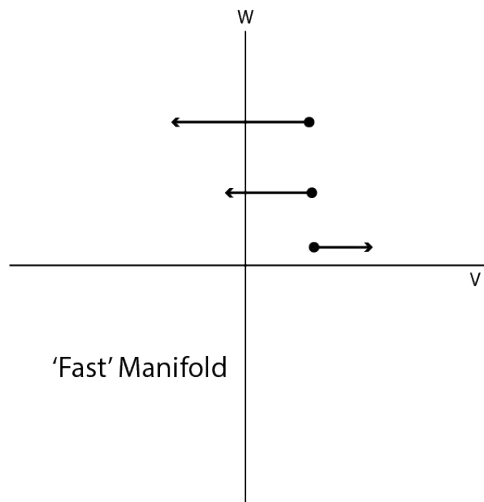
HH



FHN

'Fast' and 'Slow' Time

- $\tau = \epsilon t$ gives us a short and long time scale: change of variables to get two manifolds

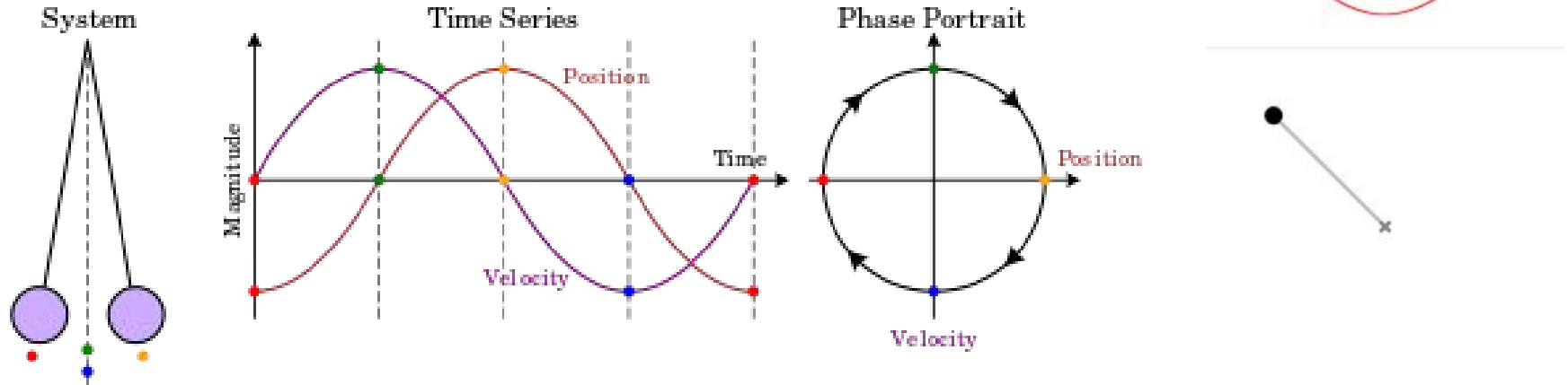


$$\begin{cases} \dot{V} &= V - V^3 - W + I \\ \dot{W} &= 0, \end{cases}$$

$$\begin{cases} 0 &= V - V^3 - W + I \\ \dot{W} &= (V - W). \end{cases}$$

Phase Space

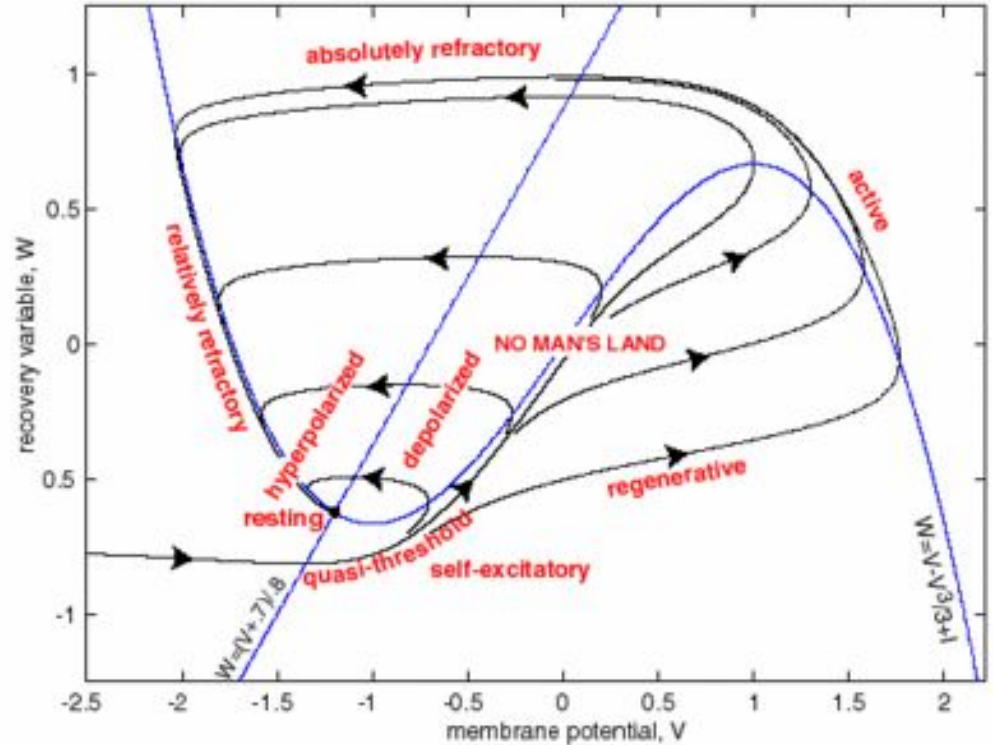
- Plotting all the possible system states:



- A powerful tool in complicated systems

The FitzHugh-Nagumo Phase Portrait

- Blue lines are “nullclines,” lines where one state variable isn’t changing
- Notice: similarity to manifolds
- Equilibrium at intersection(s)
- Black lines are time traces of system with different stimuli

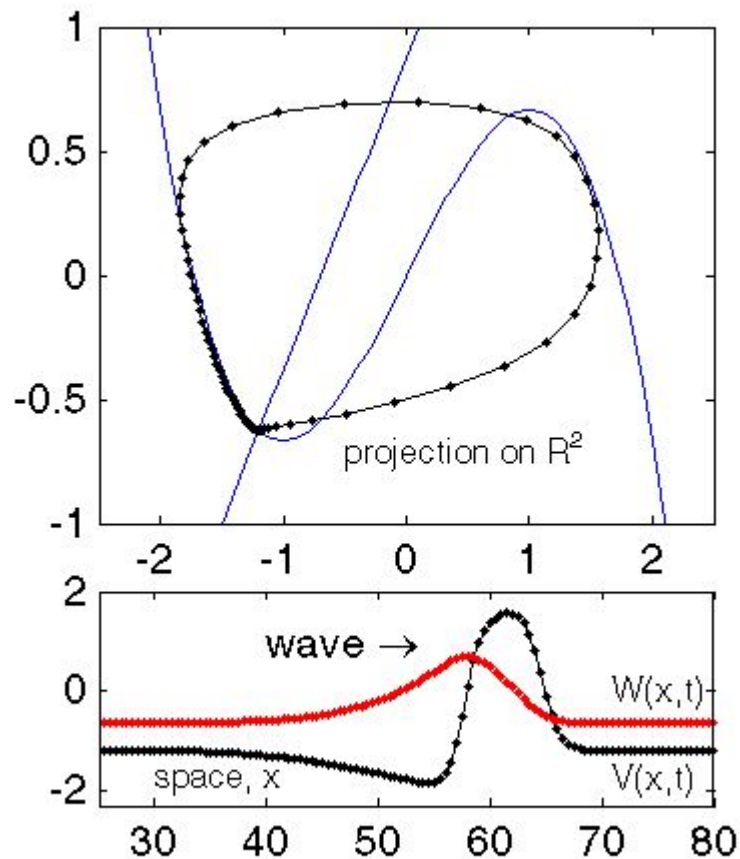


Traveling Waves

- Can our old traveling wave solution work here?

$$\dot{V} = V_{xx} + \underbrace{V - \frac{V^3}{3} - W + I}_{\text{familiar nonlinear bits, } f(V)}$$

- Oscillator + Diffusion Term
 - Biologically: this wave is propagating by ion diffusion so this kind of makes sense...
 - “Reaction-Diffusion Equation”
 - This form gets used all over the place



Traveling Waves: Good and Bad News

If we're hoping for

$$V(t, x) = \Phi(x - ct)$$

then this must be true

$$-c\Phi' = \Phi'' + f(\Phi)$$

And if *that* solution is bounded, then our traveling wave soln. can be physically reasonable

But myelin = spatially variant capacitance?



We have tools for estimating this grossness:



$$\begin{aligned}(K(x)V_x)_x &= \int V(y)K(x-y)dy \\ &= V_{n+1} + V_{n-1} - 2V_n\end{aligned}$$

Further Analysis + Applications

- Stability Problem
 - If we poke our stable waveform (imperfect axon, imperfect model)-- will the entire signal fall apart or fix itself and successfully reach the next neuron?
- Neural Prostheses
 - Retinal devices, cochlear implants, and neurally-controlled artificial limbs
 - These devices mimic how neurons send signals to the brain and each other
- Networks of Neurons
 - Mathematically interesting connections between how individual neurons fire & network fn.
 - Epilepsy research

Thank you!

(Shoutout: particularly you, Aaron...)