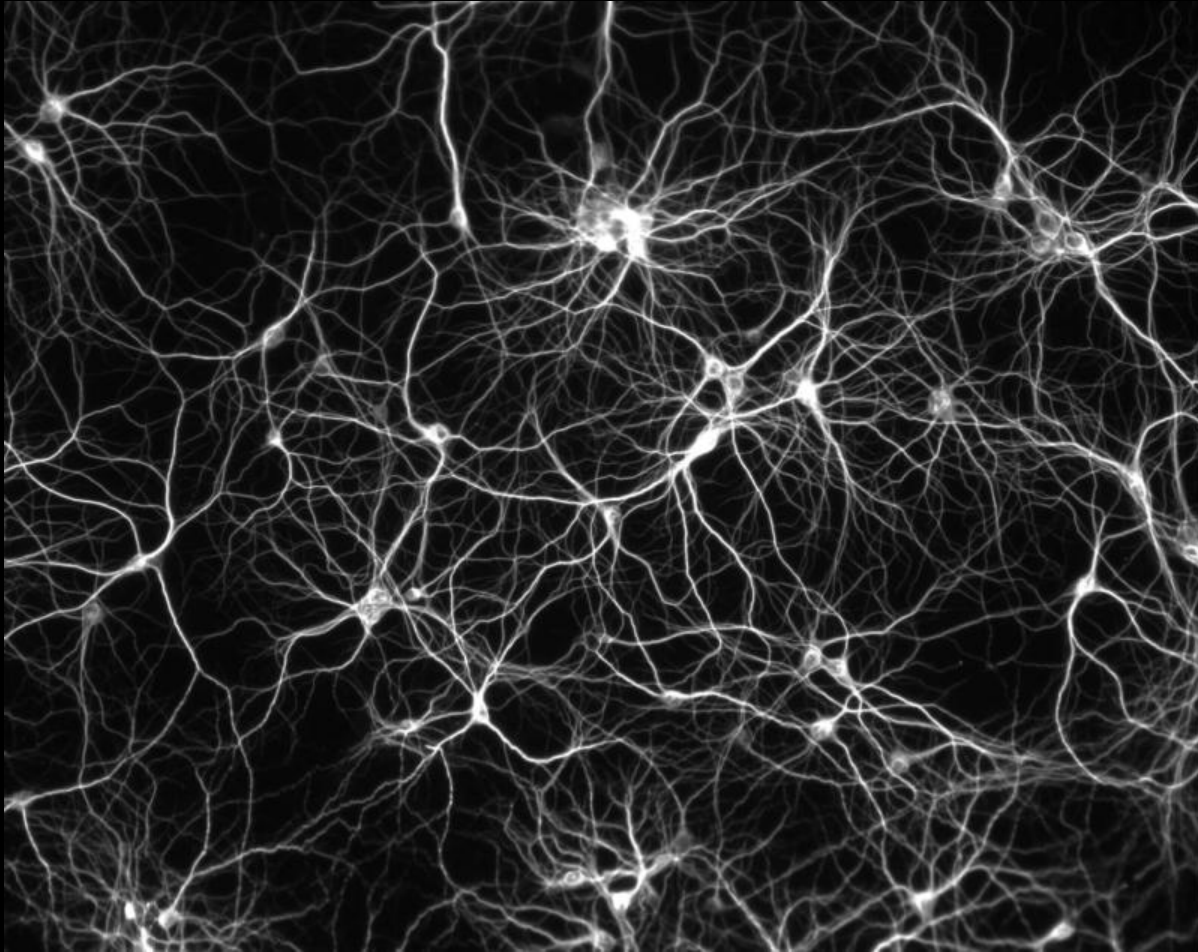


# Neural Physics: an overview



Cooper Mellema

# Topics

- Electrophysiology
- Channel Biophysics
- Information theory in neural networks

# Brief overview of Electrophysiology

- Describes the electrical properties of neurons and neural tissues
- Neurons are negative compared to the intercellular space
- Signals are transduced as a rolling wave of depolarization
- Neuron can be modeled as an RC circuit with different components of R and C modeled as differential equations

# How a neuron fires

- Low sodium inside, low potassium outside
- First, an input opens a number of sodium channels, causing the cell to depolarize
- Sodium channels are gated, and more open with sufficient depolarization, leading to a positive feedback loop of depolarization, then shut off
- Potassium channels open, leading to repolarization



# Hodgkin-Huxley Equation

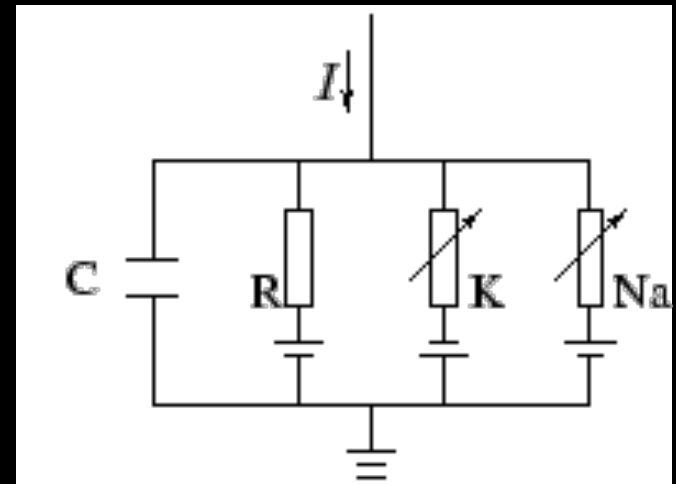
- 4 components: membrane capacitance, sodium resistance, potassium resistance, and resistance of all other channels

- First, we split the current into different components for later

$$I(t) = I_C(t) + \sum I_k(t)$$

- Then, we set up a differential equation whose parameters we can measure

$$C dV/dt = -\sum I_k(t) + I(t)$$



# Hodgkin-Huxley 2

- Each component of the current  $I_k(t)$  is described by a differential equation in turn:

$$\Sigma I_k = g_{Na} m^3 h (V - V_{Na}) + g_K n^4 (V - V_K) + g_L (V - E_L)$$

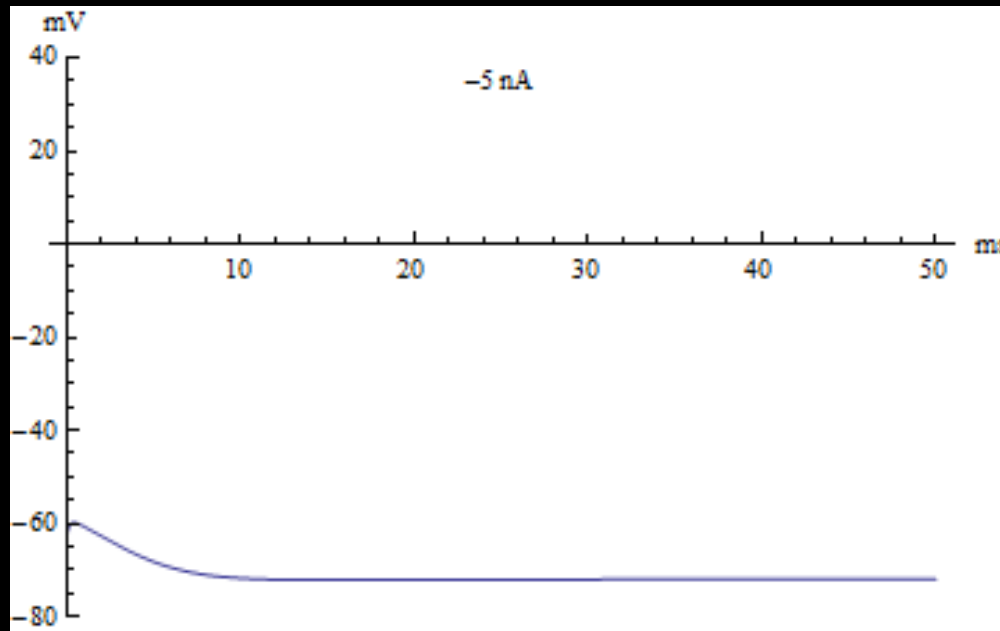
m and h represent the gating of sodium channels, n represents the gating of potassium channels. These are governed by the following equations:

$$dm/dt = f(V)(1-m) - g(V)m$$

$$dn/dt = f_2(V)(1-n) - g_2(V)n$$

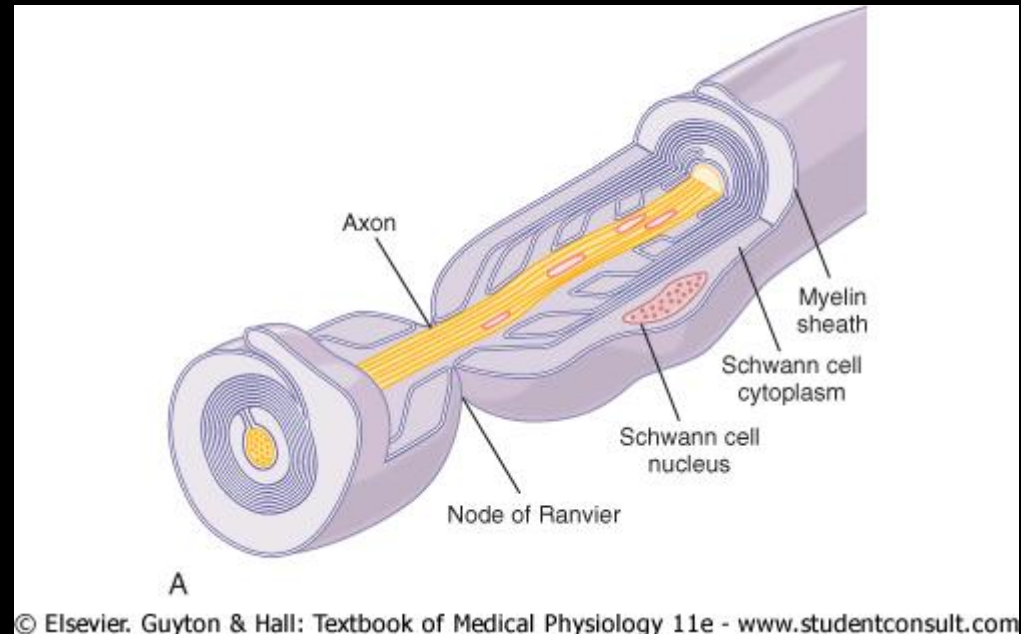
$$dh/dt = f_3(V)(1-h) - g_3(V)h$$

Result: Simple spiking model that has a frequency dependent on the input current



# Improvements on Hodgekin-Huxley:

- Add in additional channels rather than just “other resistances” (ex: chloride can be important)
- Rewrite equations in terms of both voltage *and* concentration
- Apply electronic cable theory to axons along with periodic myelination
- Many more, must decide what is useful





# Example Extension: Rose-Hindmarsh model (from X. Zhao et al.)

$$C dV/dt = -g(V)(V - V_{Na}) - g_{LR} R(V - V_K) - H + I_{ext}$$

$$dR/dt = -1/\tau_R [R - R_{\infty}(V)] \\ (V - V_H)]$$

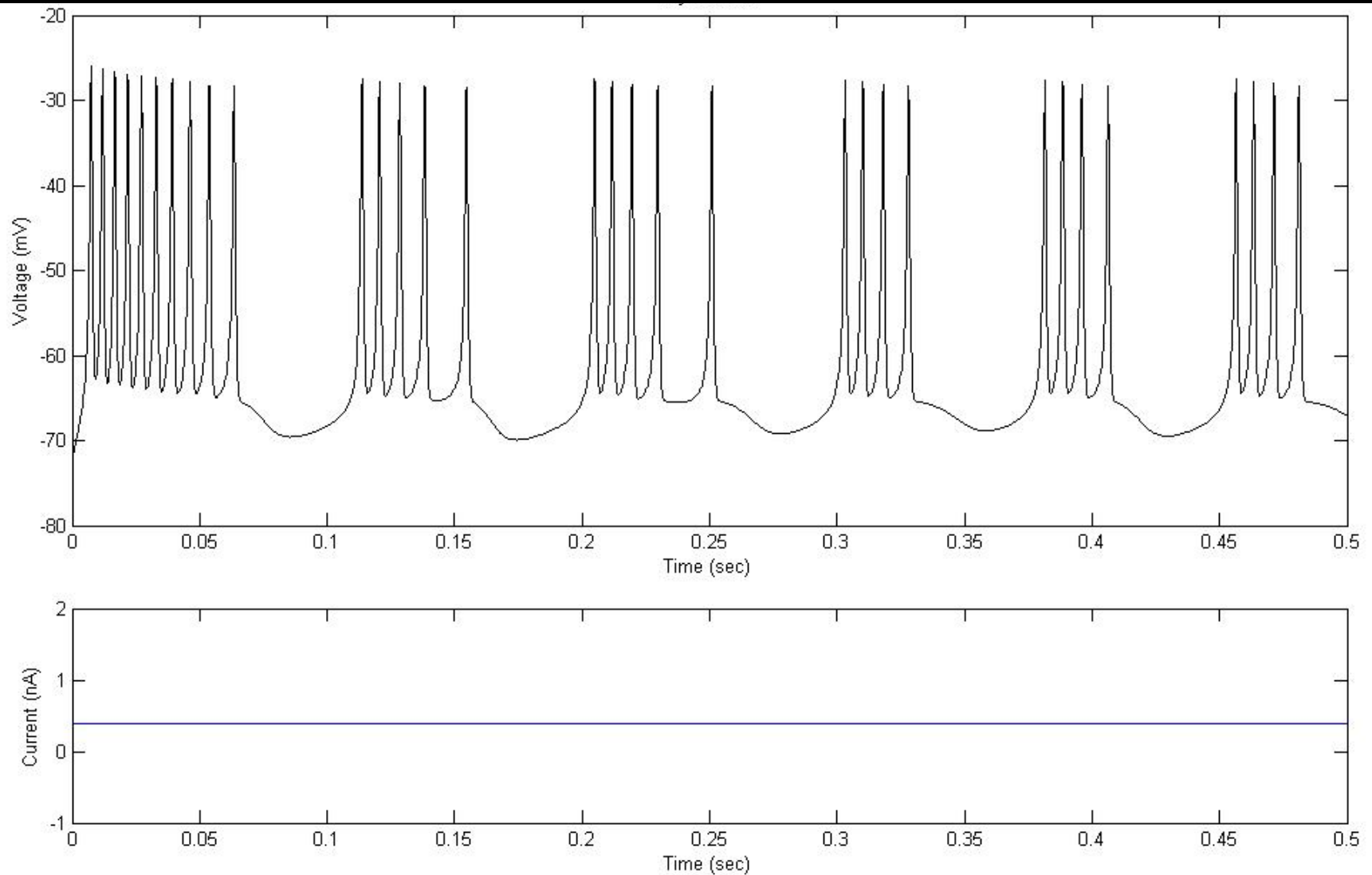
$$dH/dt = -1/\tau_H [H - g_H(V - V_H)]$$

$$g(V) = v_0 + v_1 V + v_2 V^2 \\ V^4)^2$$

$$R_{\infty}(V) = 0.17 + r_2 (V -$$

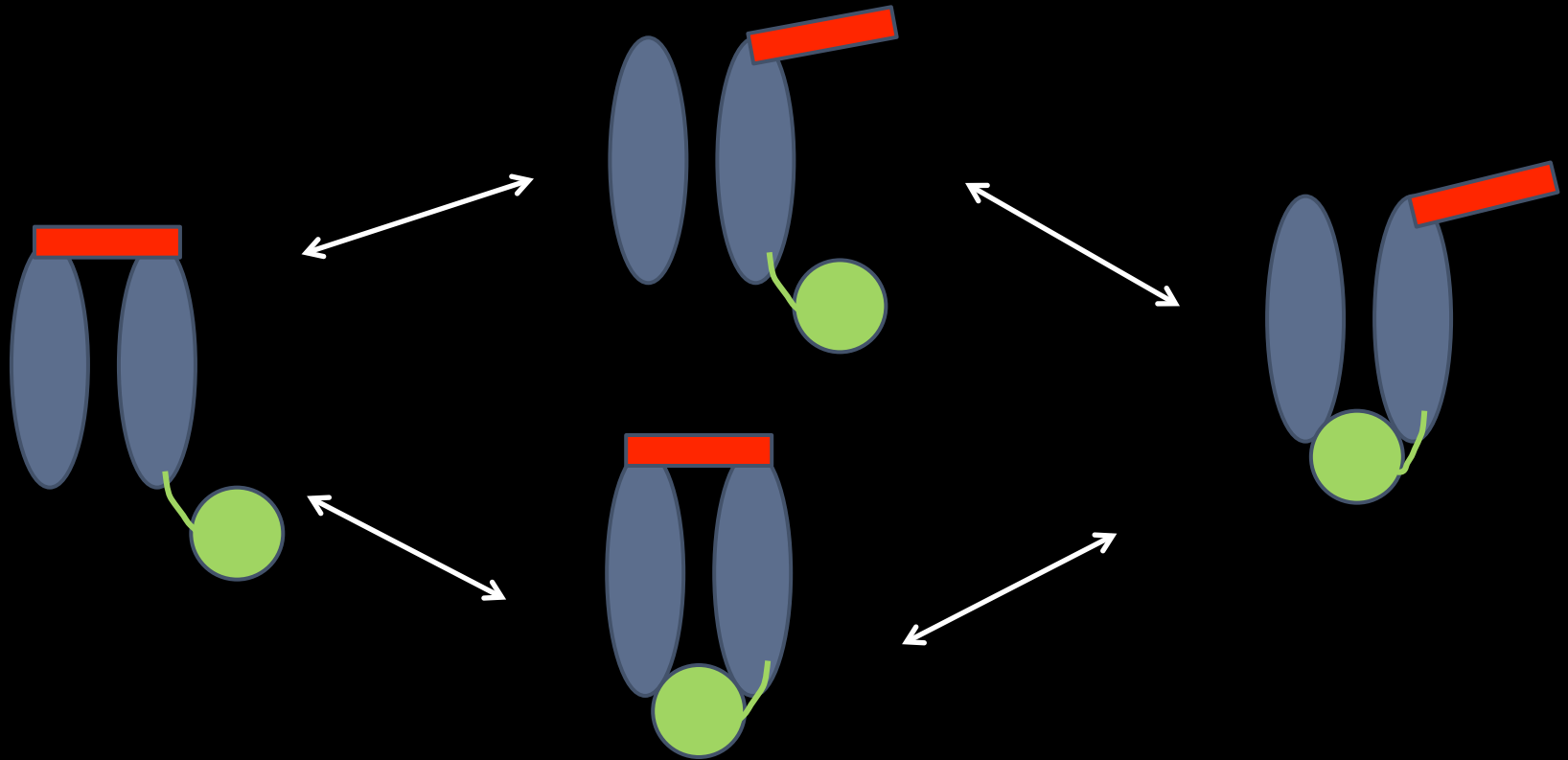
- Specific example of Hodgekin-Huxley with less general solutions

# Result: bursting pattern

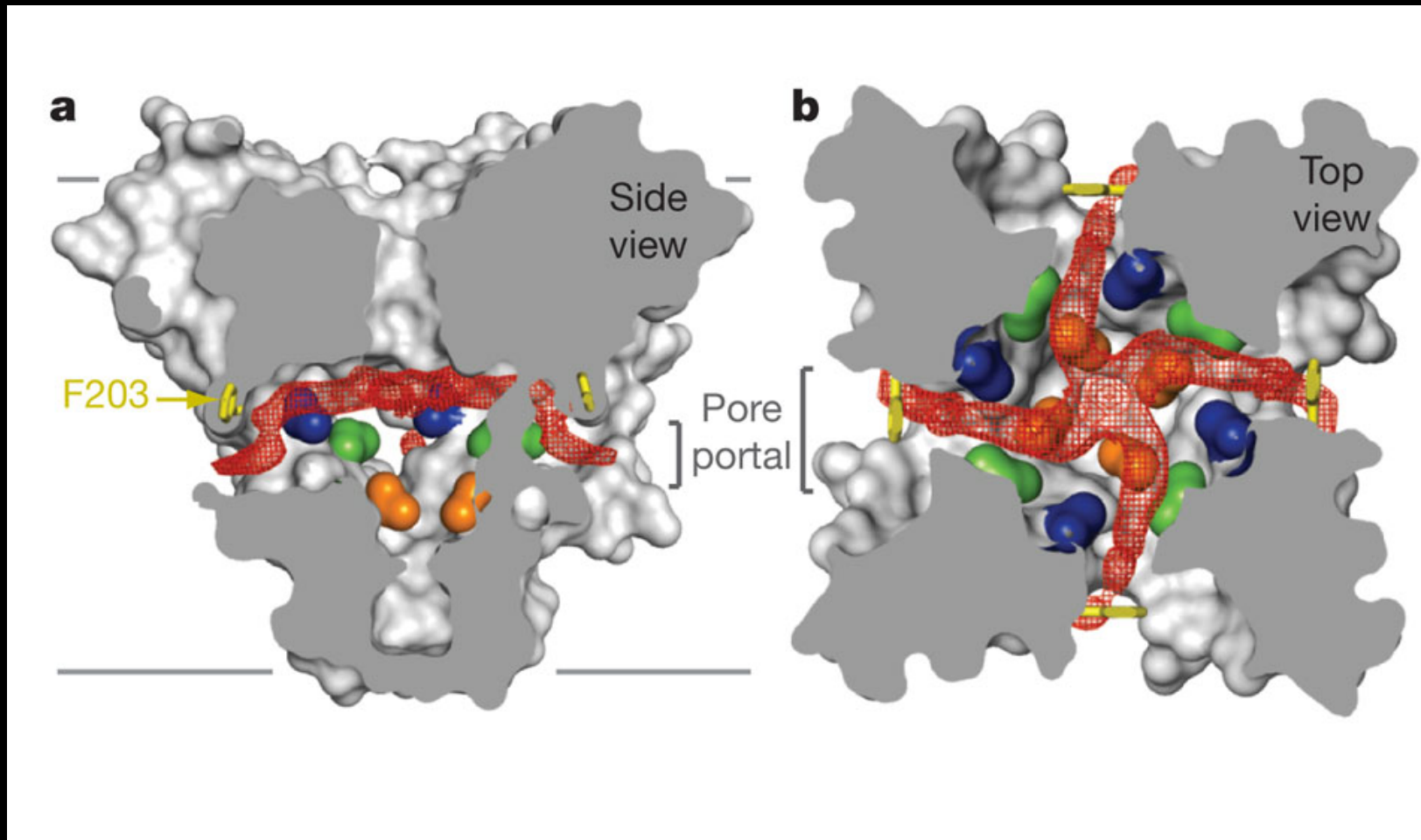


# Single Channel biophysics

- Sodium channel has 2 gates, and therefore 4 states, probability to go between states governed by voltage and temp: statistical mechanics approach

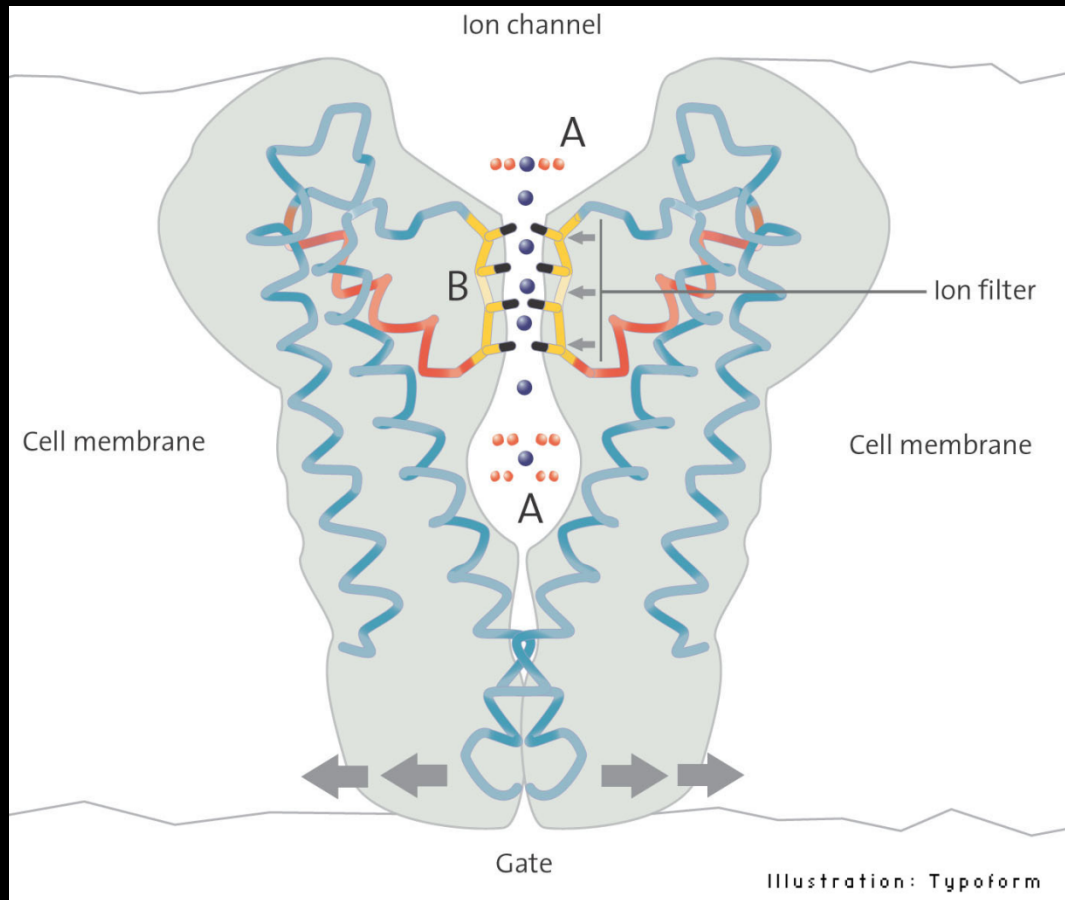


# Actual structure

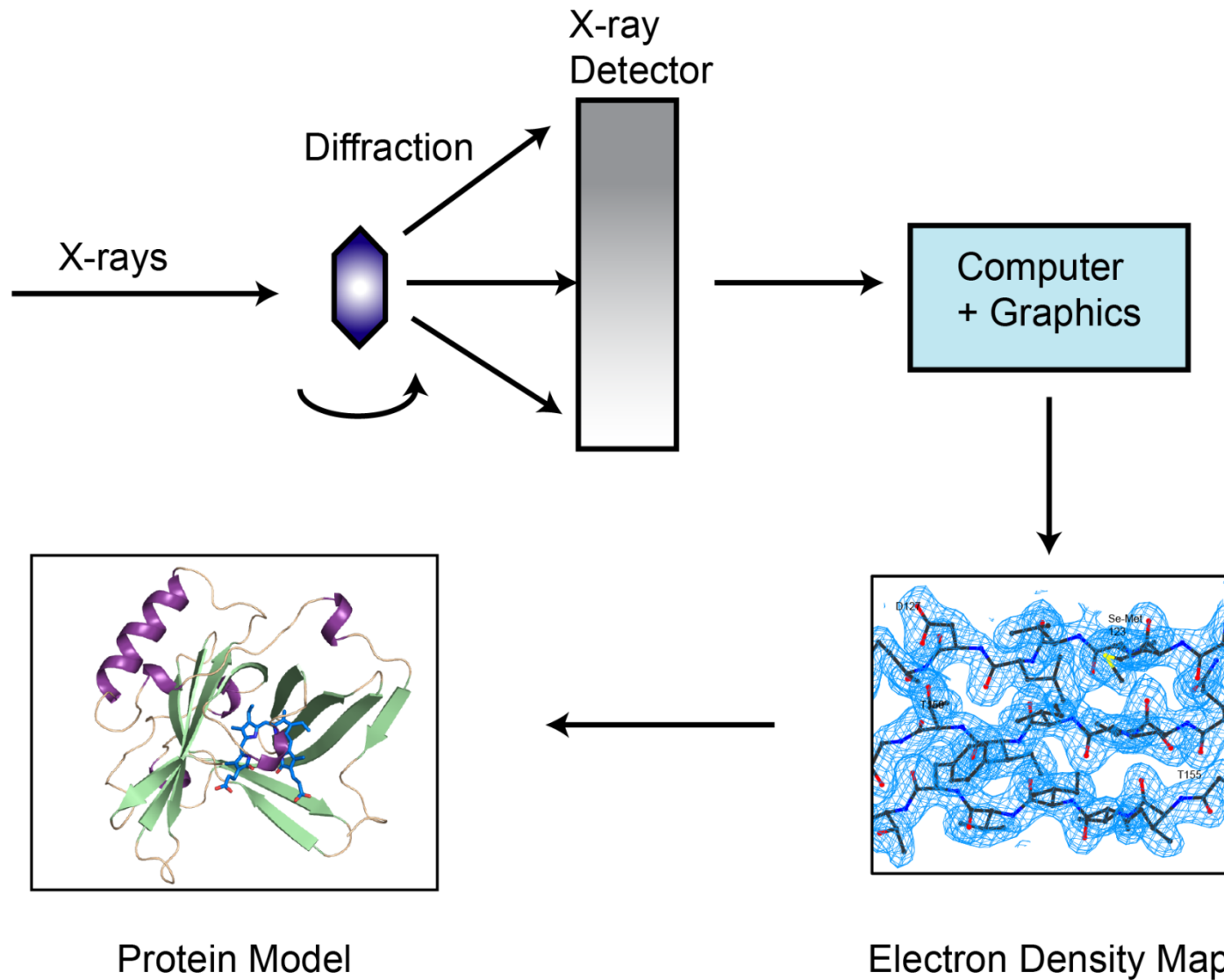


# Mechanism of Sodium Channel Activity

- Inner “neck” is positively charged
- With a change in voltage, neck twists open or closed
- Sodium let through by size exclusion, governed primarily by hydration energy

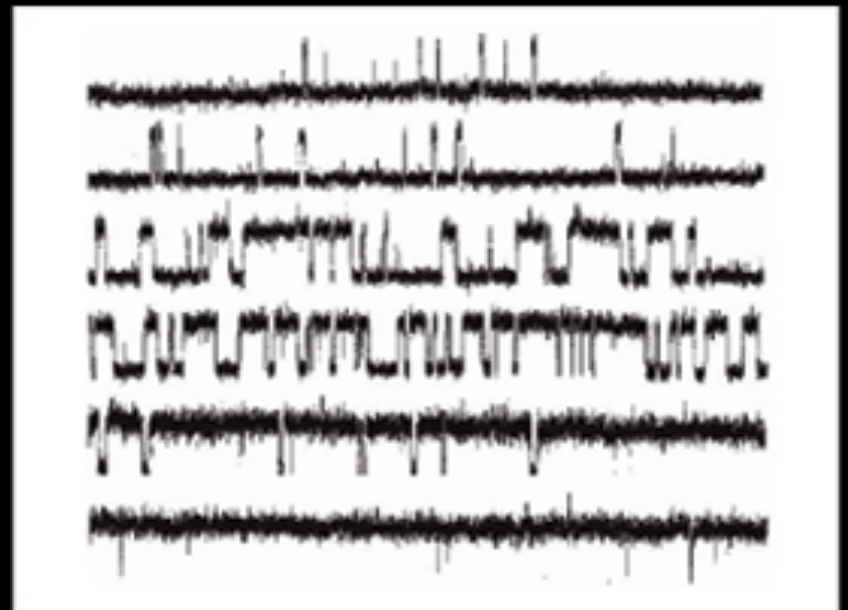


# Overview of the X-ray Crystallographic Method



# Patch Clamp

- Allows for determinations of single channel behavior



# Information theory

- Need to be able to quantify information able to be encoded by the brain
- Multiple approaches: I am going to cover the method where “information” is a change in entropy and look at a specific example from Urban and Padmanabhan (2010)
- Individual neurons vary within a network, and highly heterogeneous neurons can encode up to 2 times more information than a purely homogeneous system
- This means that intrinsic diversity is important for neuronal coding, and not just biological imprecision



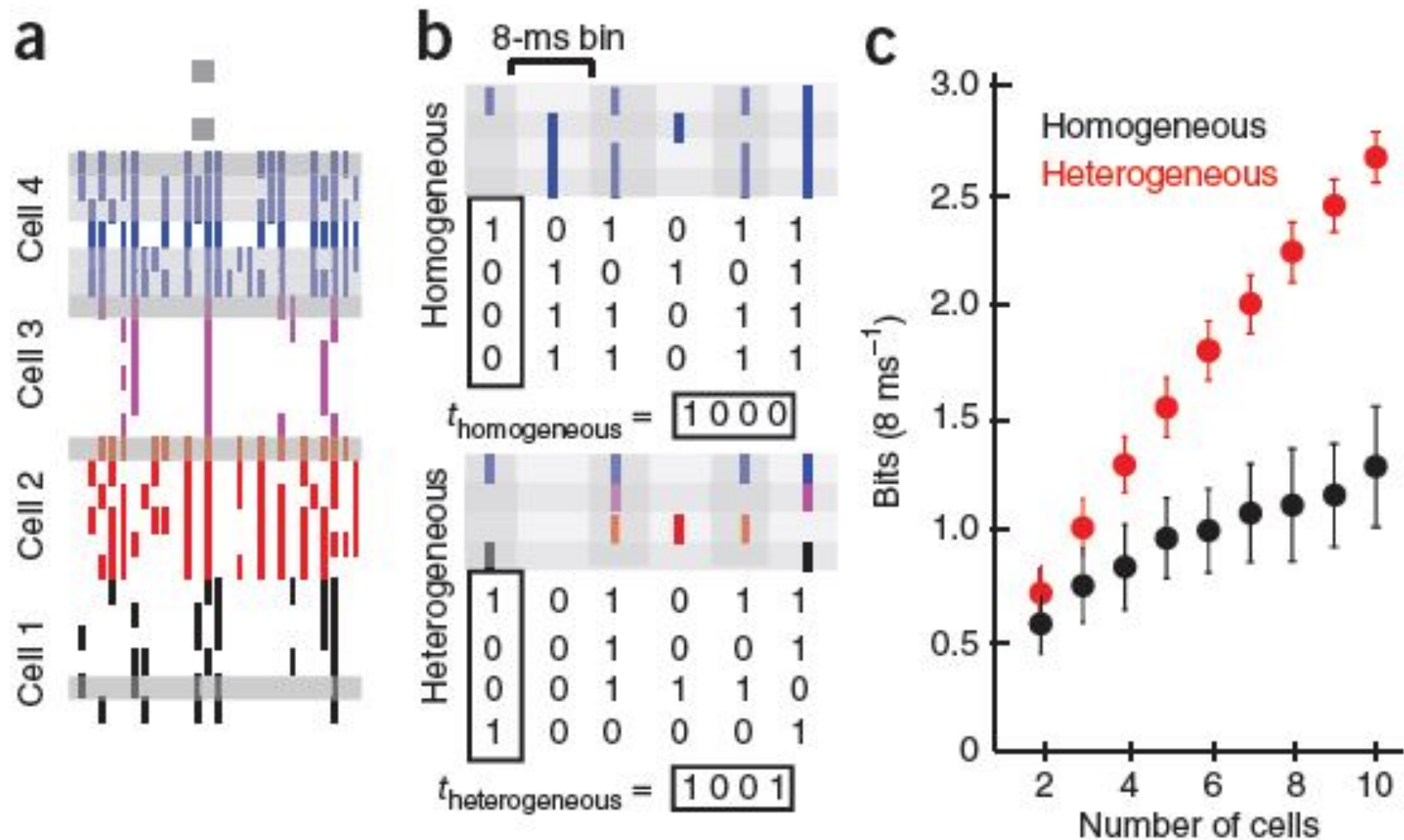
# Quantifying “Information”

- Using the statistical definition of entropy,

$$S = -A \sum_n P(n) * \ln[P(n)]$$

- $P(n)$  is the probability of a certain “word” of 0’s and 1’s based on the distribution of 0’s and 1’s generated.
- Information here is defined as entropy minus a noise factor

# Information Transfer



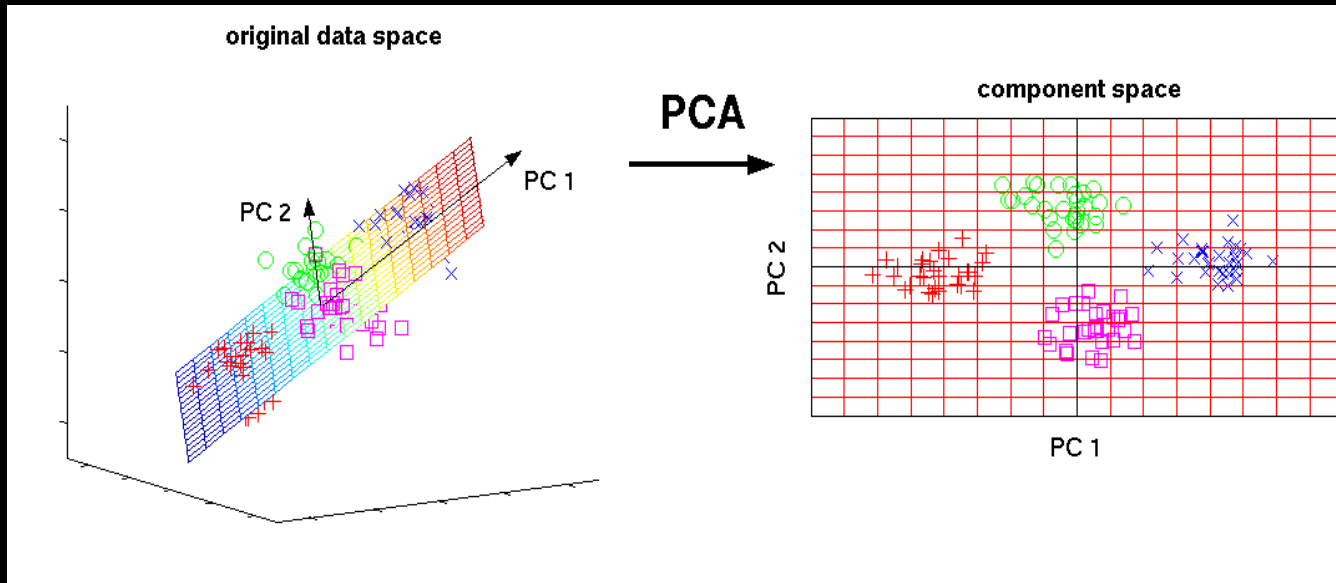
# Overcoming Noise and Chaotic behavior:

- Neurons are incredibly noisy, and sorting information from noise is incredibly difficult
- Solutions:
  - Matrix algebra analysis of overdetermined systems
  - Dimensional reduction techniques, such as PCA (principal component analysis)
  - Smoothing and filtering data

# Overdetermined Systems

- Problem when trying to decode neuronal systems and modeling them
- Far more solutions than unknowns, sometimes orders of magnitude more solutions than unknowns
- Approximate solution (ordinary least squares) for  $\mathbf{Ax}=\mathbf{b}$ :  
find Min. of  $\|\mathbf{Ax}-\mathbf{b}\|$   
given by  $\mathbf{x}=(\mathbf{A}^t\mathbf{A})^{-1}\mathbf{A}^t\mathbf{b}$

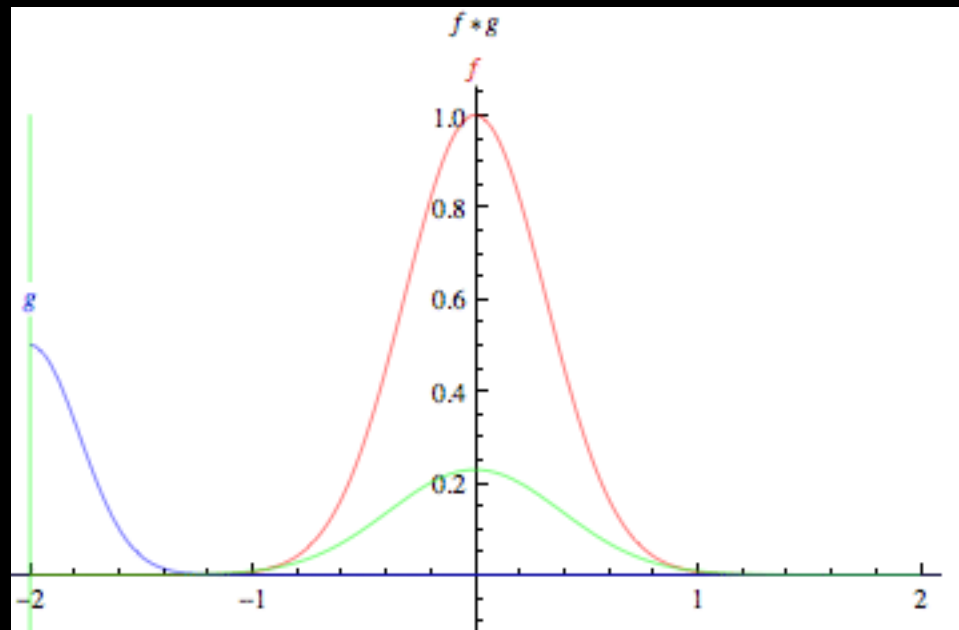
# PCA: Principal Component Analysis



- Repeatedly project a high dimensional space onto a lower dimensional space while maximizing the variance
- Principal components are eigenvectors of covariance matrix, these are our basis vectors

# Smoothing and Filtering

- In practice, much of the calculations and computations are simpler if you smooth the data
- Basic filters can do anything from averaging windows of data to convoluting Gaussians with data



Questions