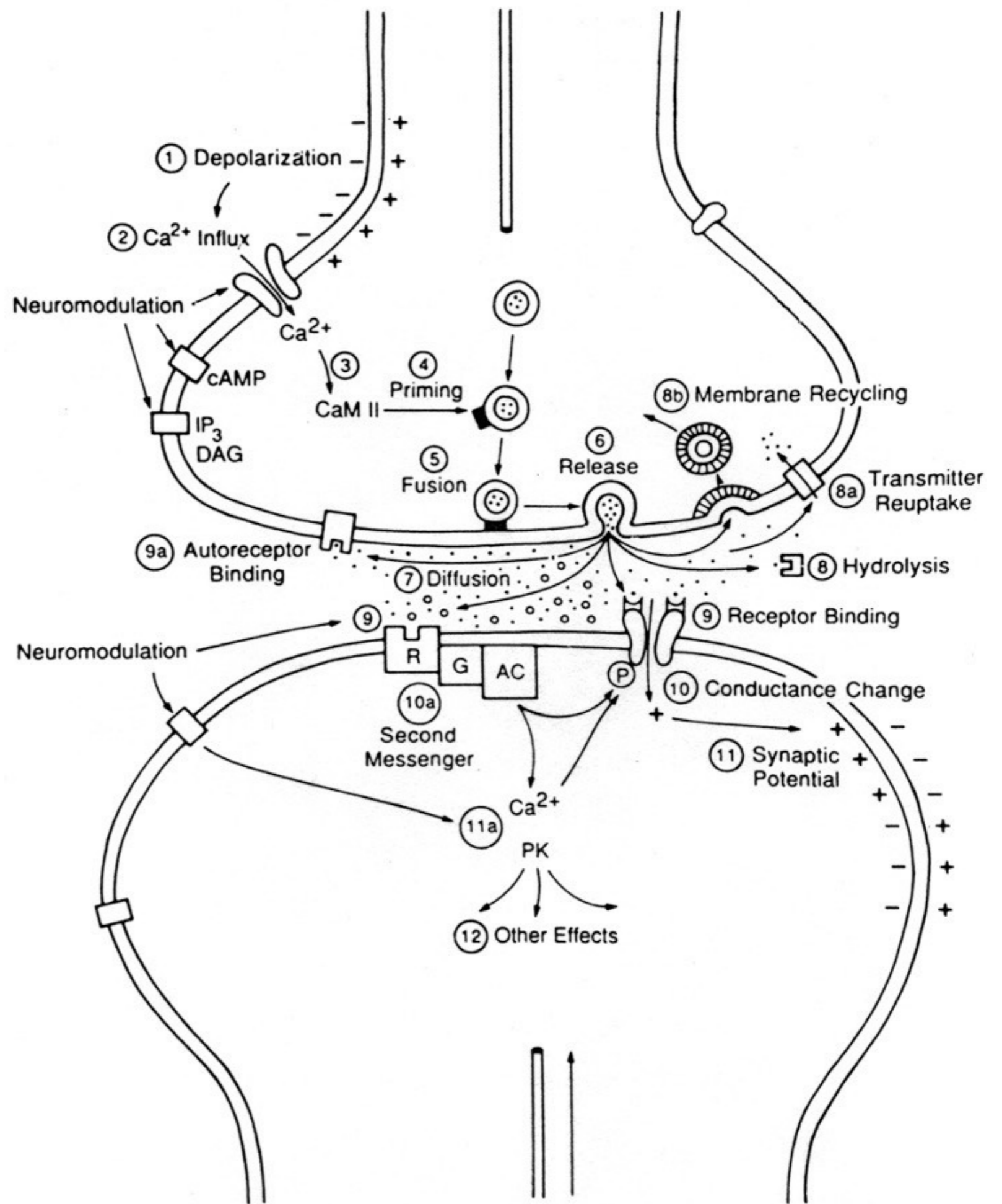
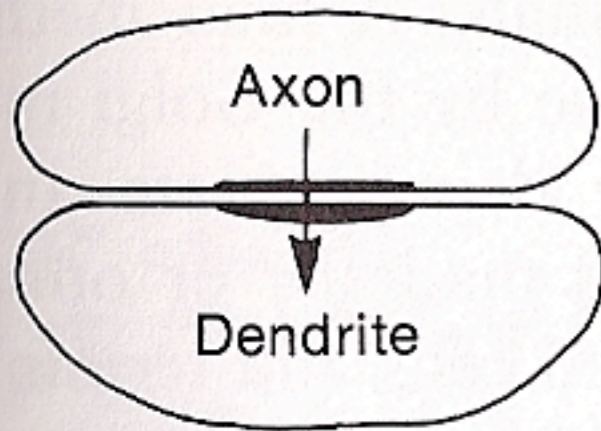


Anatomy of a synapse

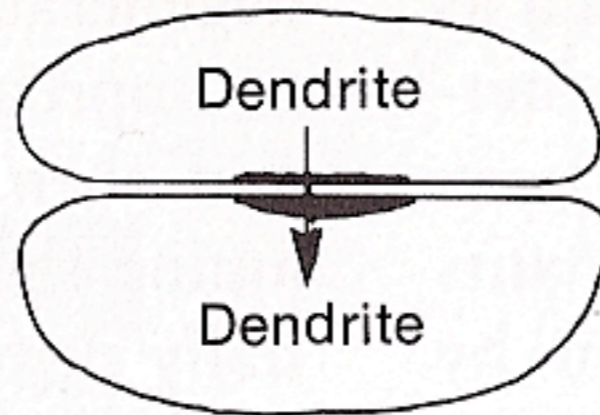


THE SYNAPSE

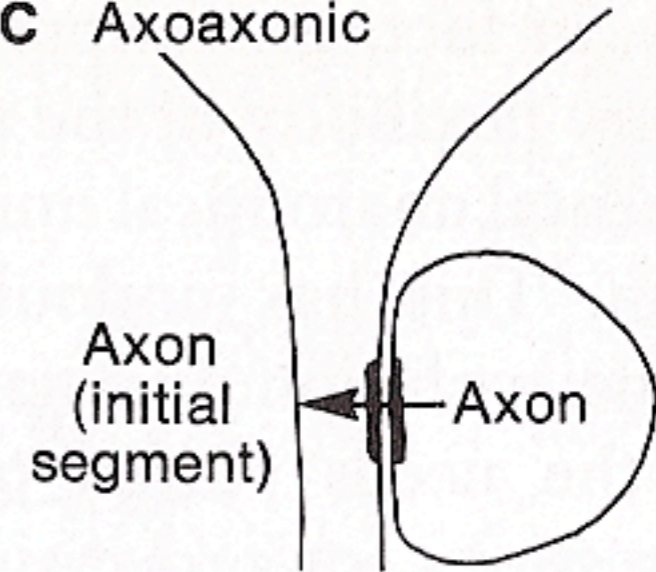
A Axodendritic



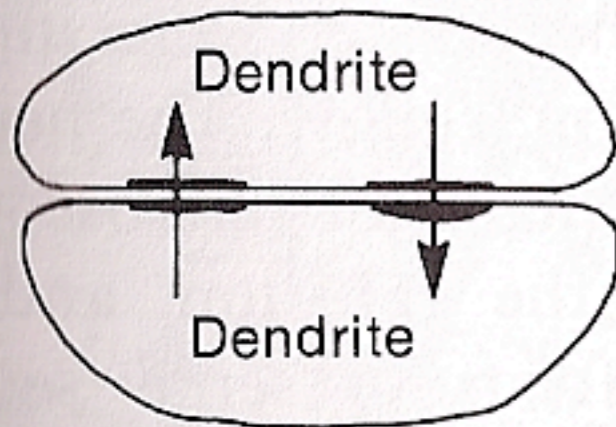
B Dendrodendritic



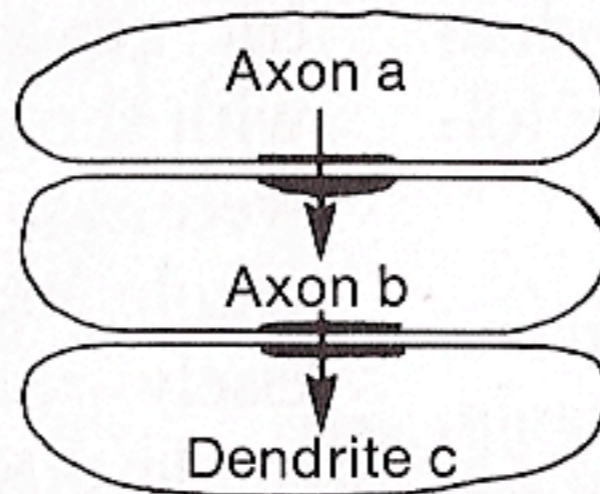
C Axoaxonic



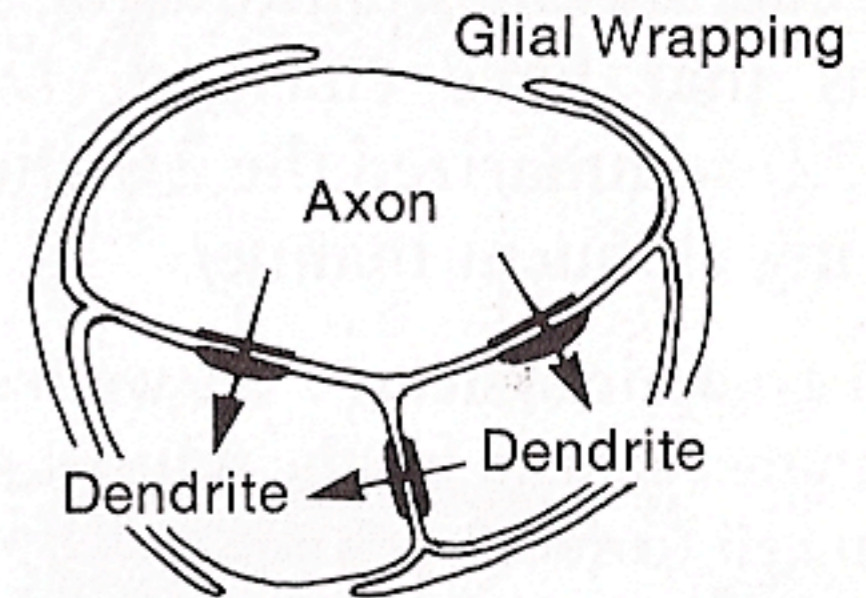
D Reciprocal Synapses



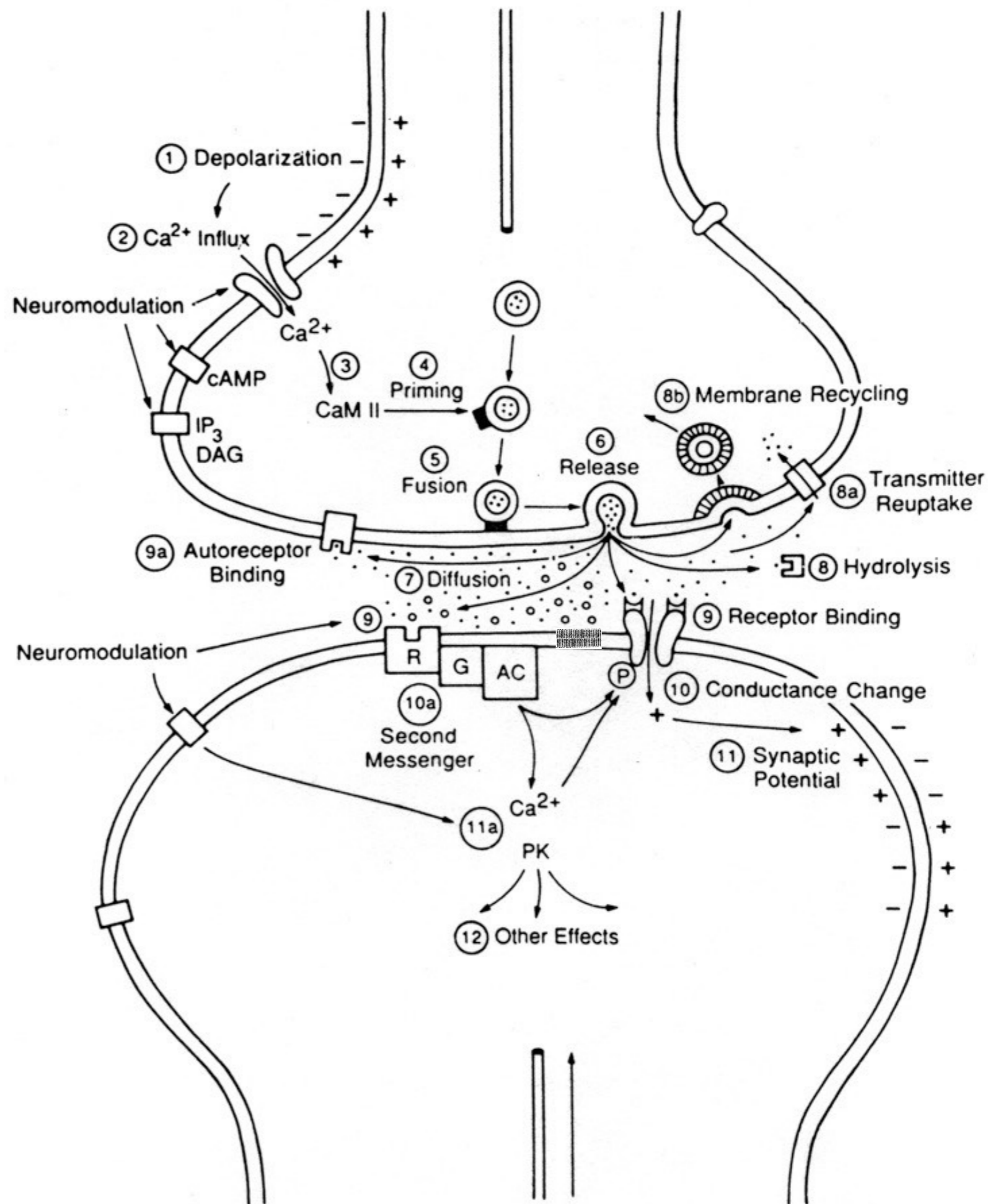
E Serial Synapses
(Axoaxodendritic)



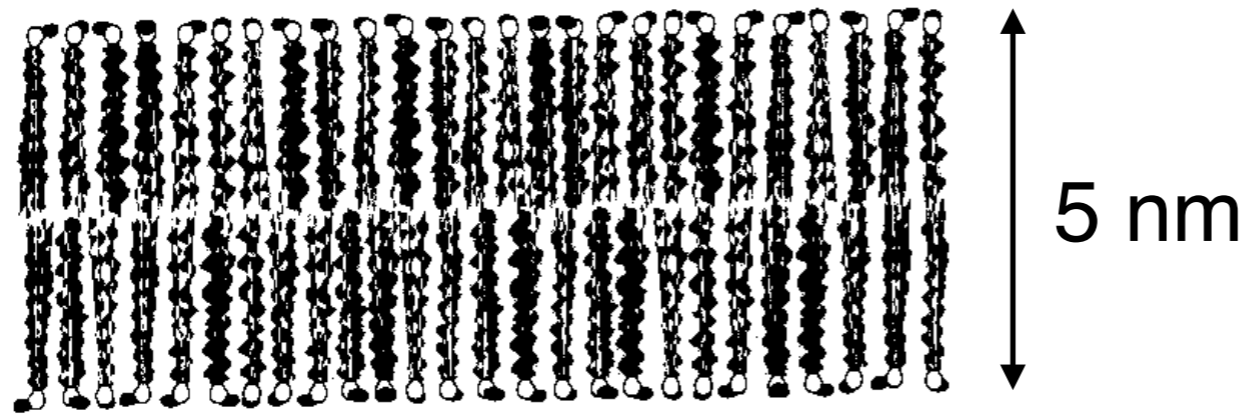
F Synaptic Glomerulus



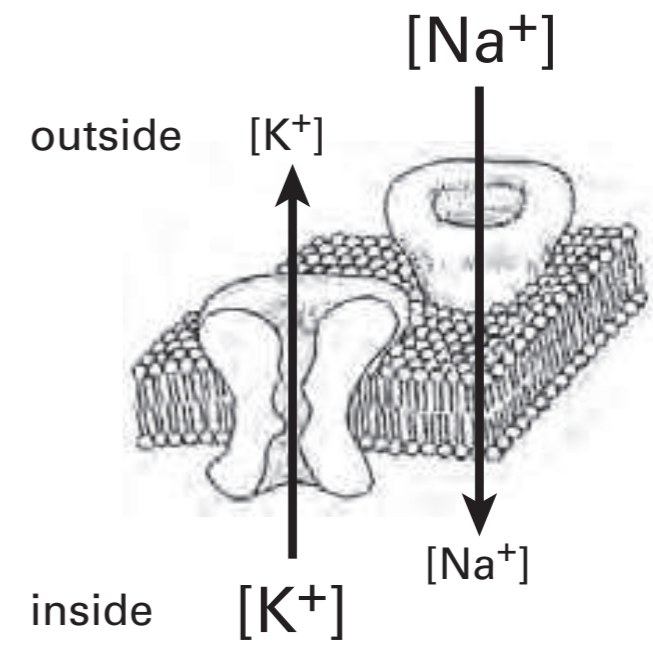
Anatomy of a synapse

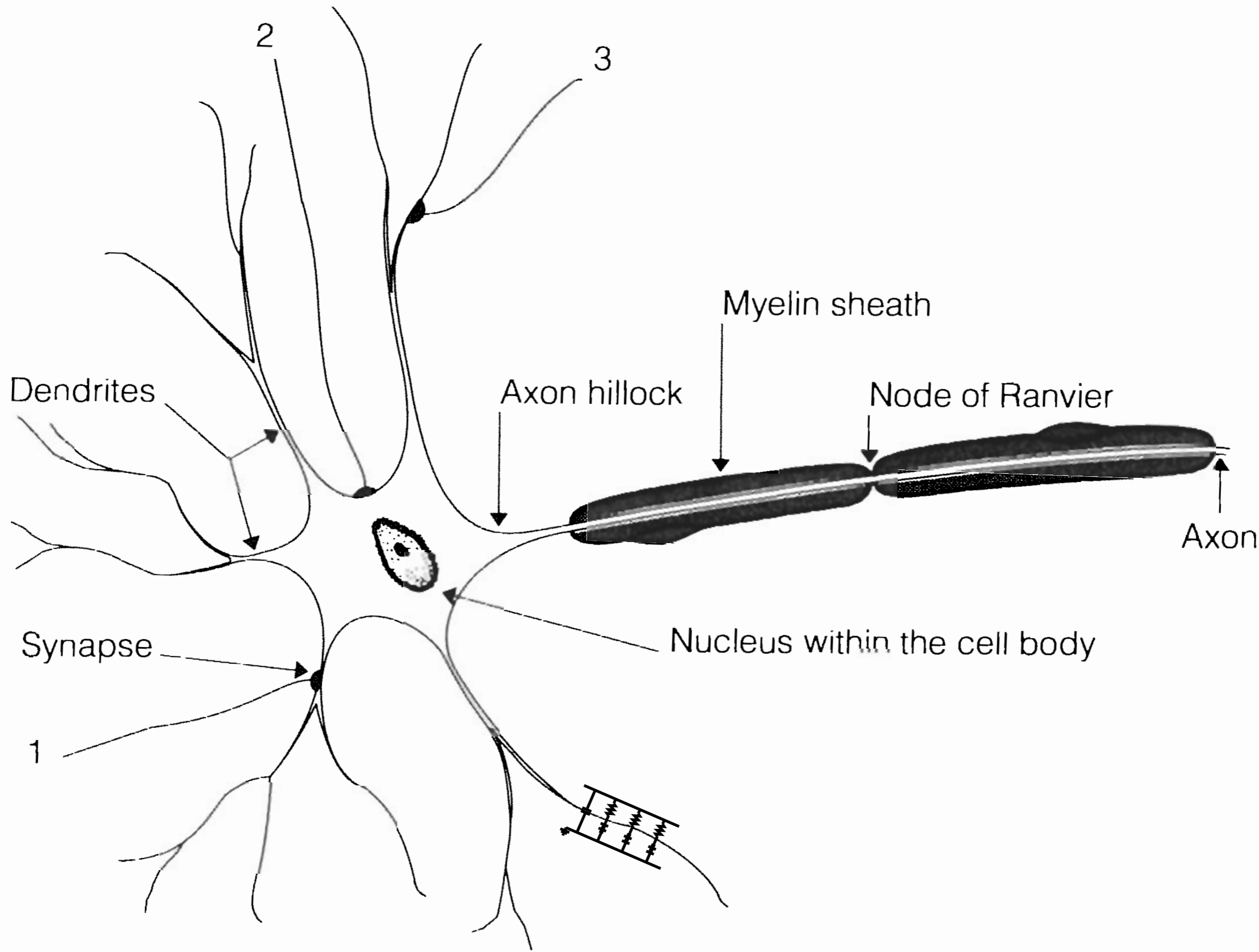


neuron membrane

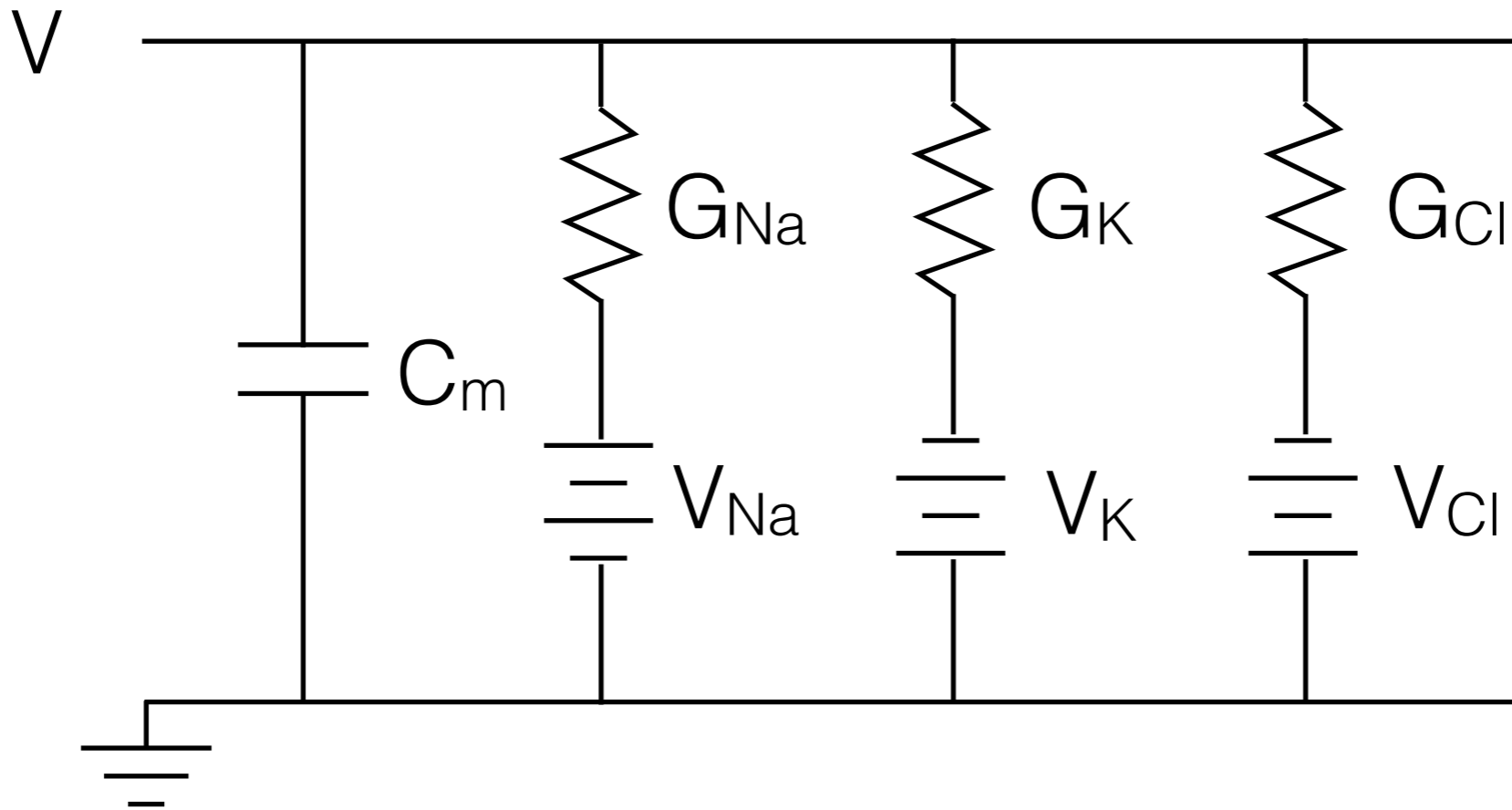


ion channels





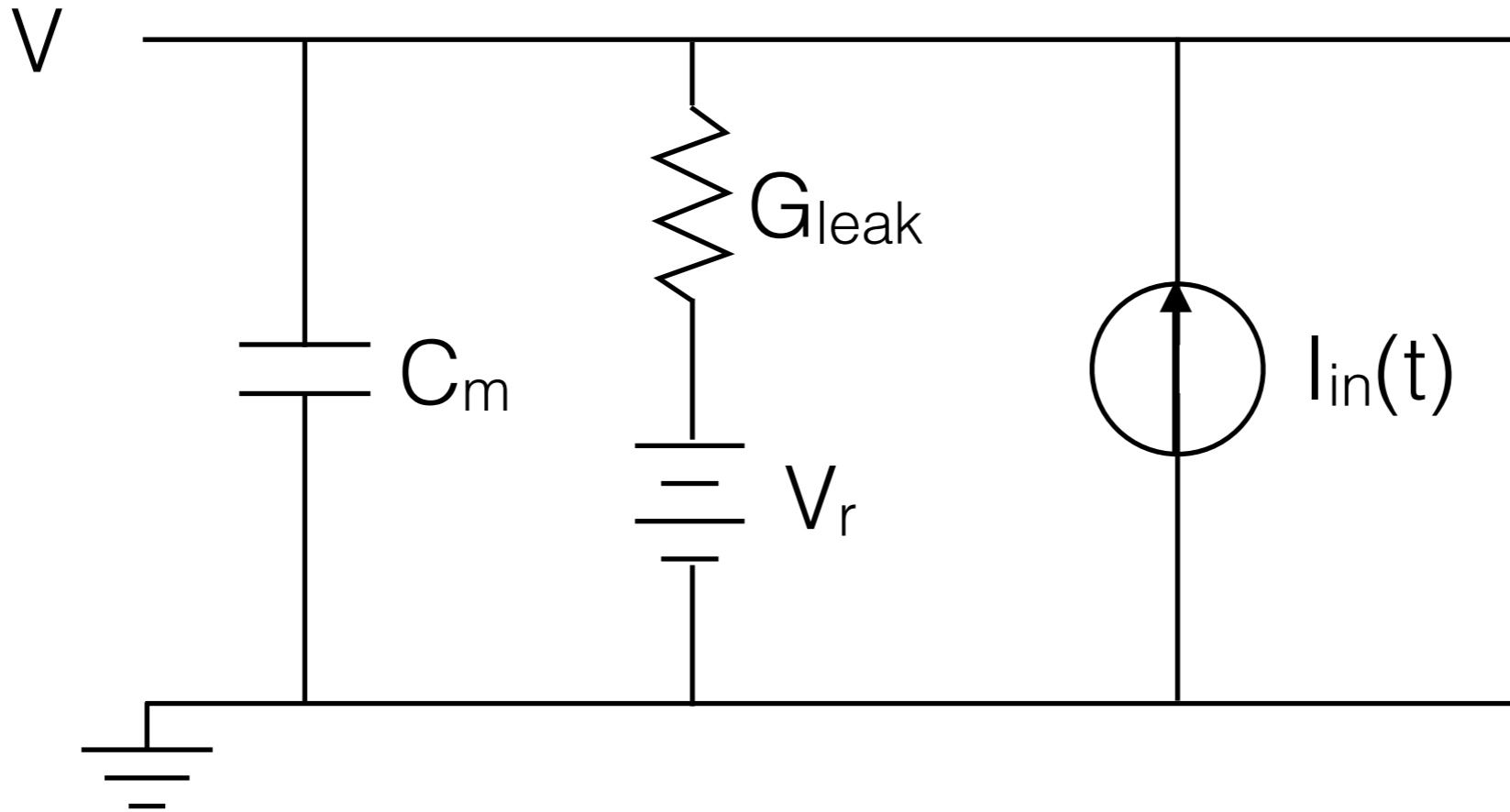
Membrane equation



$$\tau \frac{dV}{dt} + V = \frac{V_{Na} G_{Na} + V_K G_K + V_{Cl} G_{Cl}}{G_{total}}$$

$$G_{total} = G_{Na} + G_K + G_{Cl} \quad \tau = \frac{C_m}{G_{total}}$$

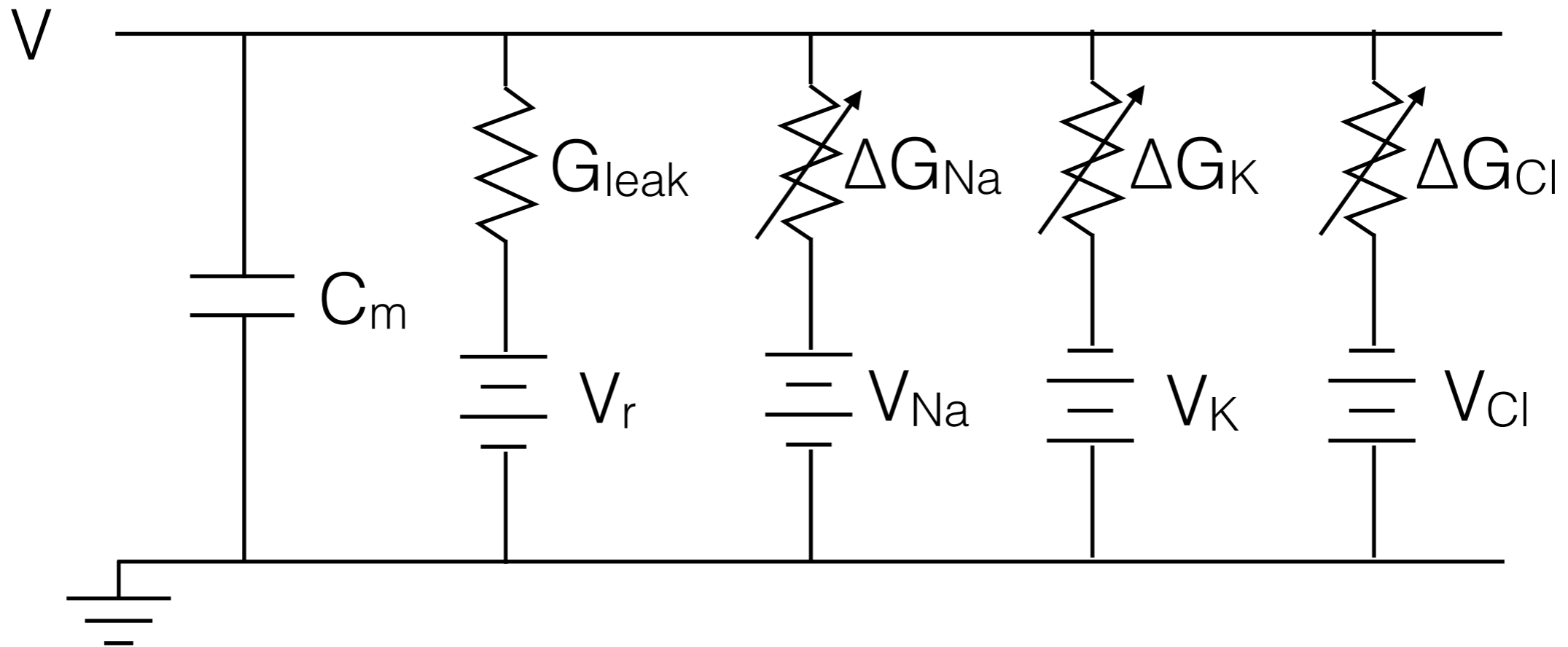
Membrane with input current



$$\tau \frac{dV}{dt} + V = V_r + \frac{1}{G_{\text{leak}}} I_{\text{in}}(t)$$

$$G_{\text{leak}} = G_{Na} + G_K + G_{Cl} \quad \tau = \frac{C_m}{G_{\text{leak}}}$$

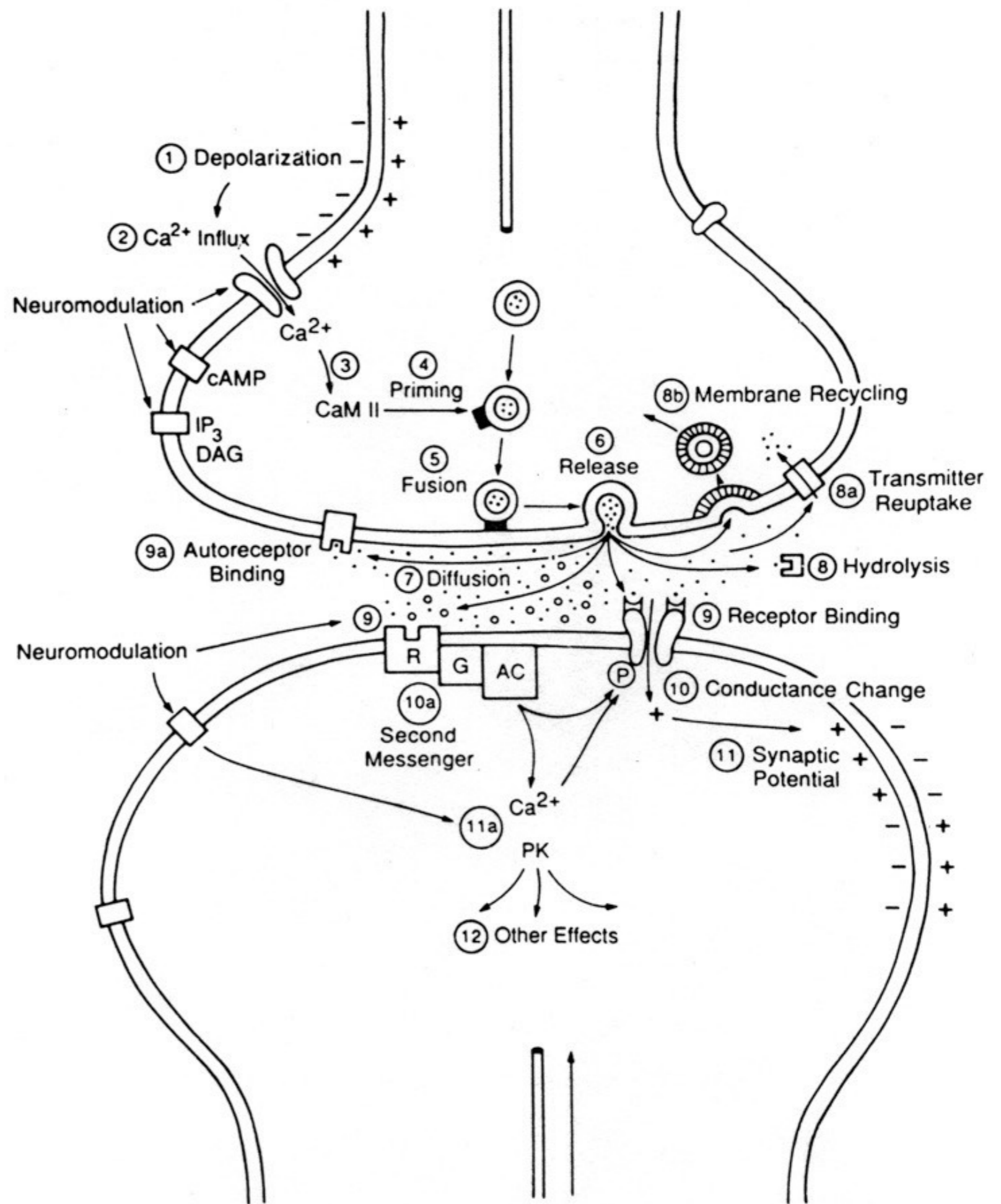
Membrane with synaptic inputs



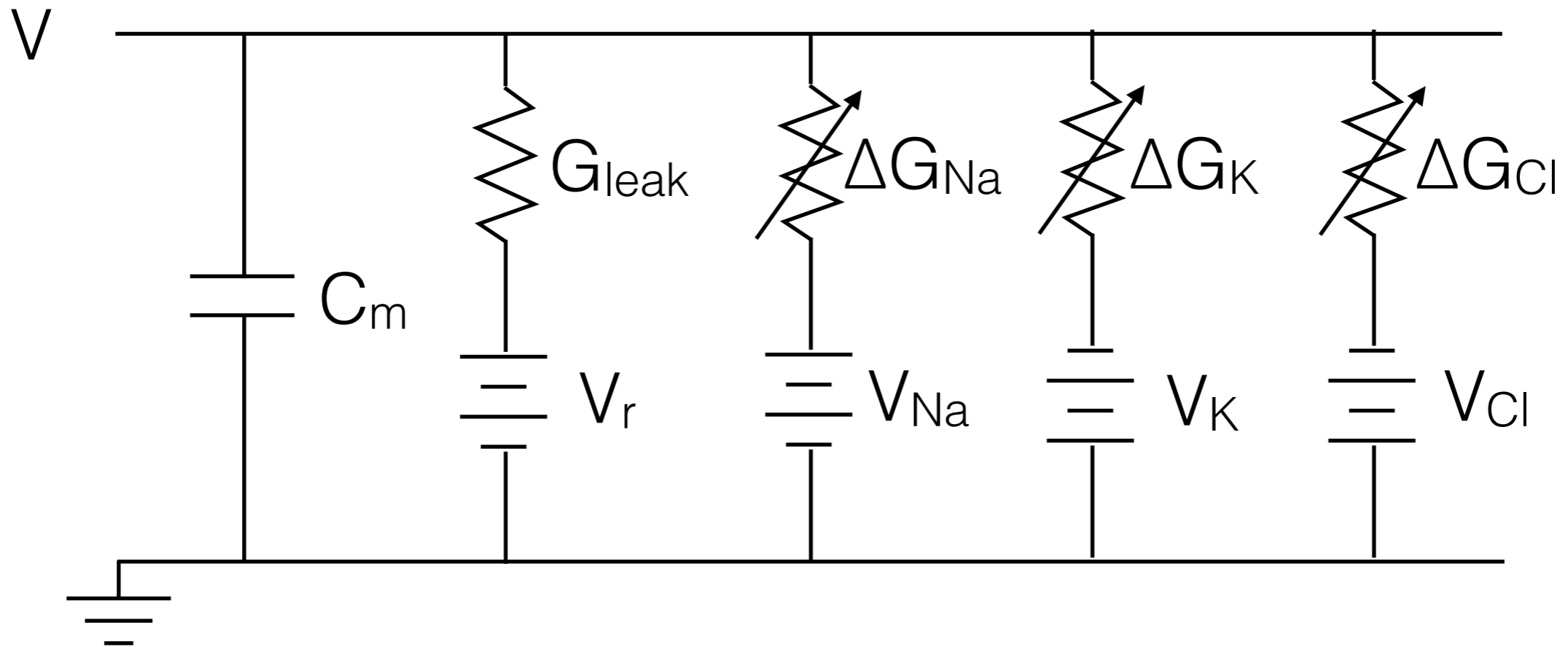
$$\tau \frac{dV}{dt} + V = \frac{V_r G_{\text{leak}} + V_{\text{Na}} \Delta G_{\text{Na}} + V_{\text{K}} \Delta G_{\text{K}} + V_{\text{Cl}} \Delta G_{\text{Cl}}}{G_{\text{total}}}$$

$$G_{\text{total}} = G_{\text{leak}} + \Delta G_{\text{Na}} + \Delta G_{\text{K}} + \Delta G_{\text{Cl}} \quad \tau = \frac{C_m}{G_{\text{total}}}$$

Anatomy of a synapse

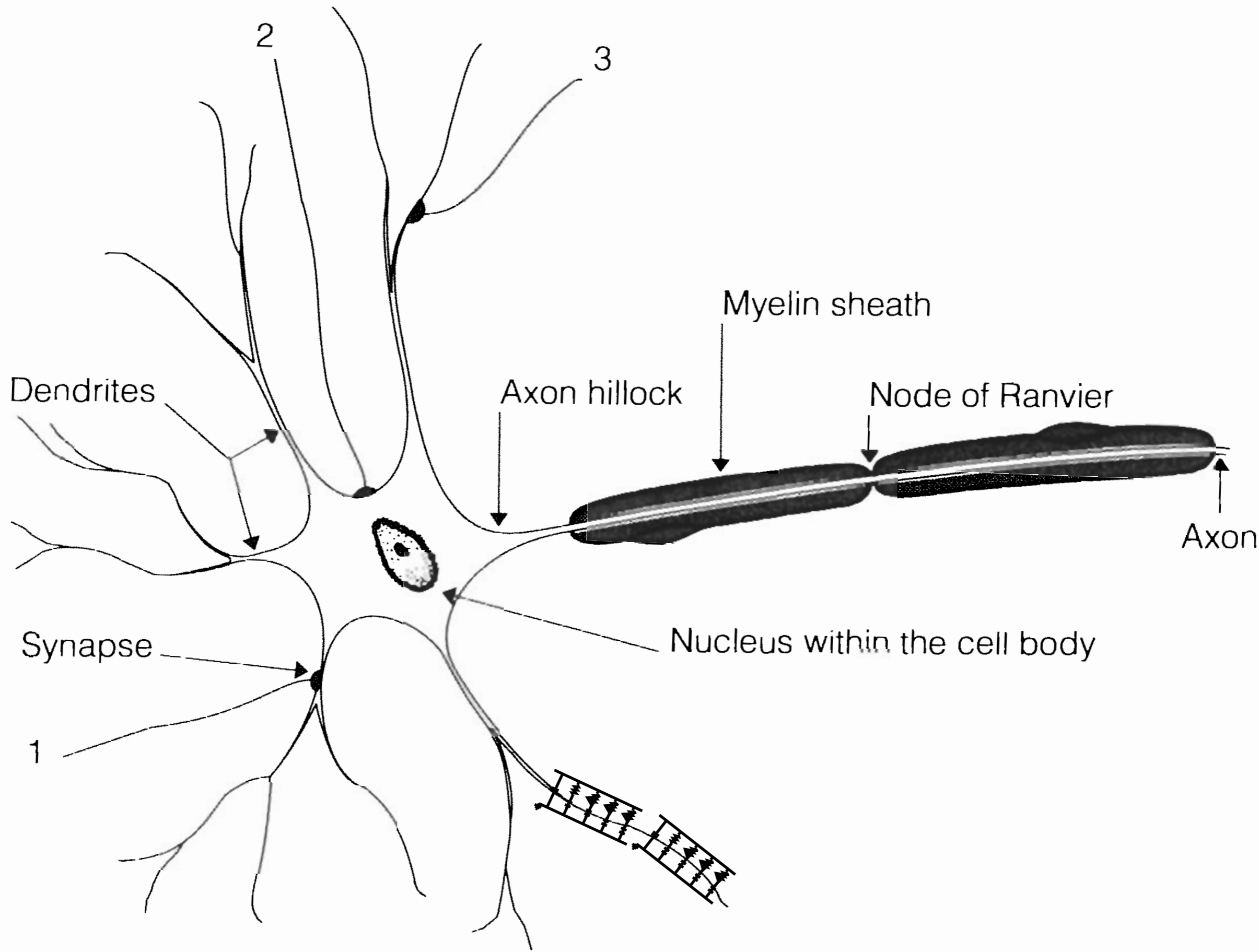


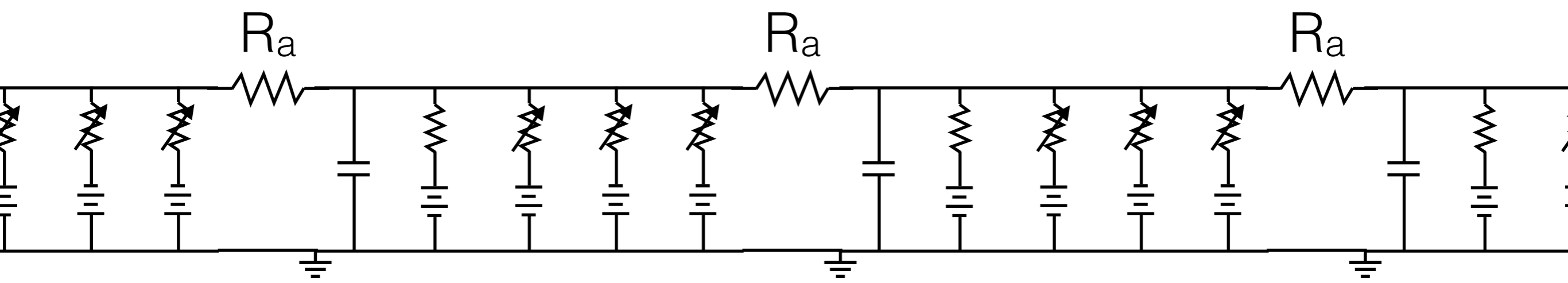
Membrane with synaptic inputs



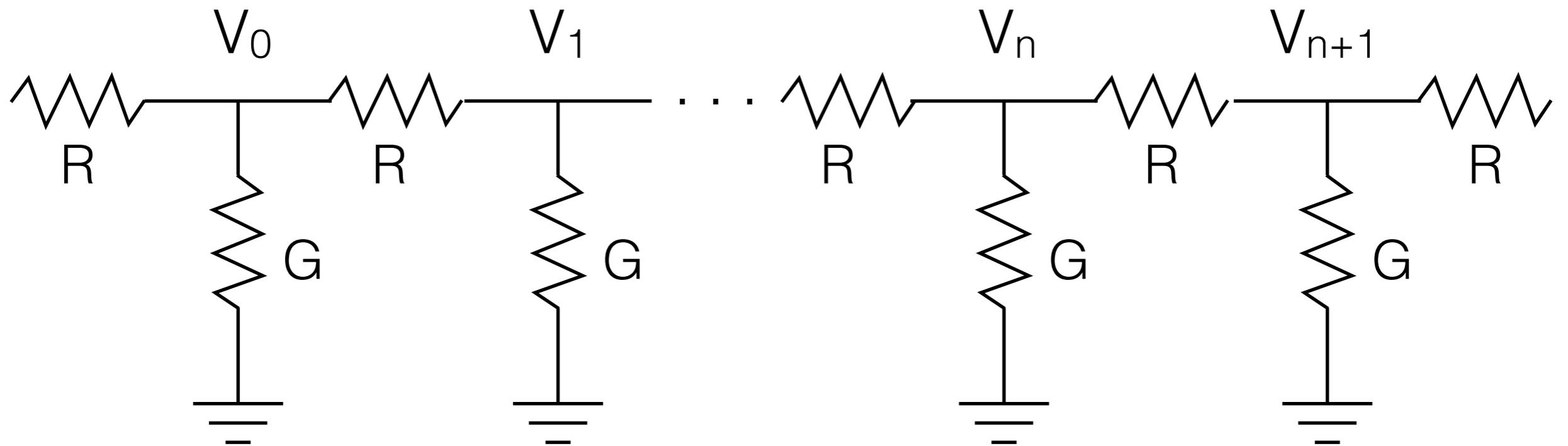
$$\tau \frac{dV}{dt} + V = \frac{V_r G_{\text{leak}} + V_{\text{Na}} \Delta G_{\text{Na}} + V_{\text{K}} \Delta G_{\text{K}} + V_{\text{Cl}} \Delta G_{\text{Cl}}}{G_{\text{total}}}$$

$$G_{\text{total}} = G_{\text{leak}} + \Delta G_{\text{Na}} + \Delta G_{\text{K}} + \Delta G_{\text{Cl}} \quad \tau = \frac{C_m}{G_{\text{total}}}$$





Resistive network



$$\frac{dV}{dx} = -I R$$

$$\frac{dI}{dx} = -V G$$

}

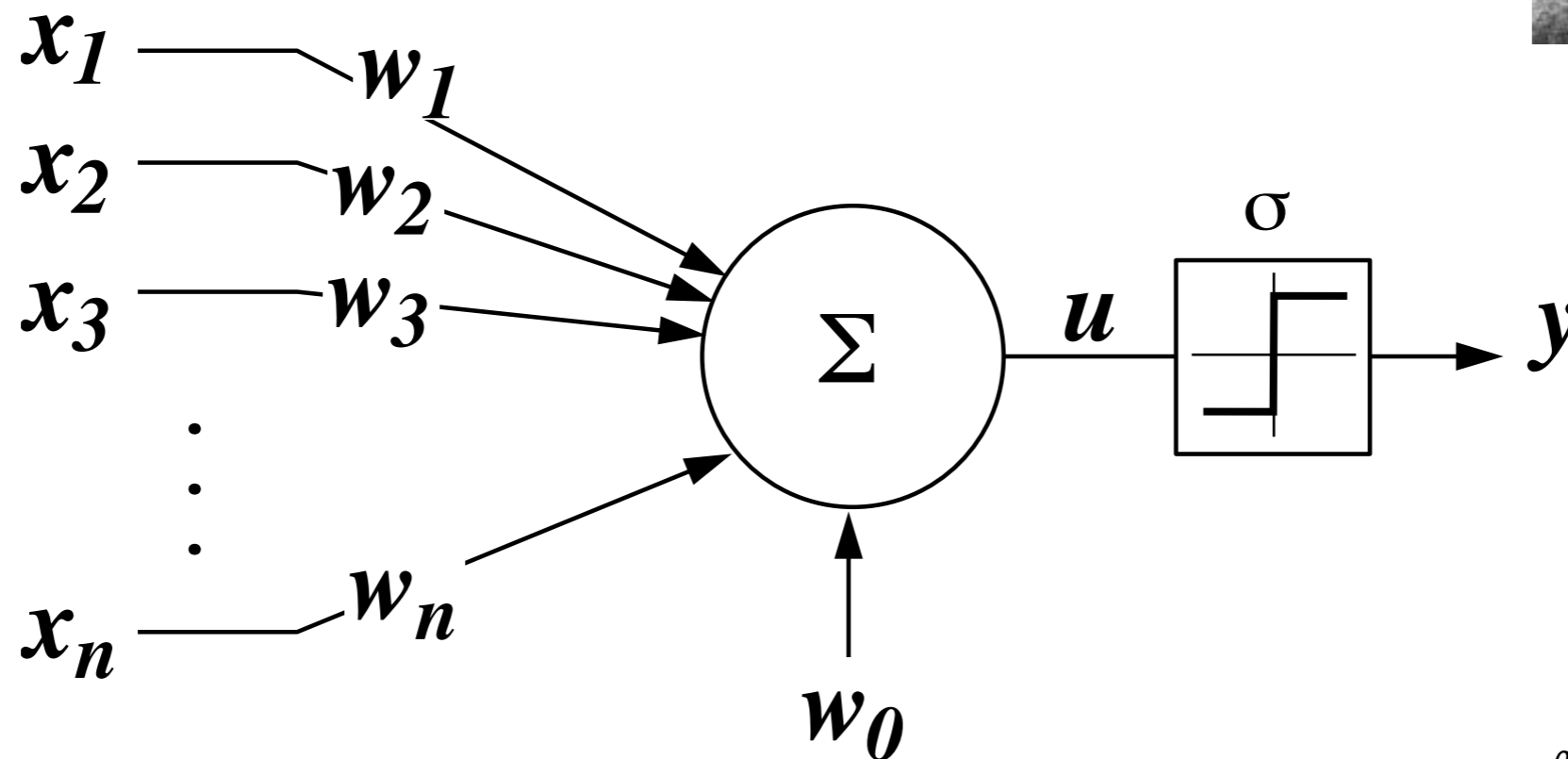
$$\frac{d^2 V}{dx^2} = R G V$$

$$V = V_0 e^{-\frac{|x|}{L}}$$

$$L = 1/\sqrt{R G}$$

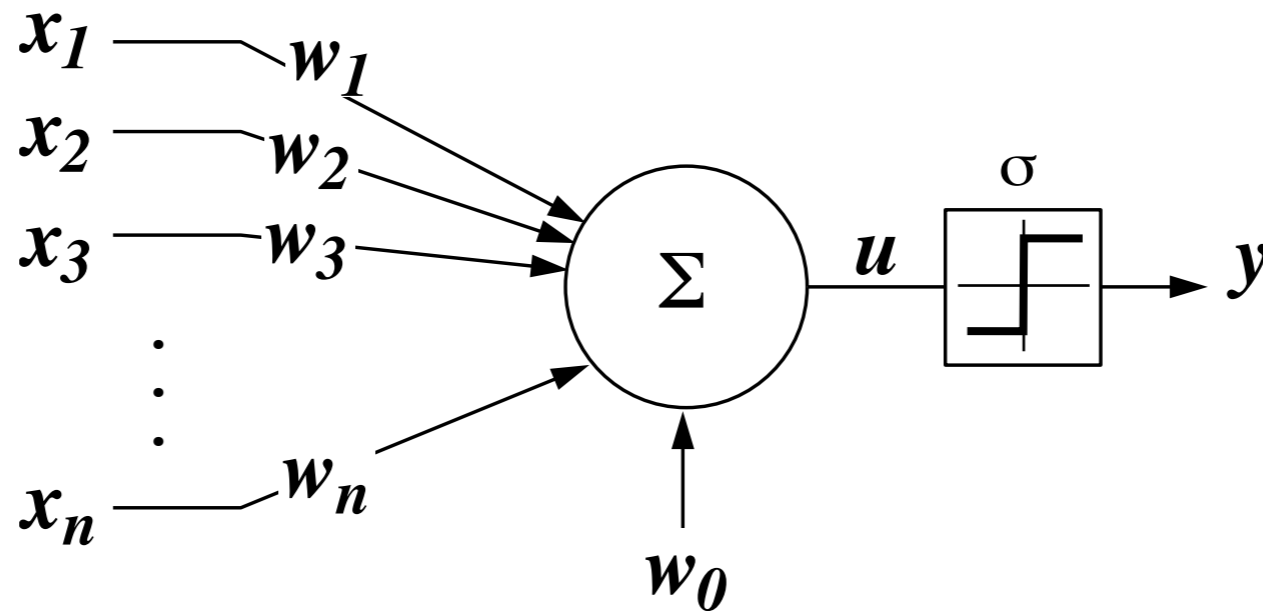
Perceptron model

(Rosenblatt, ca. 1960)



$$u = w_0 + \sum_{i=1}^n w_i x_i$$
$$y = \sigma(u)$$

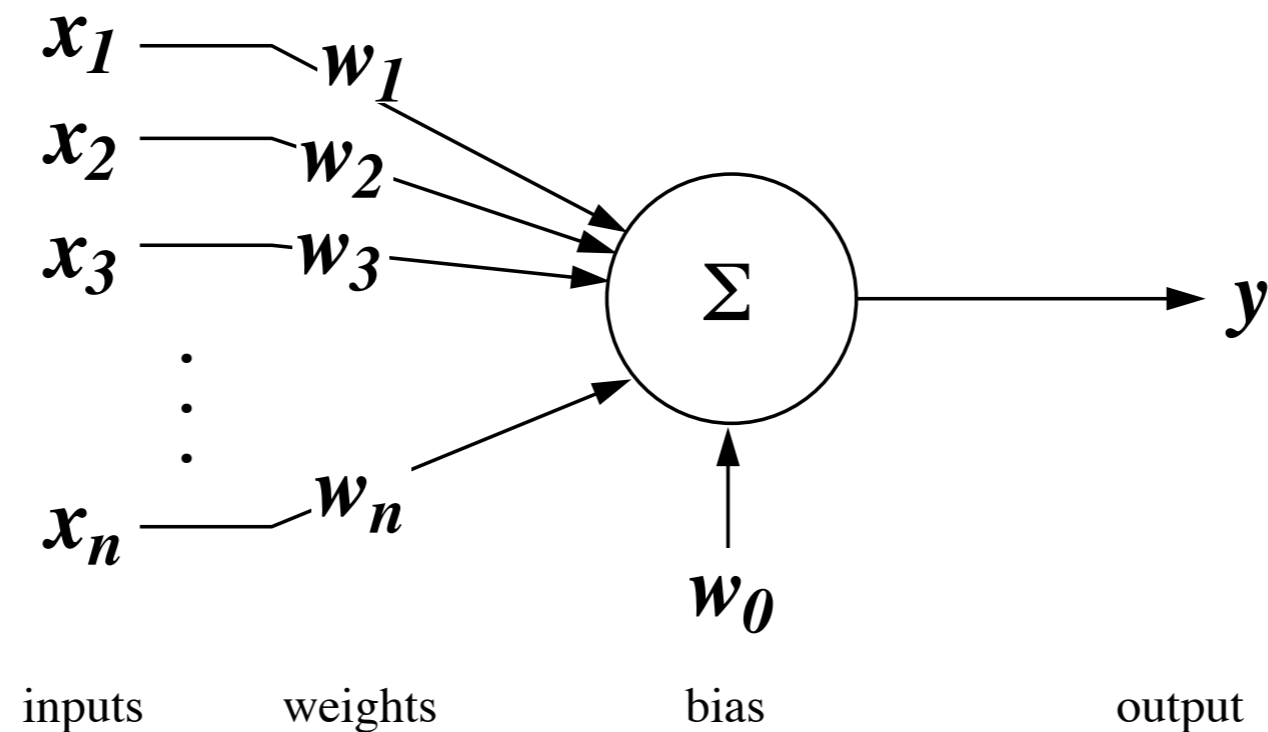
Perceptron learning rule (Rosenblatt 1962)



$$\Delta w_k = \begin{cases} 2\eta T^{(\alpha)} x_k^{(\alpha)} & y^{(\alpha)} \neq T^{(\alpha)} \\ 0 & \text{otherwise} \end{cases}$$

$$= \eta (T^{(\alpha)} - y^{(\alpha)}) x_k$$

Linear neuron learning rule (Widrow & Hoff 1960)



Objective function



Learning rule

$$E = \frac{1}{2} \sum_{\alpha} \left[T^{(\alpha)} - y^{(\alpha)} \right]^2$$

$$\begin{aligned} \Delta w_k &= -\eta \frac{\partial E}{\partial w_k} \\ &= \eta \sum_{\alpha} \delta^{(\alpha)} x_k^{(\alpha)} \end{aligned}$$

$$\delta^{(\alpha)} = T^{(\alpha)} - y^{(\alpha)}$$