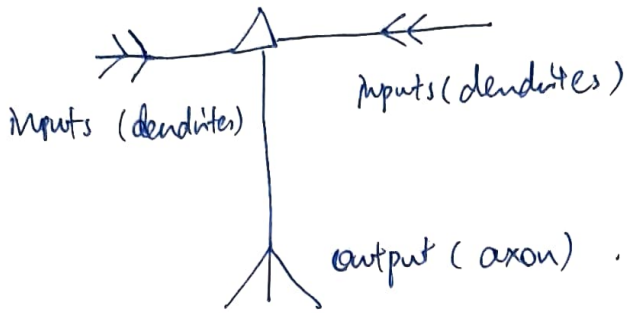


Review on Neuroscience Lecture 1

- The neuron (simplified):



- Mechanisms for neural potential:

① Neurons have semi-permeable membrane.

→ electrostatics.

→ electrostatics.

differential equation: $\frac{dn}{dh} = - \frac{mg}{kT} n$ constant.

(The rate at which the density of atoms changes at every instant of height is inversely proportional to the density of atoms present at that height.)

Solution :

Thus, we have the Boltzmann Relationship :

the density of atoms at a particular height is proportional to $e^{-\text{potential energy of each atom}/kT}$.

⇒ Boltzman v/s → ion distribution across neural membrane

→ electric field across hemol membrane

and V (voltage) $\propto Q$ (charge).

i.e., $V = C_m Q$

constant, "membrane capacitance".

electrode applies current



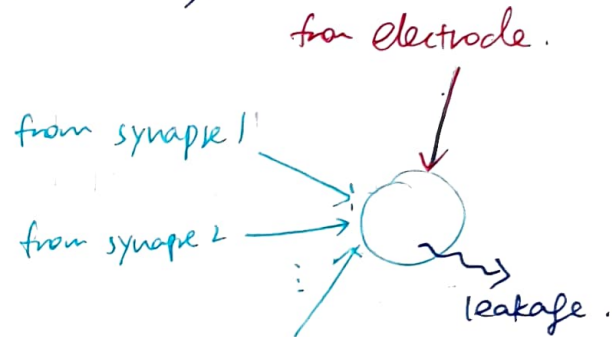
$$V = C_m Q \text{ (from ①) .}$$

$$\text{Cm } \frac{dV}{dt} = \frac{dQ}{dt} = I \leftarrow \text{the current}$$

③. Activation potentials propagate along the axon, until they reach a synapse.

2 types of synapses — excitatory
 \ inhibitory

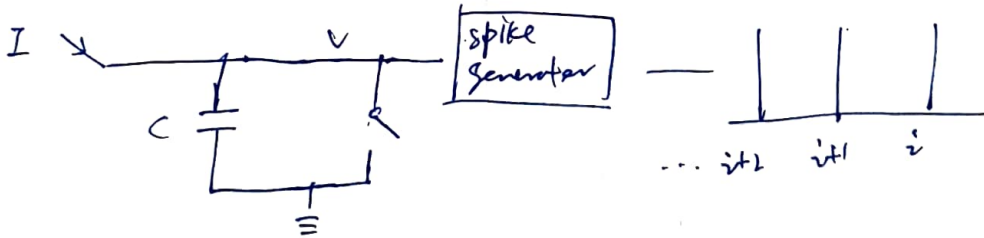
What happened here is modelled by =



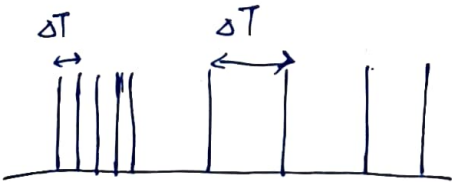
(a) $C_m \frac{dv}{dt} = I_{\text{total}}$ (sum of all input currents).
 $= -\text{leakage} + \text{from synapses} + \text{from electrode}$.
 $= -i_m + I_{\text{synapses}} + I_e/A$

$$= -g_{\text{leakage}} (V - E_i) + g_{\text{synaptic}} (V - E_i) + I_e / A.$$

(b) inegrate-and-fire circuit.

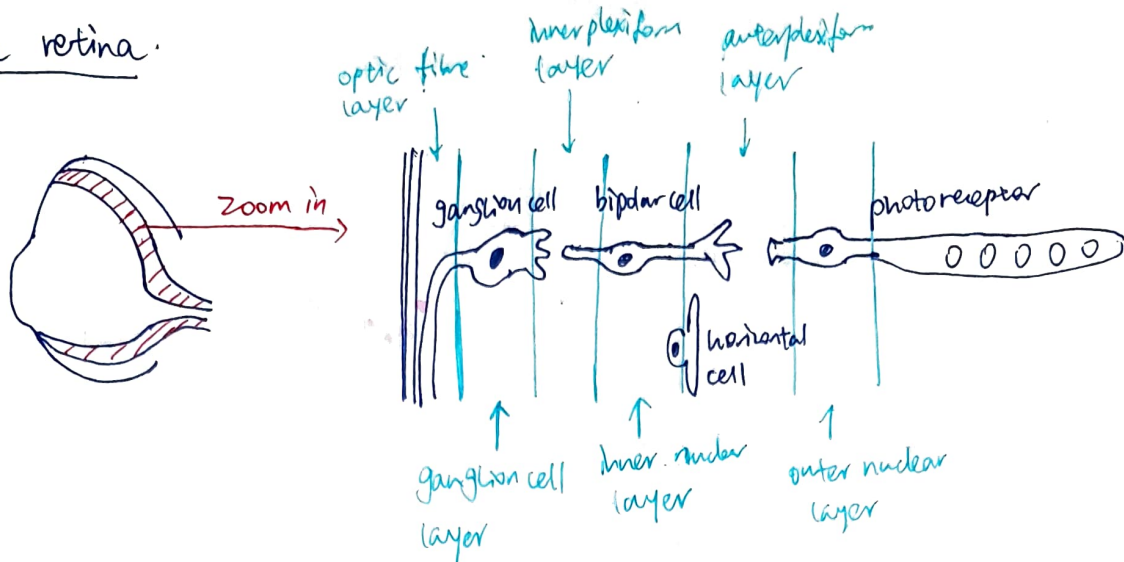


⑤ The spike generated in ③b encodes information.



$$f = \frac{1}{\Delta T}.$$

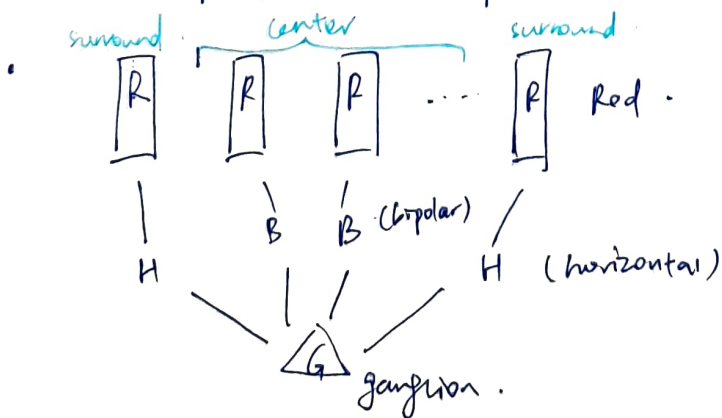
The retina.



2 types of retinal ganglion cell (RGC):

visual stimuli	on-center RGC	off-center RGC

- Each ganglion cell has a receptive field.
- Depending on its type (on-center or off-center), the ganglion cell responds with (outputs) different spikes from the receptive field.



- This arrangements enable the RGC to detect contrast, and hence identify edges in image.

① Each RGC is a "simple cell".

The firing response of each simple cell is selective to the specific orientation. \rightarrow each simple cell detects one edge of a particular orientation.

② convolution: Sum of output spikes produced by many simple cells.
 \rightarrow sum of all edges detected.

Formally, given two spikes $h(t)$ and $f(t)$, the result of convolution:

$$y(t) = h(t) * f(t) \\ = \int_{-\infty}^{+\infty} h(t-\lambda) f(\lambda) d\lambda$$

If $h(t) = 0 \forall t < 0$ and $f(t) = 0 \forall t < 0$, then

$$y(t) = \int_0^t h(t-\lambda) f(\lambda) d\lambda.$$

③ pooling: "pooling operation" takes place at the complex cells.

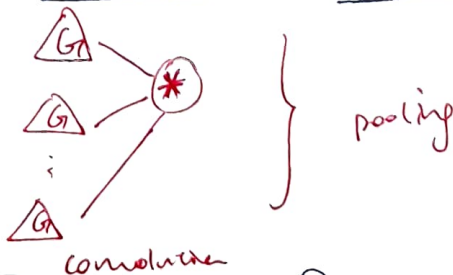
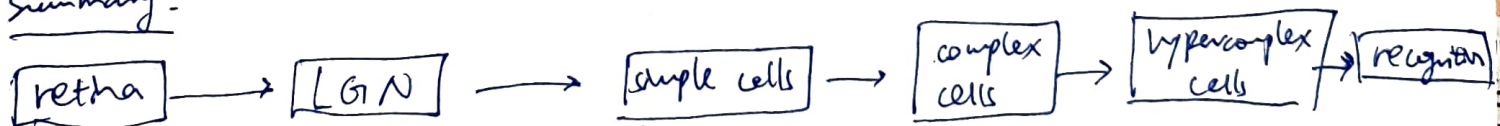
This produces output of edges in generalized position.

(i.e. makes the model invariant to local translation).

④ Hypercomplex cells etc.

\downarrow
recognize the object.

Summary:



① images \rightarrow ② edges \rightarrow ③ edges in generalized position \rightarrow ④ objects.