

Highlights of our journey thus far...

♦ Neuroscience Review

Neurons, synapses, and brain regions

♦ Neural Encoding

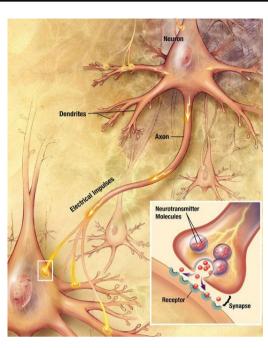
- ❖ What makes a neuron fire? (STA, covariance analysis)
- Poisson model of spiking

♦ Neural Decoding and Information Theory

- Stimulus discrimination and signal detection
- Population decoding and Bayesian estimation
- Information and neural coding principles

♦ Single Neuron Models

- RC circuit model of membrane
- Hodgkin-Huxley and compartmental models
- ❖ Integrate-and-fire and simplified neuron models



How do neurons connect to form networks?

They use synapses!

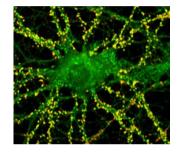
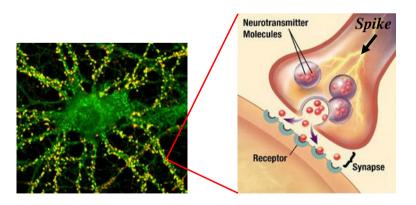


Image Source: Wikimedia Commons

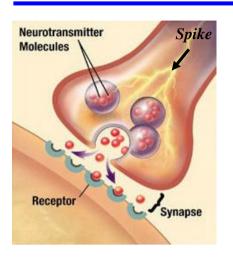
What do synapses do?



Increase or decrease postsynaptic membrane potential

1

An Excitatory Synapse

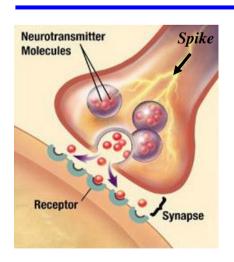


Input spike →
Neurotransmitter release
(e.g., Glutamate) →
Binds to receptors →
Ion channels open →
positive ions (e.g. Na+)
enter cell →
Depolarization
(increases local
membrane potential)

5

Image Source: Wikimedia Commons

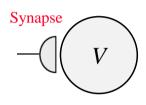
An Inhibitory Synapse



Input spike →
Neurotransmitter
release (e.g., GABA)
→ Binds to receptors
→ Ion channels open
→ positive ions (e.g.,
K+) leave cell →
Hyperpolarization
(decreases local
membrane potential)

We want a *computational* model of the effects of a synapse on the membrane potential V

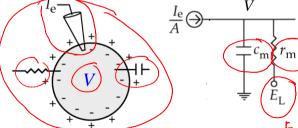




How do we do this?

7

Flashback RC Circuit Model of the Membrane



$$c_{\rm m} \approx 10 \, \rm nF/mm^2$$

$$r_{\rm m} \approx 1 \, {\rm M}\Omega \, {\rm mm}^2$$

$$C_{\rm m} = c_{\rm m} A$$

$$R_{\rm m} = r_{\rm m}/A$$

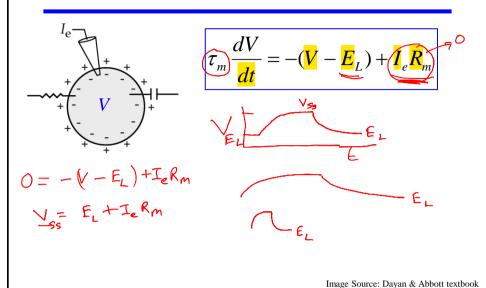
$$c_m \frac{dV}{dt} = -\frac{(V - E_L)}{r_m} + \frac{I_e}{A}$$
 or equivalently:

$$\tau_m = r_m c_m = R_m C_m$$
 is the membrane time constant

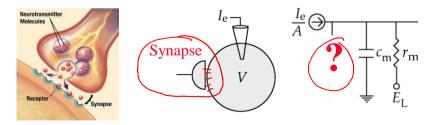
$$\tau_m \frac{dV}{dt} = -(V - E_L) + I_e R_m$$

Image Source: Dayan & Abbott textbook

What is this equation really saying?

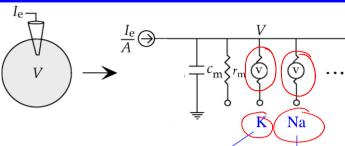


How do we model the effects of a synapse on the membrane potential V?





Hodgkin-Huxley Model



$$\tau_m \frac{dV}{dt} = -i_m r_m + I_e R_m$$

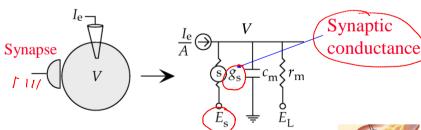
$$i_m = (1/r_m)(V - E_L) + g_{K,\text{max}}n^4(V - E_K) + g_{Na,\text{max}}m^3h(V - E_{Na})$$

$$E_L = -54 \text{ mV}, E_K = -77 \text{ mV}, E_{Na} = +50 \text{ mV}$$

11

Image Source: Dayan & Abbott textbook

Modeling Synaptic Inputs



$$\tau_m \frac{dV}{dt} = -((V - E_L) + g_s(V - E_s)r_m) + I_e R_m$$



 $g_s = g_{s,\text{max}} P_{re} P_s$ — Probability of postsynaptic channel opening (= fraction of channels opened)

Probability of transmitter release given an input spike

12

Basic Synapse Model



$$g_s = g_{s,\max} P_{rel} P_s$$

- ightharpoonup Assume $P_{rel} = 1$
- ♦ Model the effect of a single spike input on P_s

fraction of channels opened

★ Kinetic Model of postsynaptic channels:

 $\xrightarrow{\alpha_{s}}$ Open

 $\frac{dP_s}{dt} = \alpha_s (1 - P_s) - \beta_s P_s$ Opening rate

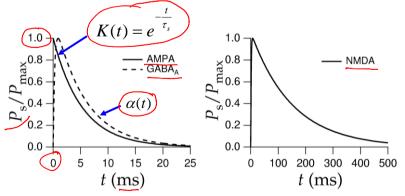
Closing rate

Fraction of channels closed

Fraction of channels open

13

What does P_s look like over time given a spike?



Exponential function gives reasonable fit for some synapses

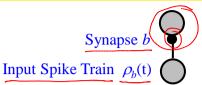
Others can be fit using "Alpha" function:

 $\alpha(t) = \frac{t}{\tau_{peak}} \cdot e^{\left(1 - \frac{t}{\tau_{peak}}\right)}$

 $\begin{array}{c|c}
1 & & \\
0 & \tau_{neel} & t
\end{array}$

Image Source: Dayan & Abbott textbook





 $\rho_b(t) = \Sigma_i \delta(t - t_i)$ (t_i are the input spike times, δ = delta function)

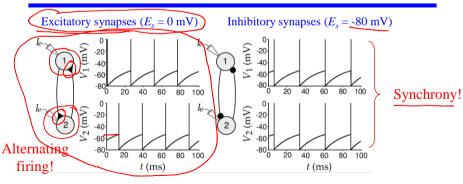
Filter for synapse
$$b = K(t)$$

Synaptic conductance at *b*:

$$g_b(t) = g_{b,\max} \sum_{t_i < t} K(t - t_i)$$

$$= g_{b,\max} \int_{-\infty}^{t} K(t-\tau) \underline{\rho_b(\tau)} d\tau$$

Example: Network of Integrate-and-Fire Neurons



Each neuron:
$$\tau_m \frac{dV}{dt} = -((V - E_L) - g_s(t)(V - E_s)r_m) + I_e R_m$$

Synapses (Alpha function)
$$E_L = -70 \text{ mV}$$
 $V_{thresh} = -54 \text{ mV}$ $\tau_m = 20 \text{ ms}$ $\tau_{peak} = 10 \text{ ms}$ $I_e R_m = 25 \text{ mV}$

Image Source: Dayan & Abbott textbook