

Block course Fall 2013
ETH course 227-1049-00L Insights into Neuroinformatics

Neuromorphic Engineering, with Biological and Silicon Retinas

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Teaching assistants



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Minhao Yang



Outline of course

- The biology of the retina
 - How retinas uses adaptive photoreceptors and horizontal cells, together with bipolar cells, to compute rectified local contrast
 - The Physiologist's Friend Chip
 - The Dynamic Vision Sensor
- Silicon technology and the operation of a single transistor
 - CMOS vs. complementary channels in neurons
- Neuromorphic Engineering (NE)
 - Context of electronics (synchronous logic)
 - Motivation for NE by contrasting computers and brains

Reading



ELSEVIER

Available online at www.sciencedirect.com



ScienceDirect

Current Opinion in

Neurobiology

Neuromorphic sensory systems

Shih-Chii Liu and Tobi Delbruck



ELSEVIER

Available online at www.sciencedirect.com

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Vision Research 44 (2004) 2083–2089

Vision
Research

www.elsevier.com/locate/visres

A silicon early visual system as a model animal

Tobi Delbrück *, Shih-Chii Liu

Institute of Neuroinformatics, University of Zürich and ETH Zürich, Winterthurerstrasse 190, CH-8057 Zürich, Switzerland

Received 3 July 2003; received in revised form 7 March 2004

Practical work with

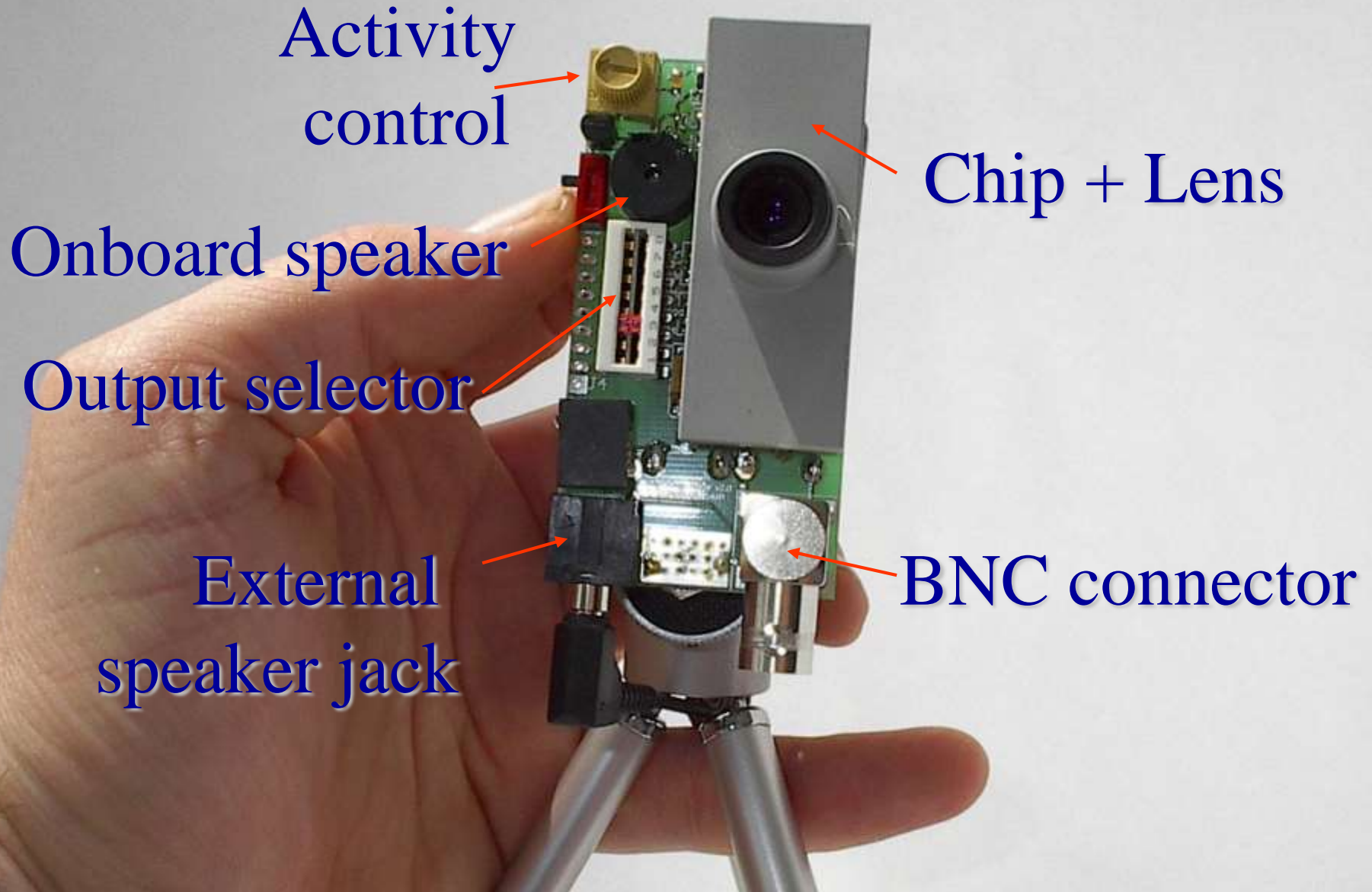
Physiologist's Friend Chip



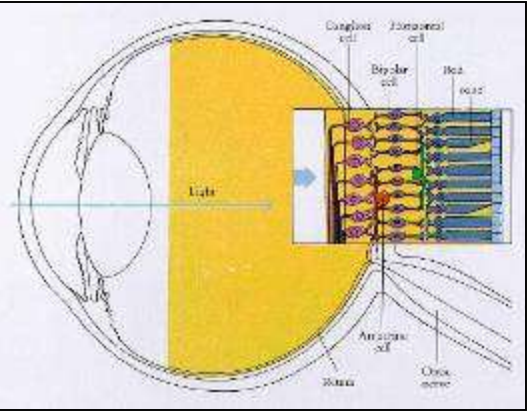
Dynamic Vision Sensor
(DVS) silicon retina



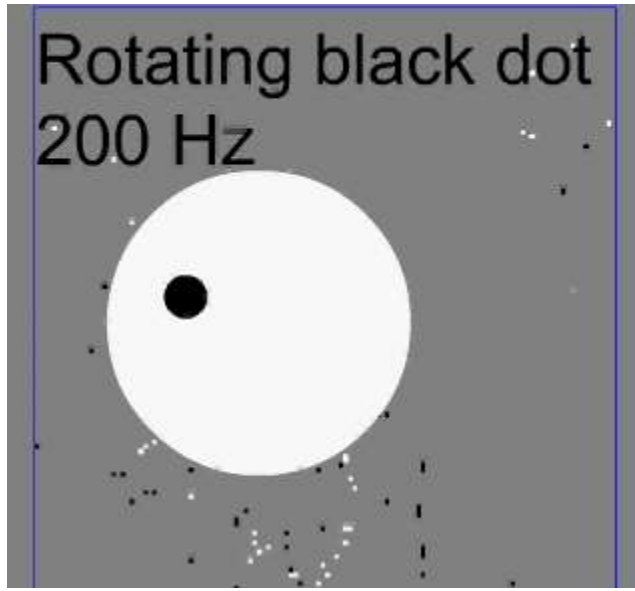
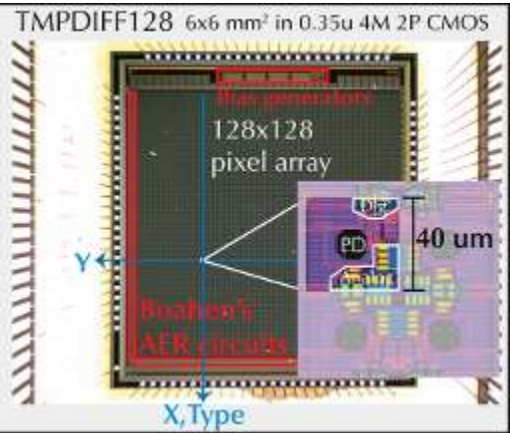
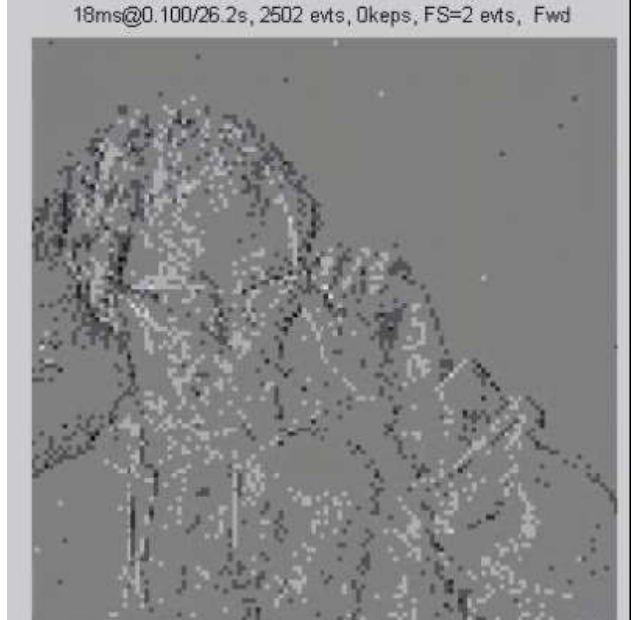
The Physiologist's Friend Chip



Temporal Contrast Dynamic Vision Sensor (DVS)



1. This silicon retina **asynchronously** outputs **pixel address-events**.
2. Each event represents a fixed **temporal contrast** ($\Delta \log I$), corresponding to change in scene reflectance.



1. Models transient pathway in retina.
2. Reduces redundancy
3. Responds quickly and preserves timing
4. Has wide dynamic range

Quiz and Grading

- You will have 30 minutes to individually answer a quiz at the end of the course
- Your grade will be based on a combination of quiz and your group report

Schedule

Day 1 (13:00-17:00)

- Lecture on biological retina
- Tutorial on using Physiologist's Friend and DVS
- Start first practical (split into groups and select either Physio Friend or DVS for first exercise)
- Read parts of Physio Friend paper and Neuromorphic Sensory Systems

Day 2 (morning, 9:00-12:00)

- Lecture on electronics and transistors
- Practical work
- Finish reading papers

Afternoon (13:30-17:00)

- Lecture continued
- Finish practical work, write reports
- Quiz

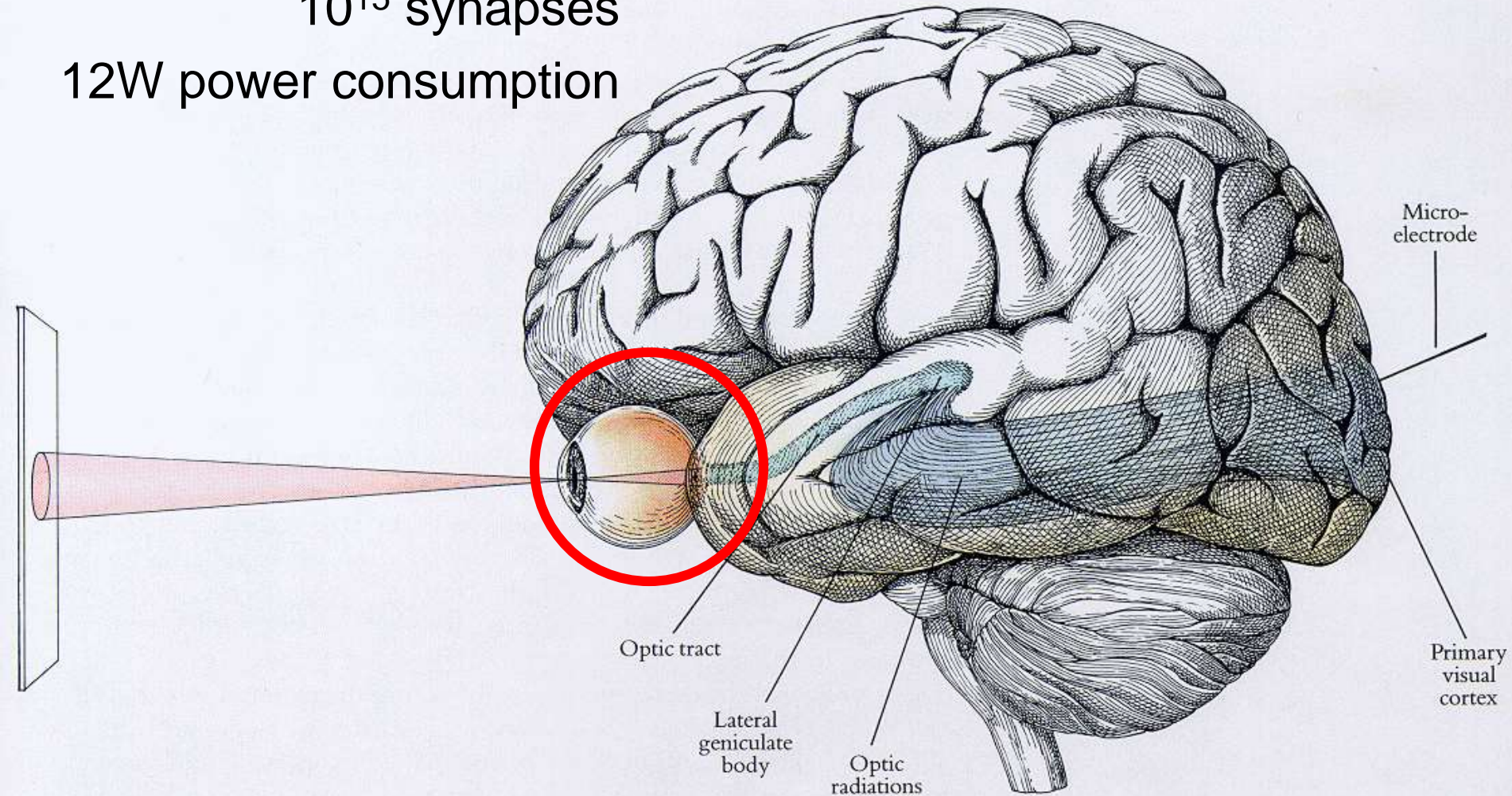
Biological Retinas

How do we see?

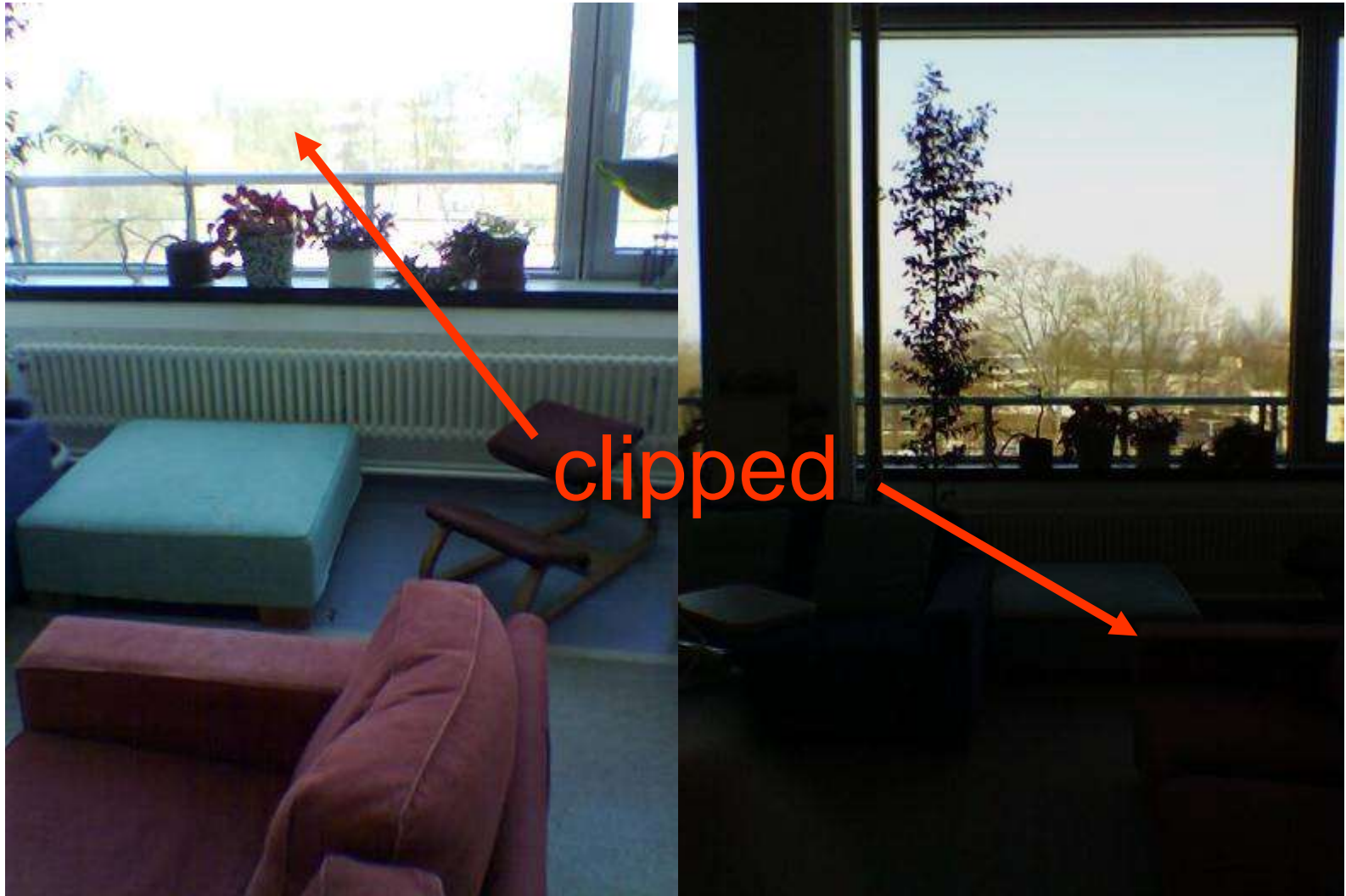
10^{11} neurons

10^{15} synapses

12W power consumption



Is your eye a camera?



Light ranges

1 lux of sunlight is about 10^4 photons/ $\mu\text{m}^2/\text{sec}$

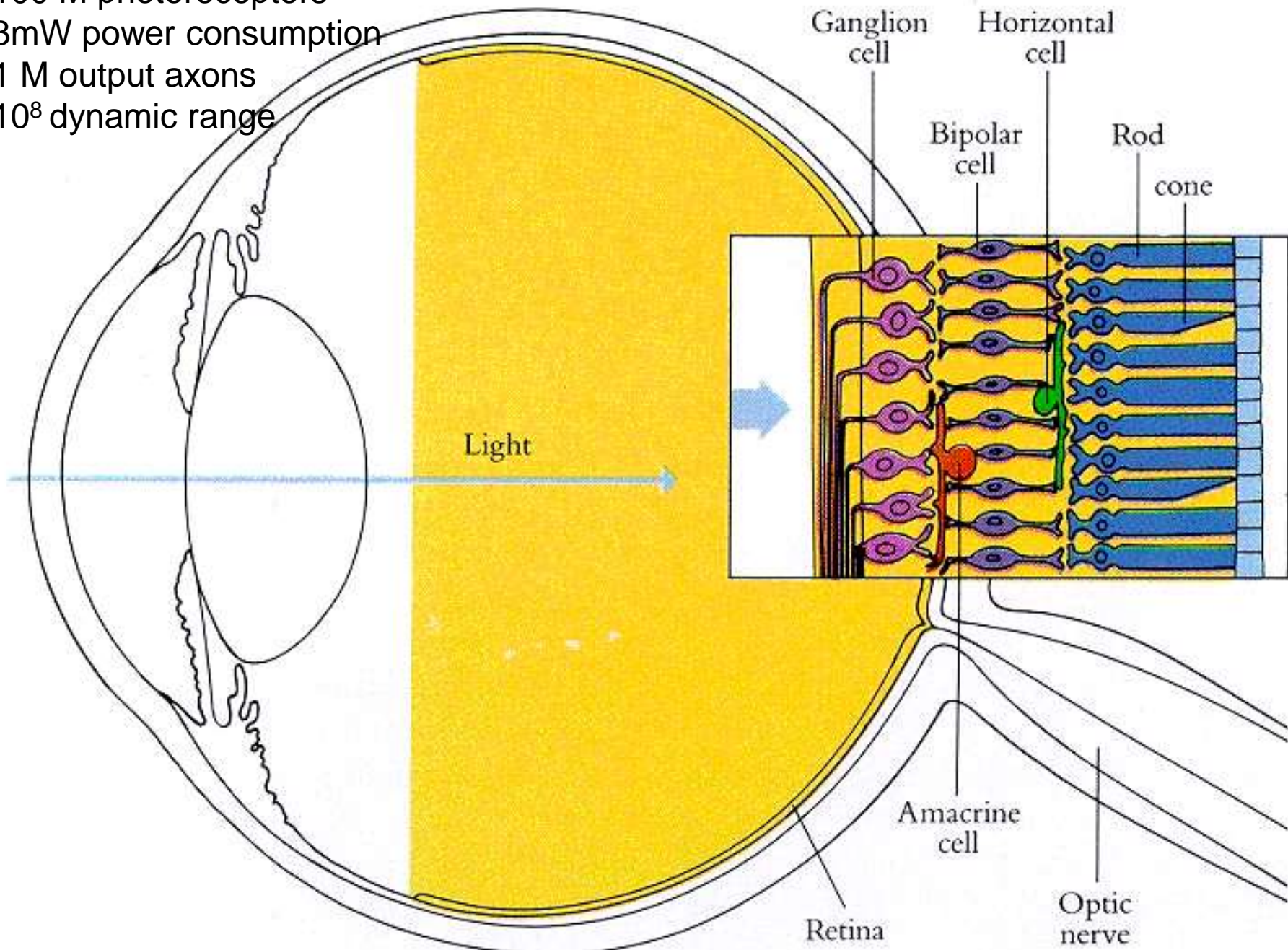


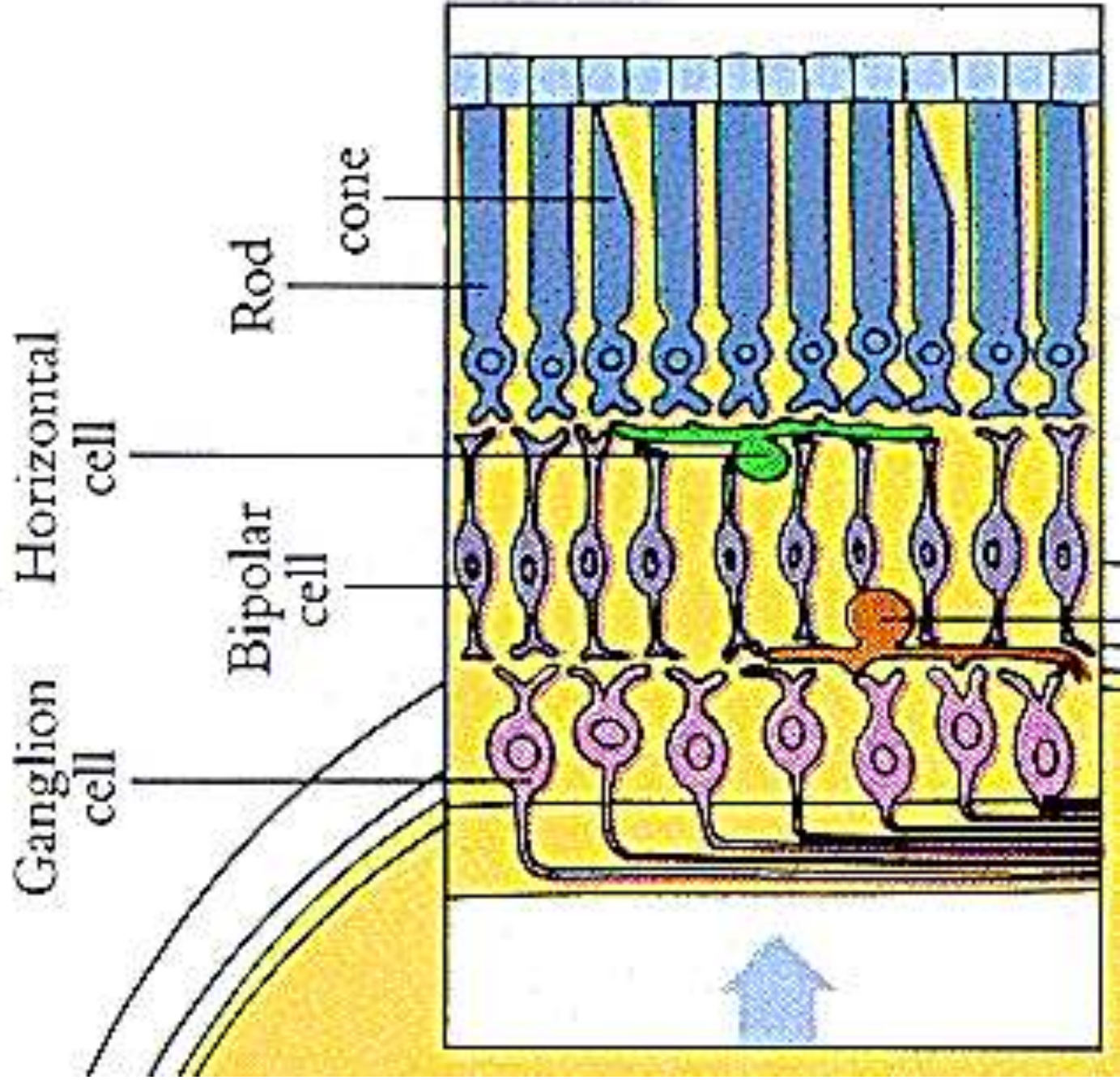
Direktes Sonnenlicht	100'000 Lux
Sonniger Tag	10'000 Lux
Bedeckter Tag	1'000 Lux
Büro	100 Lux
Einbrechende Dämmerung	10 Lux
Dämmerung	1 Lux
Vollmond	0.1 Lux
Viertelmond	0.01 Lux
Klare mondlose Nacht	0.001 Lux
Bedeckte mondlose Nacht	0.0001 Lux

Entire range

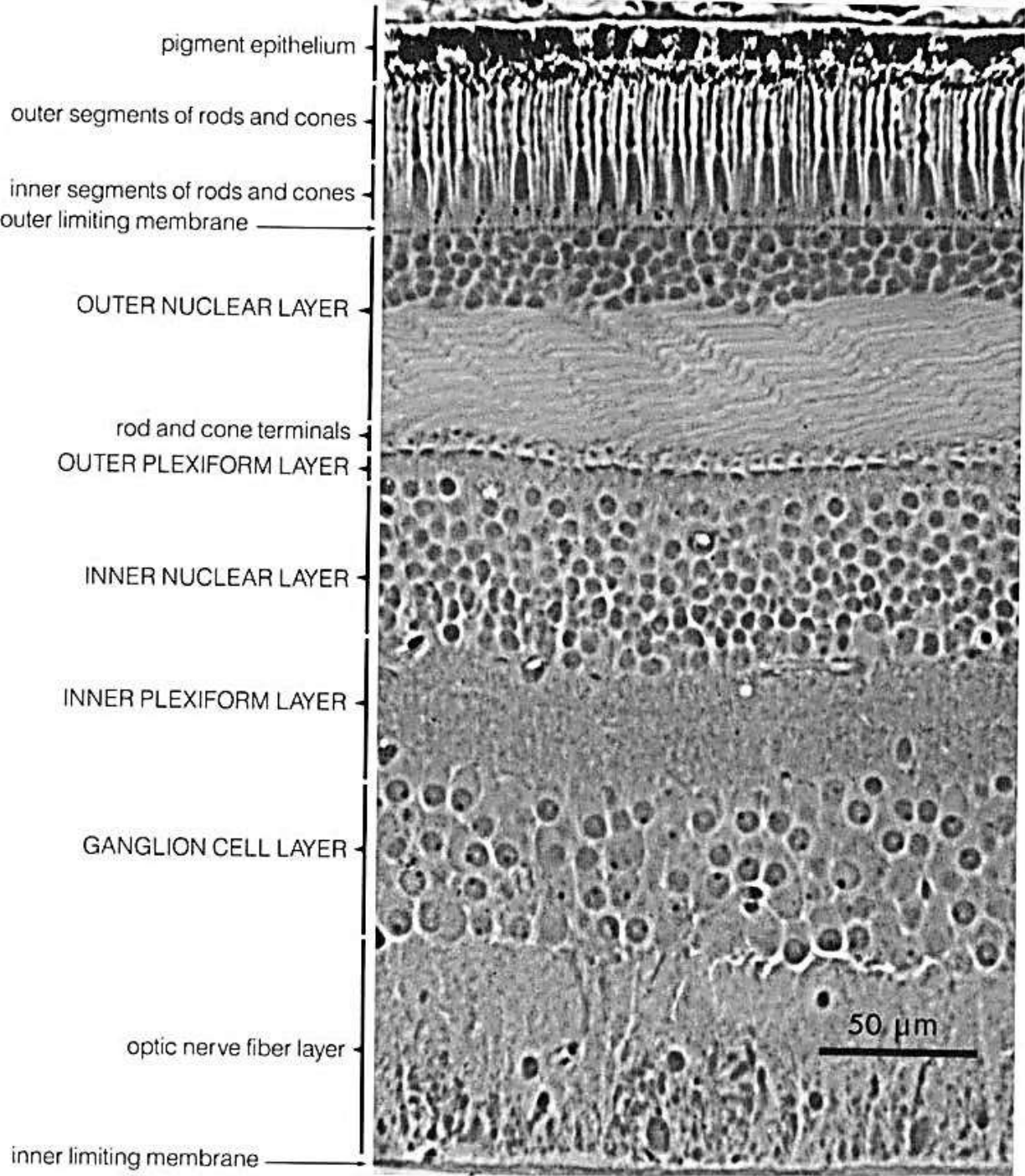
Camera range

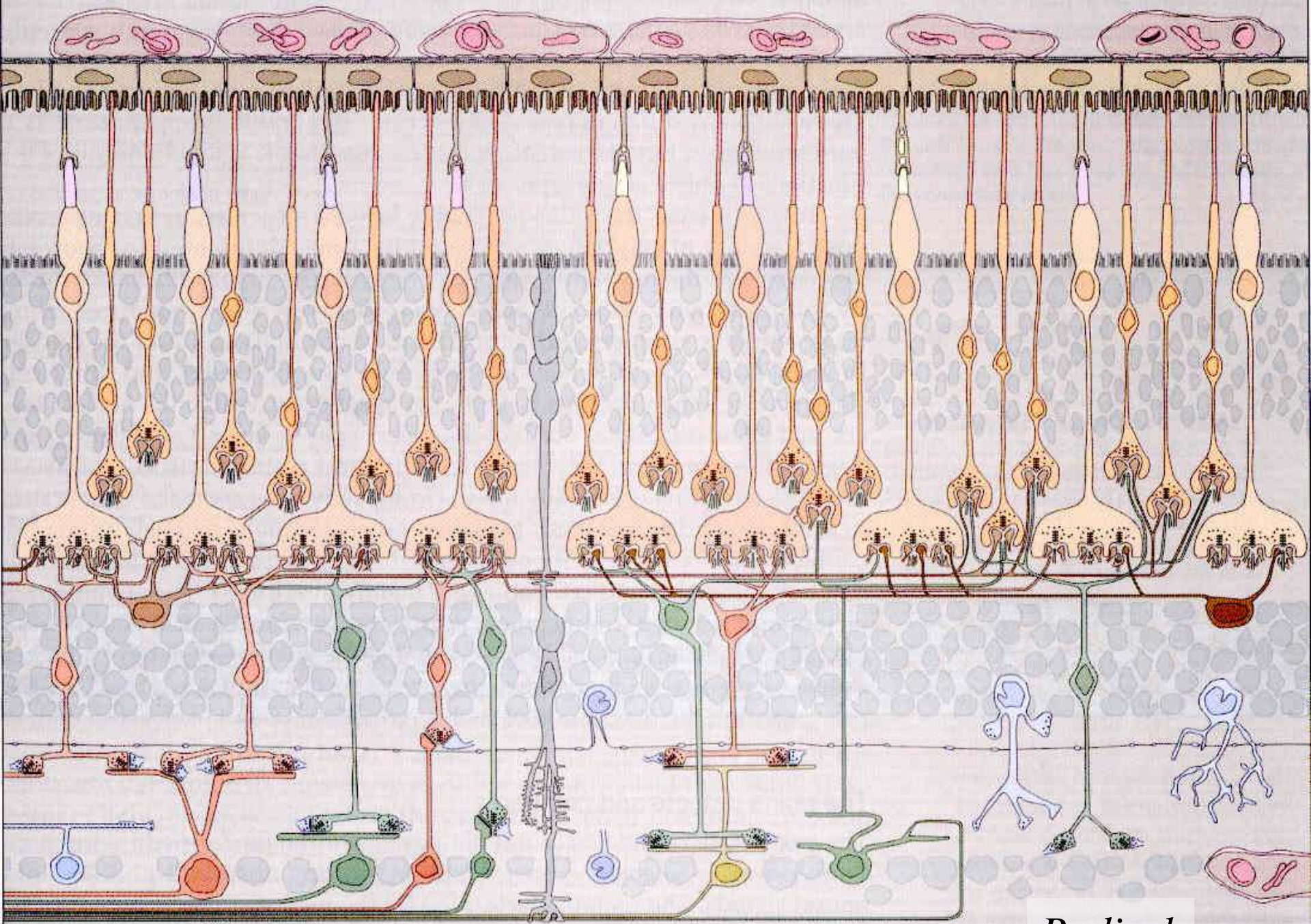
100 M photoreceptors
3mW power consumption
1 M output axons
 10^8 dynamic range



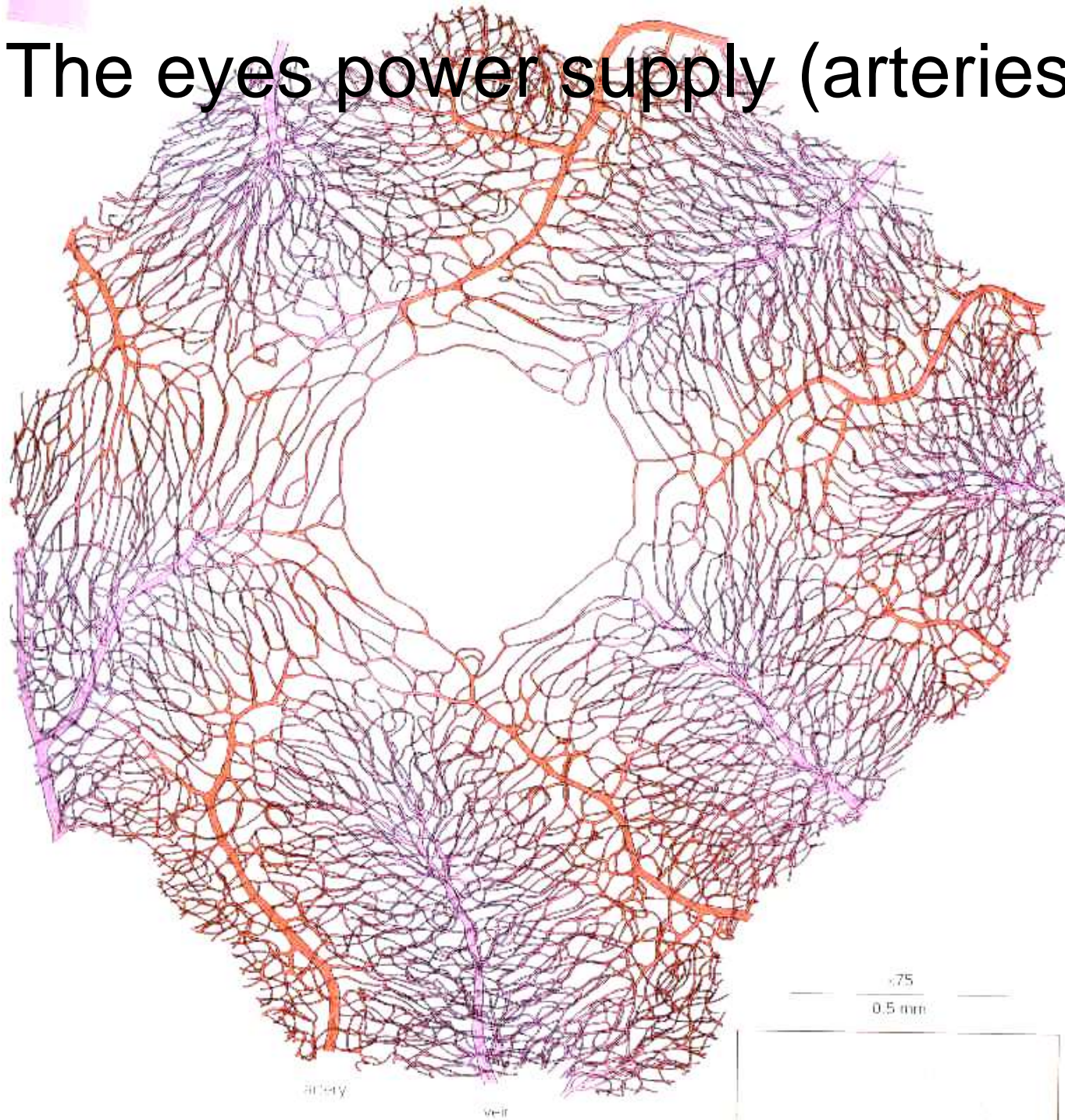


Cross section of human retina





The eyes power supply (arteries)



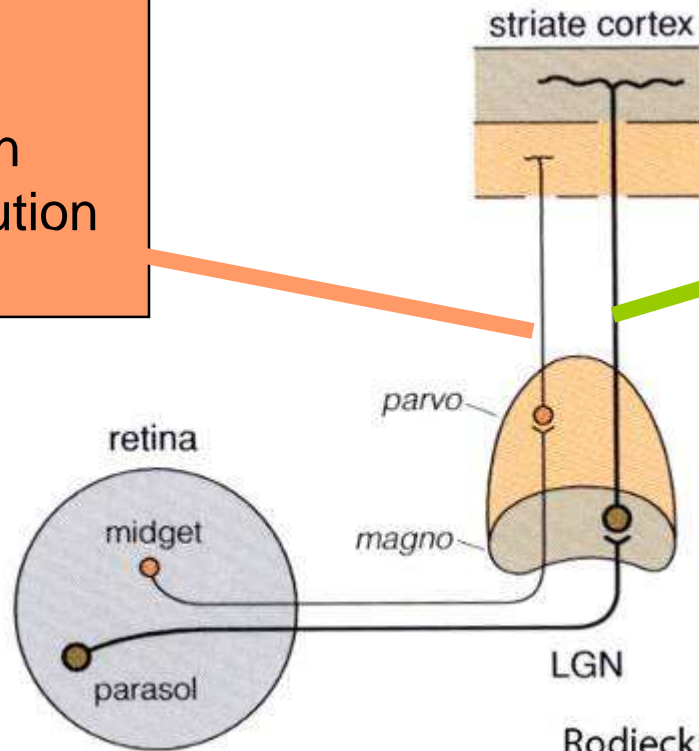
All animals (from insects to us) partition vision into **sustained** and **transient** visual pathways

Sustained (Parvo, X)
Has sustained response
Sees color
Low contrast gain
Higher spatial resolution
Slower

Transient (Magno, Y)
Blind to DC
Monochrome
High contrast gain
Lower spatial resolution
Faster

Standard image sensors do this

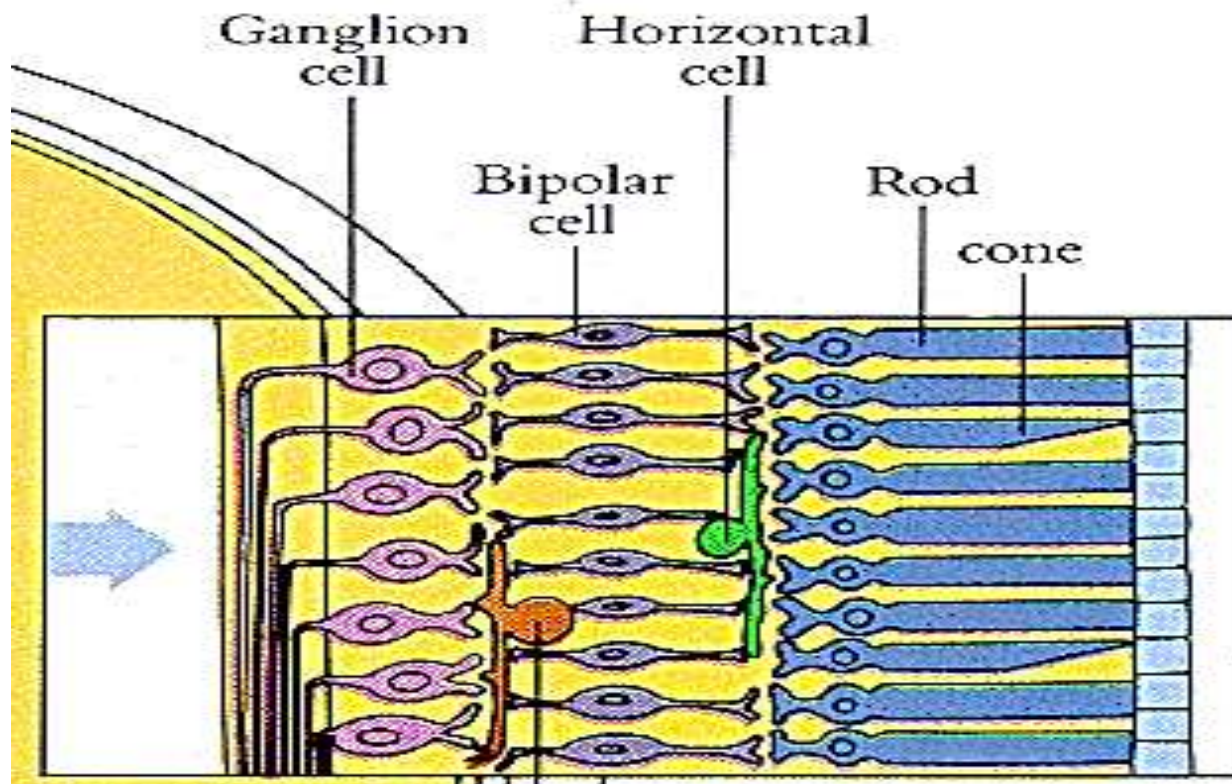
Temporal contrast sensor does this

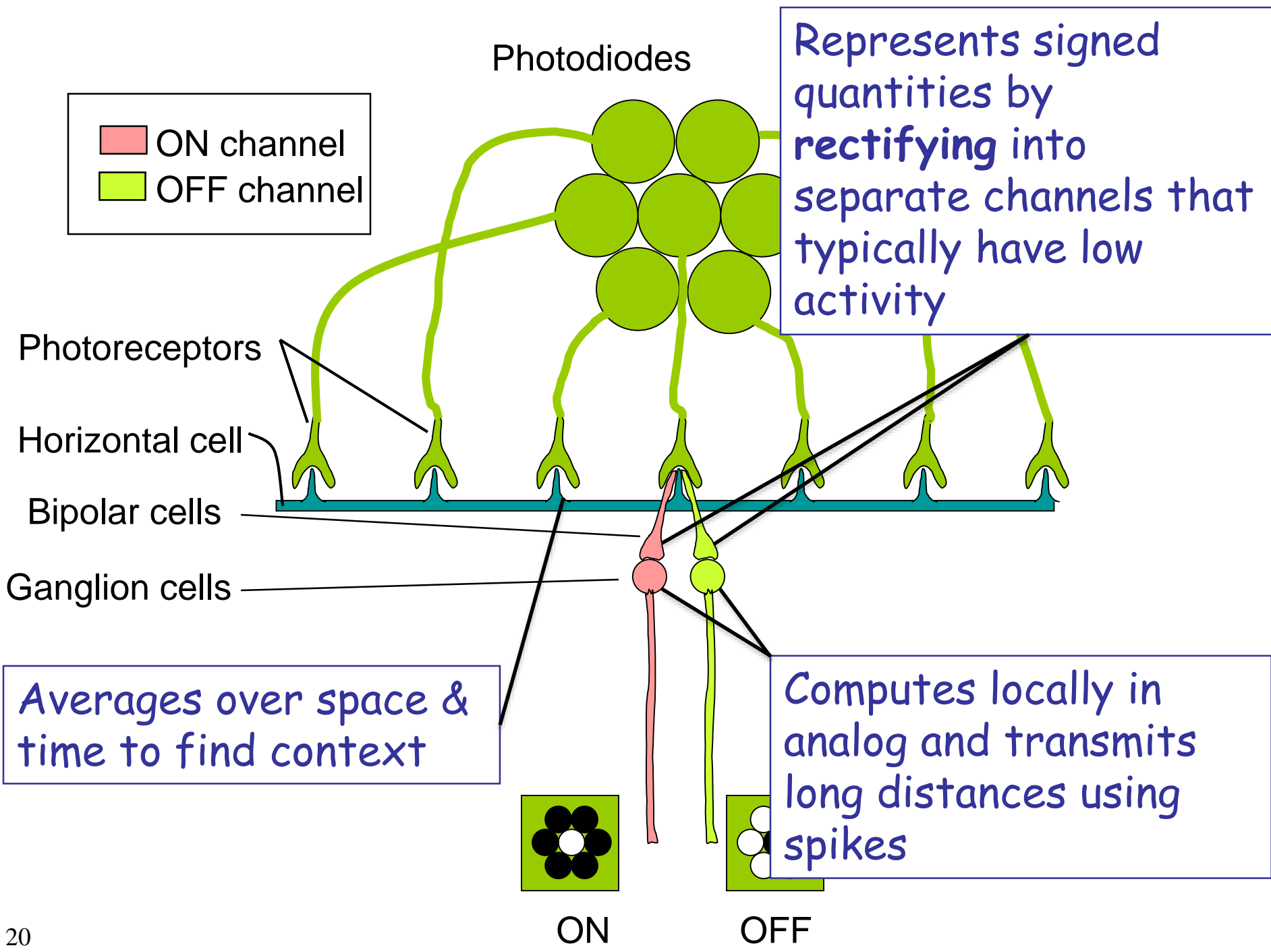


Rodieck 1998

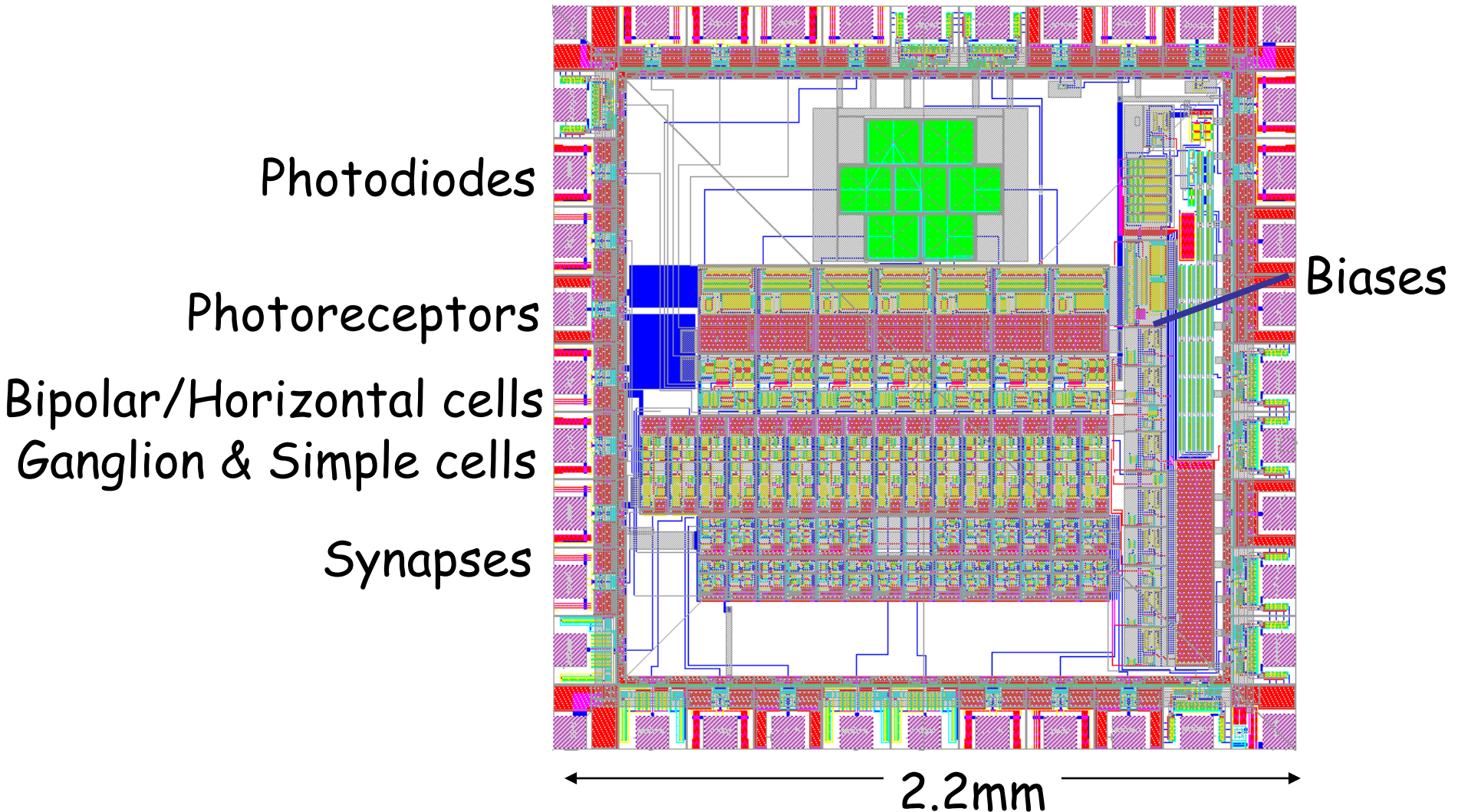
Three layer retina model

Photoreceptors – bipolar cells – ganglion cells



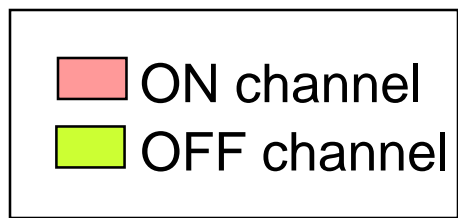


Physio Friend Layout



Photoreceptors

Outer segments (photo sensor photodiodes)

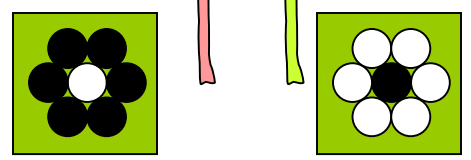
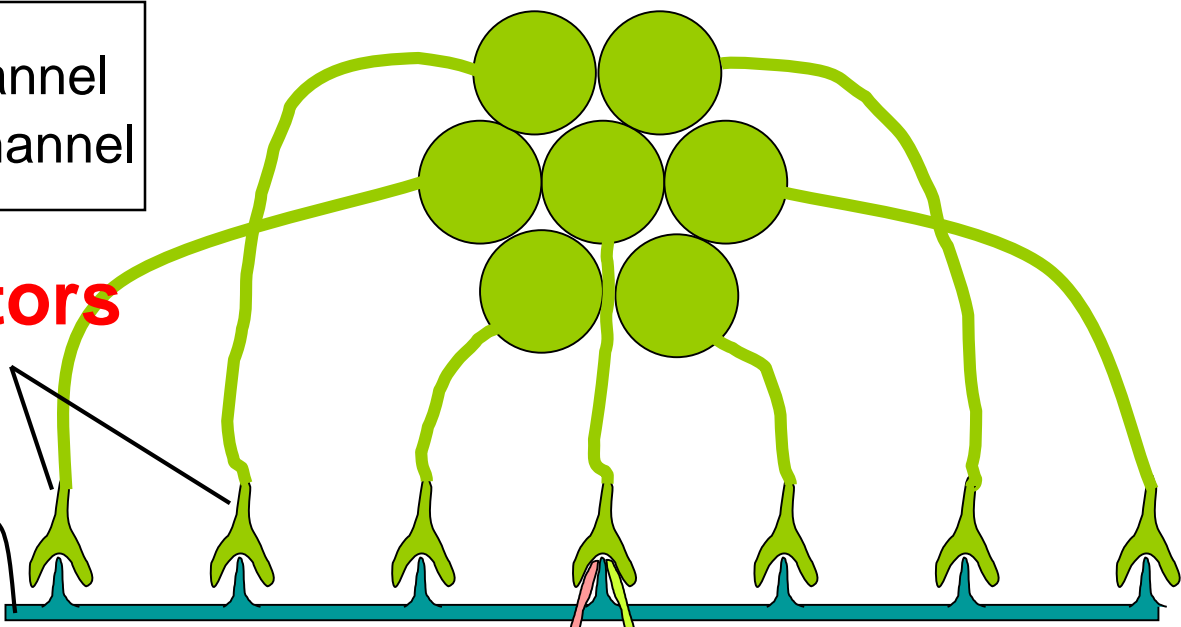


Photoreceptors

Horizontal cell

Bipolar cells

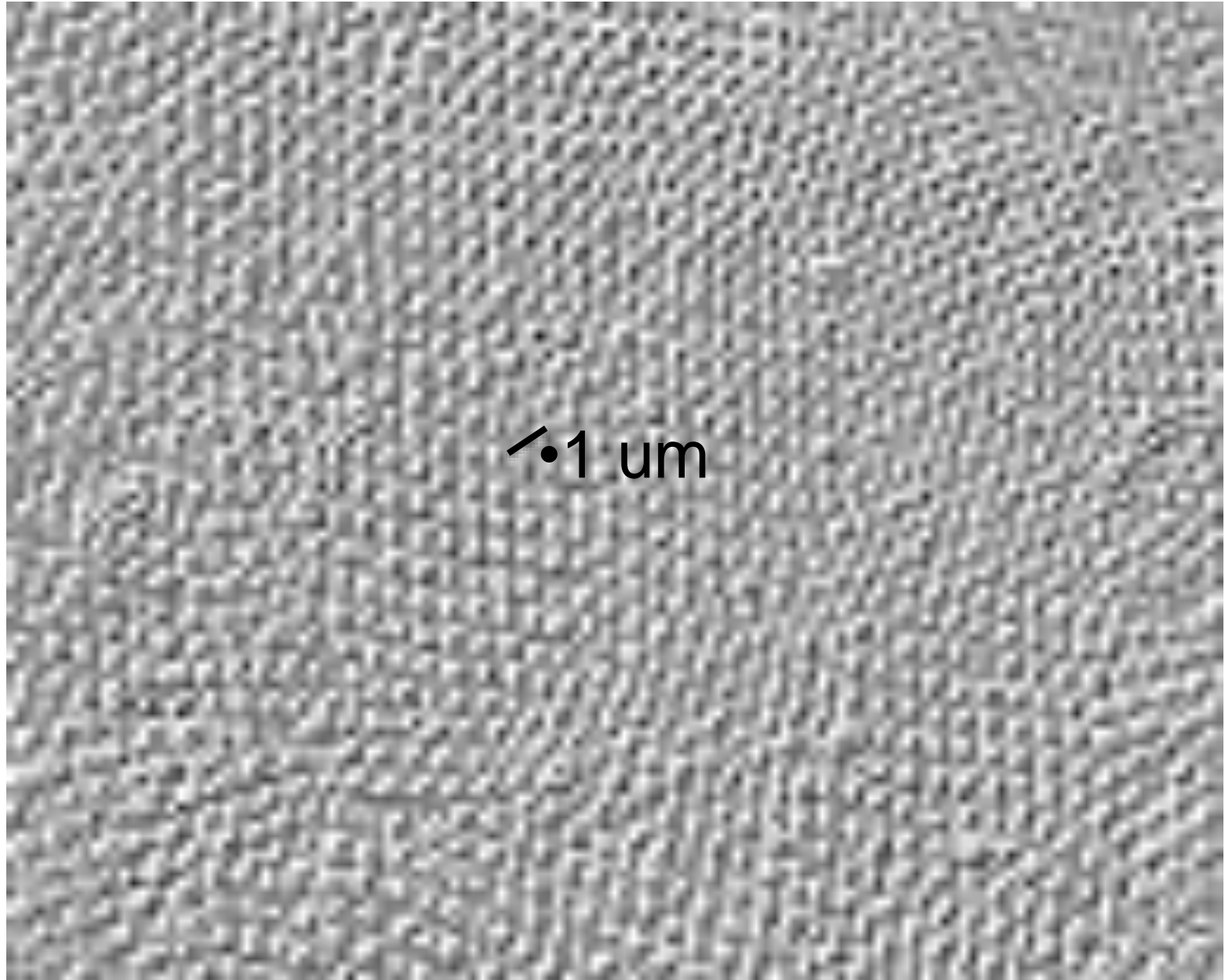
Ganglion cells

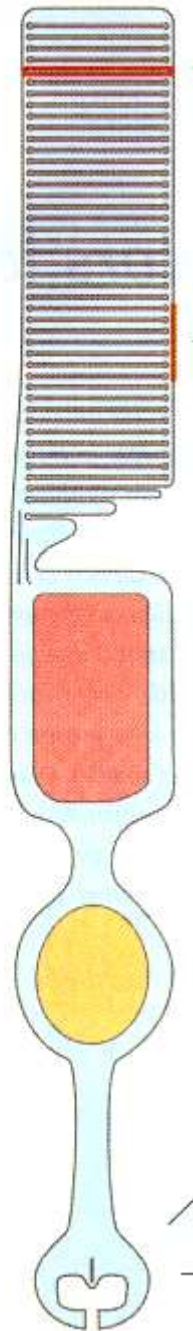


ON

OFF

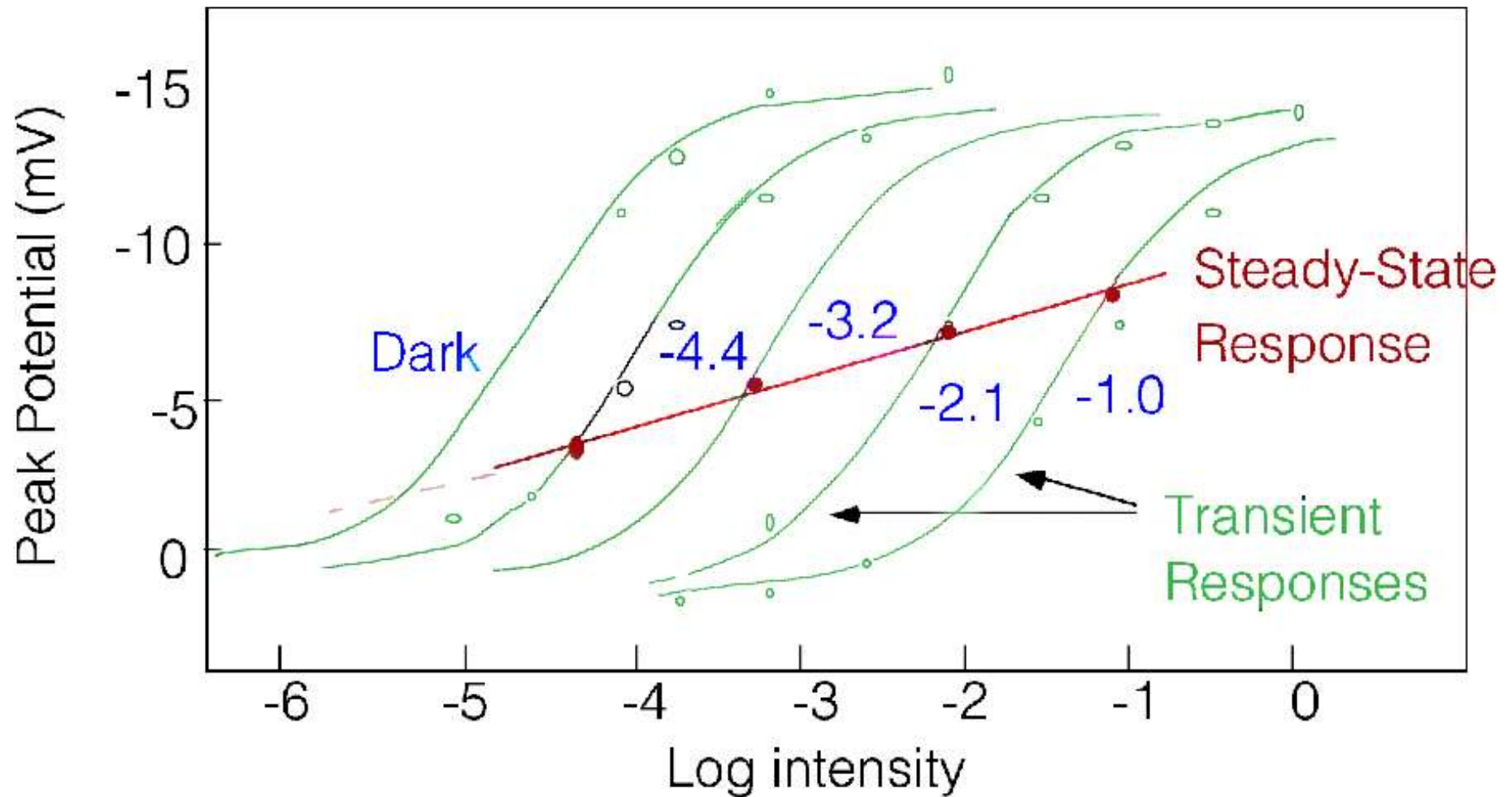
The photoreceptor mosaic in the eye





Rodieck 1998
The first steps in seeing

Biological photoreceptors adapt their operating point and gain

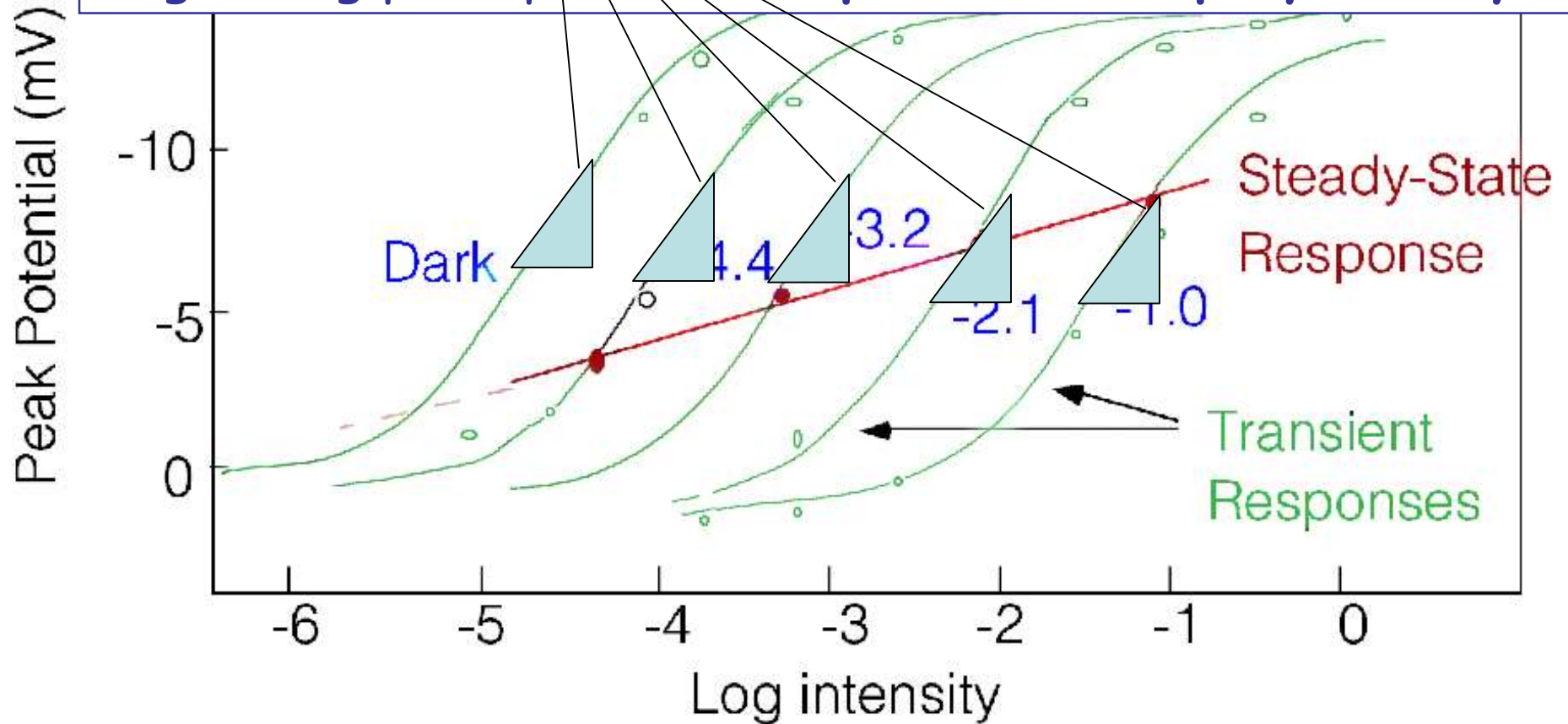


Norman & Perlman 1979

Biological photoreceptors adapt their operating point and gain

Organizing principle: Use **Context** to **Normalize** signal

Organizing principle: Use **adaptation** to **amplify novelty**

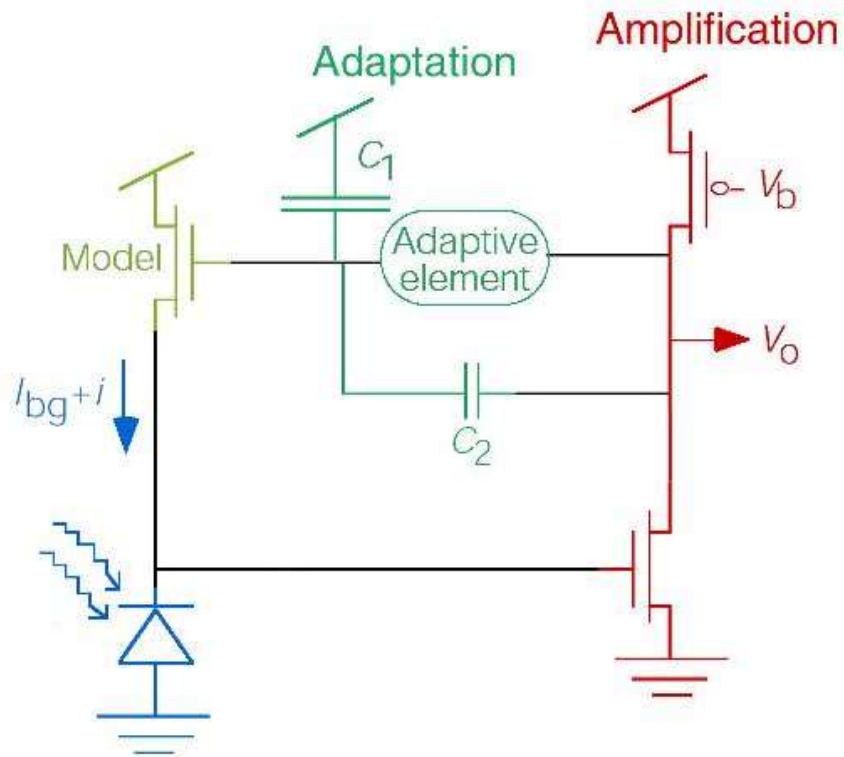


log(Intensity) is self-normalizing and automatically preserves reflectance differences, by normalizing away the constant illumination term in the product of (scene reflectance) * (illumination)

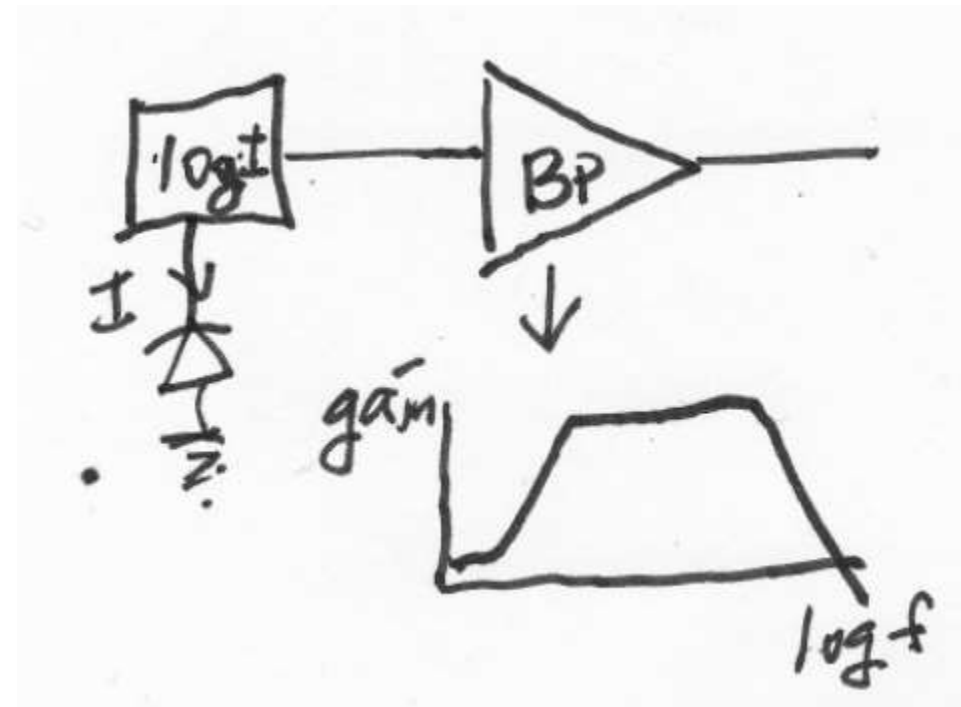
$$d(\log X) = dX/X$$

Adaptive Photoreceptor Circuit

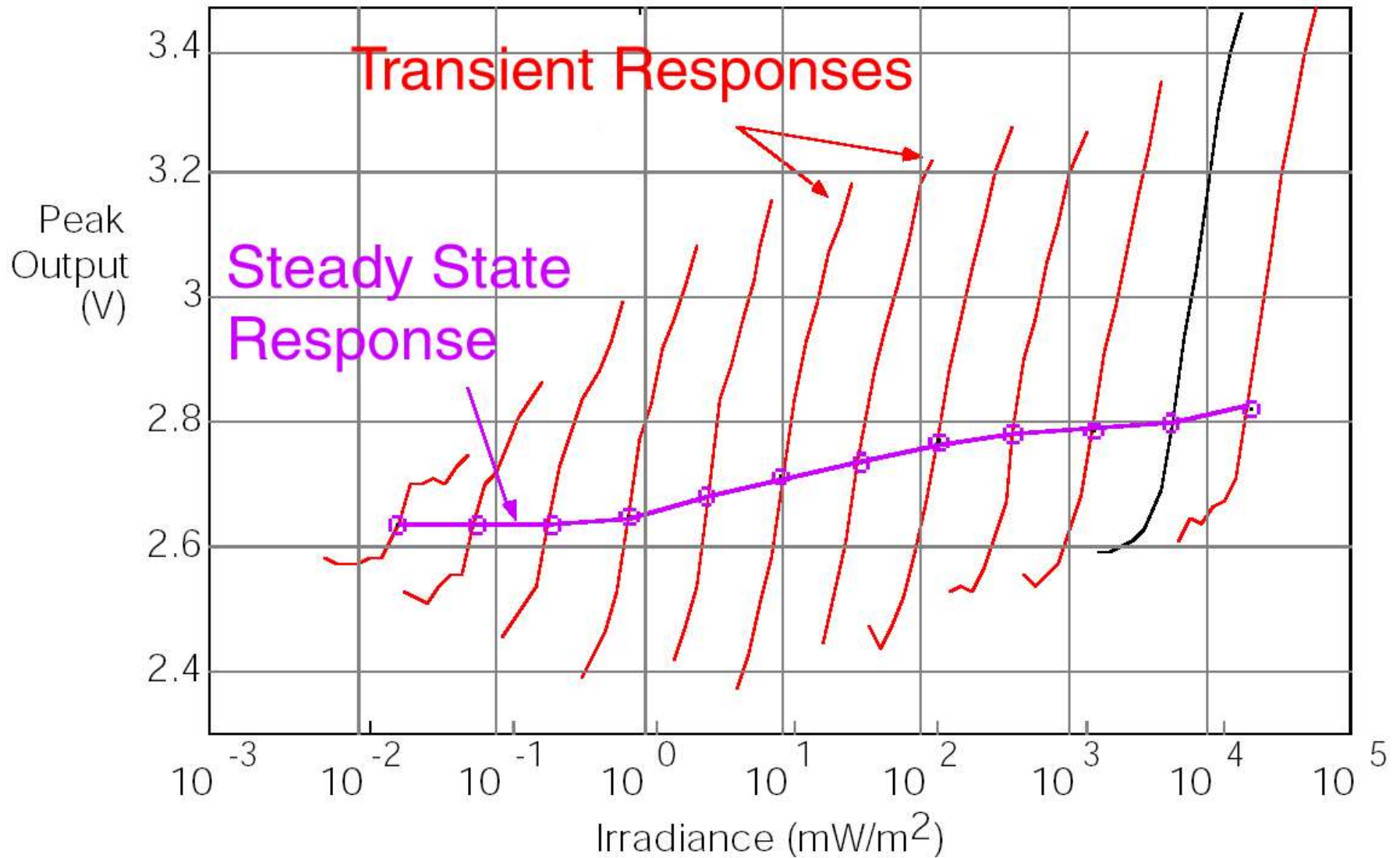
Actual circuit



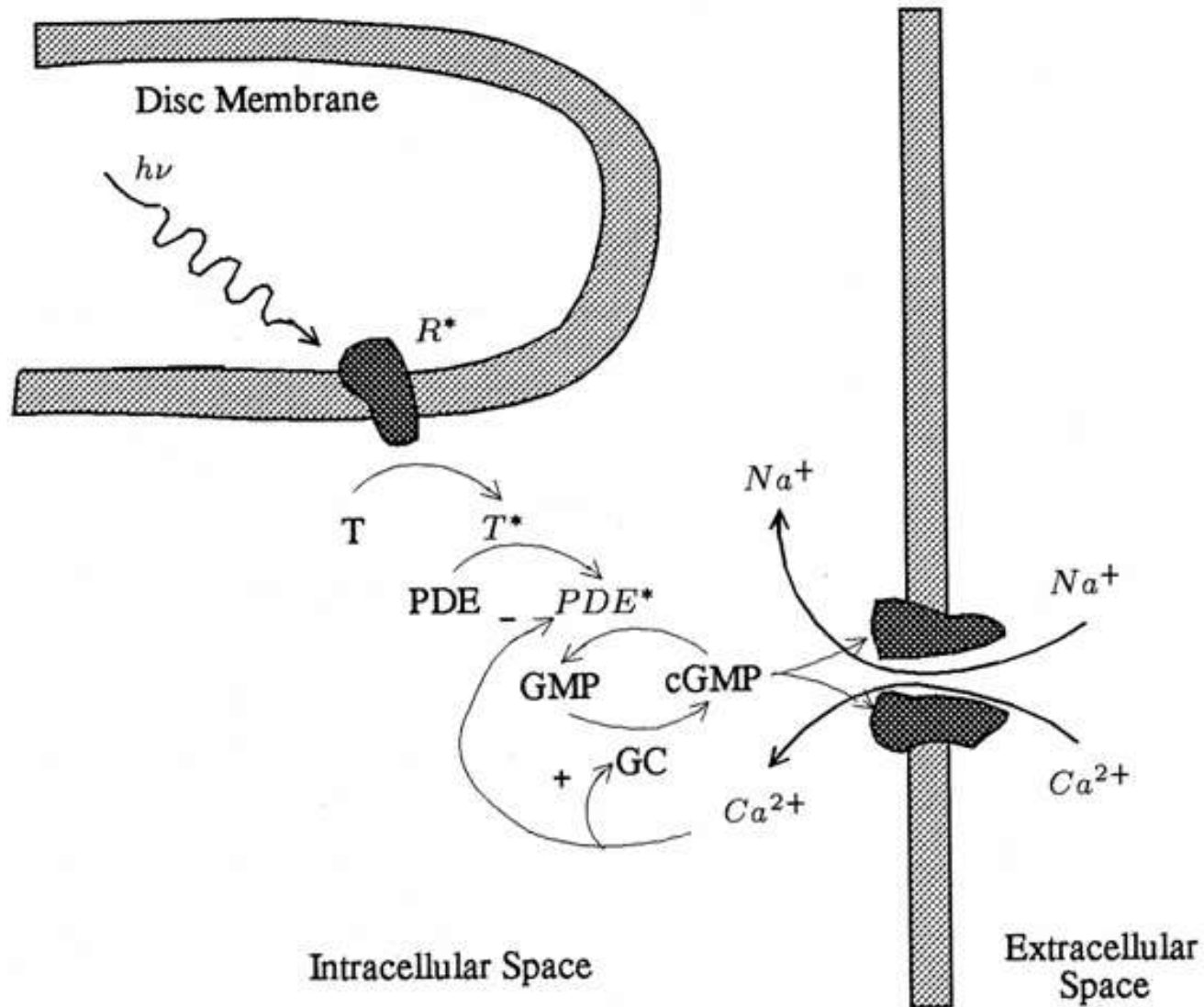
Conceptual circuit



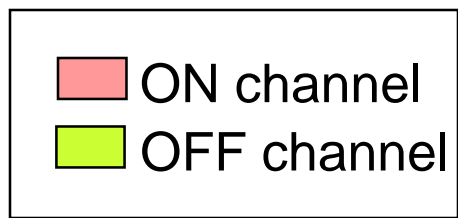
Adaptive photoreceptor responses



Biological phototransduction uses distributed chain of amplifiers



Outer segments (photo sensor photodiodes)

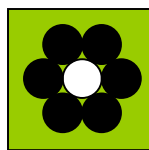
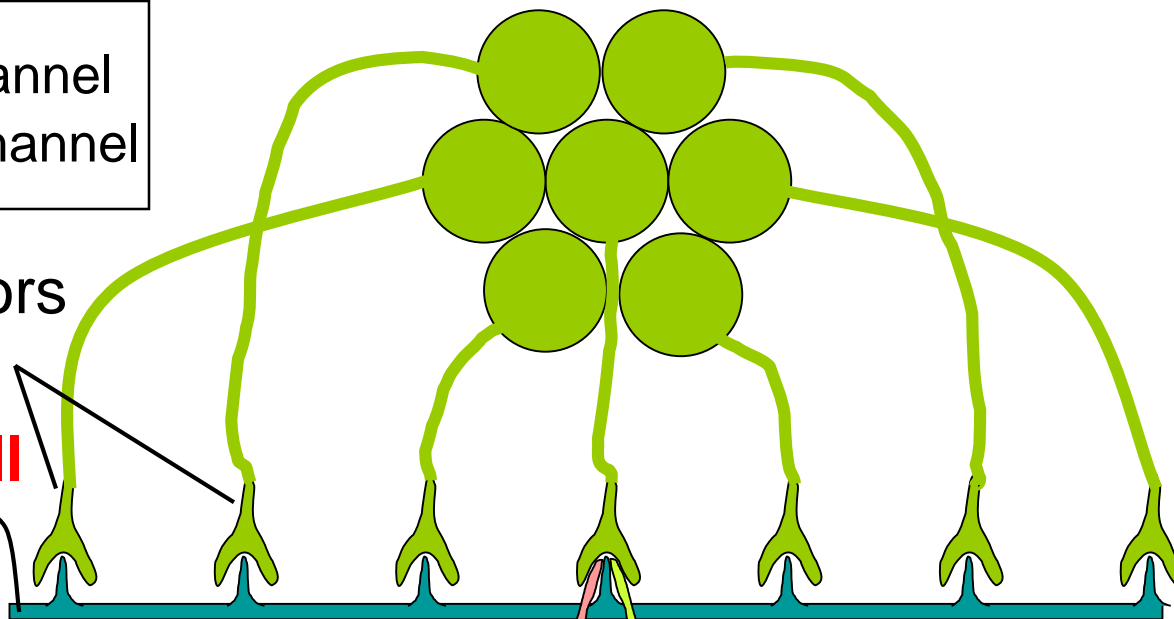


Photoreceptors

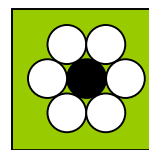
Horizontal cell

Bipolar cells

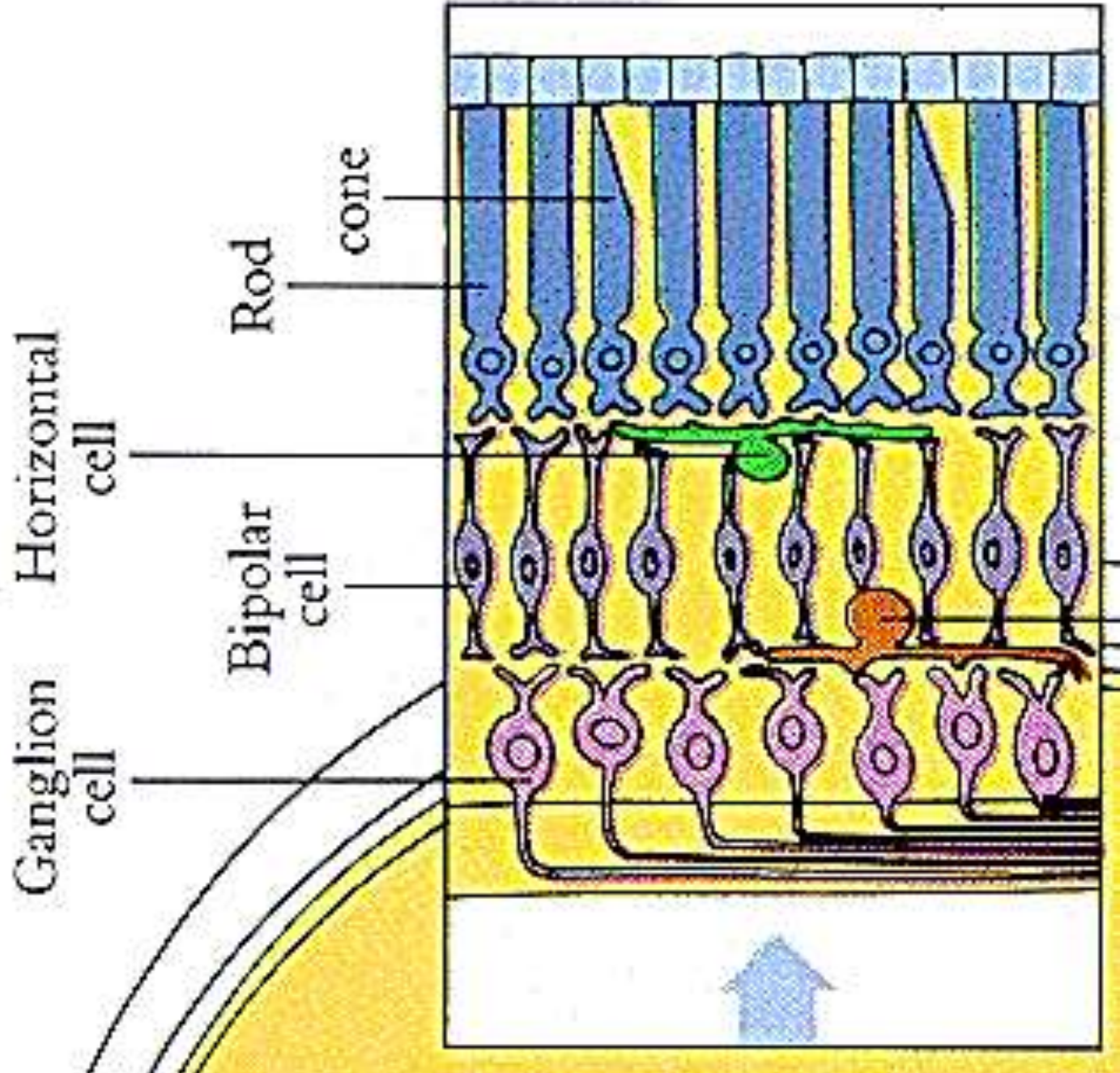
Ganglion cells

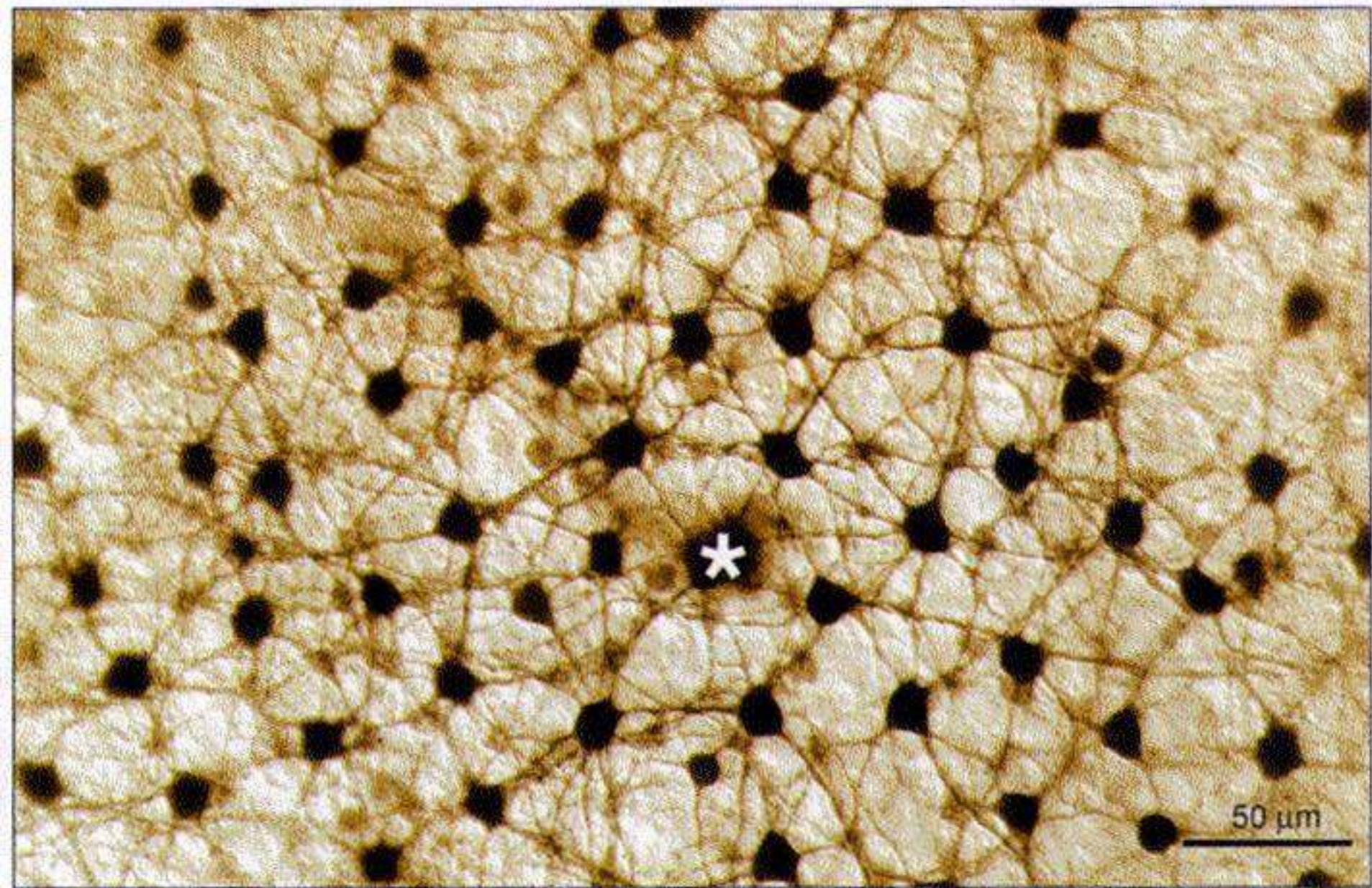


ON



OFF





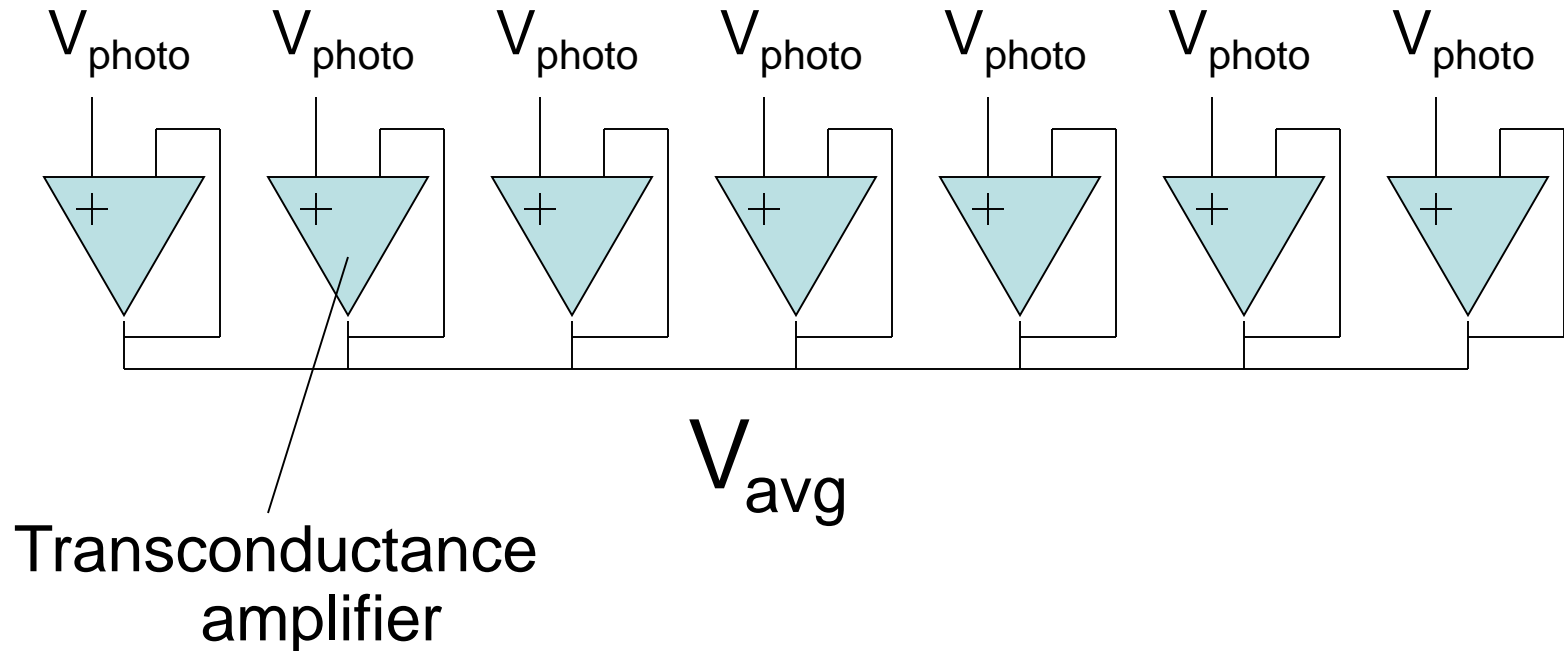
HI horizontal cells labeled following injection of one HI cell (*)

×300

after Dacey, Lee, and Stafford, 1996

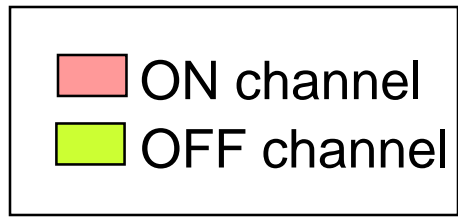
Horizontal cell

A *Follower-Aggregator* averages the photoreceptor outputs to compute the average of the inputs. This average is the *context* which is compared to the photoreceptor.



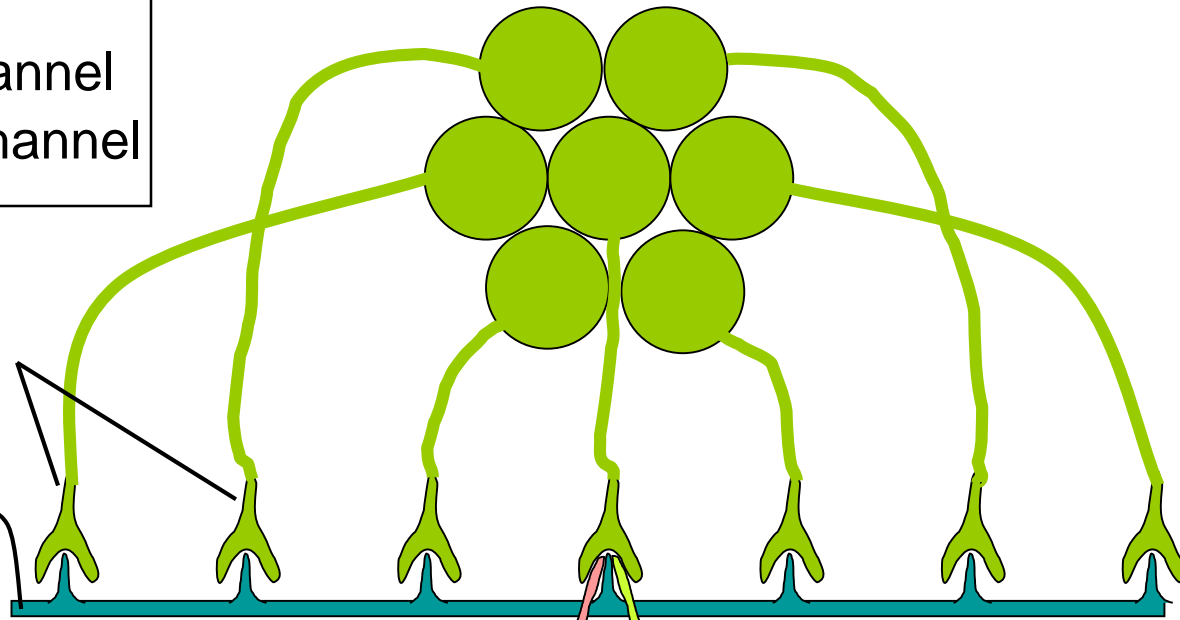
Because the follower output current saturates, the follower-aggregator computes **mean** for small signals and **median** for large signals

Outer segments (photo sensor photodiodes)



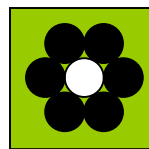
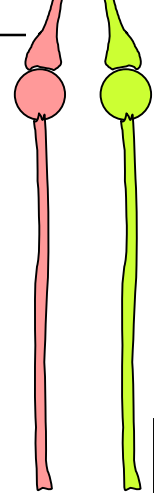
Photoreceptors

Horizontal cell

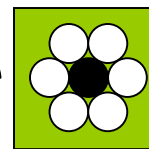


Bipolar cells

Ganglion cells

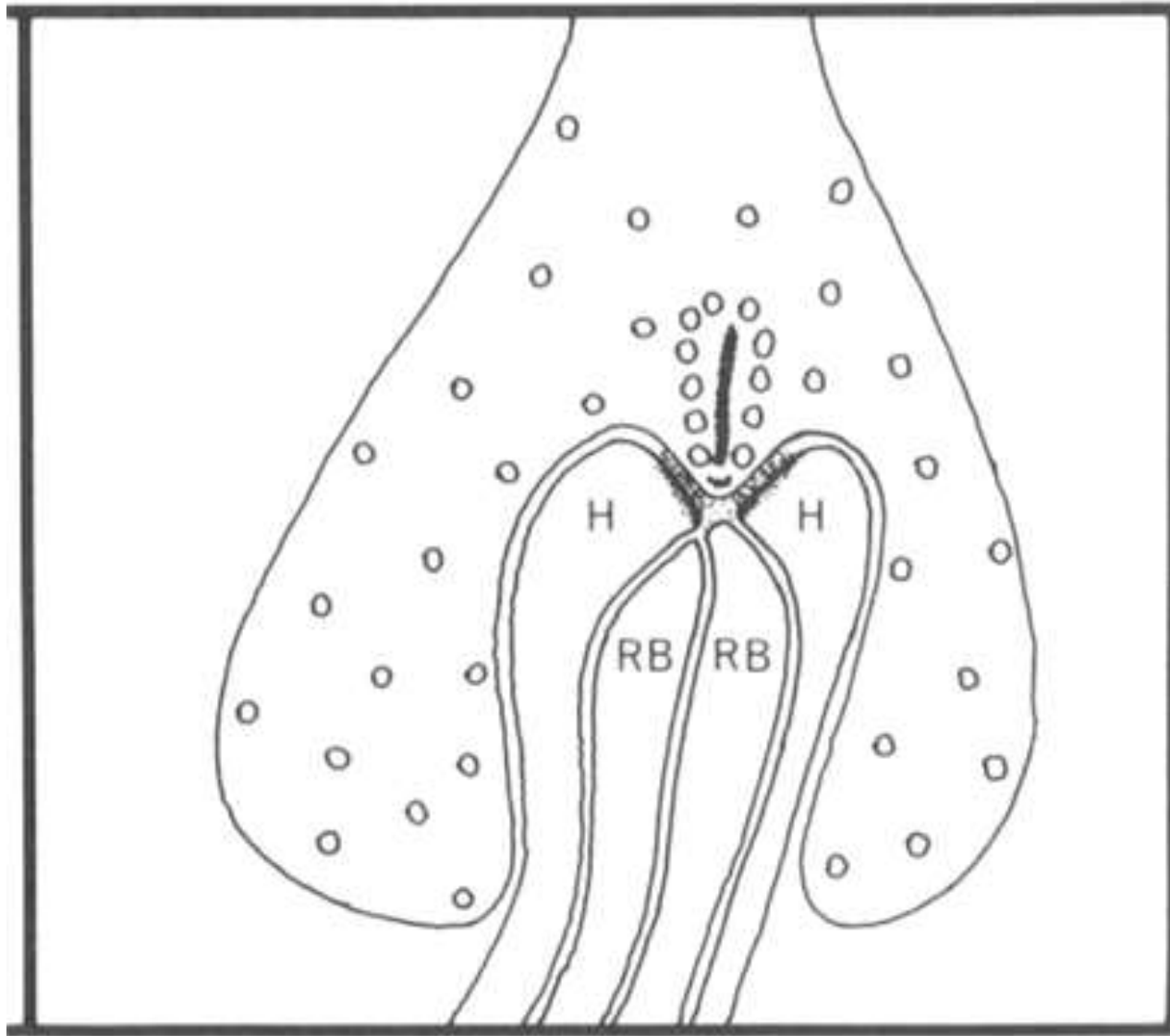


ON

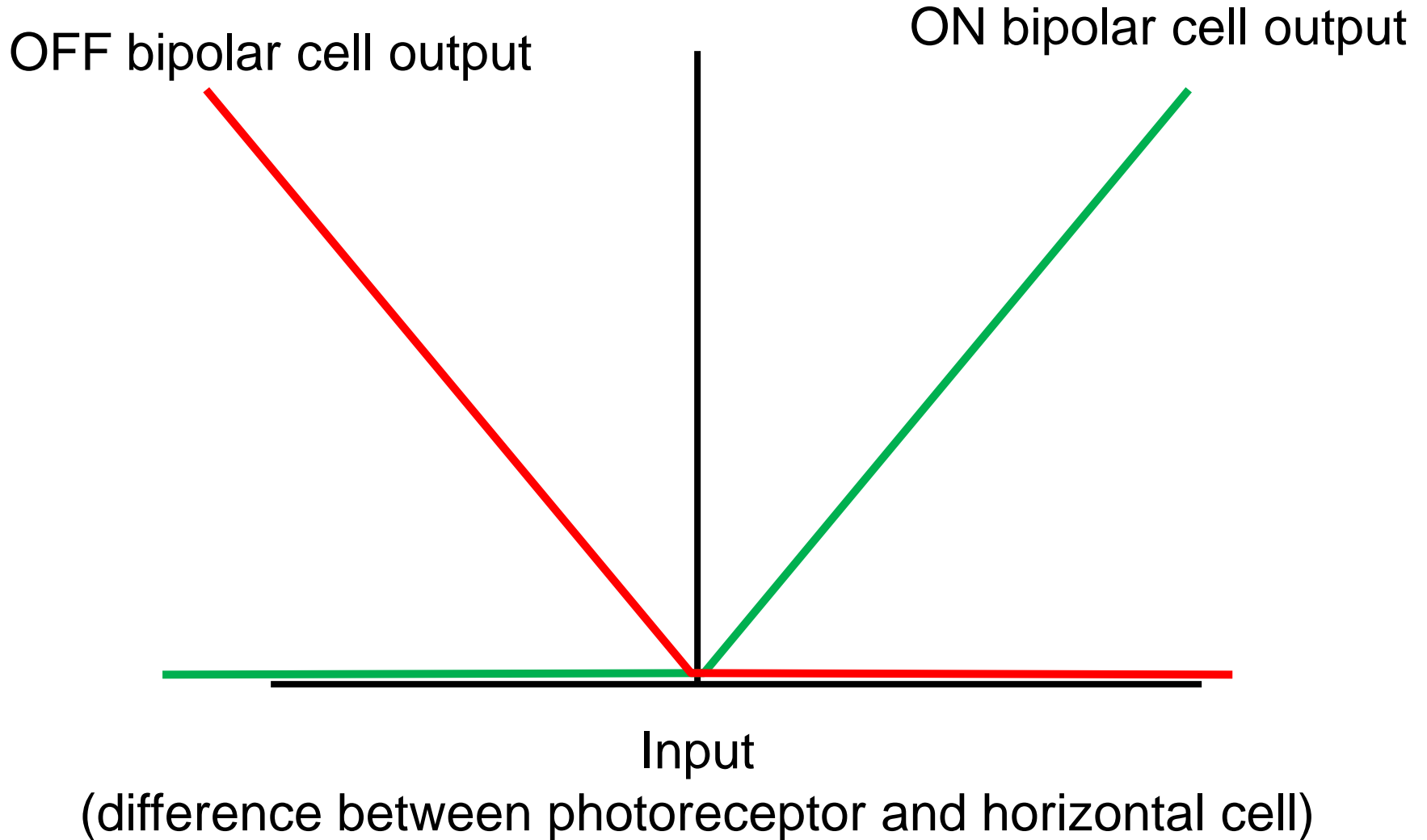


OFF

Rod-Horizontal Cell-Bipolar cell junctions

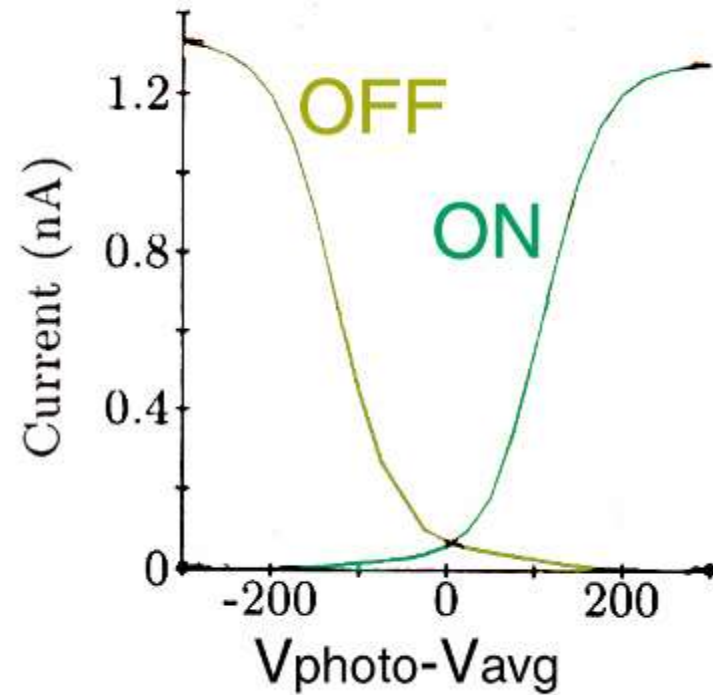
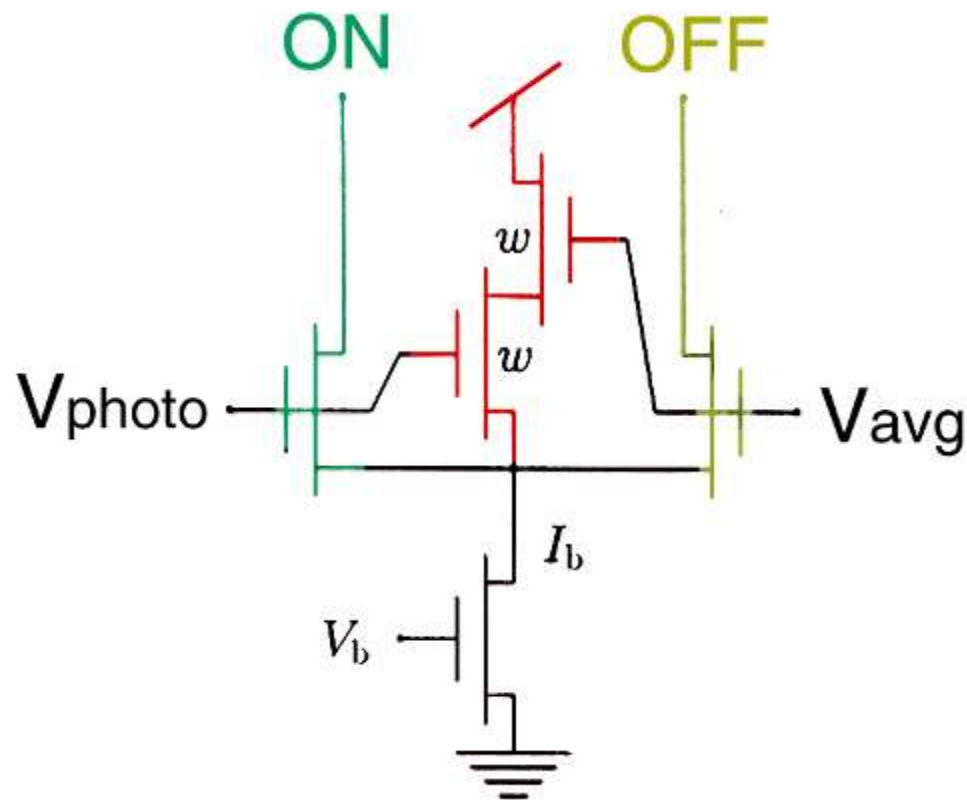


Rectification at the photoreceptor-horizontal cell-bipolar cell synapse

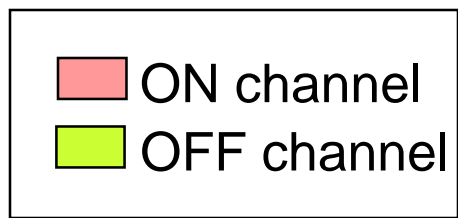


Bipolar Cell (Anti-bump circuit)

Rectifies into ON and OFF currents



Outer segments (photo sensor photodiodes)

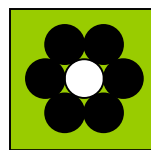
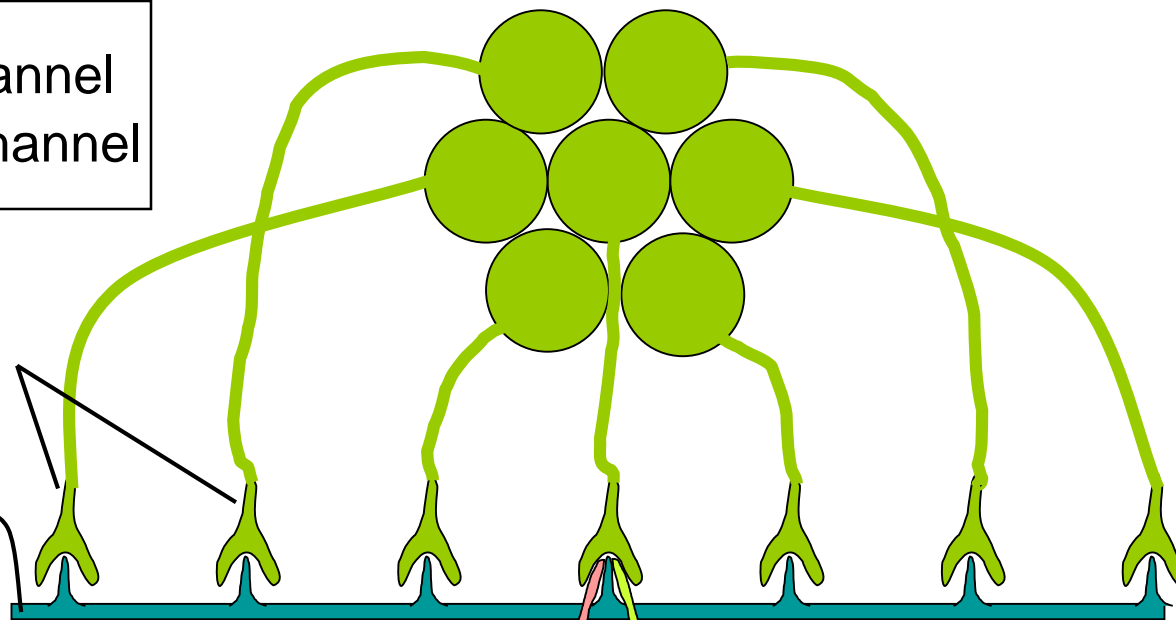


Photoreceptors

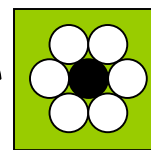
Horizontal cell

Bipolar cells

Ganglion cells

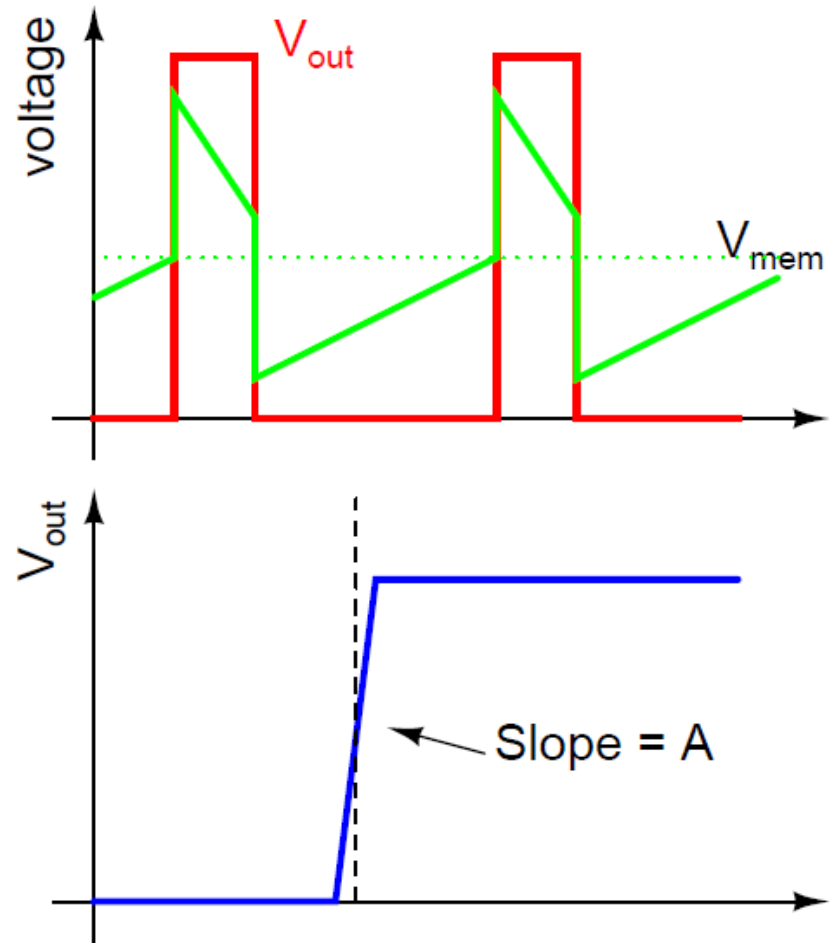
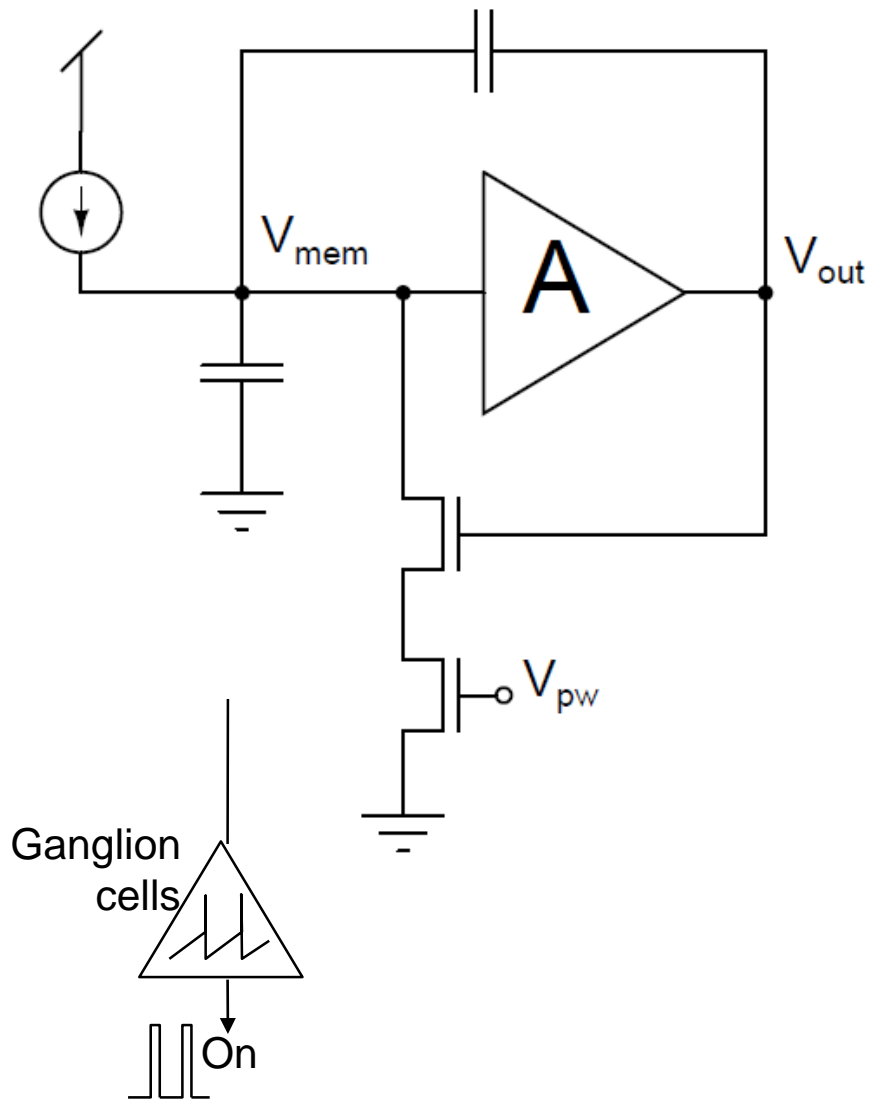


ON

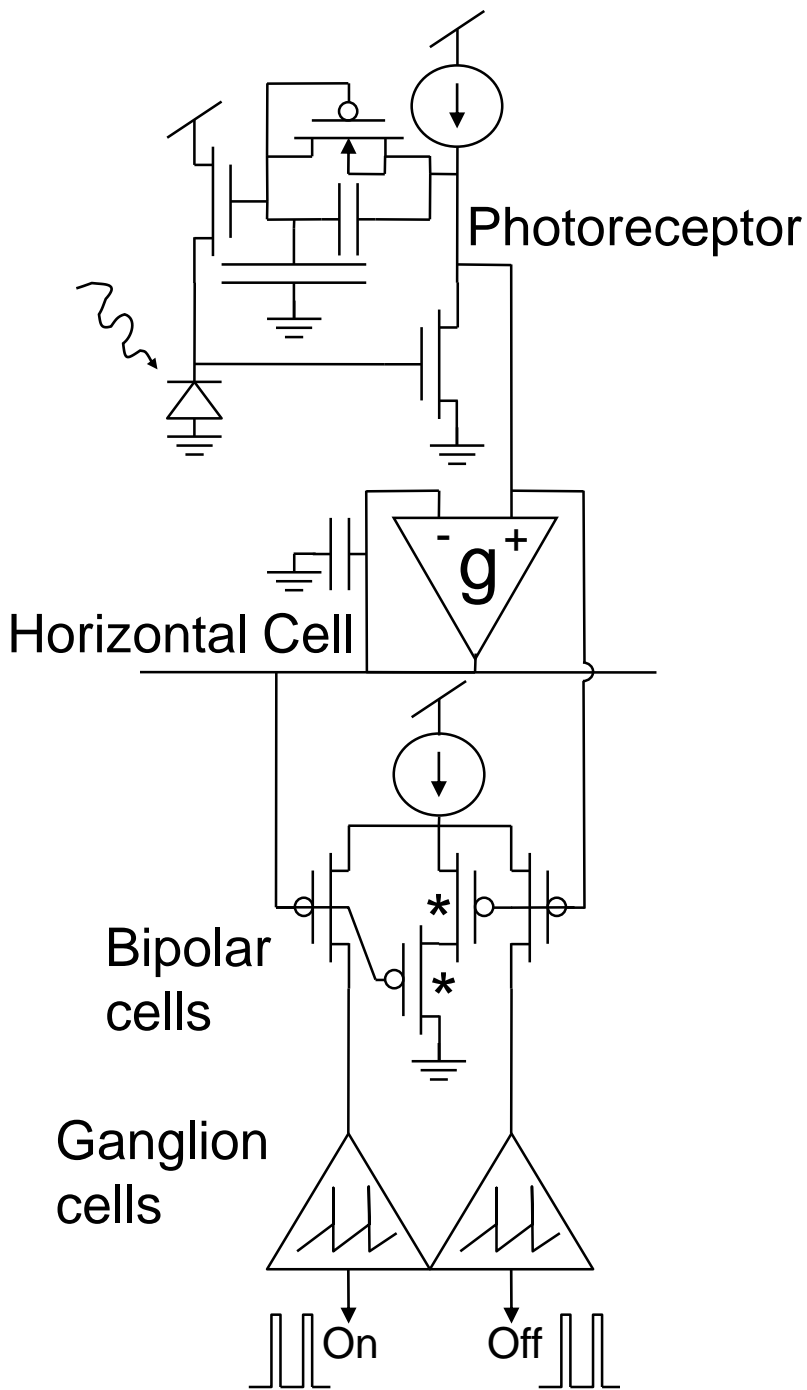


OFF

“Axon-hillock” spiking soma circuit turns the graded bipolar cell outputs into ganglion cell spikes



Complete circuit for retina part of Physiologist's Friend circuit



Neuromorphic Electronics?

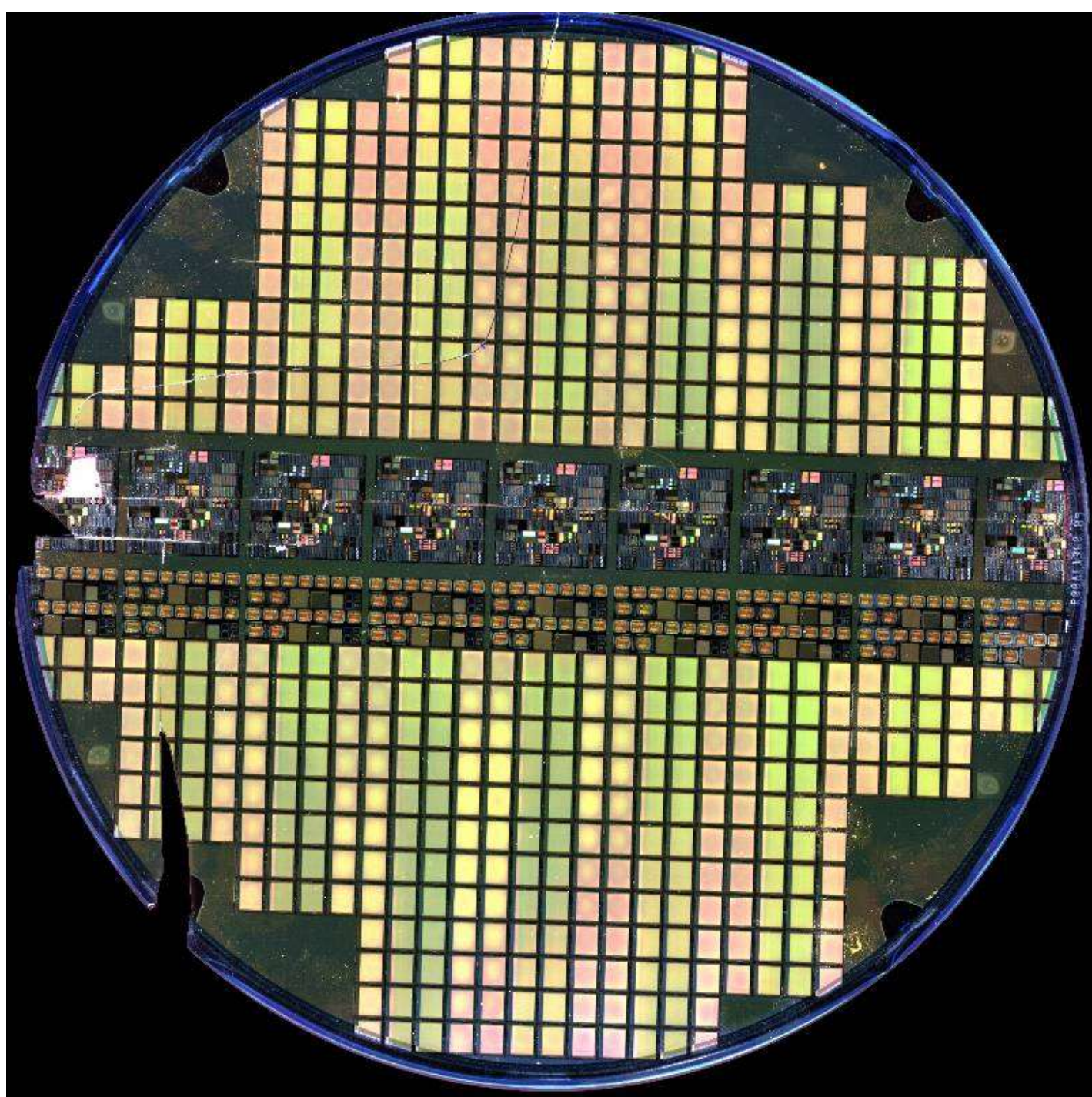
What is it all about?

The context, of silicon electronics with synchronous logic

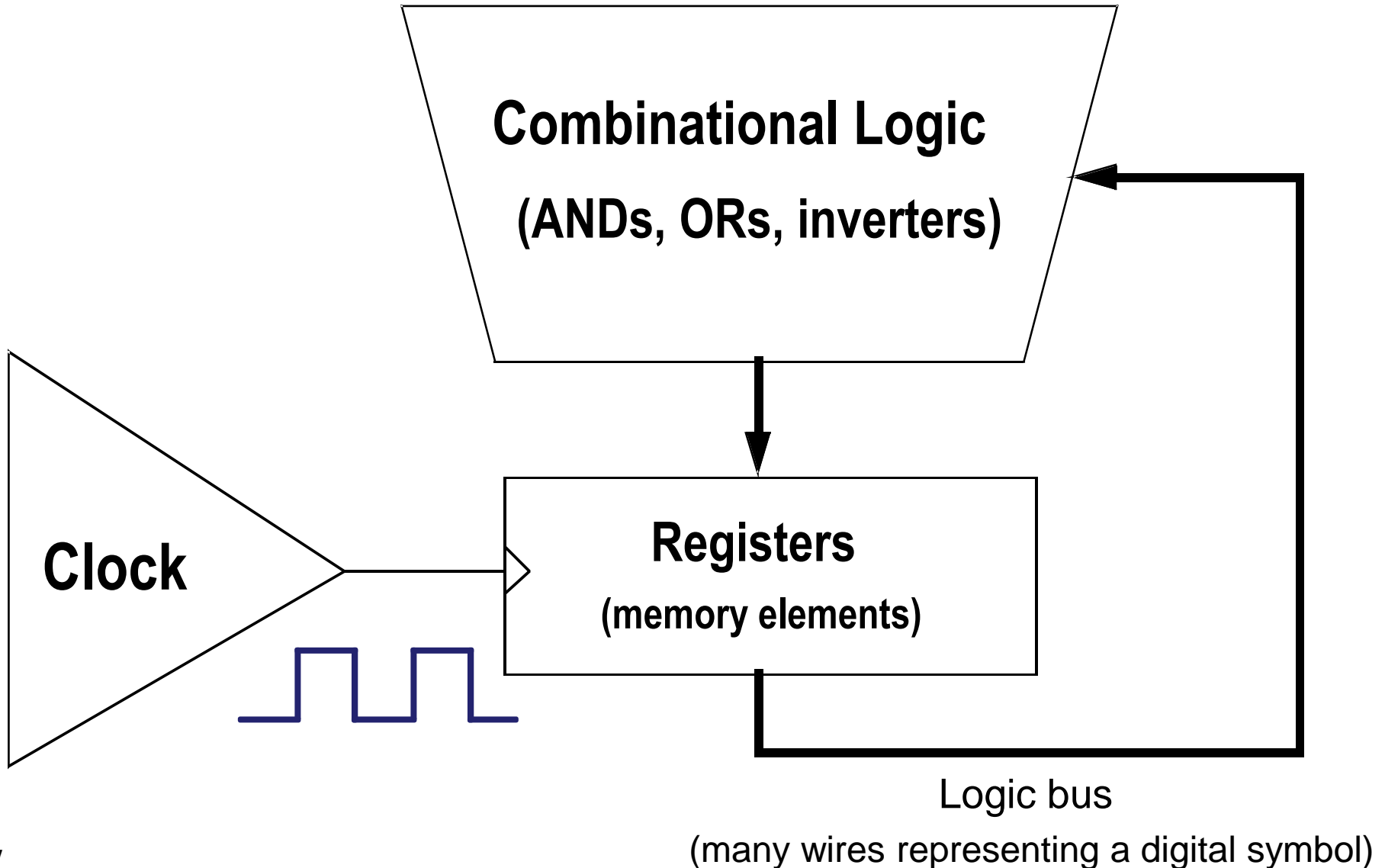
•1947

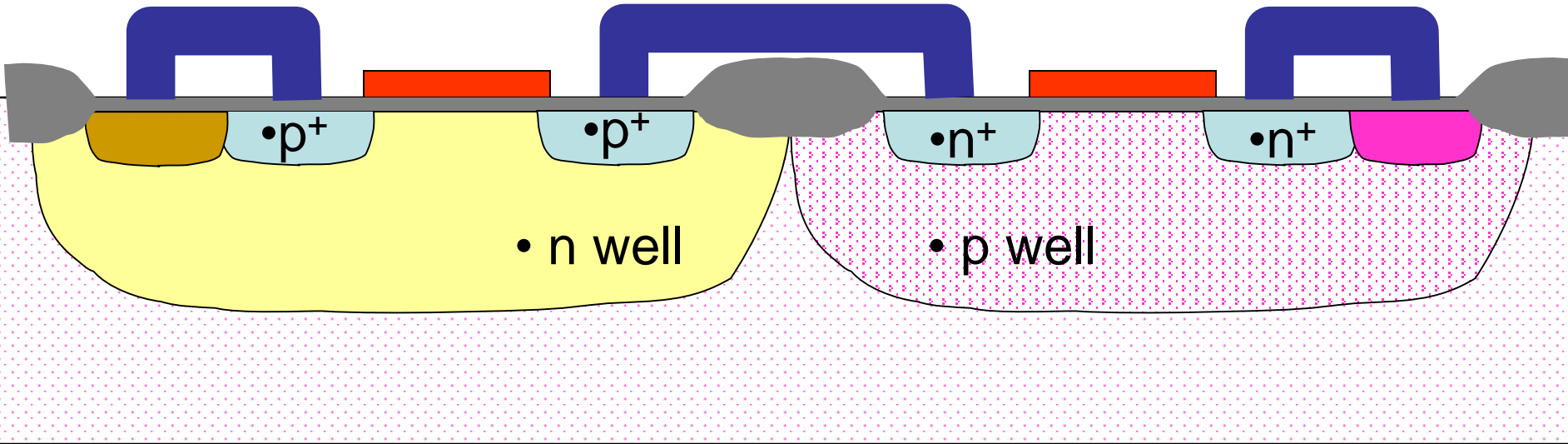
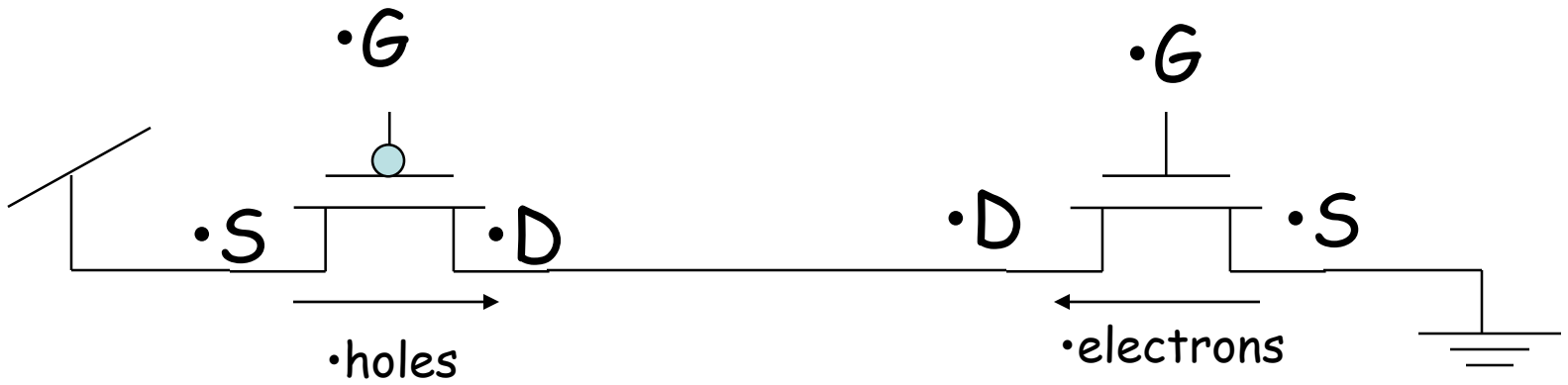


•2000

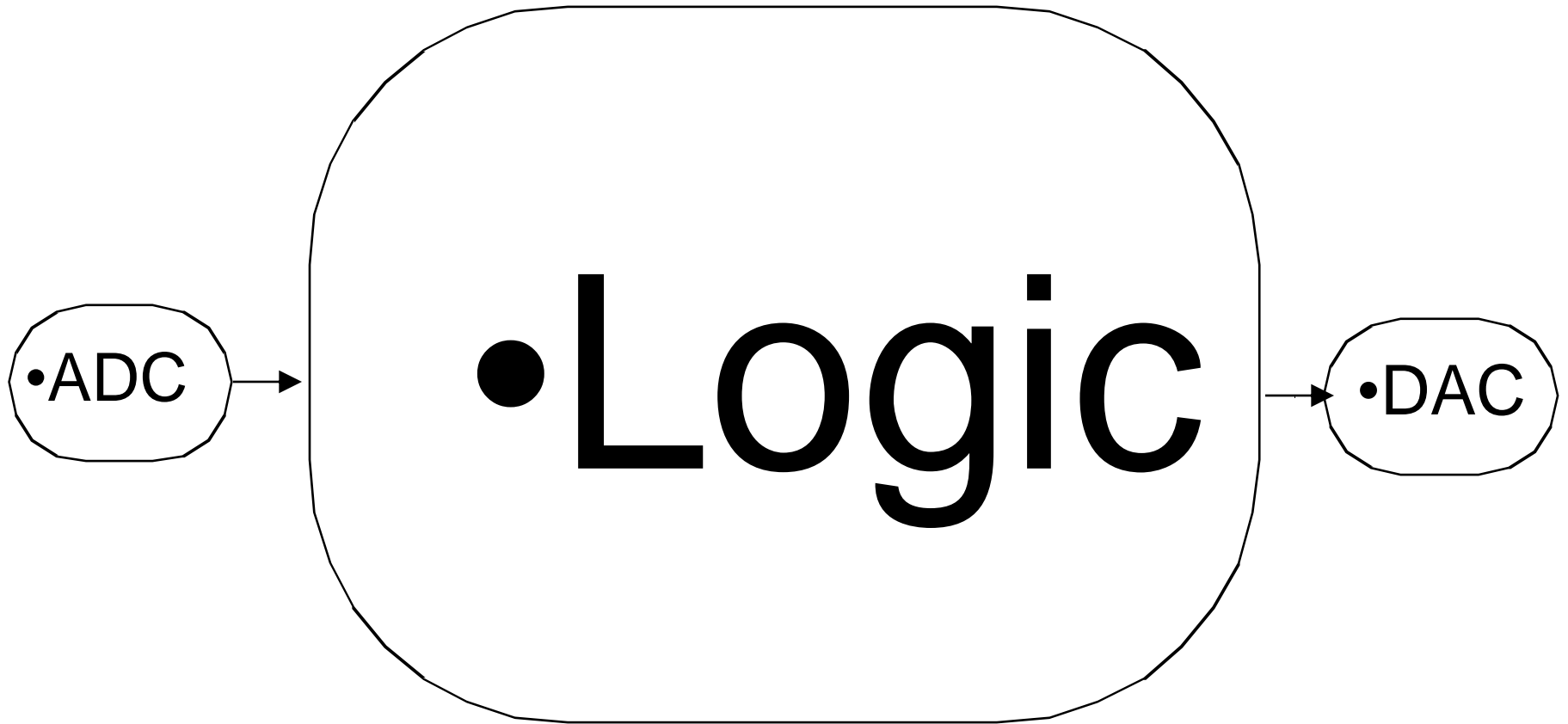


Synchronous logic is ubiquitous





Artificial real-world computation (or: How industry thinks of analog)



The motivation

• Natural computation



- Flies acrobatically
- Recognizes patterns
 - Navigates
 - Forages
- Communicates

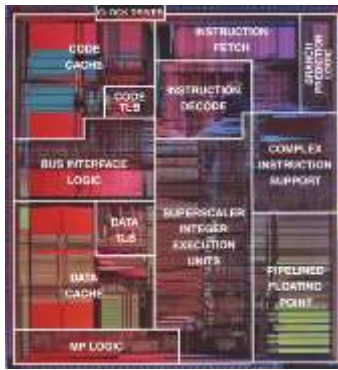
• 10^{-15} J/op

• Digital silicon 10^{-7} to 10^{-11} J/op

• 10^8 to 10^4 times as efficient as digital silicon

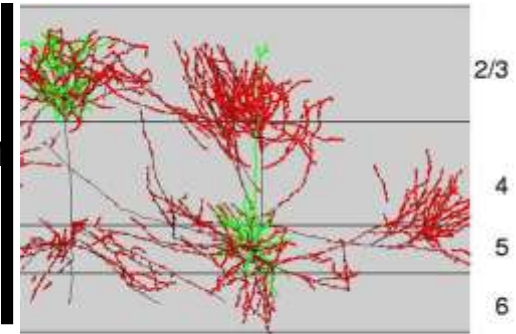
•Computer vs. Brain

•Pentium 4



•Cortex

•1mm



•Anderson et al. 2003
wm

At the system level, brains are about 1 million times more power efficient than computers. Why?

Cost of elementary operation (turning on transistor or activating synapse) is about the same. It's not some magic about physics.

Computer

Brain

Fast global clock

Self-timed

Bit-perfect deterministic logical state

Synapses are stochastic!

Computation dances: digital→analog→digital

Memory distant to computation

Memory at computation

Fast high precision power hungry ADCs

Low precision adaptive data-driven quantizers

Devices frozen on fabrication

Constant adaptation and self-modification

•1967



•C. Mead



•M. Delbruck



•P. Mueller

•E. Vittoz

•1980
s



•CSEM



•C. Koch

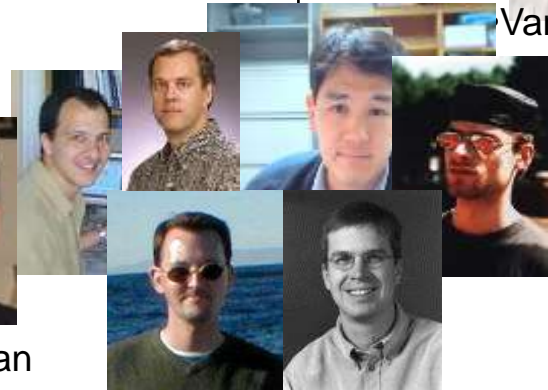


Van Schaik

•Mid to late 1990s

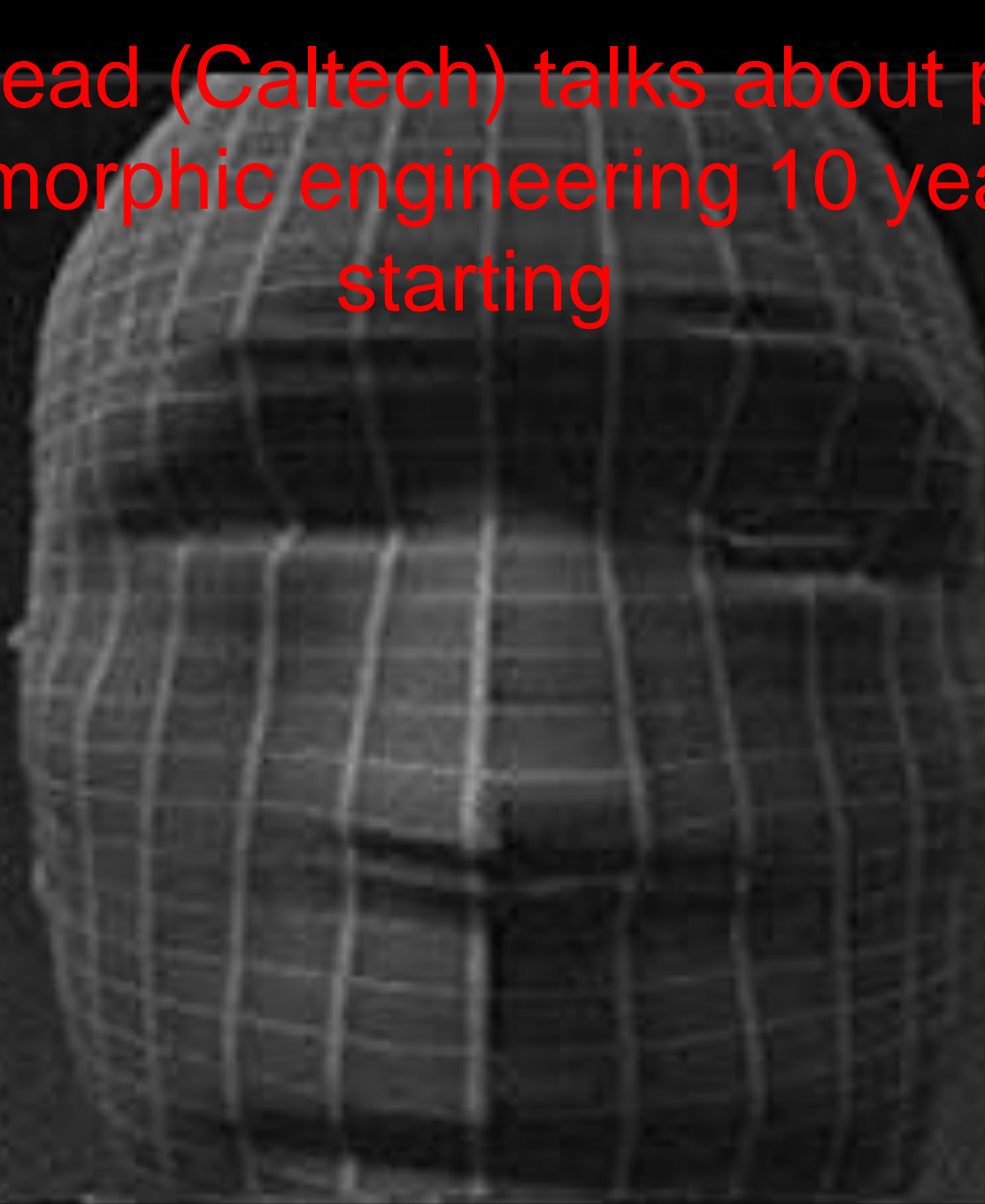


•R. Sarpeshkar •B. Minch •C. Diorio •P. Hasler •S-C. Liu •K. Boahen •P. Julian



•Liu, 2005

Carver Mead (Caltech) talks about progress
in neuromorphic engineering 10 years after
starting



The fact that we can build devices that implement the same basic operations as those the nervous system uses leads to the inevitable conclusion that we should be able to build entire systems based on the **organizing principles** used by the nervous system.

C. Mead, Proc. IEEE, 1990

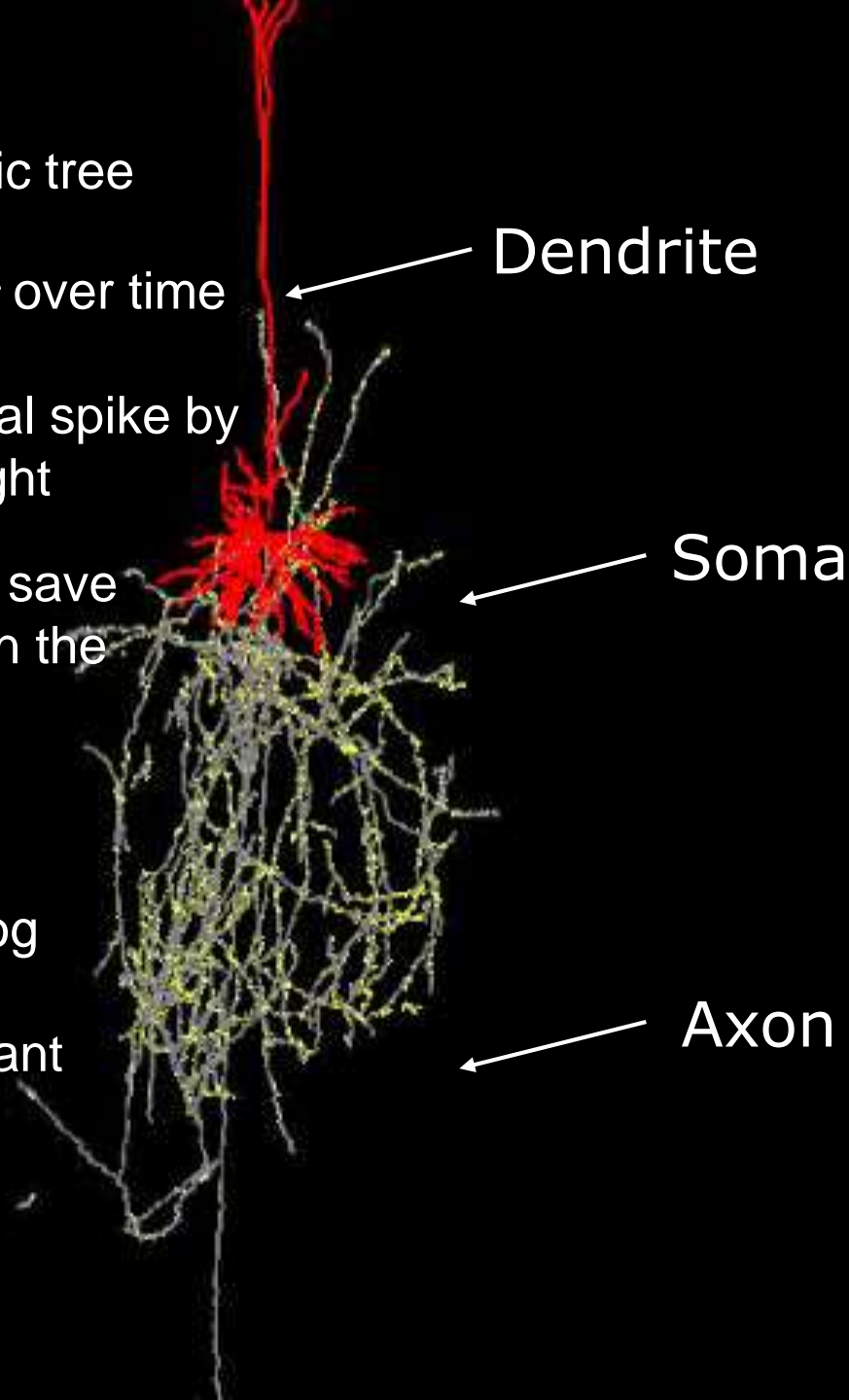
Summation on dendritic tree

Capacitance *integrates* over time

Synapses *multiply* the digital spike by a probabilistic weight

Complementary channels save power by push and pull on the membrane voltage

Dendrites do local analog computation,
Axon communicates distant digital spike events

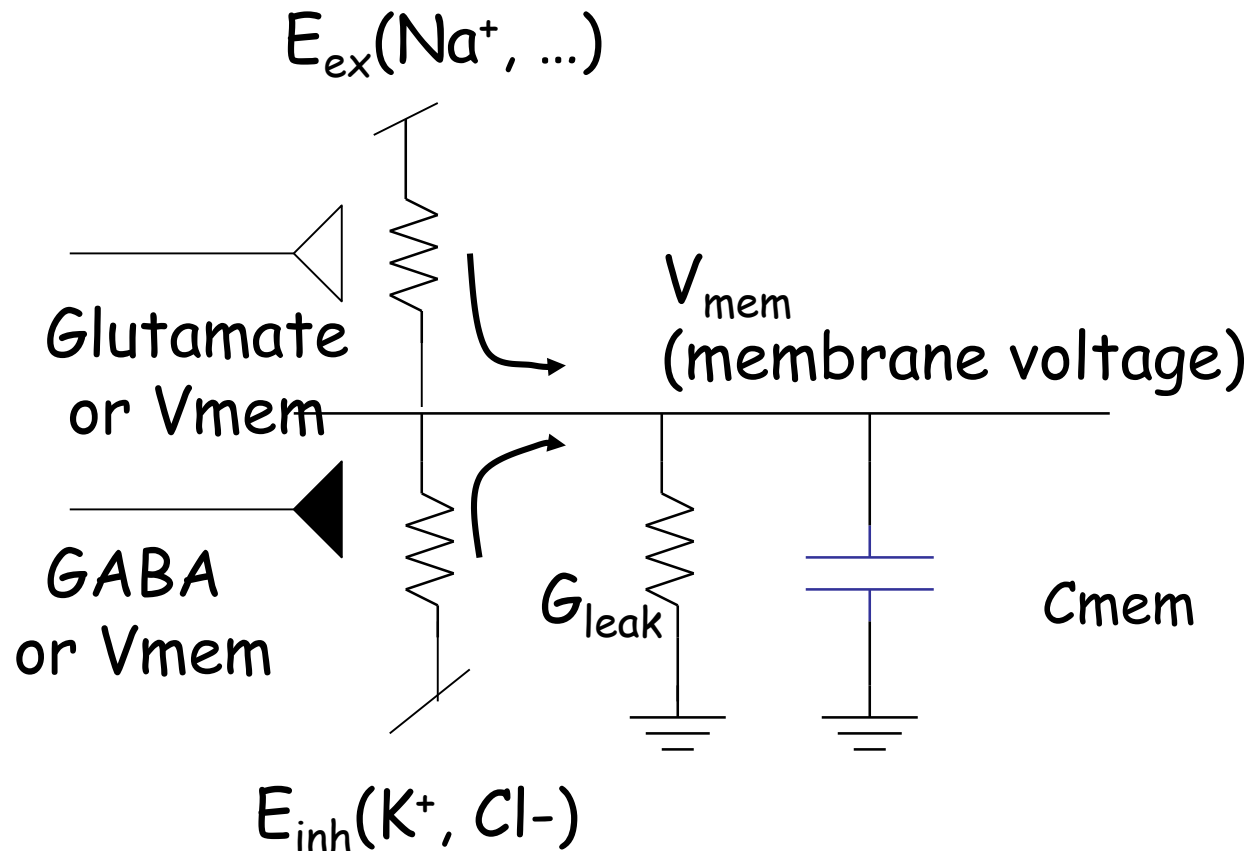


Dendrite

Soma

Axon

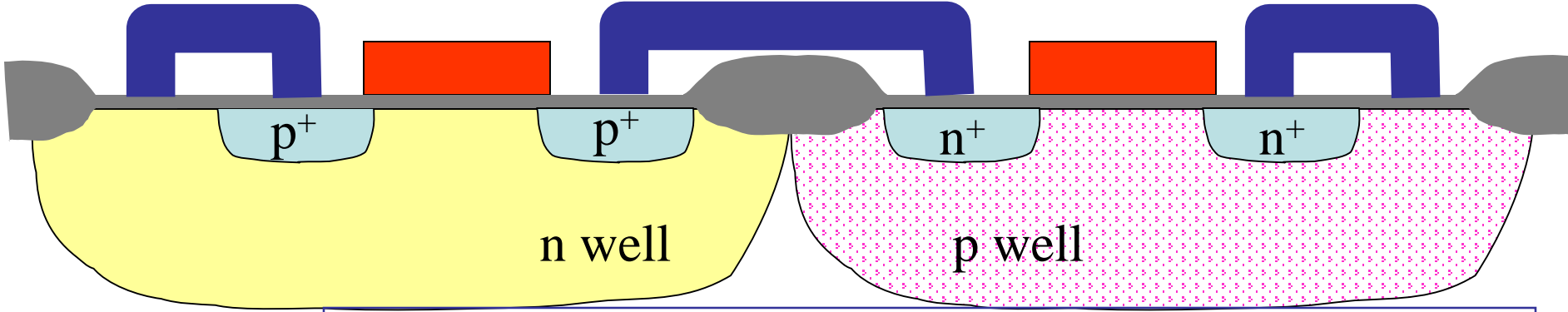
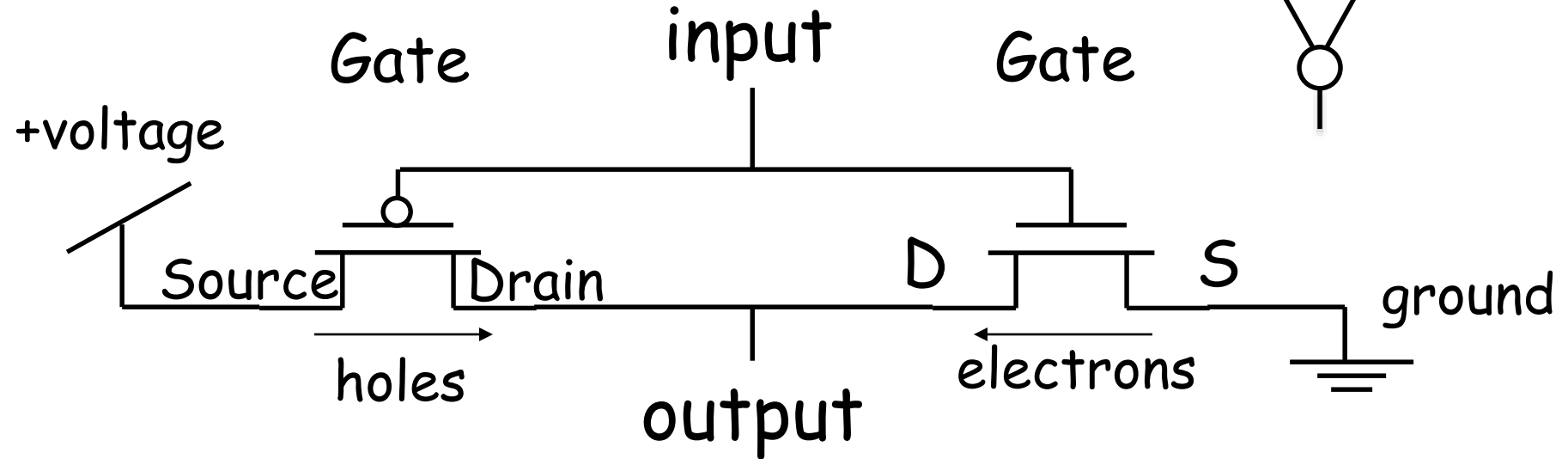
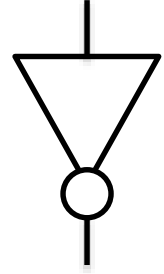
The membrane voltage is controlled by complementary transmitter or voltage gated channels



Almost no power is burned when both channels are off!

CMOS (complementary metal oxide semiconductor)

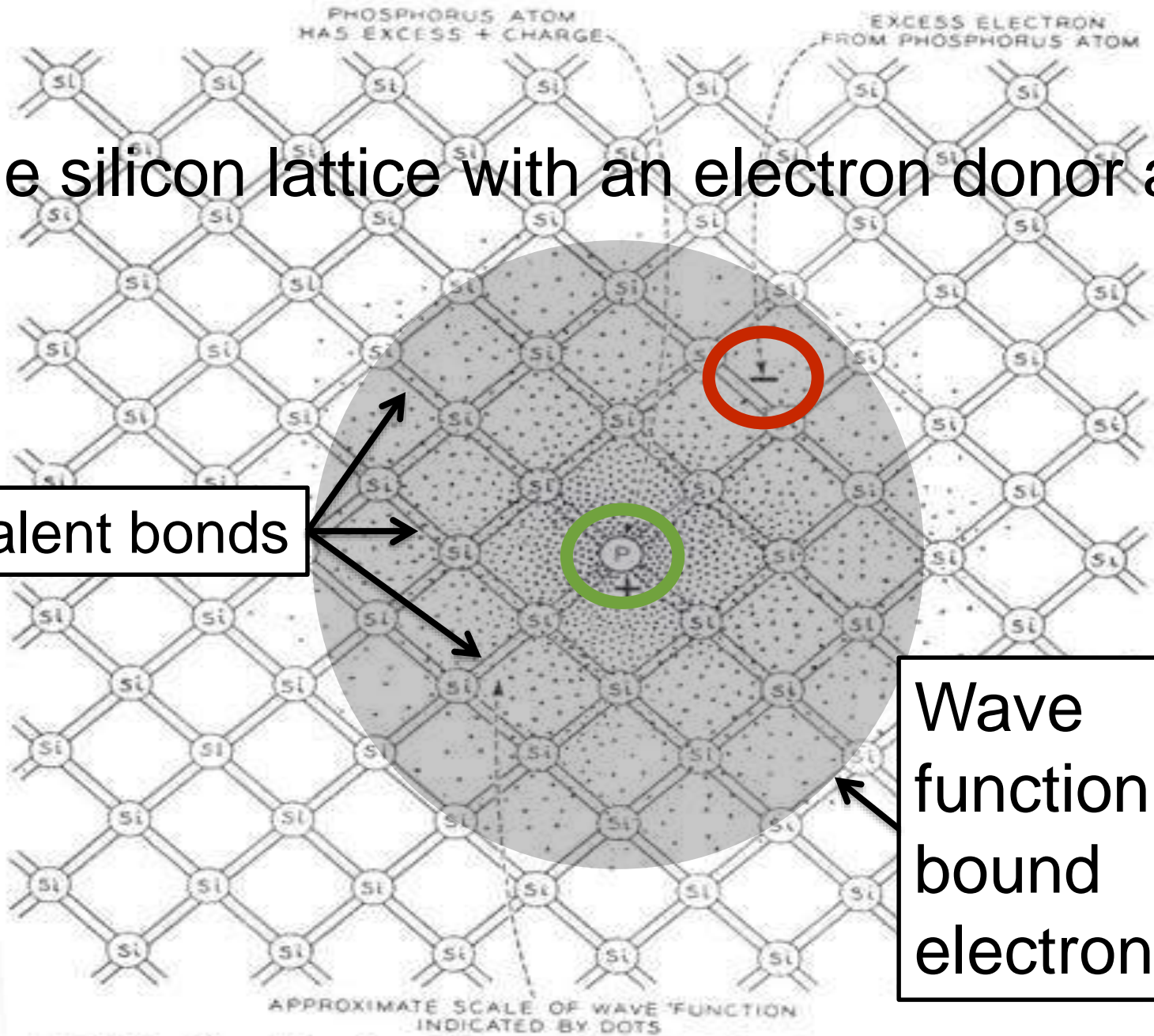
A CMOS inverting amplifier



Organizing principle:
Use complementary devices to generate gain
without burning static power

Interlude on semiconductors and transistors

The silicon lattice with an electron donor atom

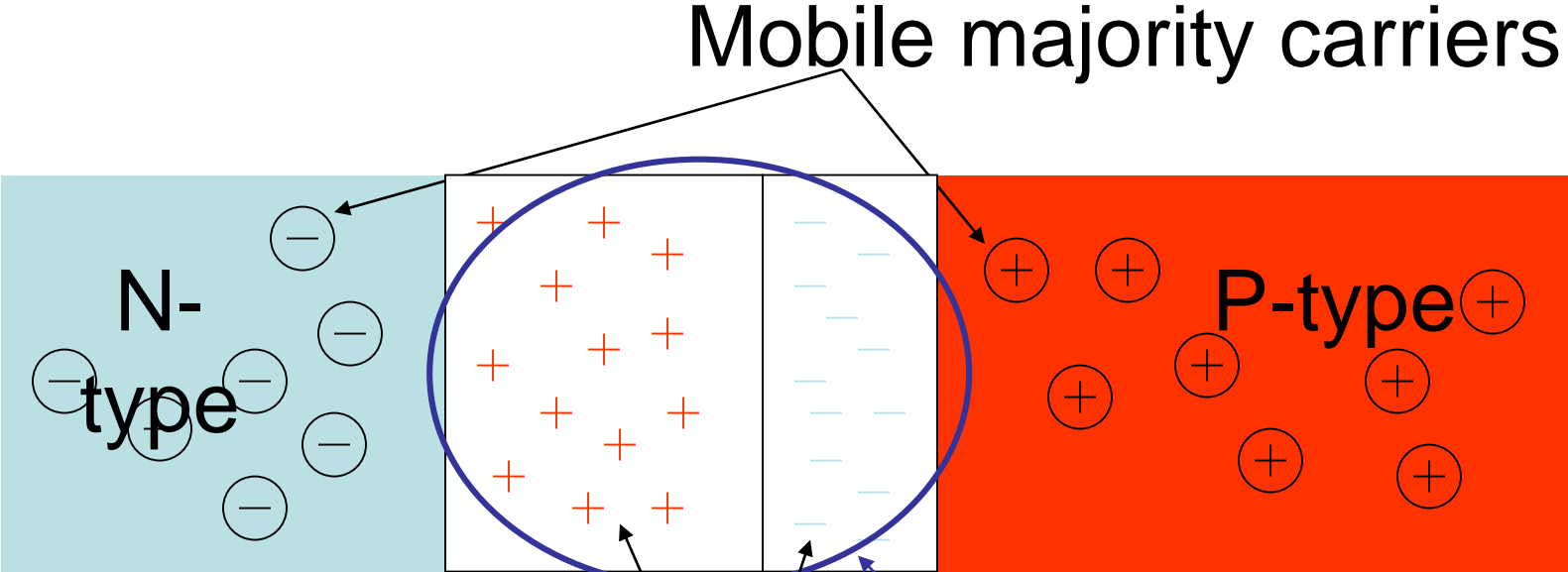


Covalent bonds

Wave function of bound electron

FIG. 1-14—Wave Function of Electron Bound to Phosphorous Atom in Silicon.

A P-N junction



Mobile majority carriers

N-type

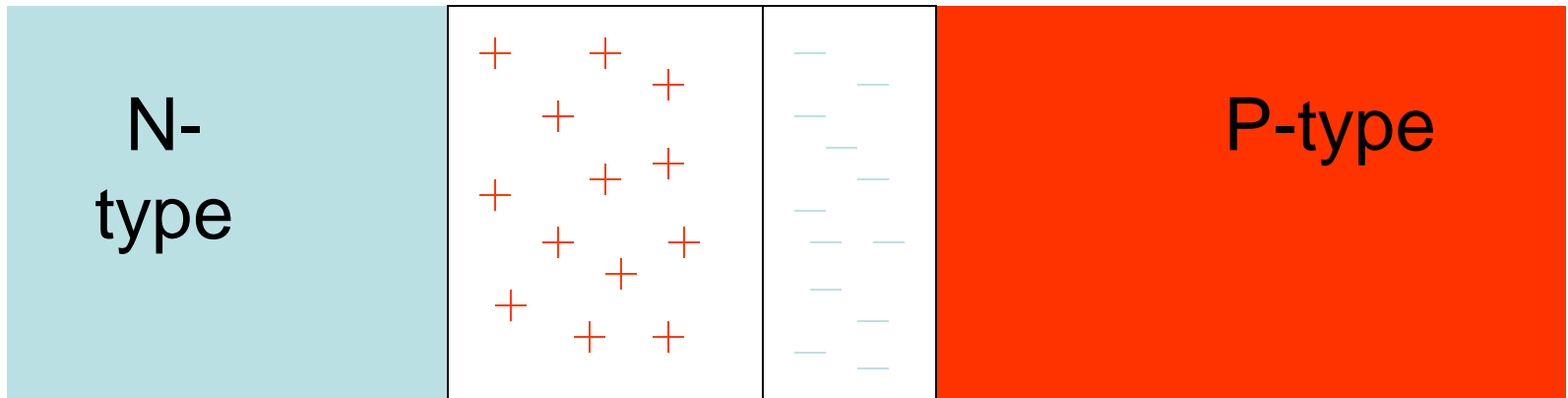
P-type

Fixed ions

“Depletion region”

“Space-charge region”

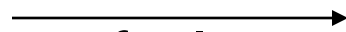
A P-N junction



Electric field



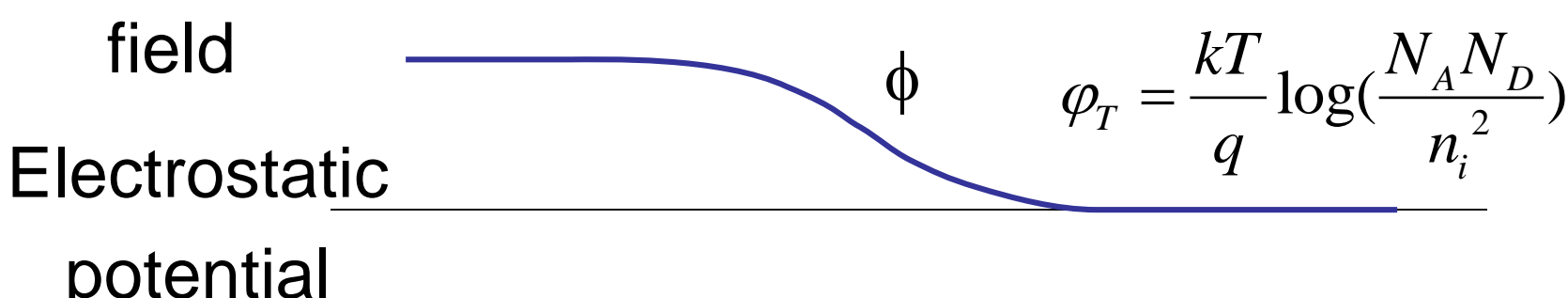
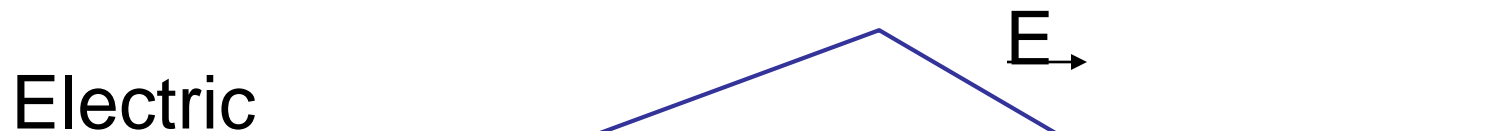
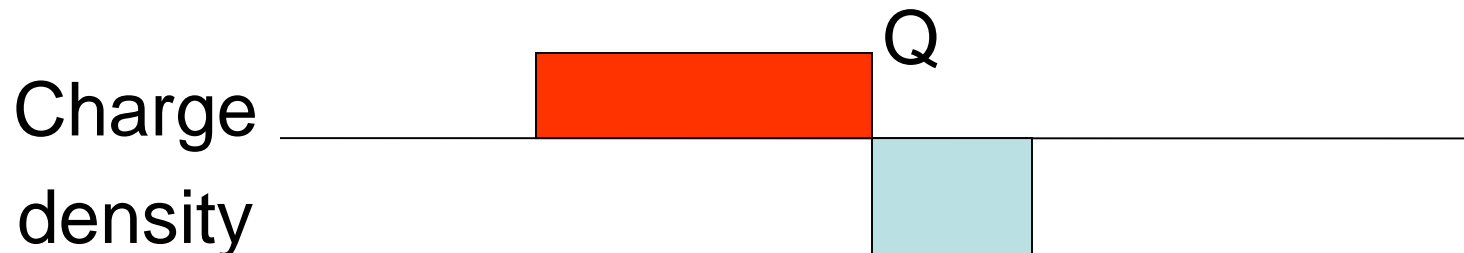
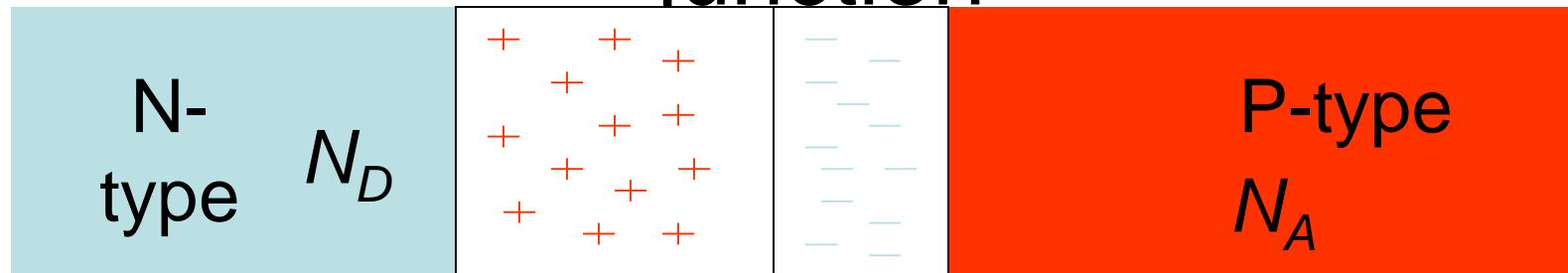
Diffusion of holes from p region



Diffusion of electrons from n region

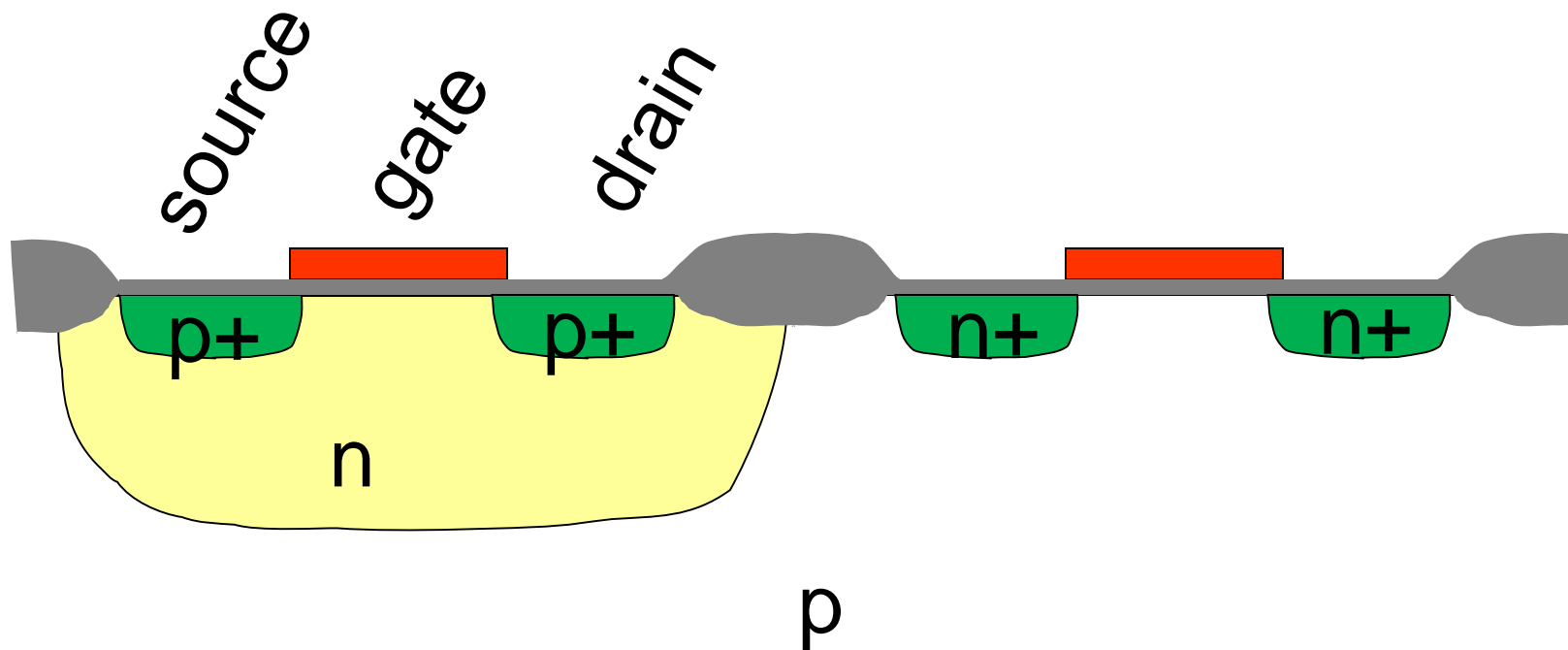
In equilibrium, $Drift = Diffusion$ for electrons *and* holes

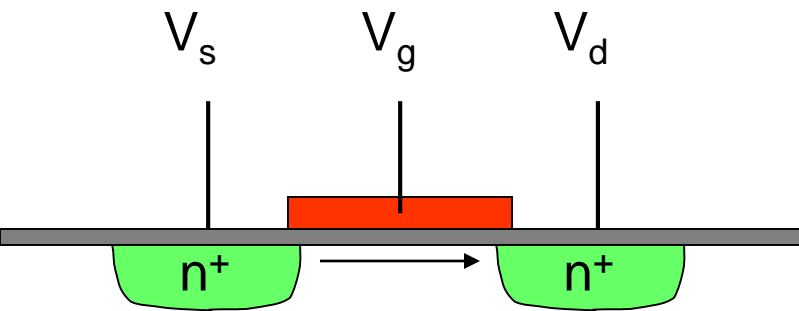
Charges, fields, and potentials in a PN junction



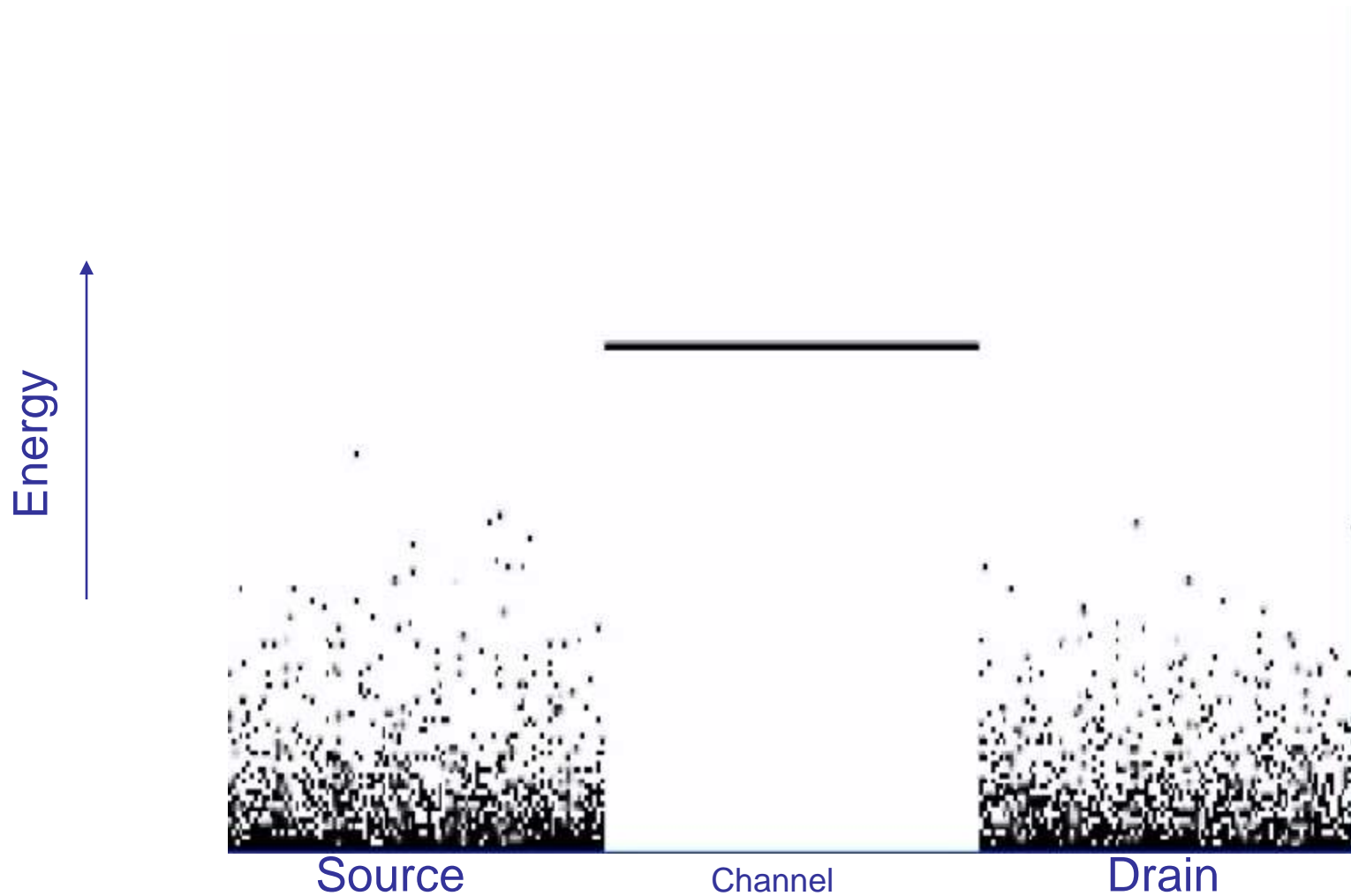
Typically, the *built-in voltage*, ϕ_T , is about 0.75V

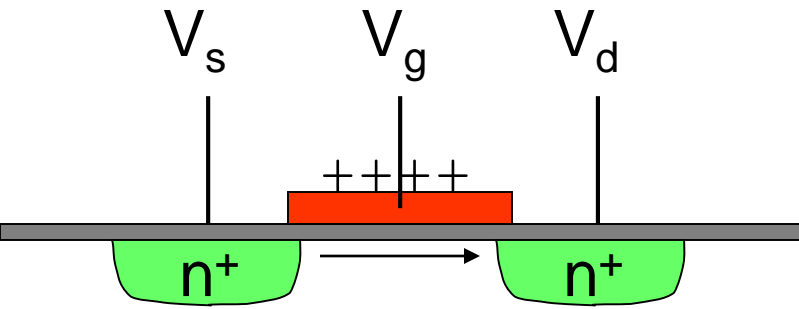
MOS transistors use insulated gates to control barrier energies at PN surface junctions at source and drain



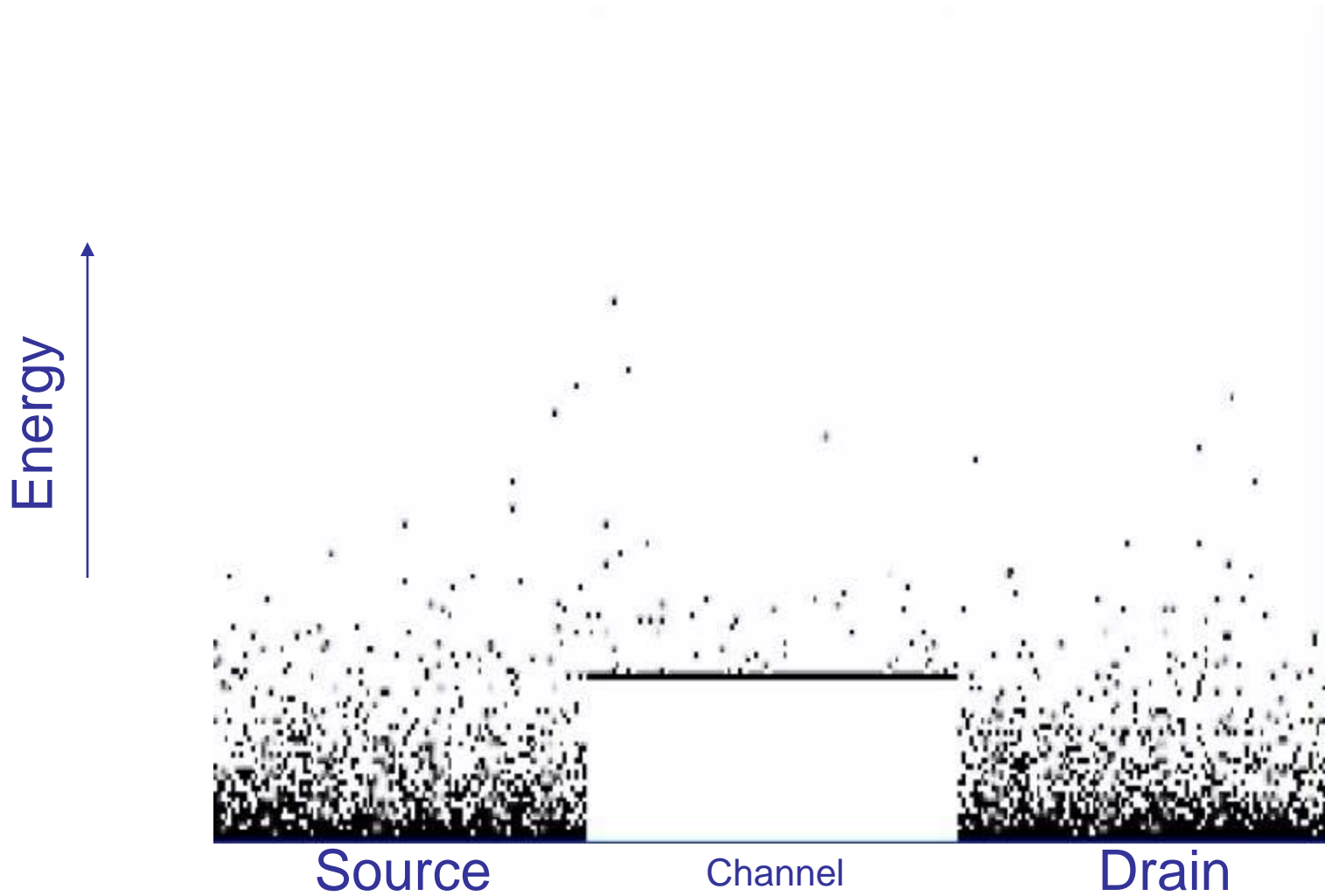


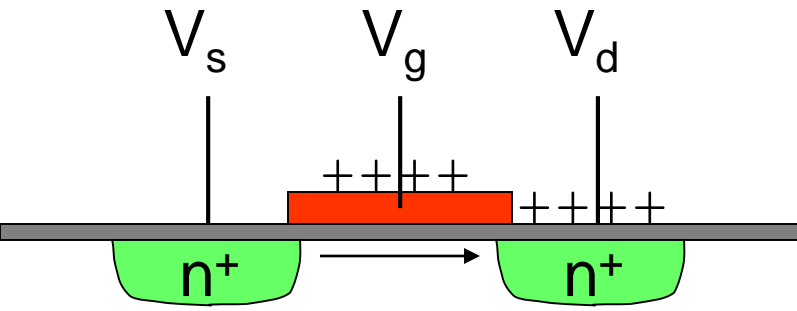
Small V_g s, V_d s=0



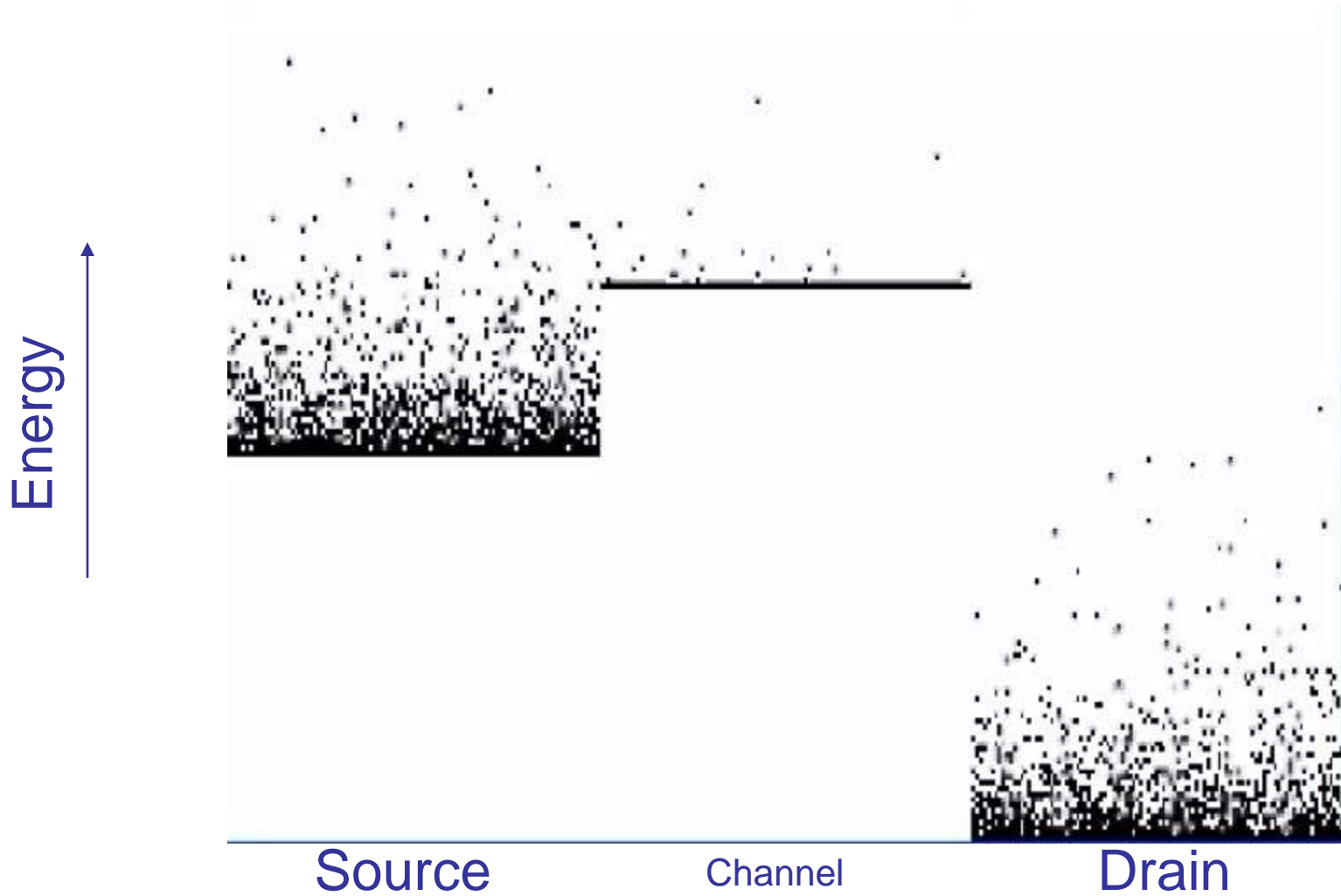


Larger V_{gs} , $V_{ds}=0$

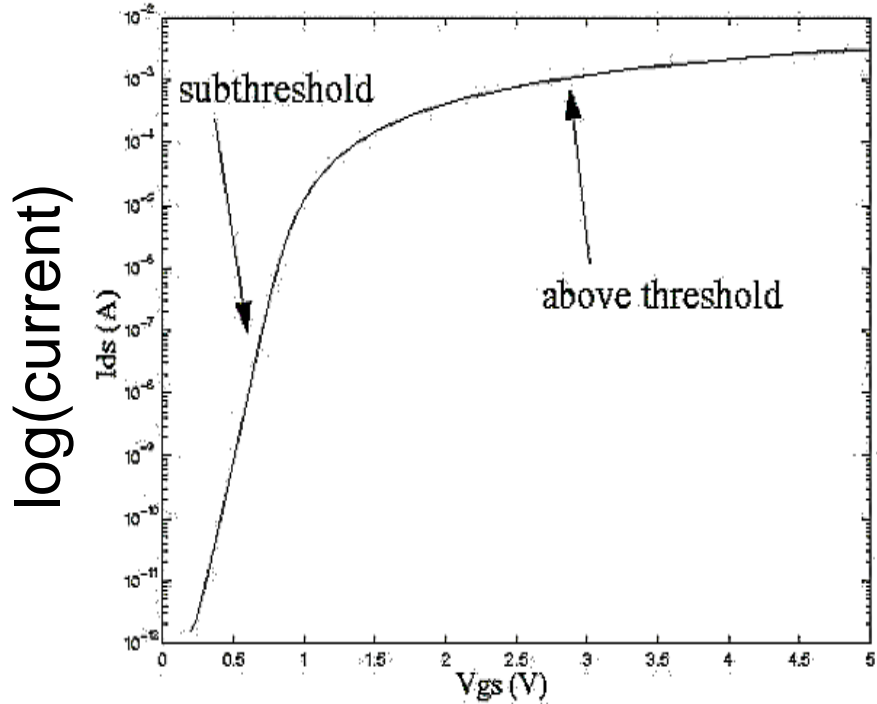
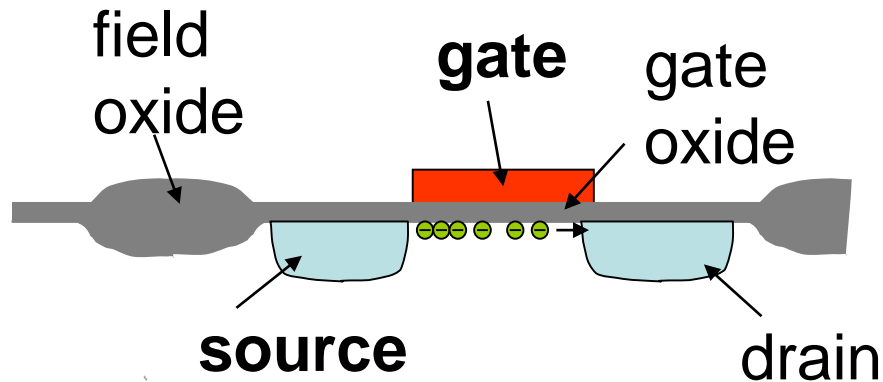




Larger V_g s,
Large V_d s

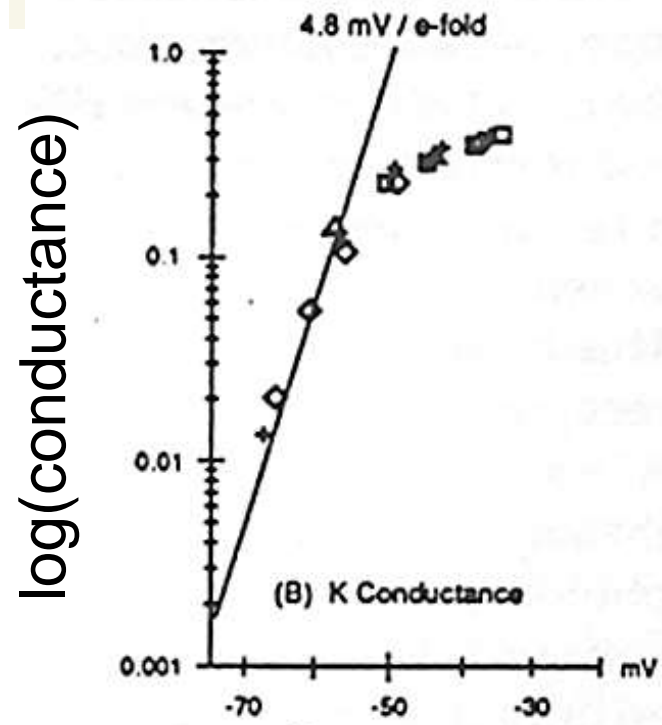
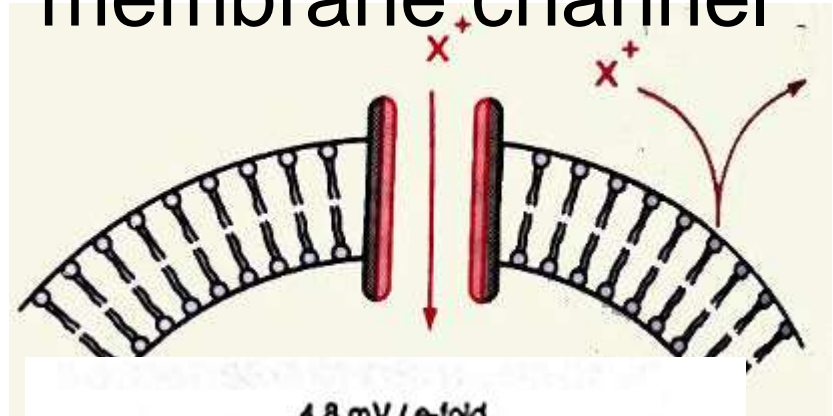


Transistor



gate-source voltage

Voltage activated membrane channel



membrane voltage

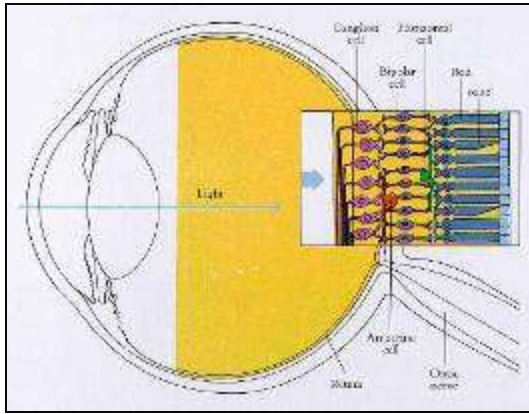
Mechanism of gain

Voltage sensitive channel conductance is exponential in membrane voltage

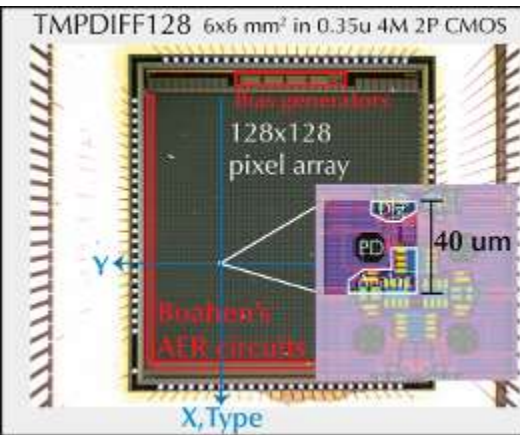
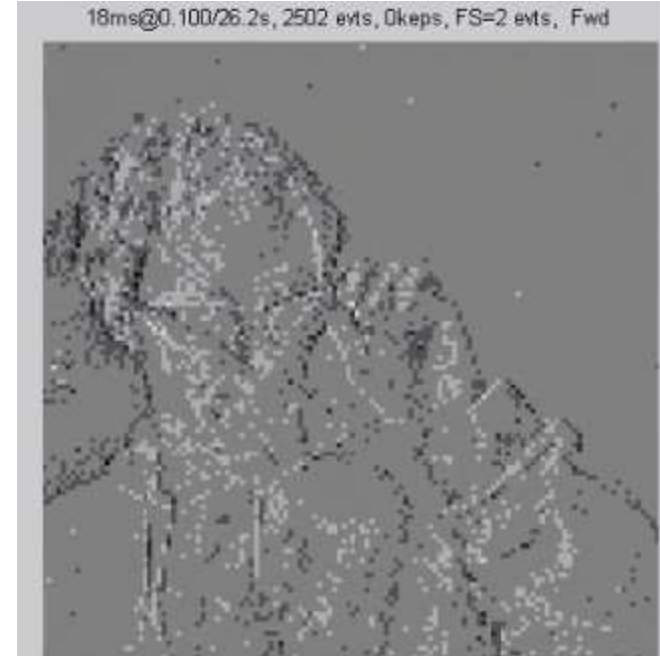
Transistor current is exponential function of gate voltage

Organizing principle:
Use controlled energy barriers (with Boltzmann energy distributions) to amplify

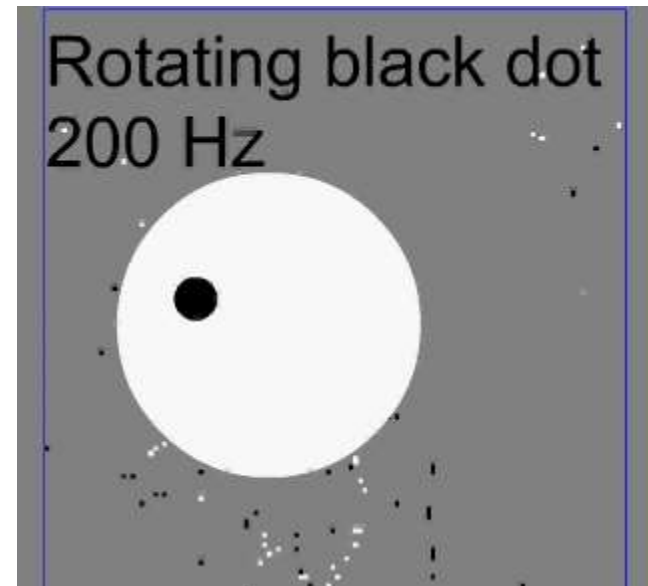
Temporal Contrast Dynamic Vision Sensor



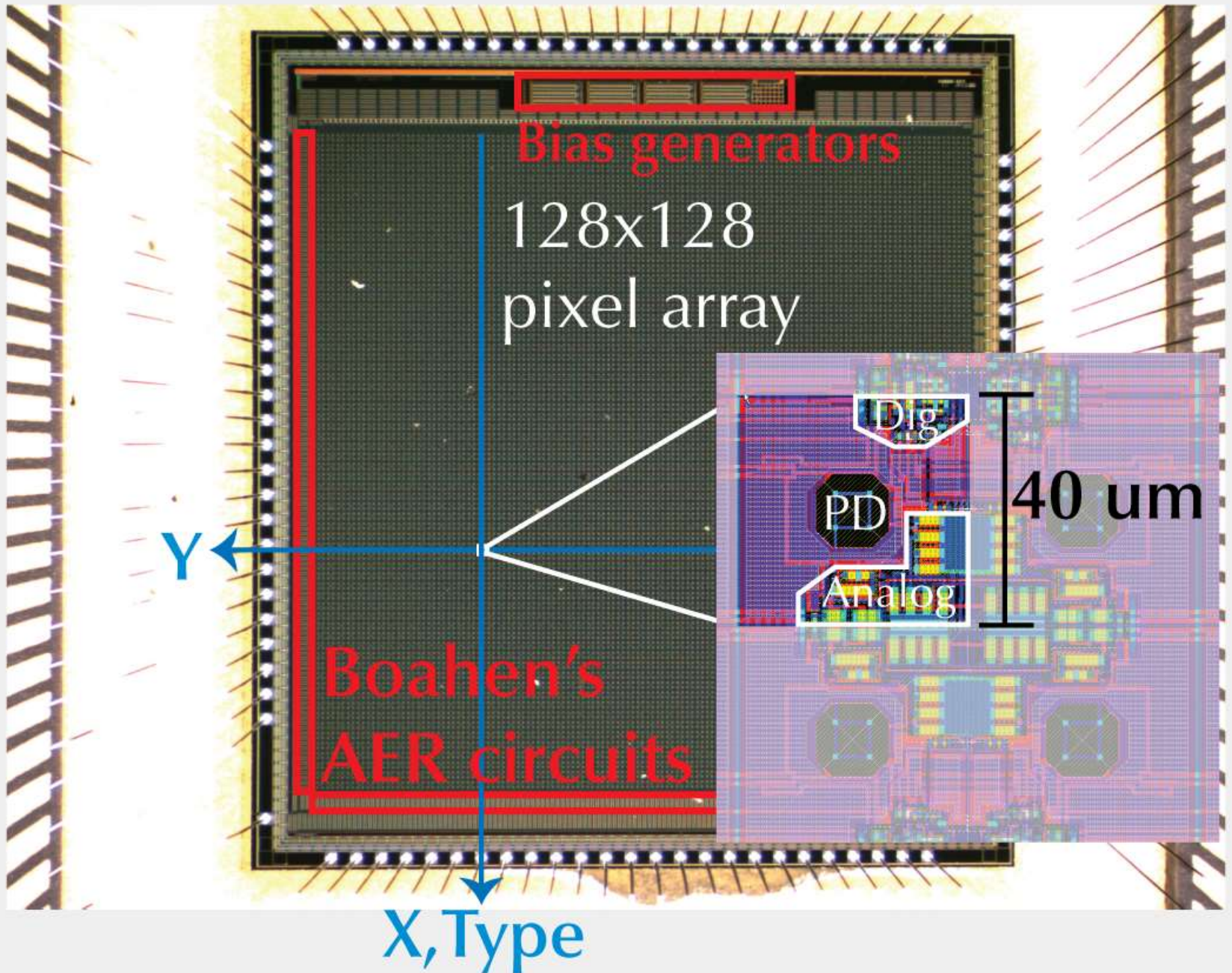
1. This silicon retina **asynchronously** outputs **spiking pixel identities**.
2. Each spike represents a fixed **temporal contrast** ($\Delta \log I$), corresponding to change in scene reflectance.



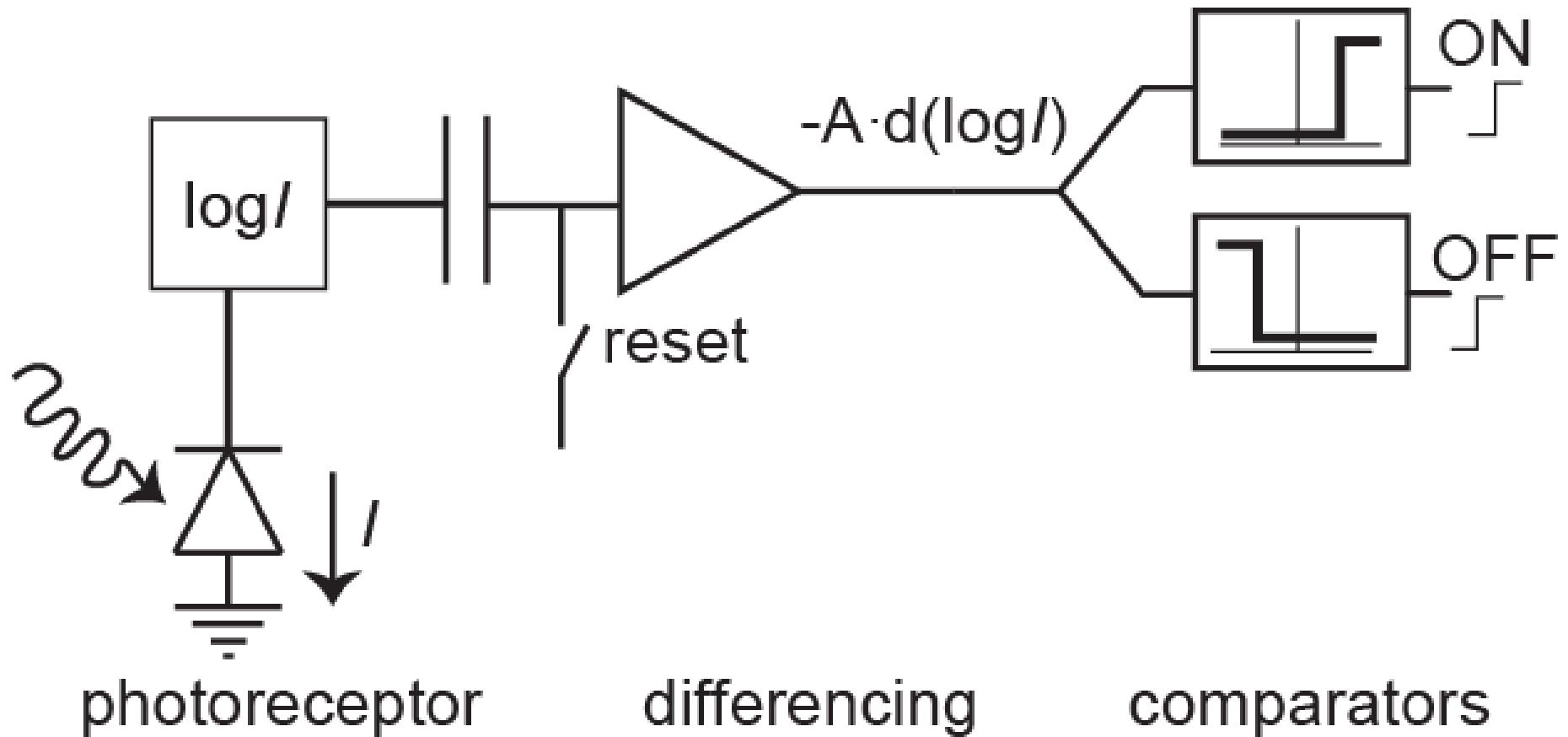
- Models transient pathway in retina.
- Reduces redundancy
- Preserves timing
- Has wide dynamic range

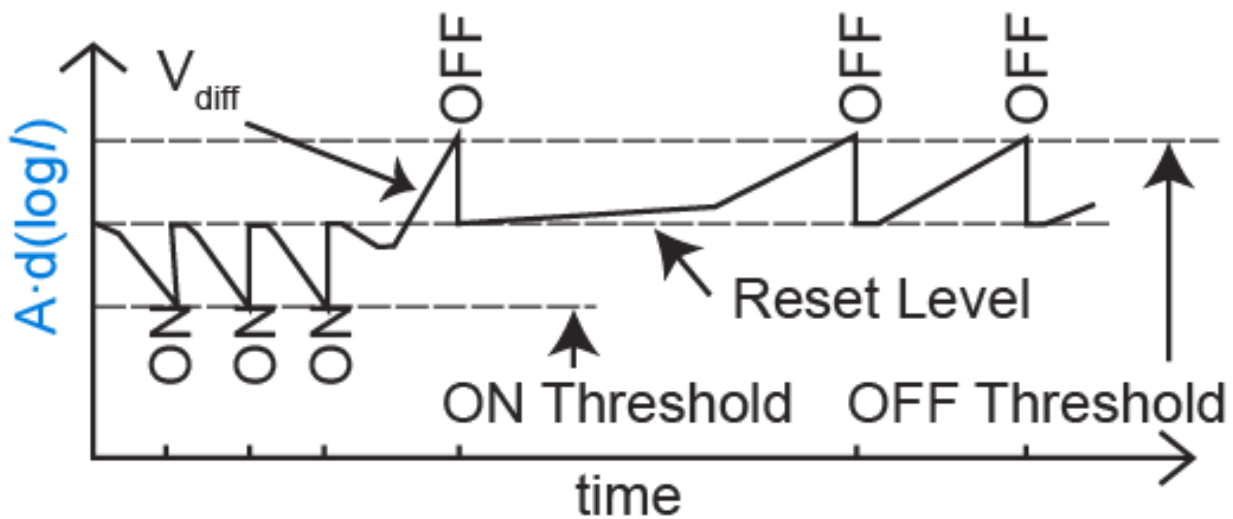
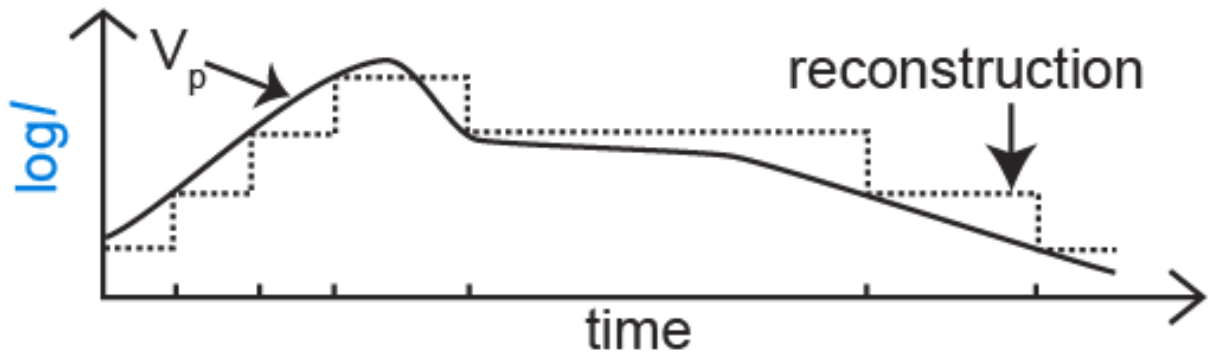
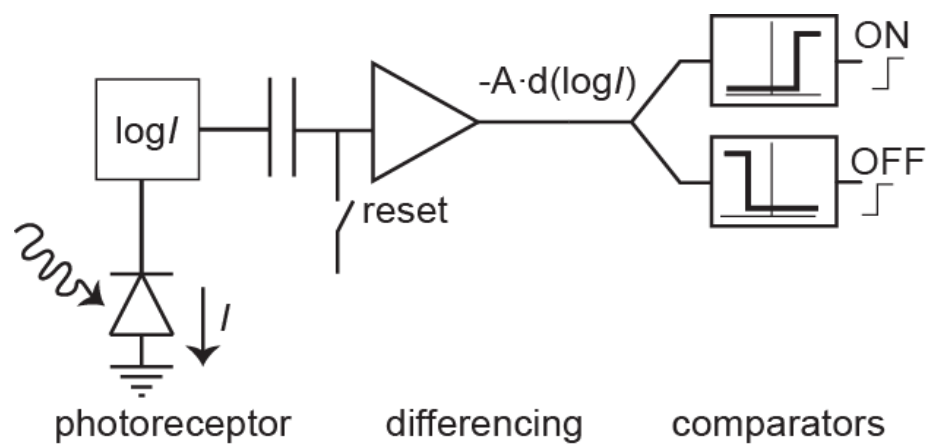


TMPDIFF128 6x6 mm² in 0.35u 4M 2P CMOS



DVS pixel architecture





Embedded DVS Pencil Balancer

Jorg Conradt, Matt Cook

3 microcontrollers, 600mW



Review

- Neuromorphic Engineering (NE)
 - Context of electronics (synchronous logic)
 - Motivation for NE by contrasting computers and brains
- Silicon technology and the operation of a single transistor
 - CMOS vs. complementary channels in neurons
- The biology of the retina
 - How retinas uses adaptive photoreceptors and horizontal cells, together with bipolar cells, to compute rectified local contrast
 - The Physiologist's Friend Chip
 - The Dynamic Vision Sensor