

Prising the secrets of energy efficiency out of brains

Simon Laughlin

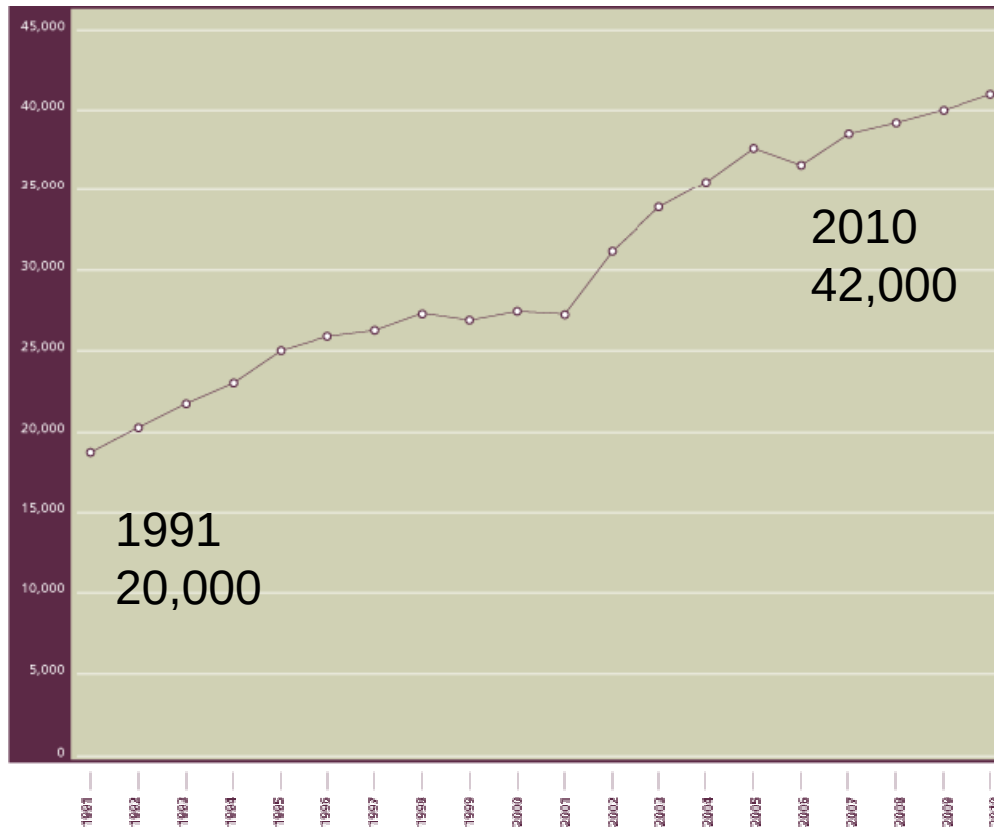
Department of Zoology

University of Cambridge UK

Principles of Neural Design,
P Sterling, S B Laughlin
MIT Press due Fall 2012

Neuroscience is BIG

Society for Neuroscience Membership | Membership Growth | 1991 – 2010



Information Processing in Neural Circuits

- Identified neurons
- In intact circuits
- Relate signals to behavioural performance

Answer

1. What?
2. How?
3. Why?

See

Design

A detailed, close-up photograph of a housefly's head, viewed from the front. The two large, rounded compound eyes are a vibrant reddish-brown color with a fine, grid-like texture. Between the eyes is a dark, sclerotized bridge. Below this bridge, the mouthparts are visible, including the central proboscis and surrounding structures. The fly's legs, which are dark and covered in fine hairs, are visible at the bottom of the frame. The background is a soft, out-of-focus grey.

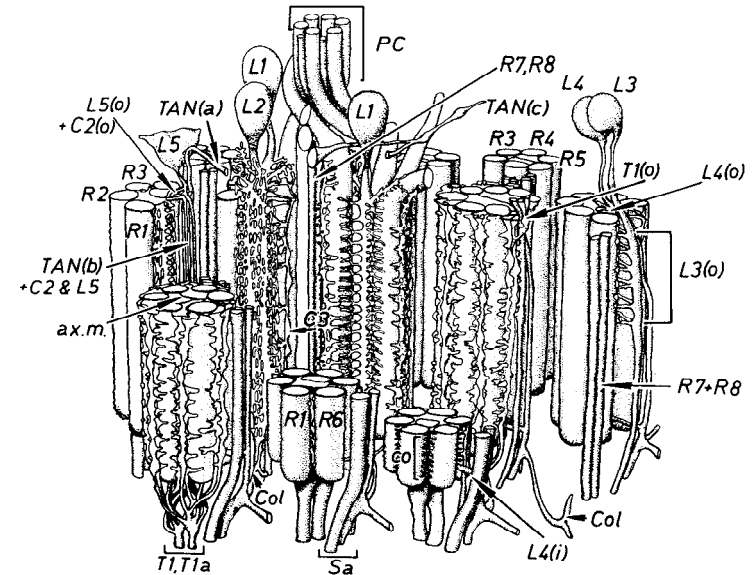
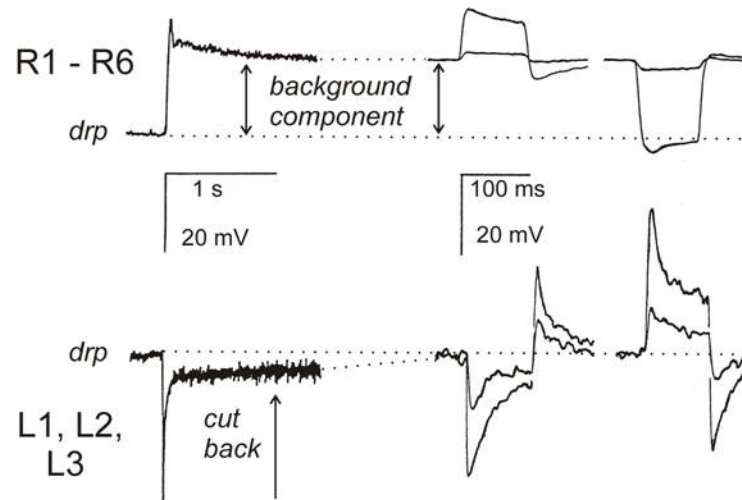
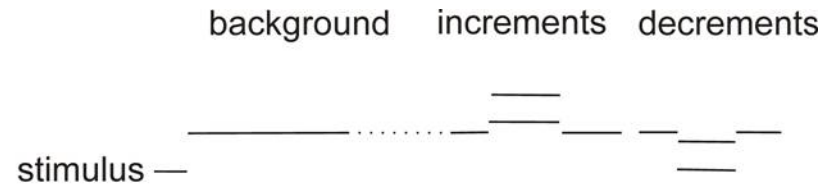
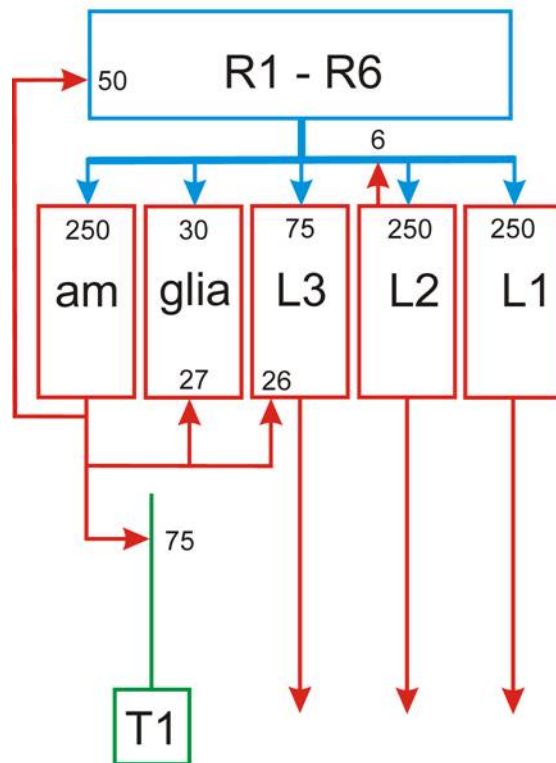


FIG. 1b.

Circuits and signals - fly compound eye



analogue signals

Secret 1

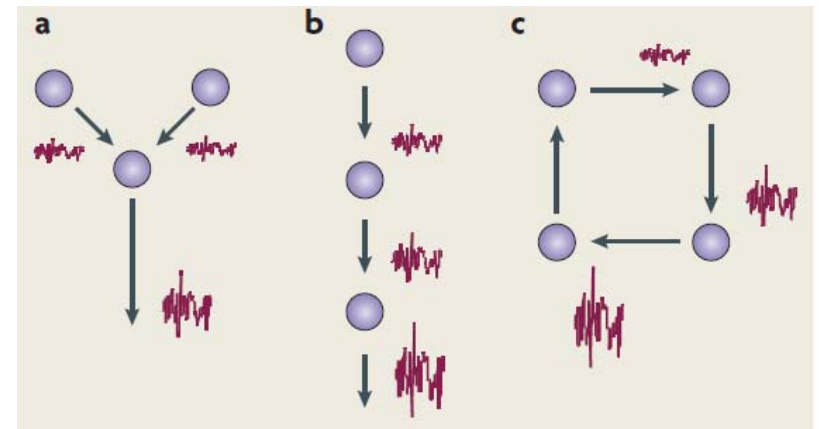
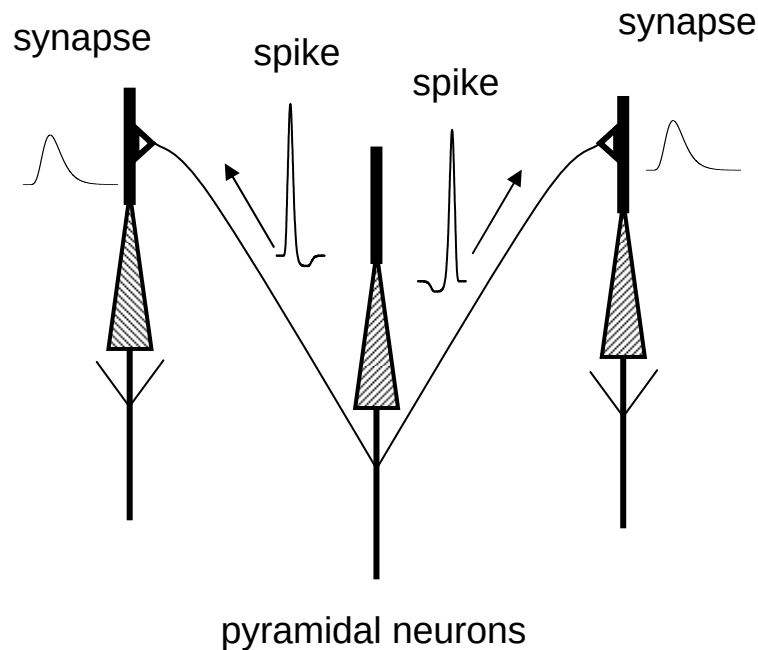
Mix analogue and
pulsatile

Hybrid computation (*mixing analog & pulsatile*) contributes to Energy Efficiency

Rahul Sarpeshkar, MIT

Sarpeshkar, R. (1998). *Analog versus digital: Extrapolating from electronics to neurobiology.* *Neural Computation* **10**, 1601-1638

His book *Ultra low power bioelectronics*, C.U.P. 2010
Independently reaches many of the conclusion of my talk



stop noise accumulating

use efficient analogue
operations to process information
directly at nodes (*rich analogue primitives*)

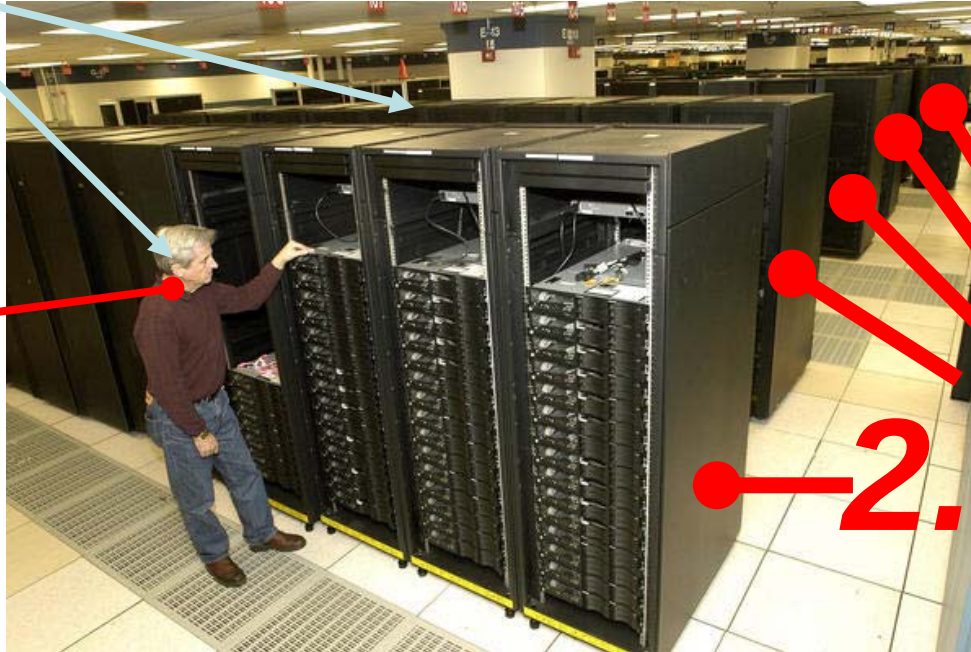
Question

Are brains efficient?

“Joe” with IBM Road Runner Supercomputer (2007)

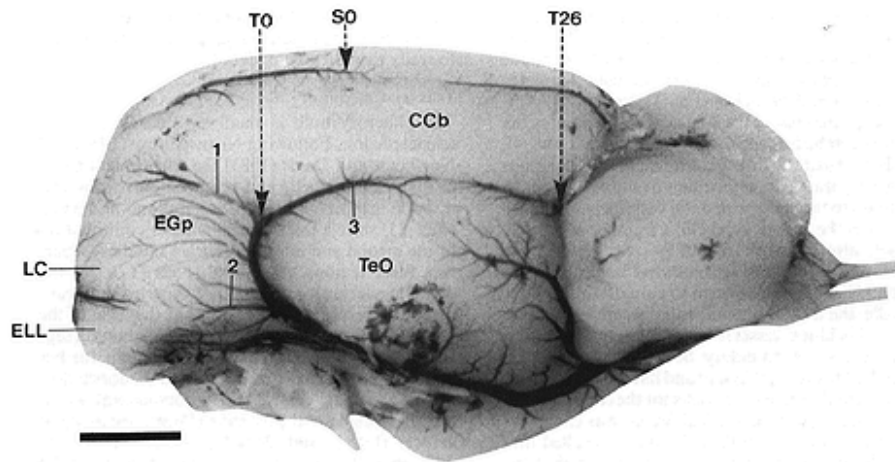
10^{15} events s^{-1}

20 watts



2.36 MW

Allocating *Materials and Space*



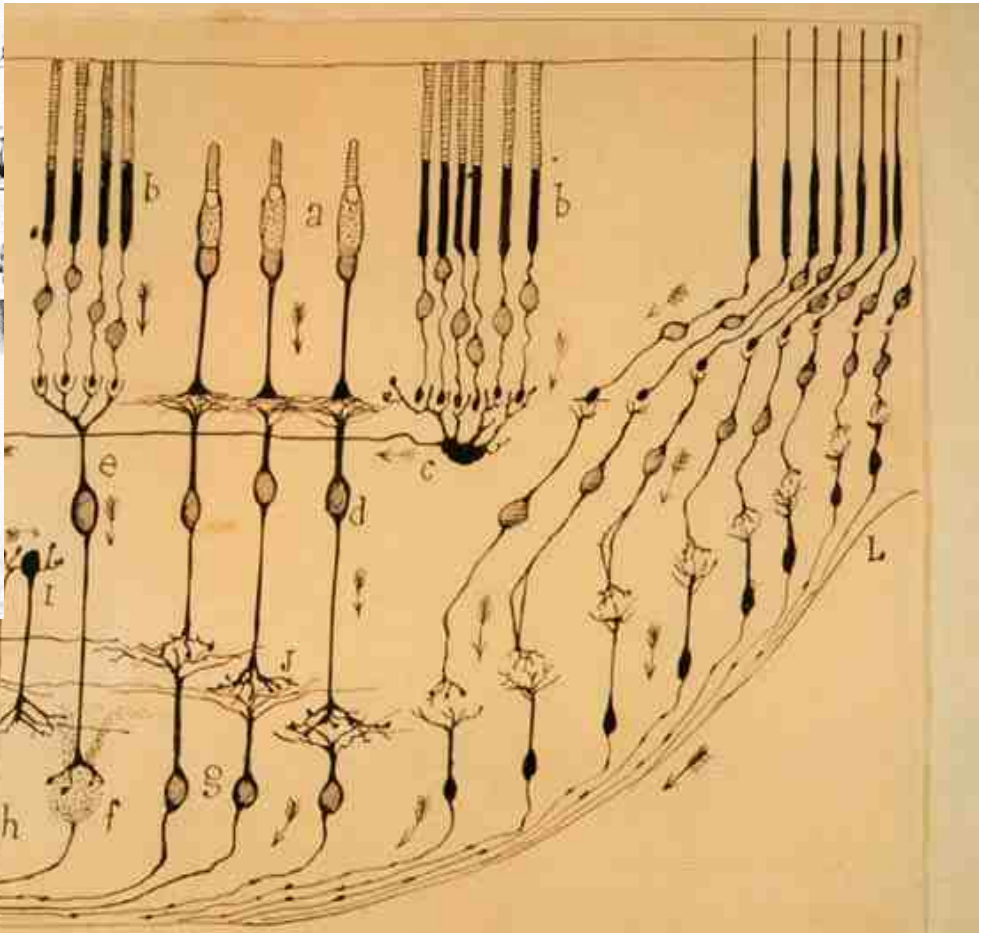
Electric fish brain



Naked mole rat

Secret 2

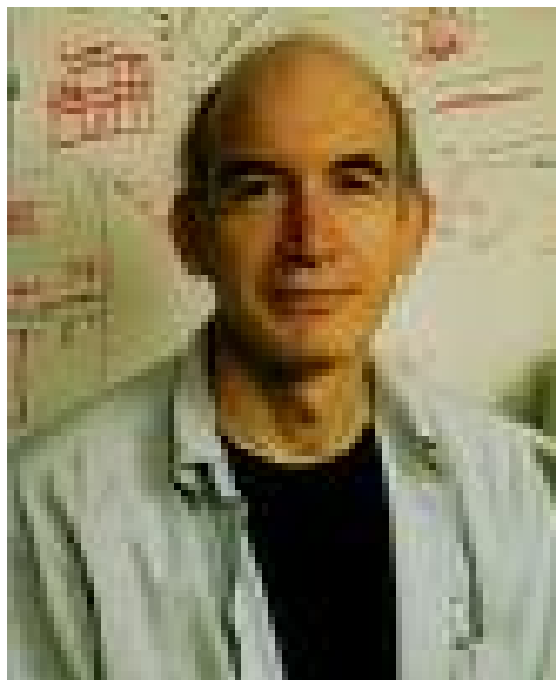
Allocate resource
according to need



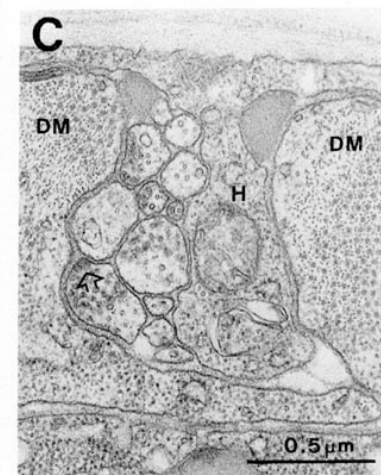
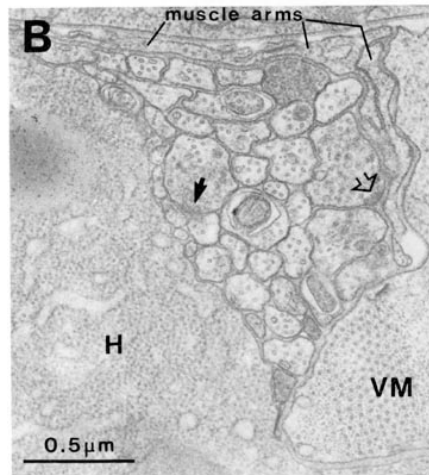
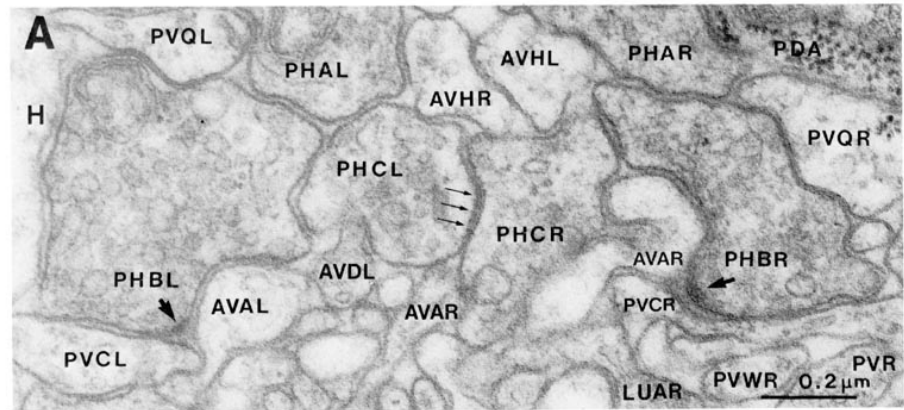
S.R. y Cajal Institute - CSIC - Madrid, Spain.

Minimal rationality

Christopher Cherniak

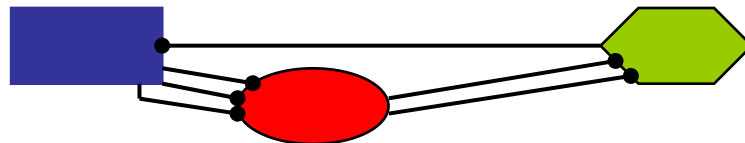
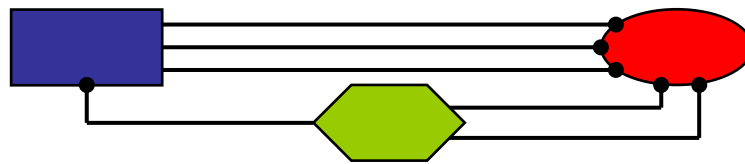


C elegans (aka “The Worm”)



Efficient layout reduces wiring costs

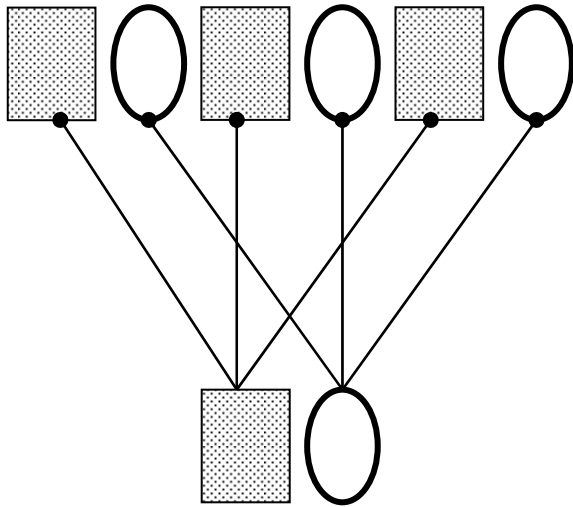
High wiring cost



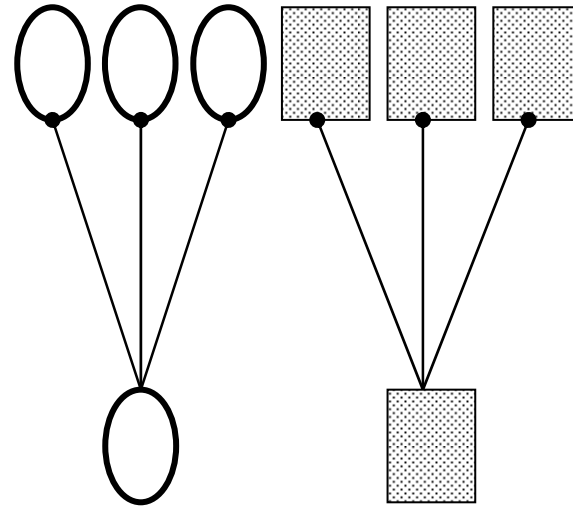
Low wiring cost

Segregating neurons according to function reduces wire length

C



D



Secret 3

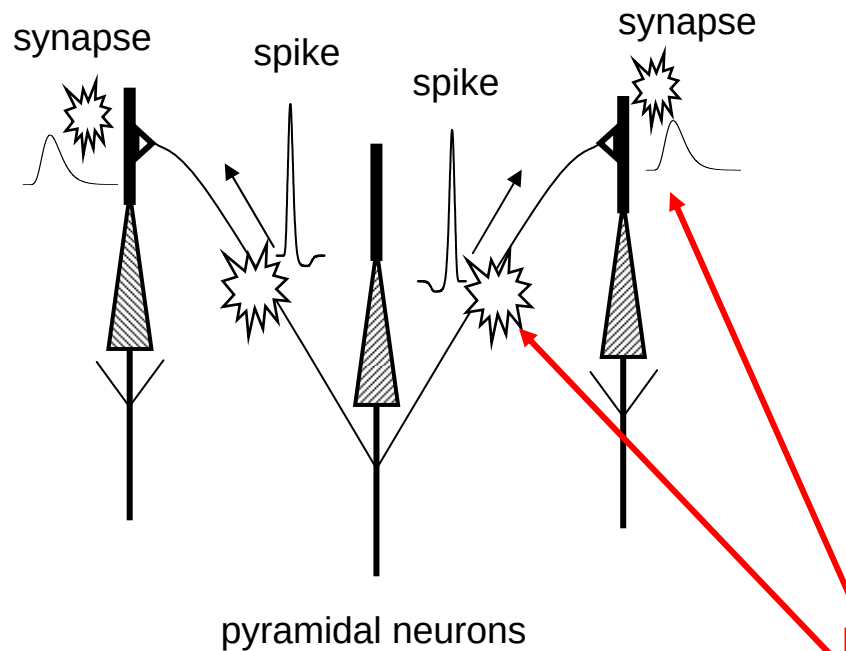
Minimise wiring costs

Brains are wired efficiently
Allocate materials efficiently

Do they need to be
Energy Efficient?

Brains need to be energy efficient because demand is high enough to limit signal traffic

Major sites of energy use

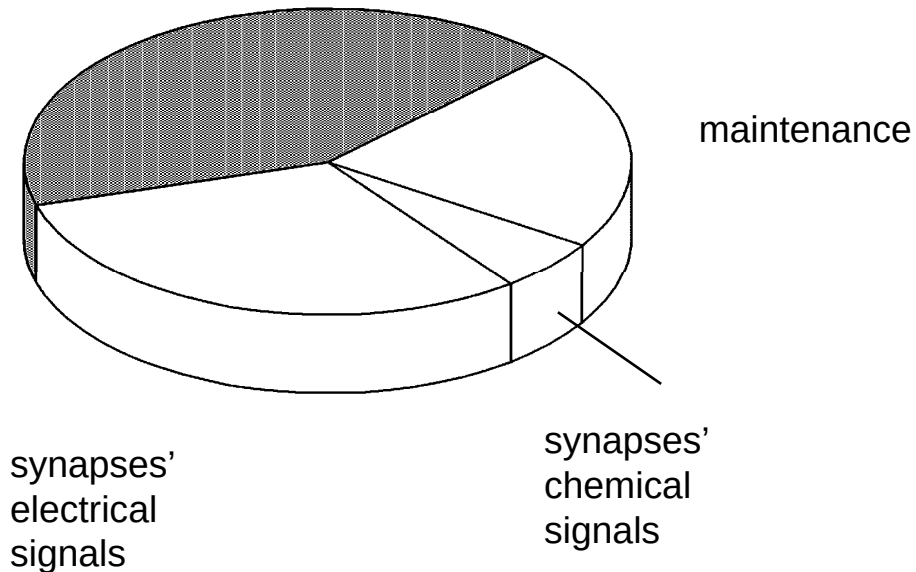


The energy consumed by cerebral cortex limits cortical signal traffic to 1 – 10 spikes/neuron/sec
Attwell & Laughlin 2001

drive membrane resistance
charge membrane capacitance

Distribution of energy costs

spikes (action potentials)



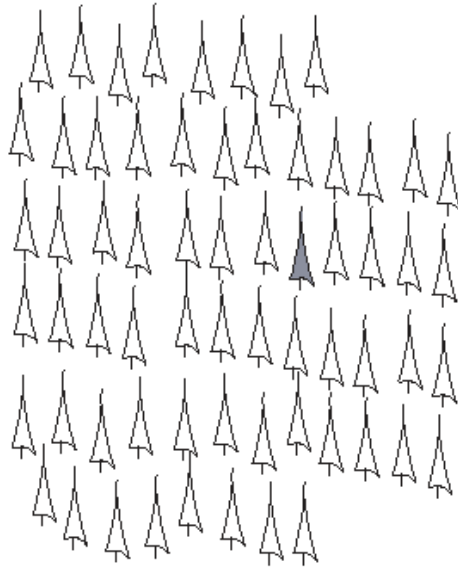
electrical – expensive
chemical - cheap

Secret 4

Distribute signals
sparsely

sparse coding is energy efficient

(after Levy & Baxter, 1996)



1 active, 63 at rest
capacity = 64

cost ratio signalling:fixed	costs		
	signalling	fixed	total
100:1	100	64	*164
10:1	10	64	74
5:1	5	64	69
1:1	1	64	65



2 active, 10 at rest
capacity = 66

costs		
signalling	fixed	total
200	12	212
20	12	*32
10	12	*22
2	12	14



4 active, 4 at rest
capacity = 70

costs		
signalling	fixed	total
400	8	408
40	8	48
20	8	28
4	8	*12

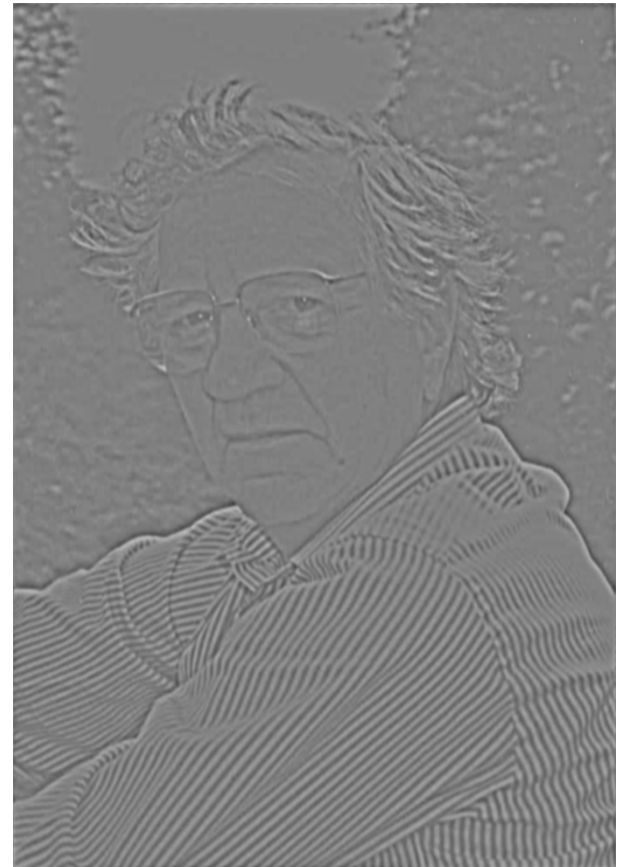
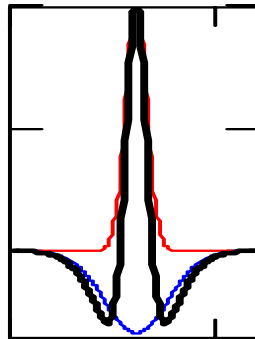
Secret 5

Send only what is
needed

“The neat packaging of information”

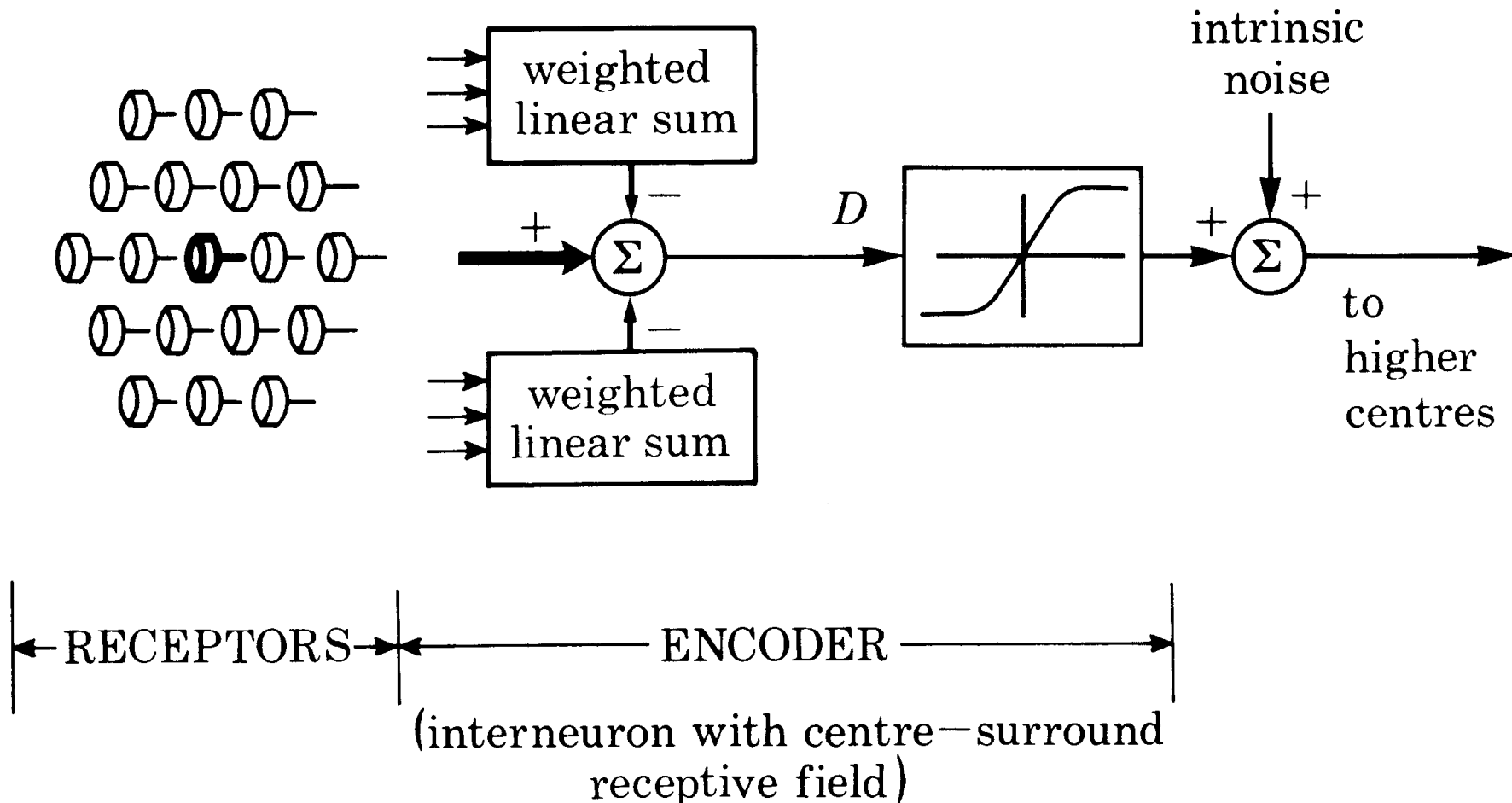
Horace Barlow 1961

*redundancy reduction as a goal of early sensory
processing*



Predictive coding

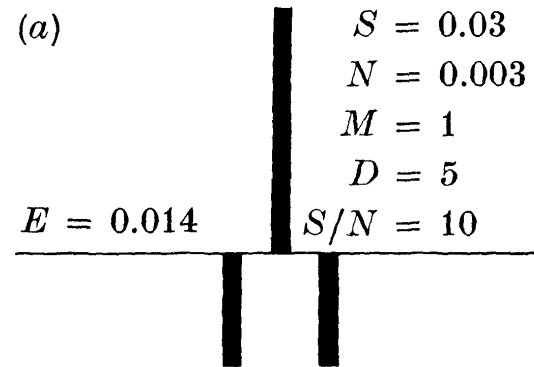
Srinivasan, Dubs and Laughlin, 1982



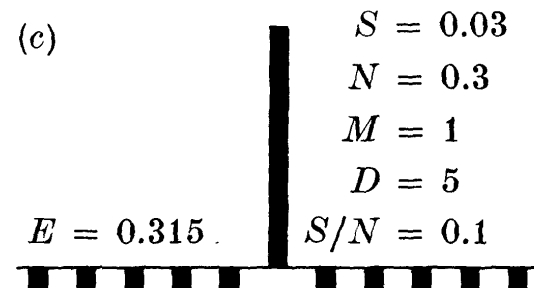
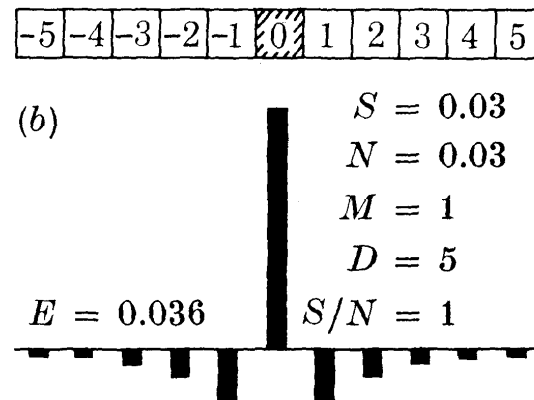
THEORY

(Srinivasan, Dubs and Laughlin, 1982)

Predictive coding
requires an
intensity dependent
surround
*Take wider
samples when input
is unreliable
(photon noise)*



Hi S/N

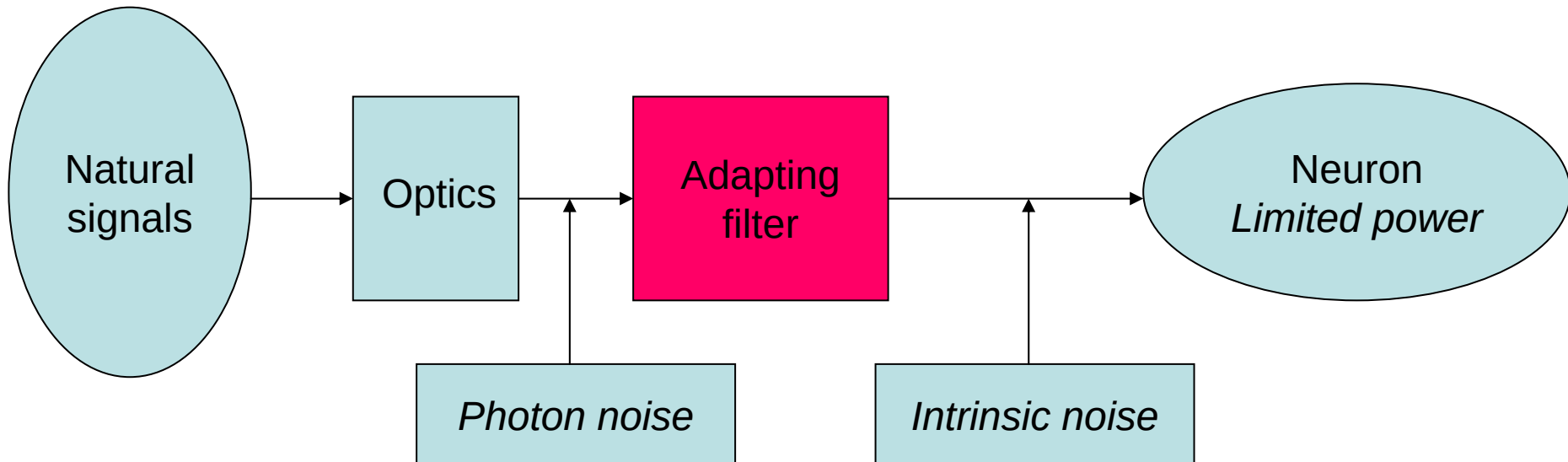


Lo S/N

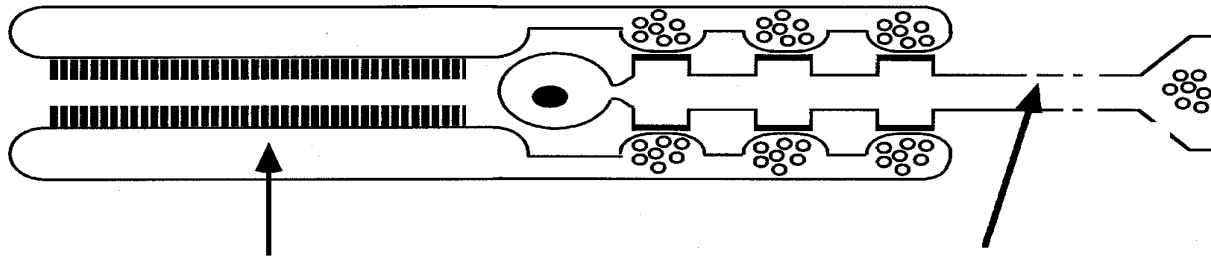
Maximise information coded in a channel limited by dynamic range and noise
(Hans van Hateren, 1992)

MODEL

Input "naturalistic" $1/f$ statistics



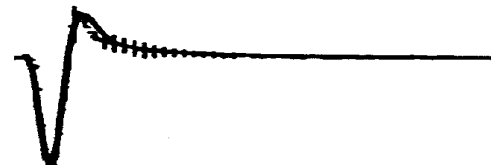
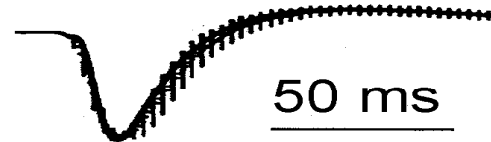
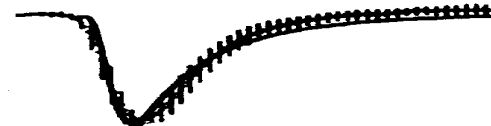
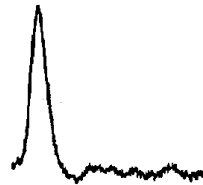
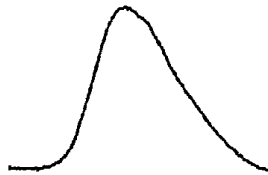
Test model



R1-6

LMC

LMC Data |
Model ____



Very dim

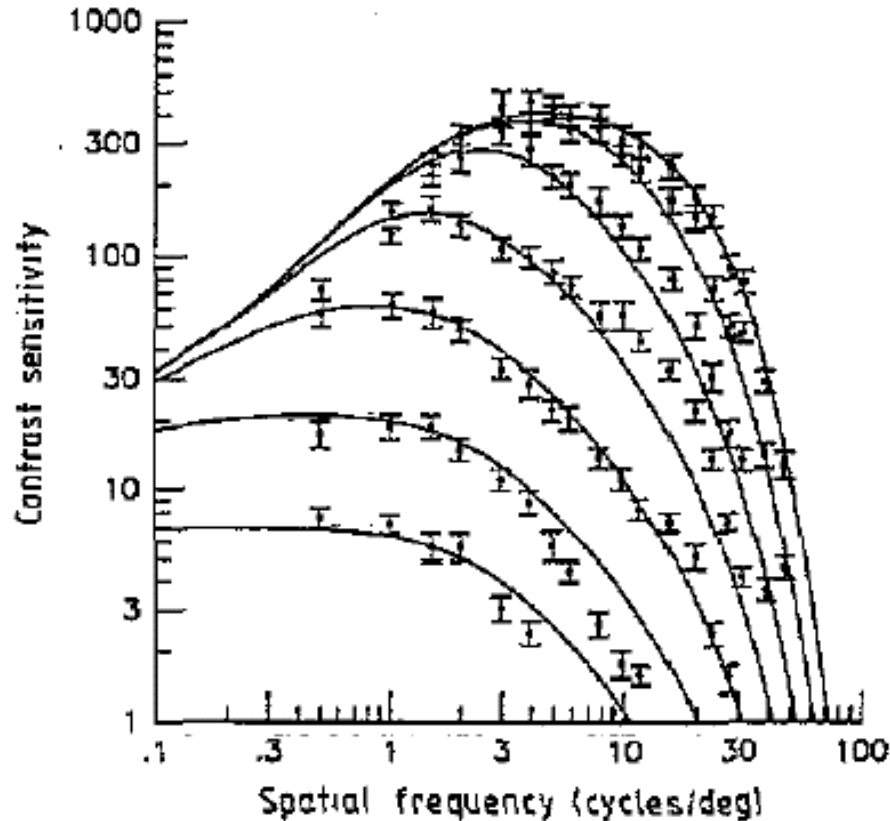
Quite dim

daylight

50 ms

Model generalises

Similar redundancy reduction model applied to human vision (Atick & Redlich, in Atick, 1992)



Data points = psychophysical data; Curves = redundancy reduction model

Secret 5

Send only what's needed

- eliminate redundancy
- improve SNR before transmission
- boil information down to "*what the destination needs to know, no less, no more*"

Secret 6

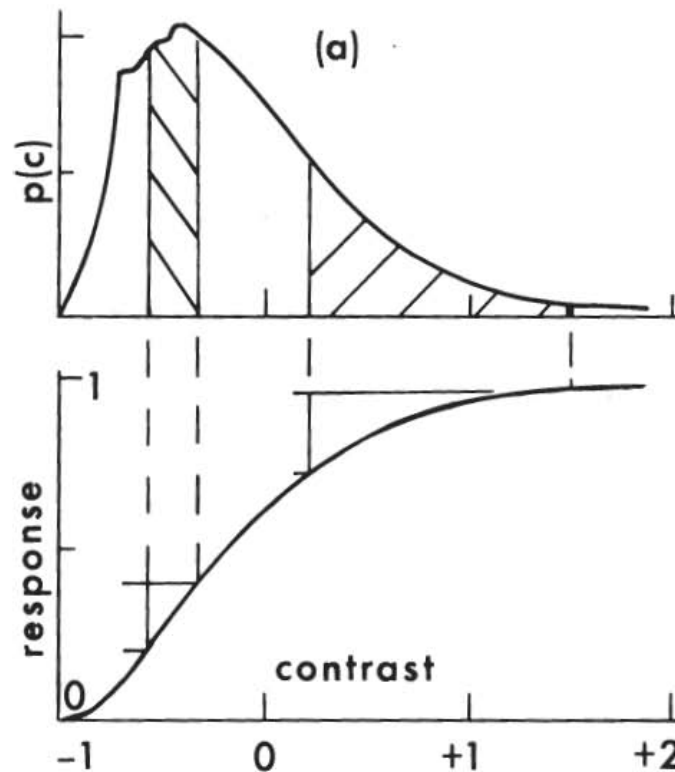
Match neural resources to natural distributions of signals and tasks. *Match* components across levels (“*symmorphosis*”)

Redundancy reducing adaptive filter matched to input SNR

Gain adapted to input amplitude distribution

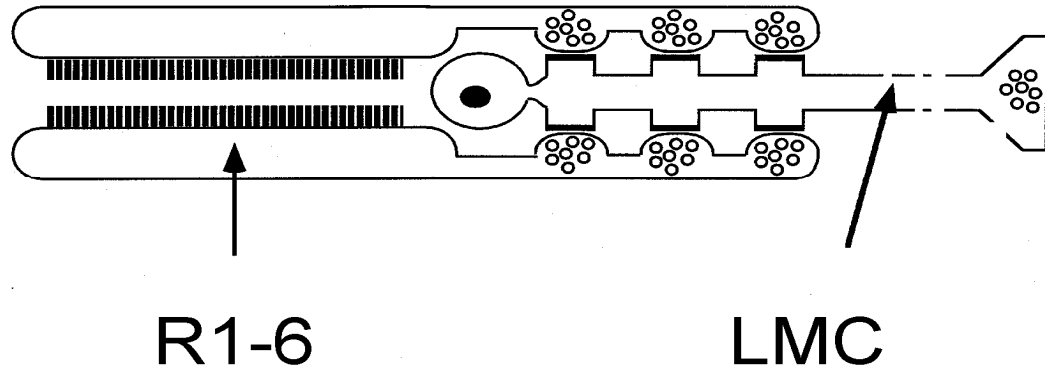
Amplifying signal to fill the response range

HISTOGRAM EQUALISATION

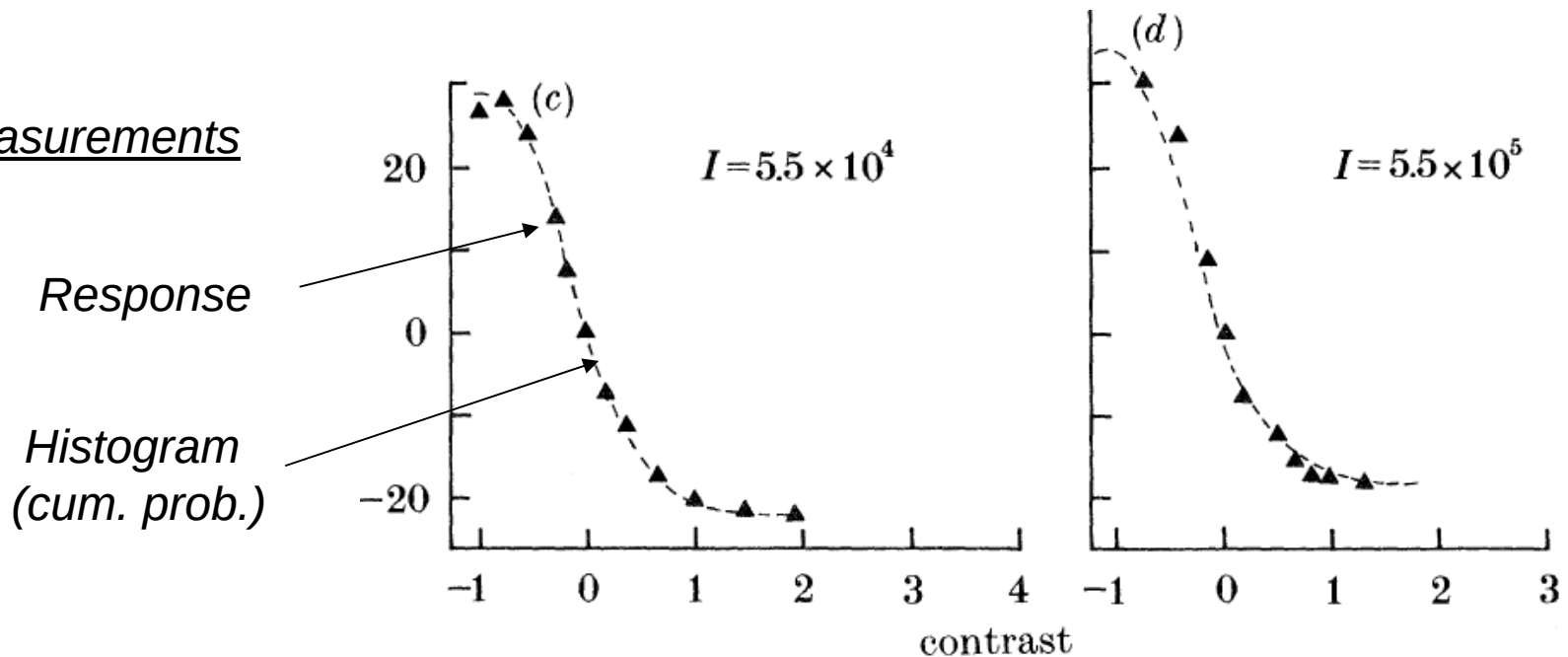


(Laughlin, 1981)

Histogram equalisation at the photoreceptor – LMC synapse



Measurements



Equalising the response histogram in *Drosophila* olfactory glomeruli

Bhandawat et al 2007

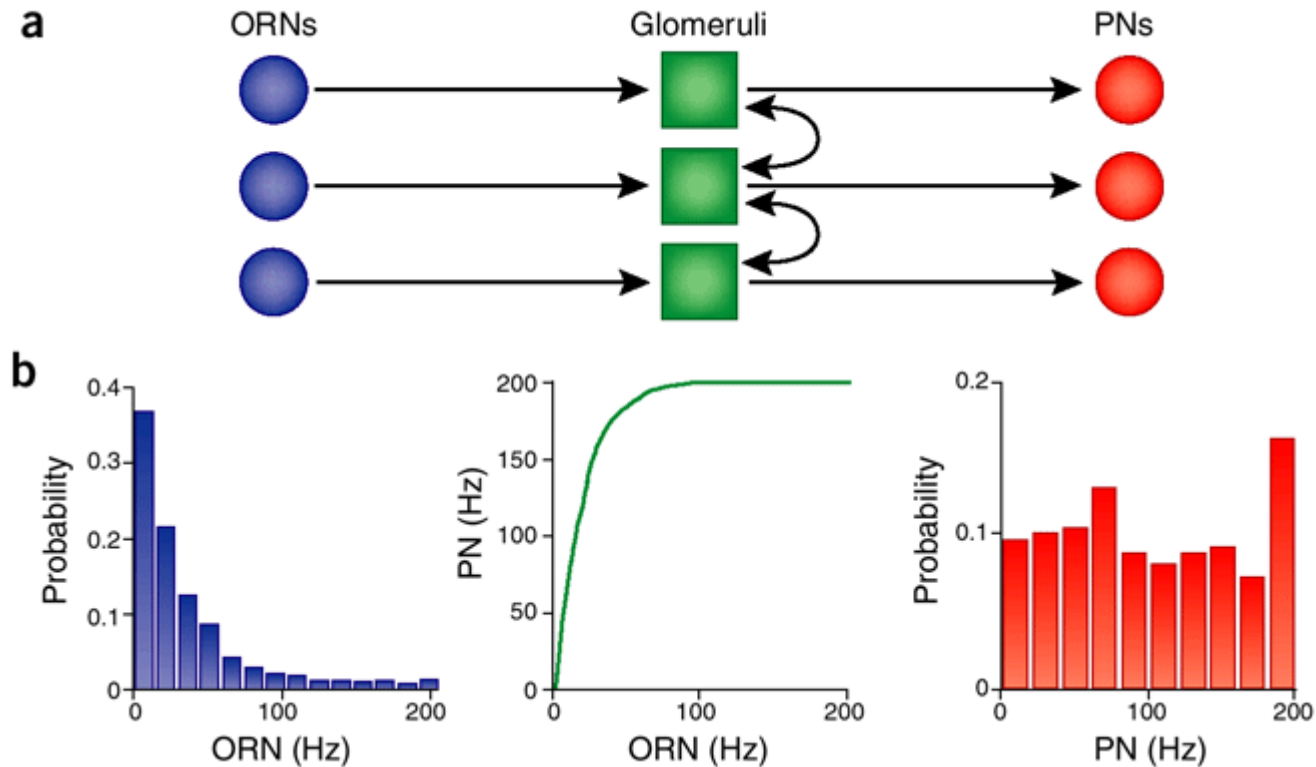


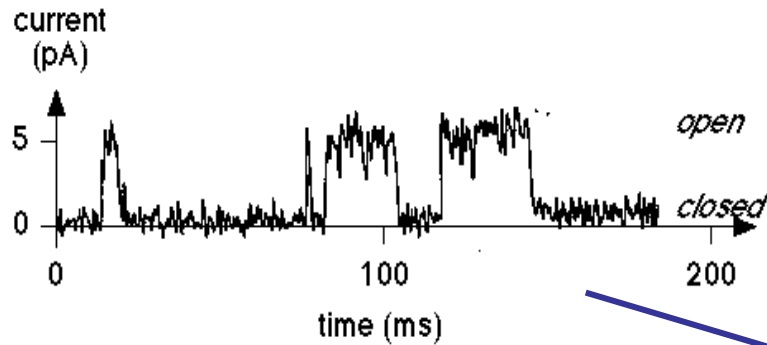
Figure from Abbott & Luo, Nature Nsci News and Views 2007

Working efficiently within device constraints

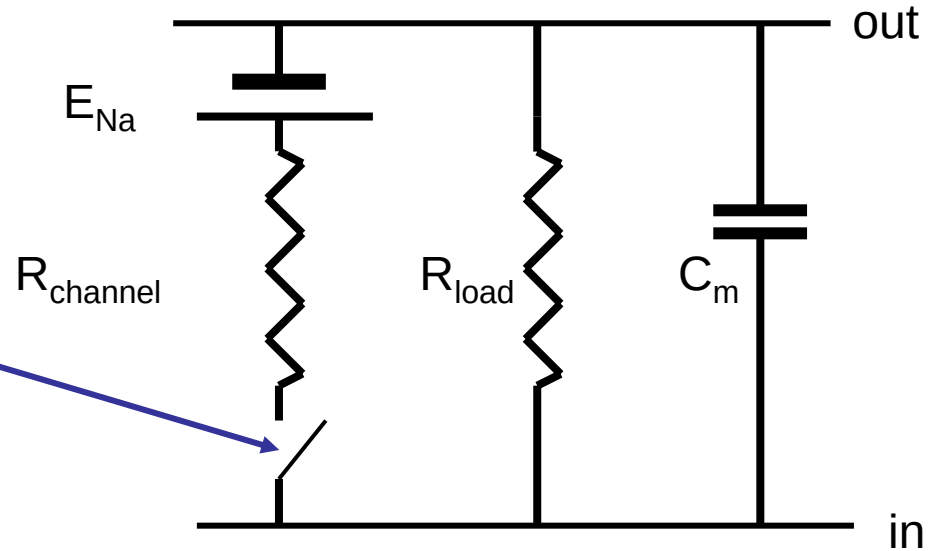
electrical signalling

Signal quality in single neurons

basic biophysics – graded responses



Noisy signals



$$R_{channel} = 10^{11} \text{ Ohms}$$

$$E = 10^{-1} \text{ V}$$

$$R_{load} = 10^9 \text{ Ohms}$$

C_m is significant

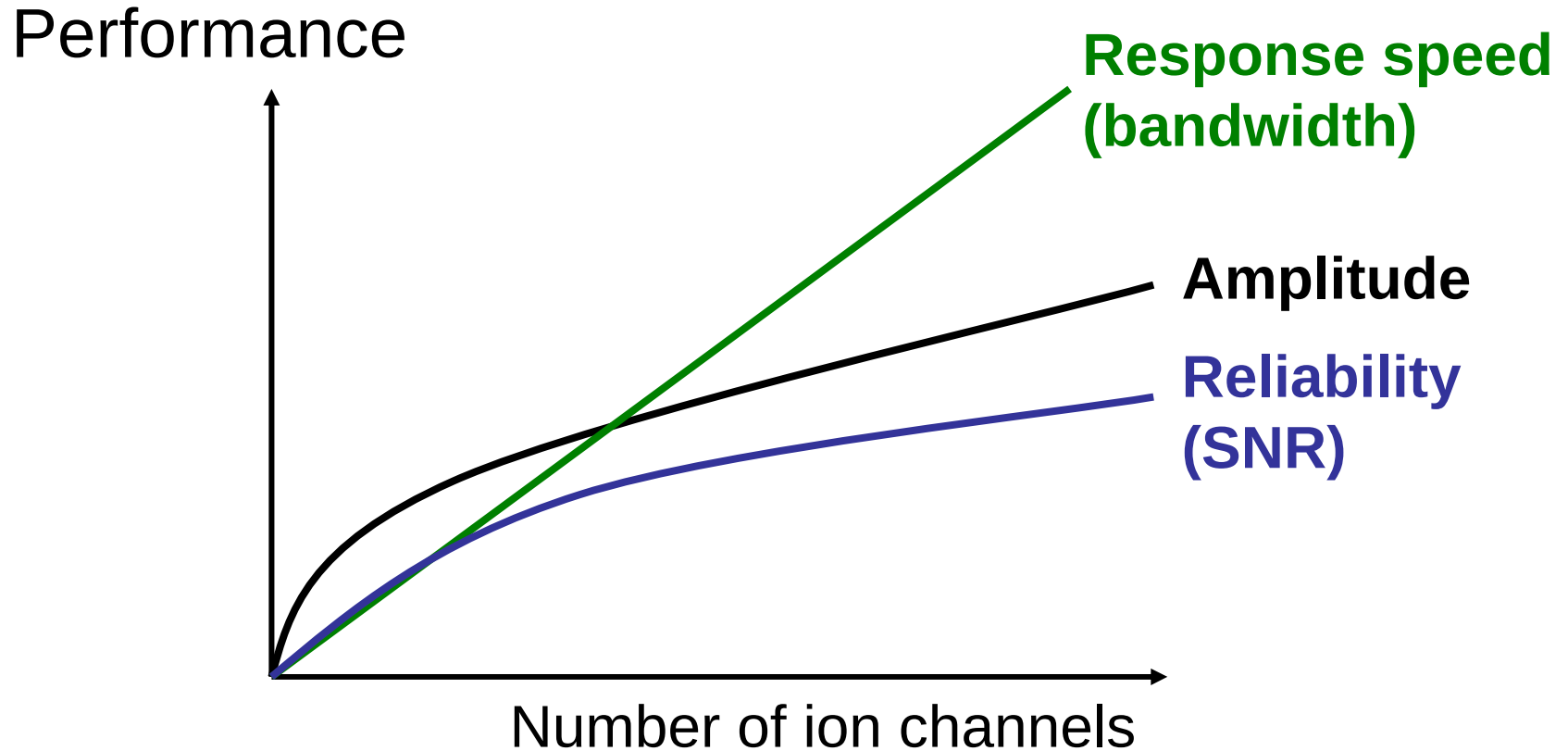
R_m is high

Small signals

Slow signals

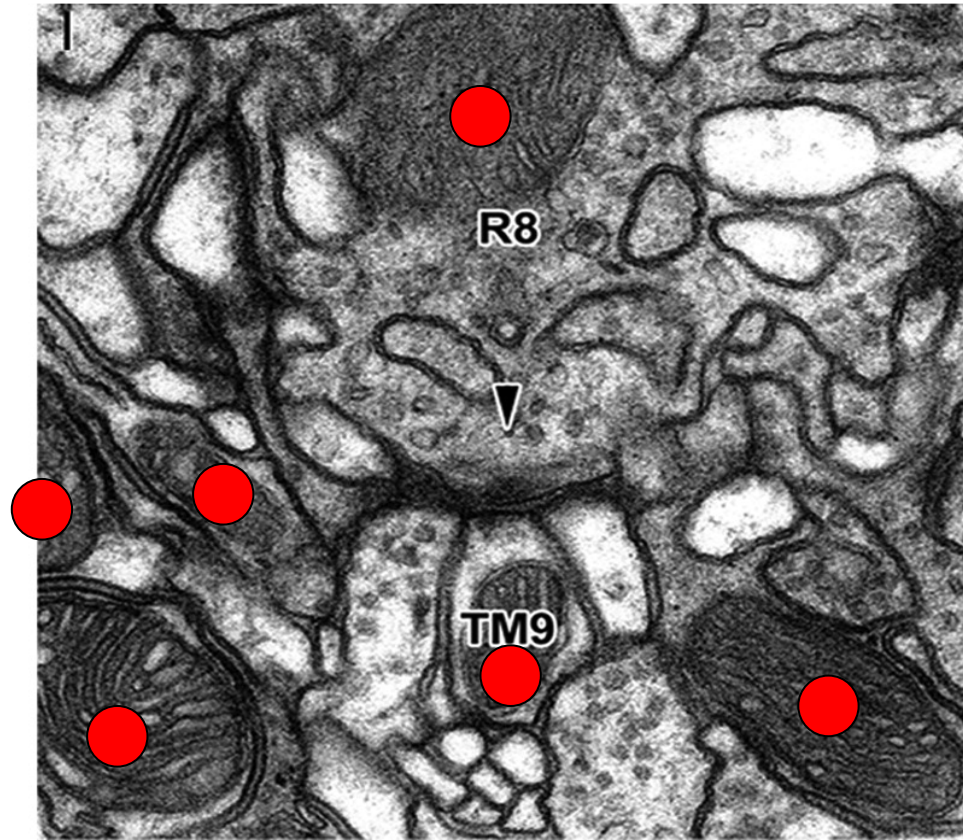
Decaying signals

More ion channels – better performance



***Energy costs;
plus space and materials***

Energy takes space mitochondria in neurons



Gao et al, *Neuron* 2008, *Drosophila* medulla

— 500 nm

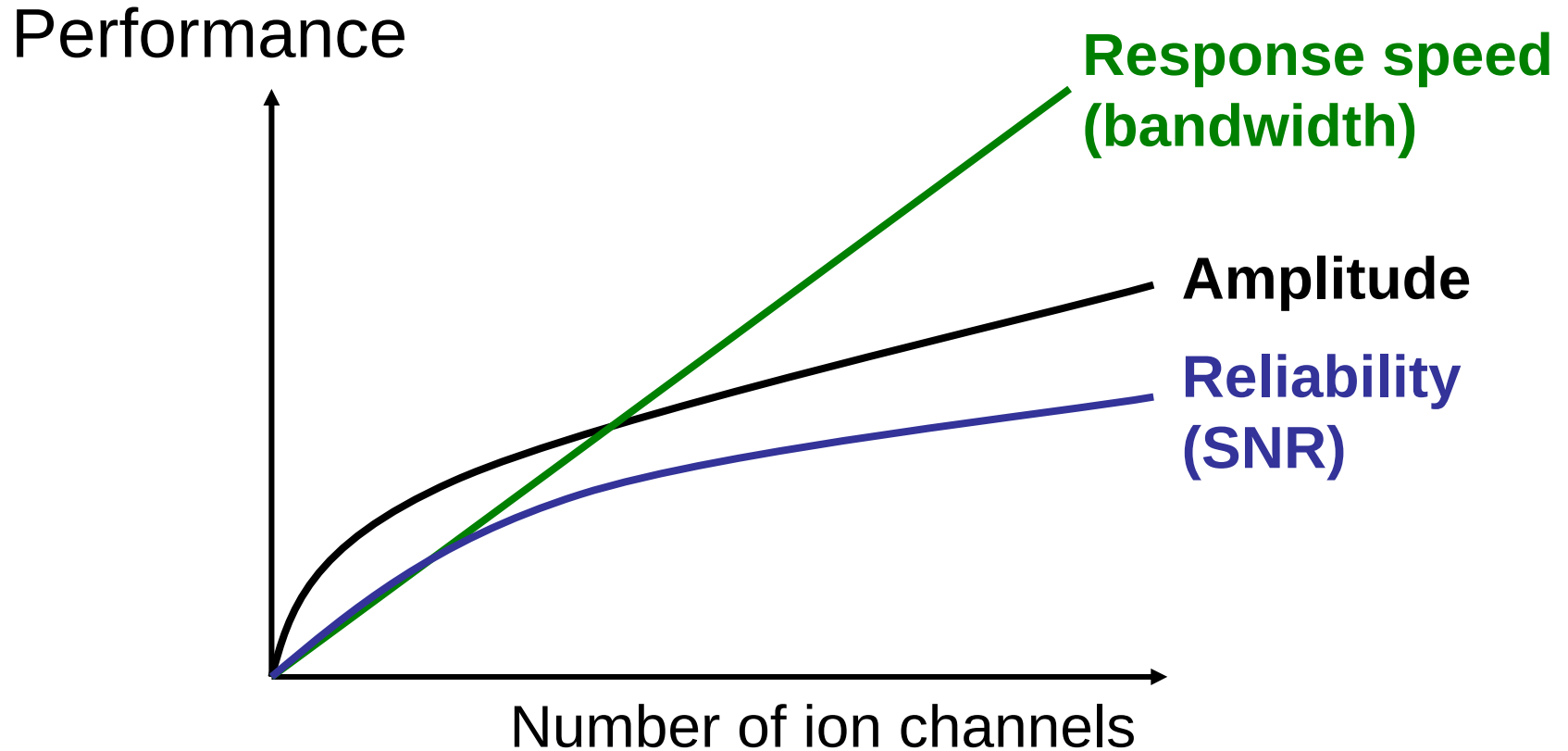
Secret 7

Reduce voltage and
increase resistance

Secret 8

send information at the lowest rate
(minimise bandwidth and precision)

More ion channels – better performance



***Energy costs;
plus space and materials***

Testing for a Law of Diminishing Returns *bit cost vs capacity (max rate)*

*Sarcophaga
carnaria*



*Calliphora
vicina*



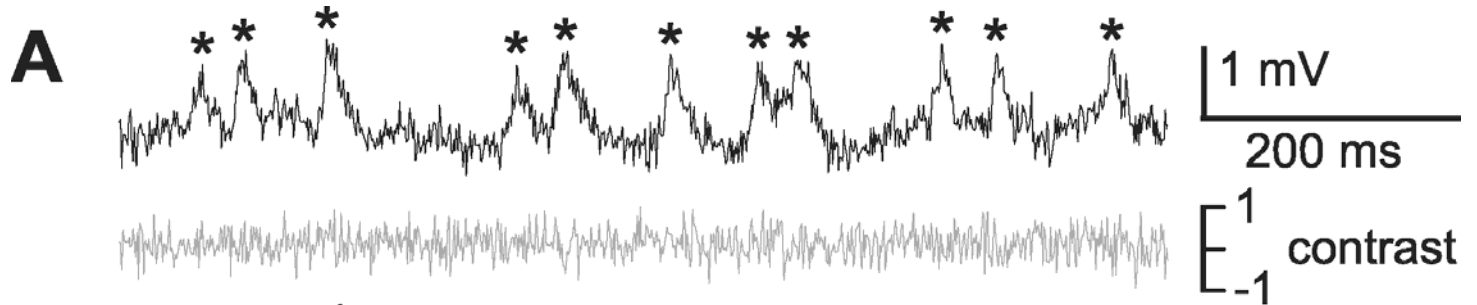
*Drosophila
melanogaster*



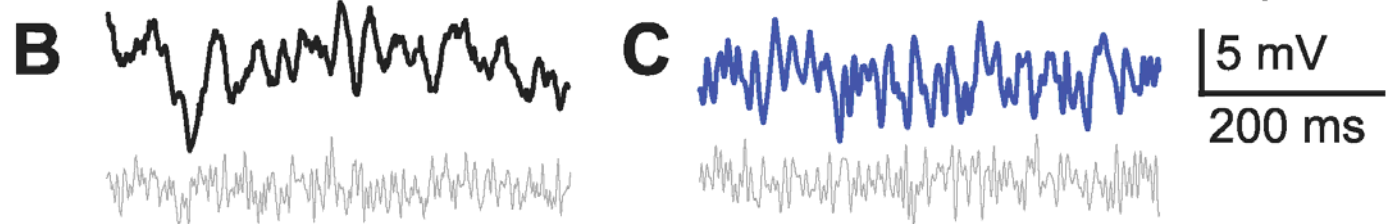
*Drosophila
virilis*

The data

Quantum
bumps



Response



White-noise
stimulus

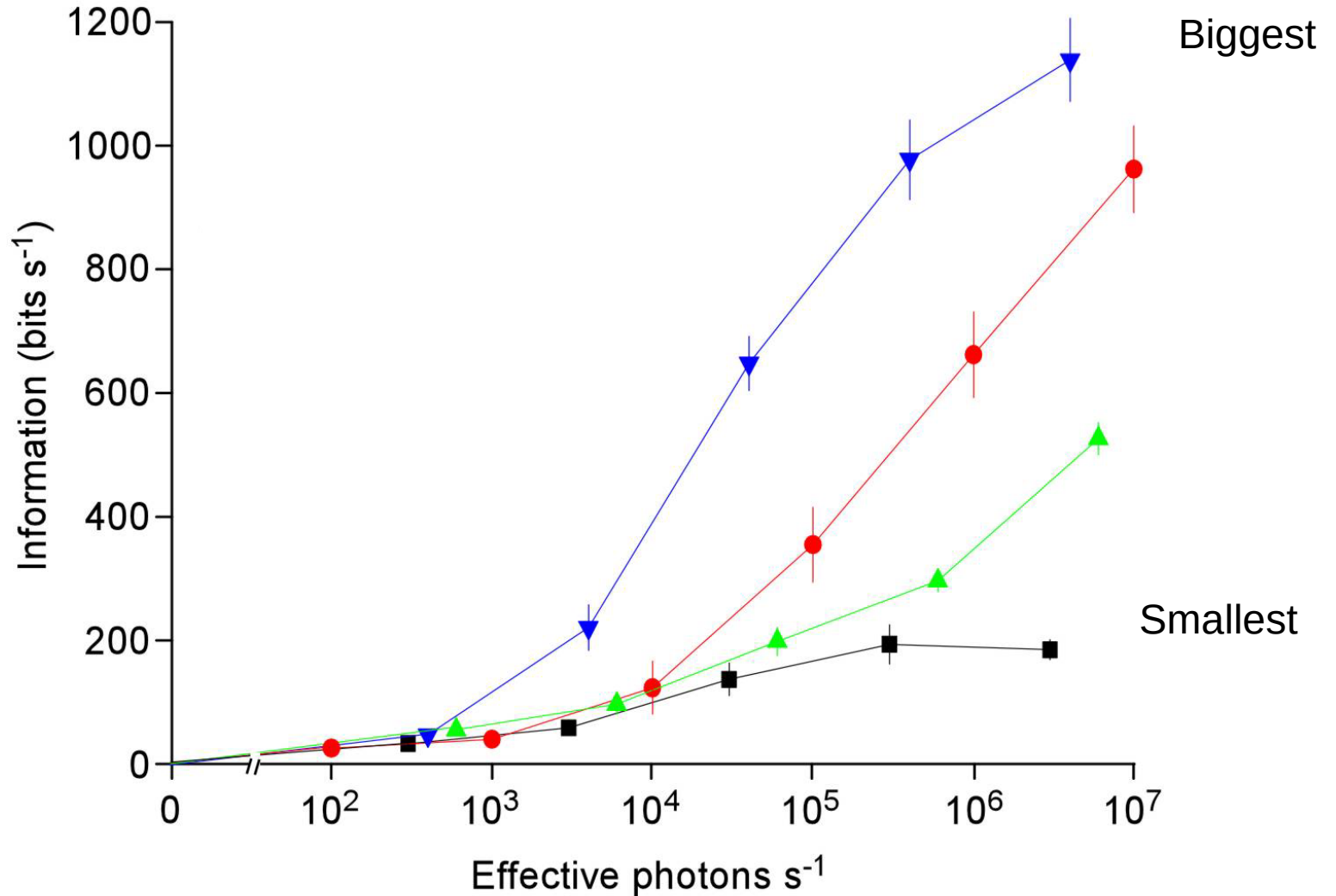
D. melanogaster

Sarcophaga

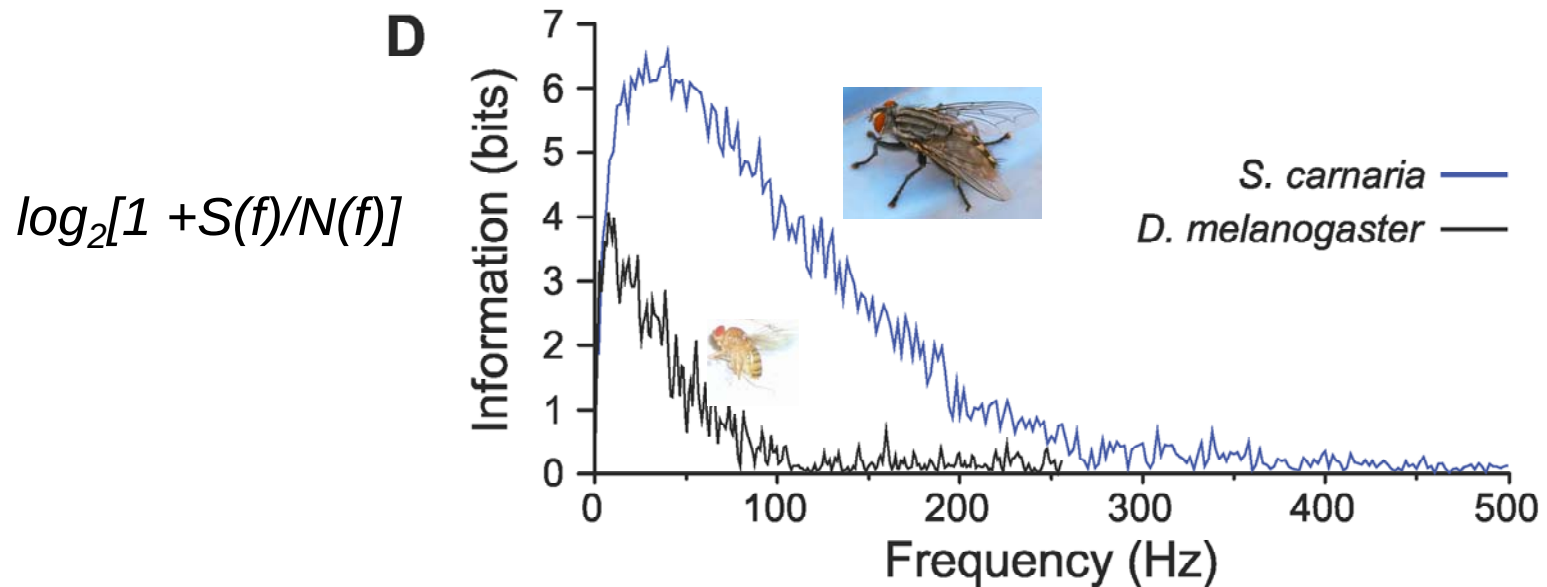
Capacity increases with bandwidth and
precision

$$I = \int_0^{co} \log_2[1 + S(f) / N(f)] \cdot df$$

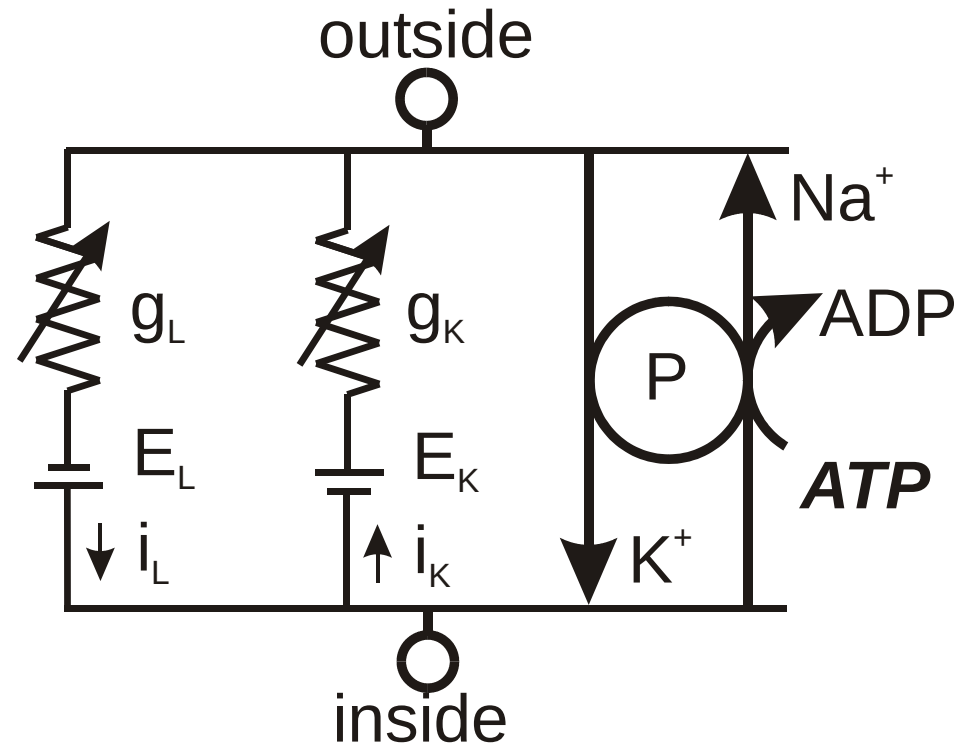
Larger cells - higher capacities



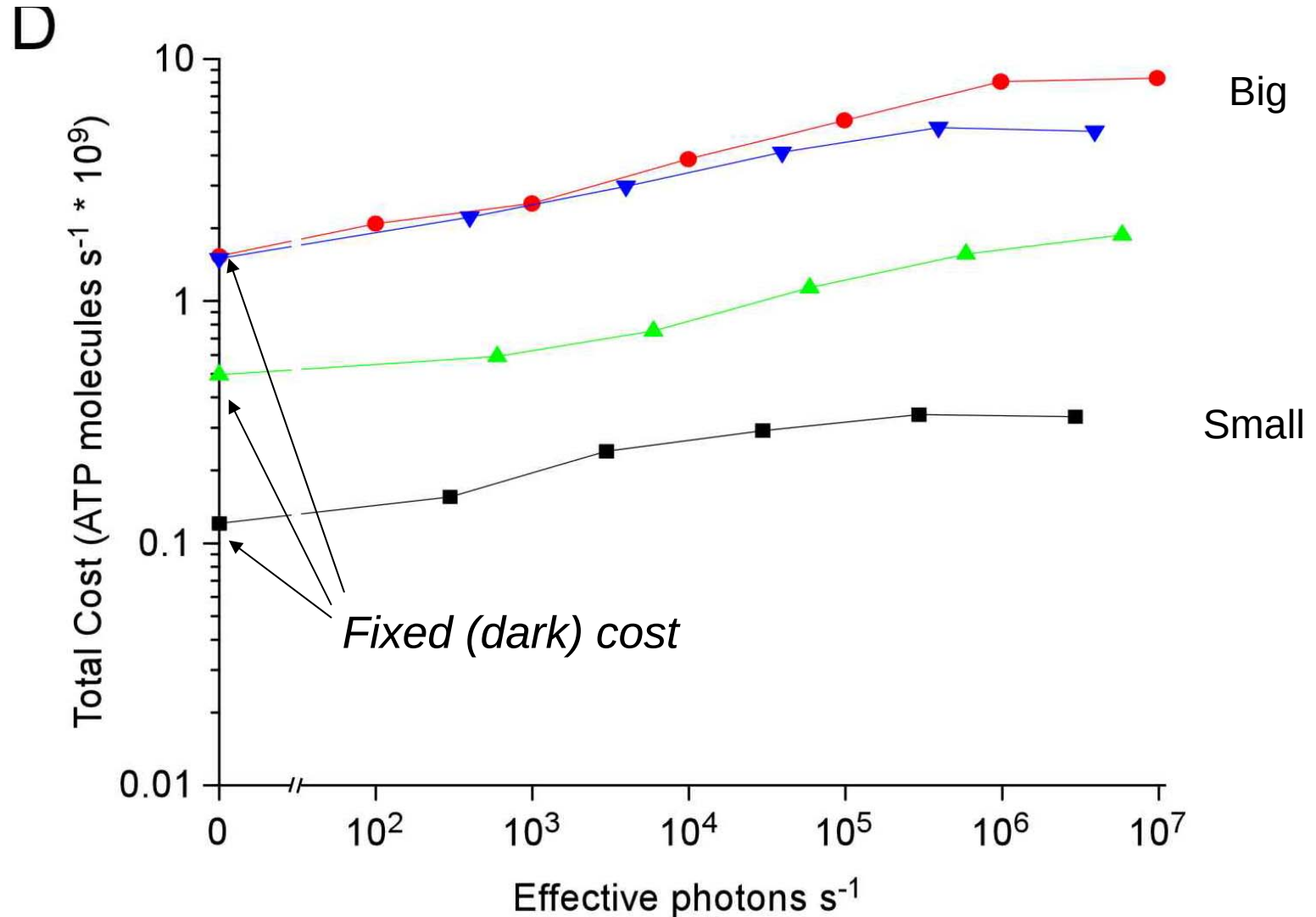
Big cell are faster and more precise



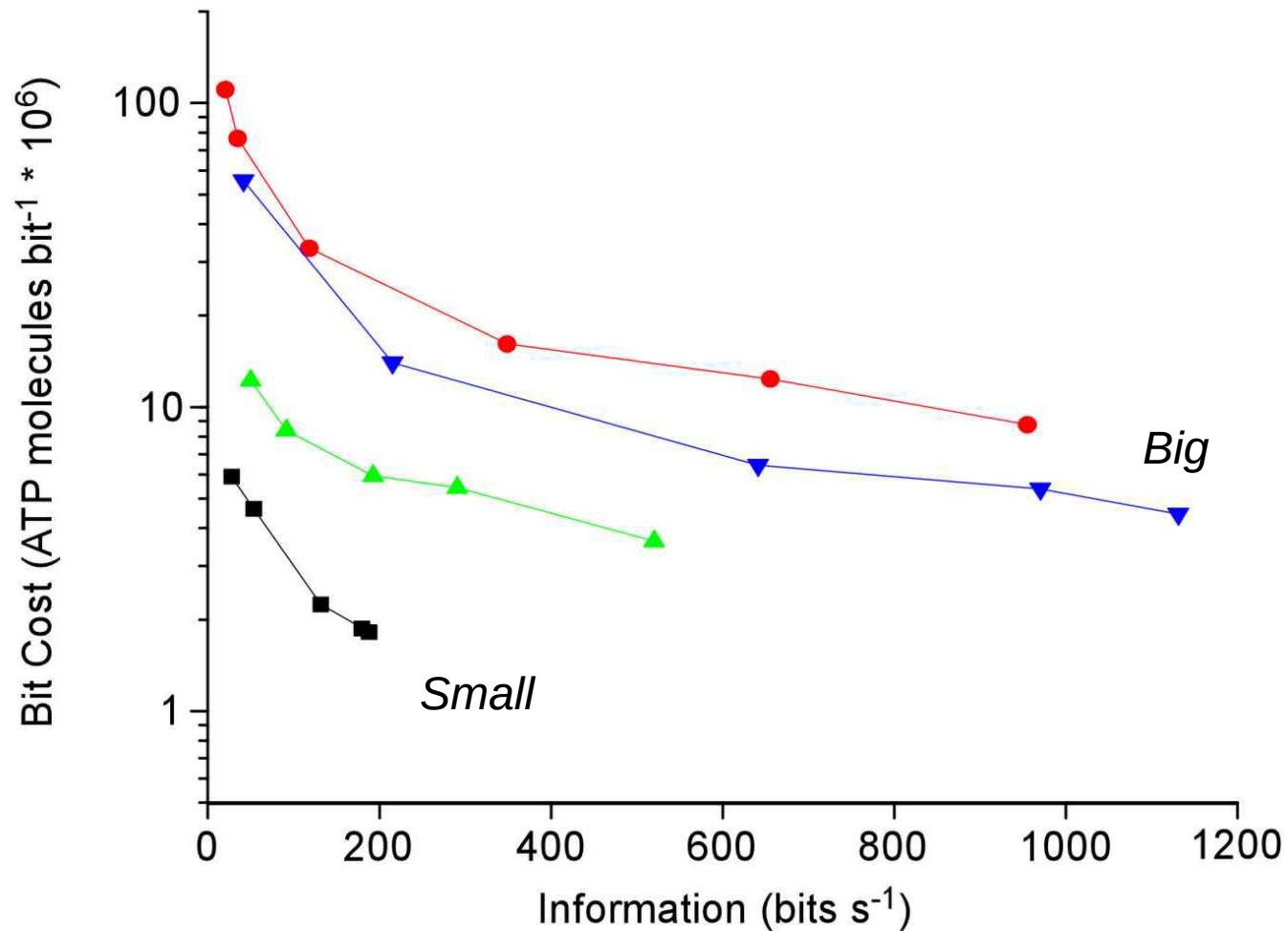
Membrane model gives energy consumption



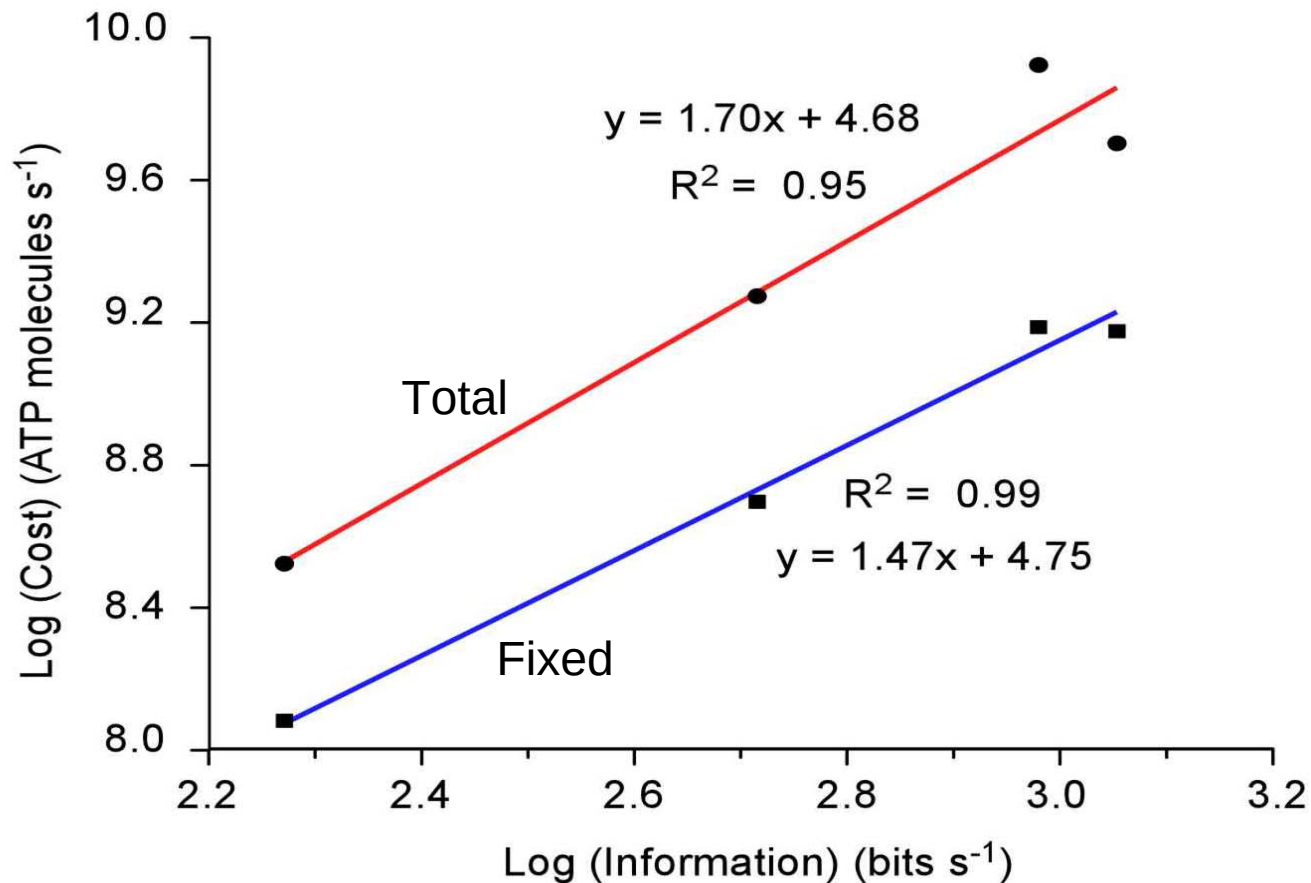
Different levels of energy consumption



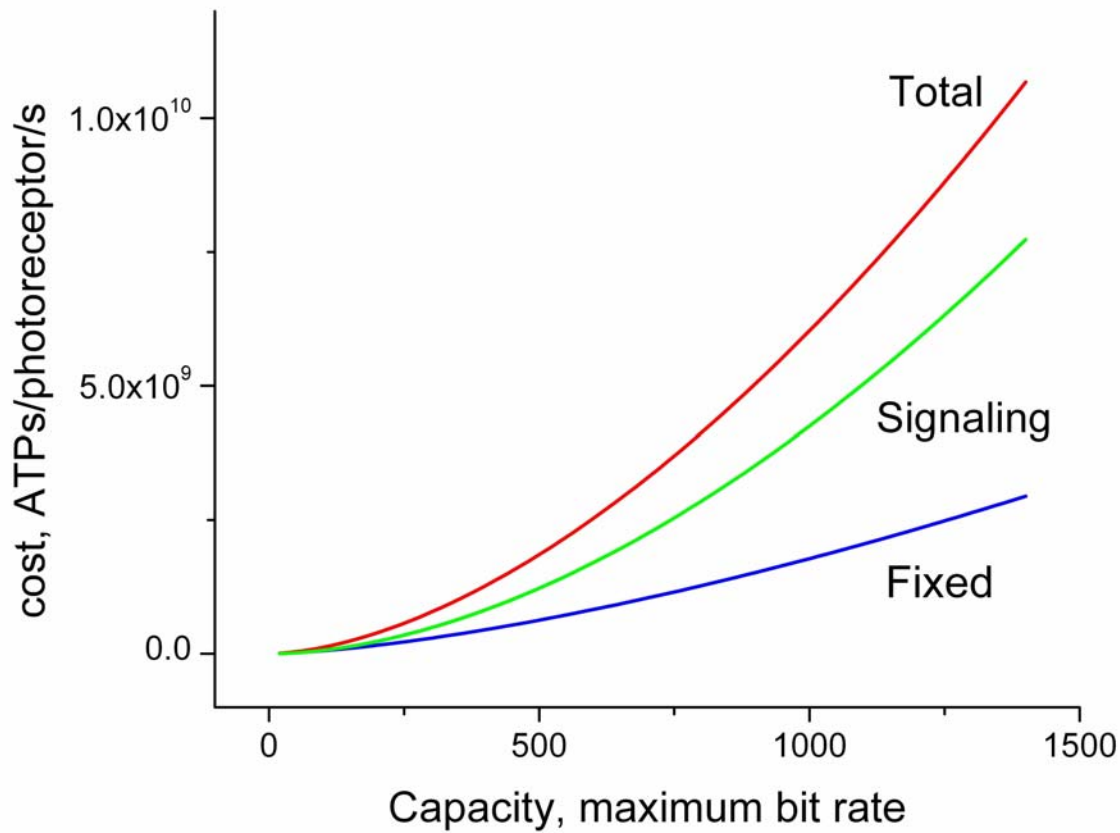
Bigger cells are less energy efficient
(they have traded economy for capacity)



Cost increases out of proportion to capacity



The Law of Diminishing Returns



How to implement

Send at the lowest rate

Distribute information and processing tasks among parallel low rate channels

sparse coding, parallel pathways

Only use high speed/precision where it is essential

small number of high speed/precision streams

larger number of low speed/precision stream

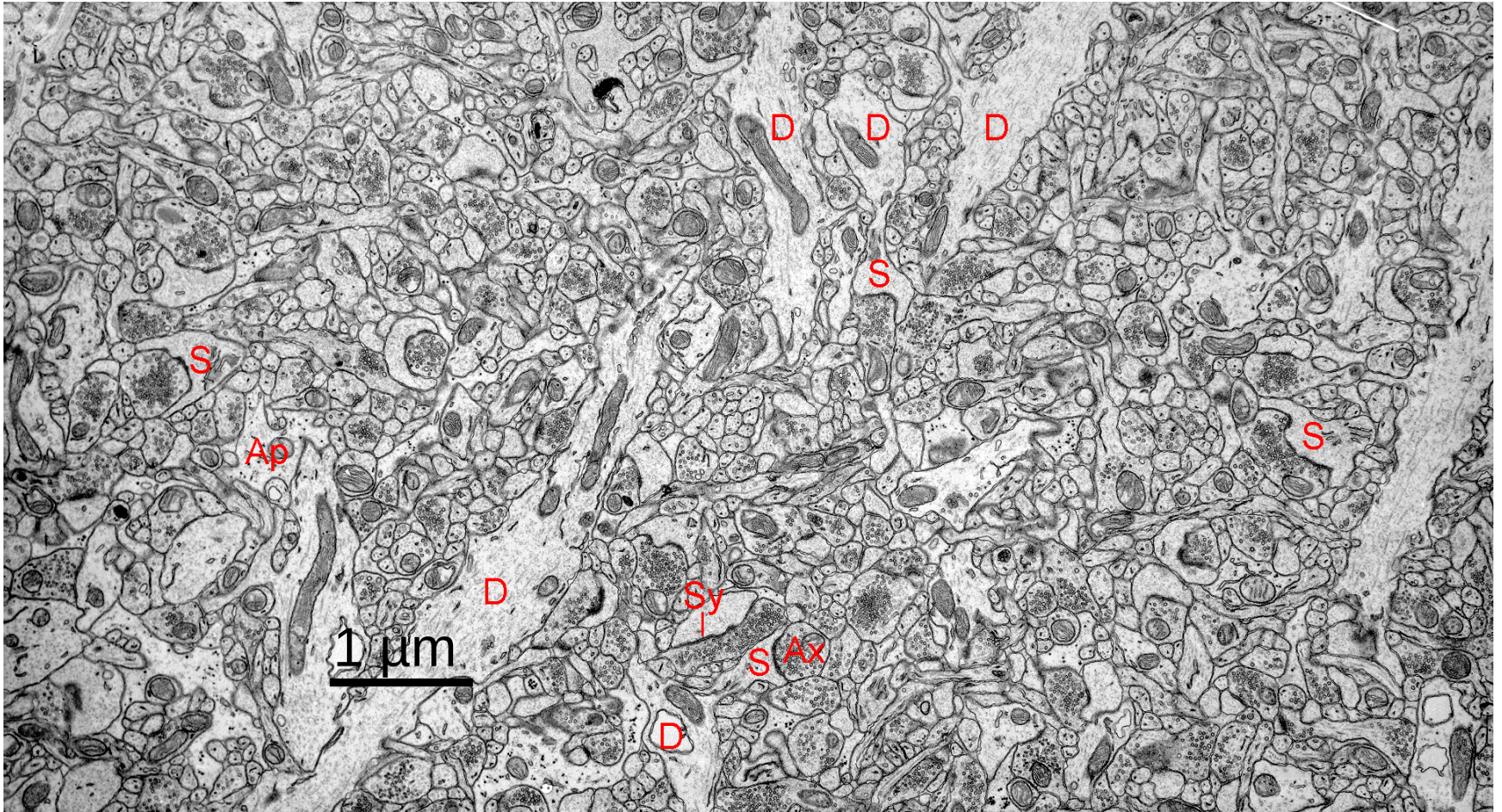
Massively parallel processing in low rate channels each dedicated to small part of the task

Secret 9

Nanofy!

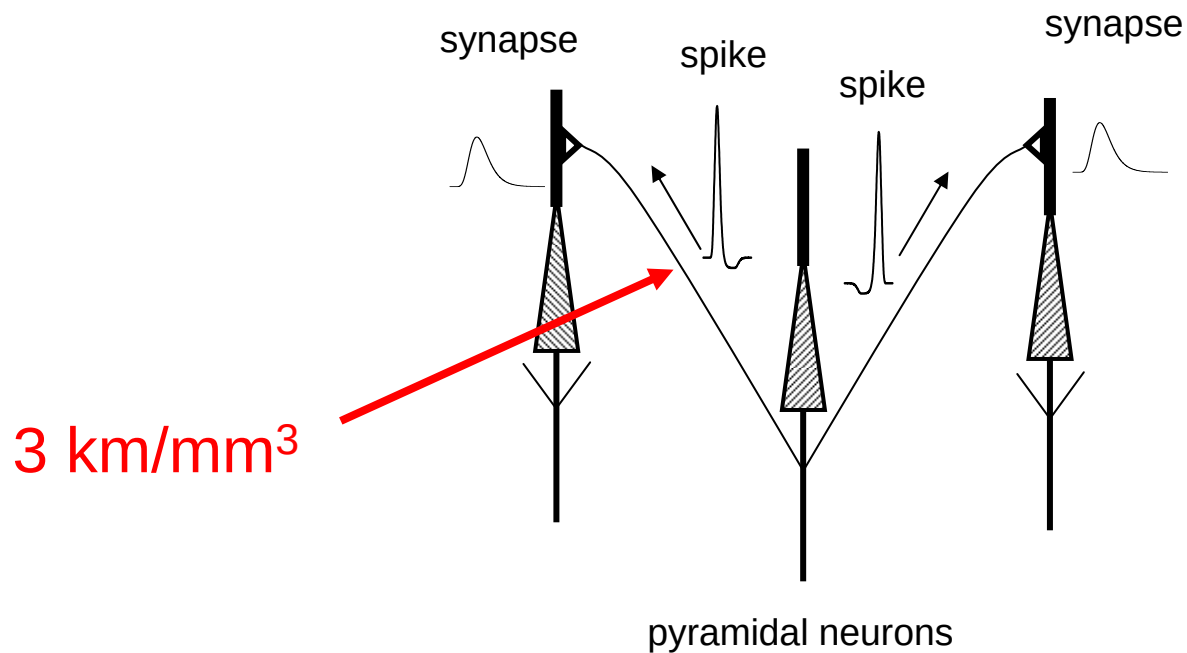
(reduce to the irreducibly small)

Miniaturisation



Dendrites (D) of CA1 pyramidal cells – EM section by Dr J. Spacek, Charles Univ Czech Rep. Visit *Synapse Web* for more details

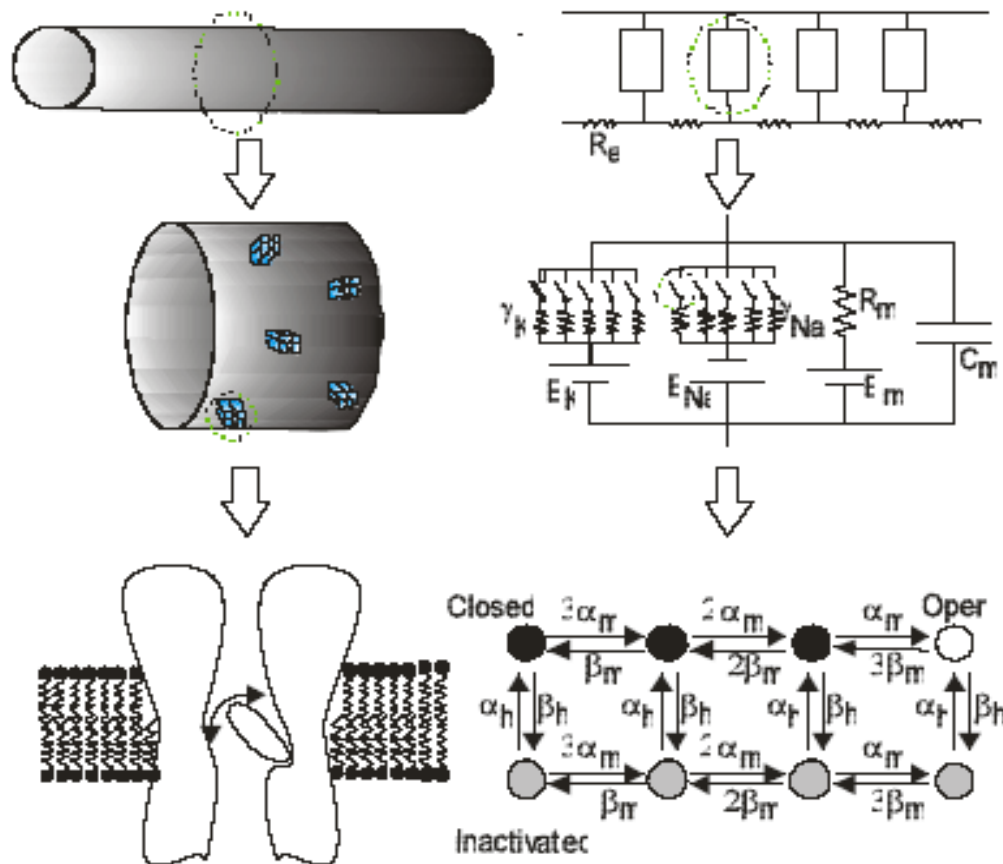
What is this spaghetti?

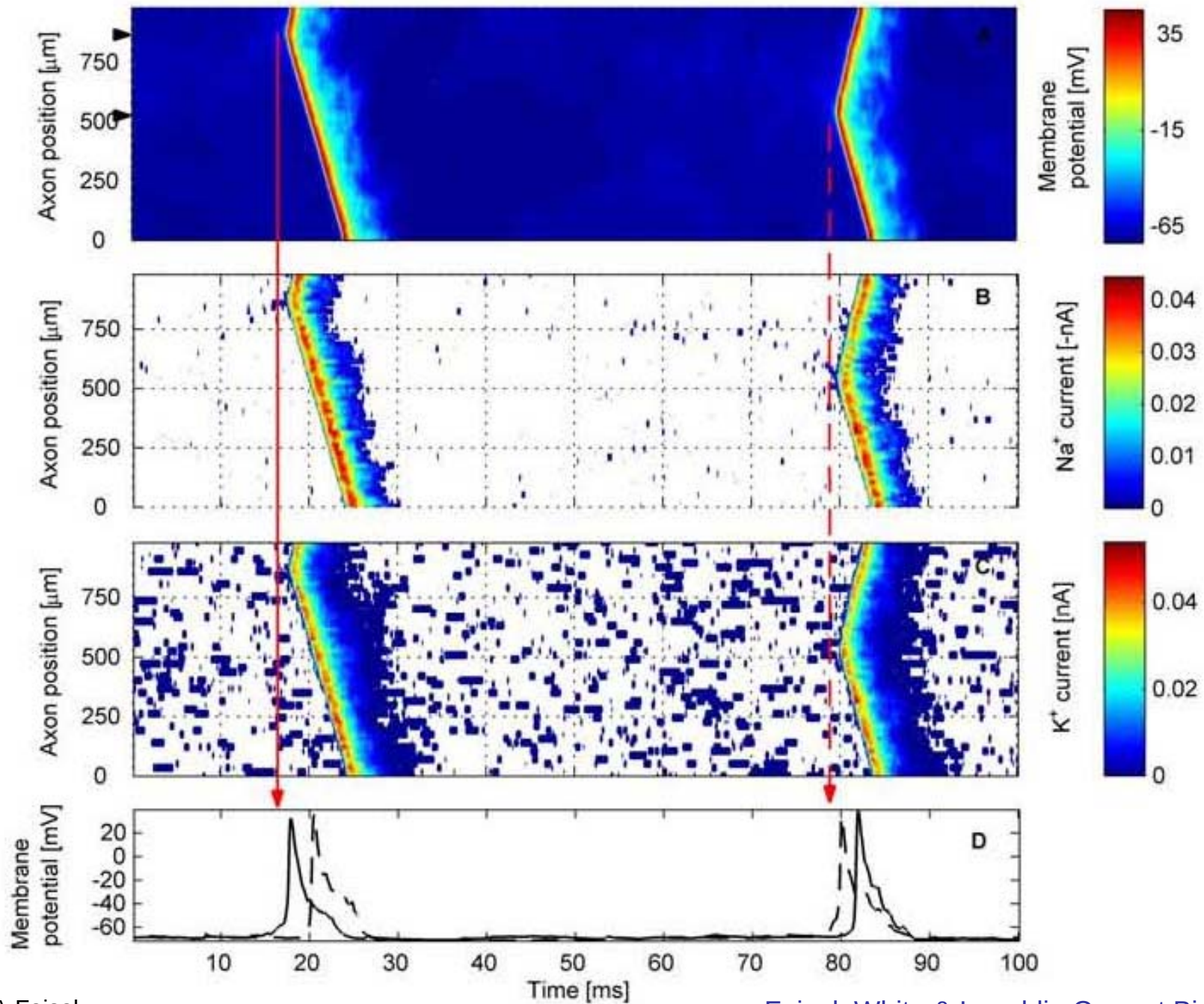


A channel noise limit to axon miniaturisation

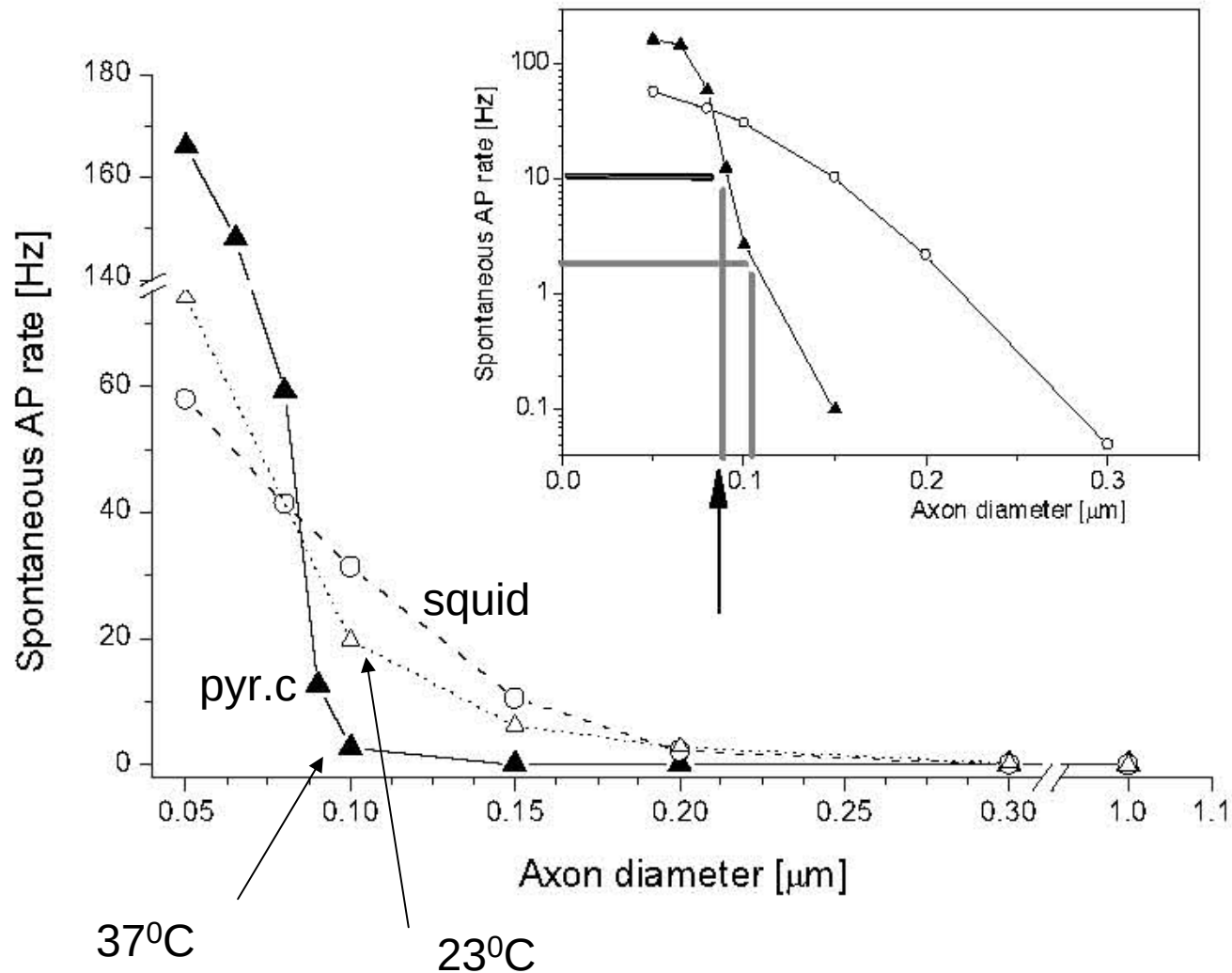
Aldo Faisal, John White and Simon Laughlin 2005

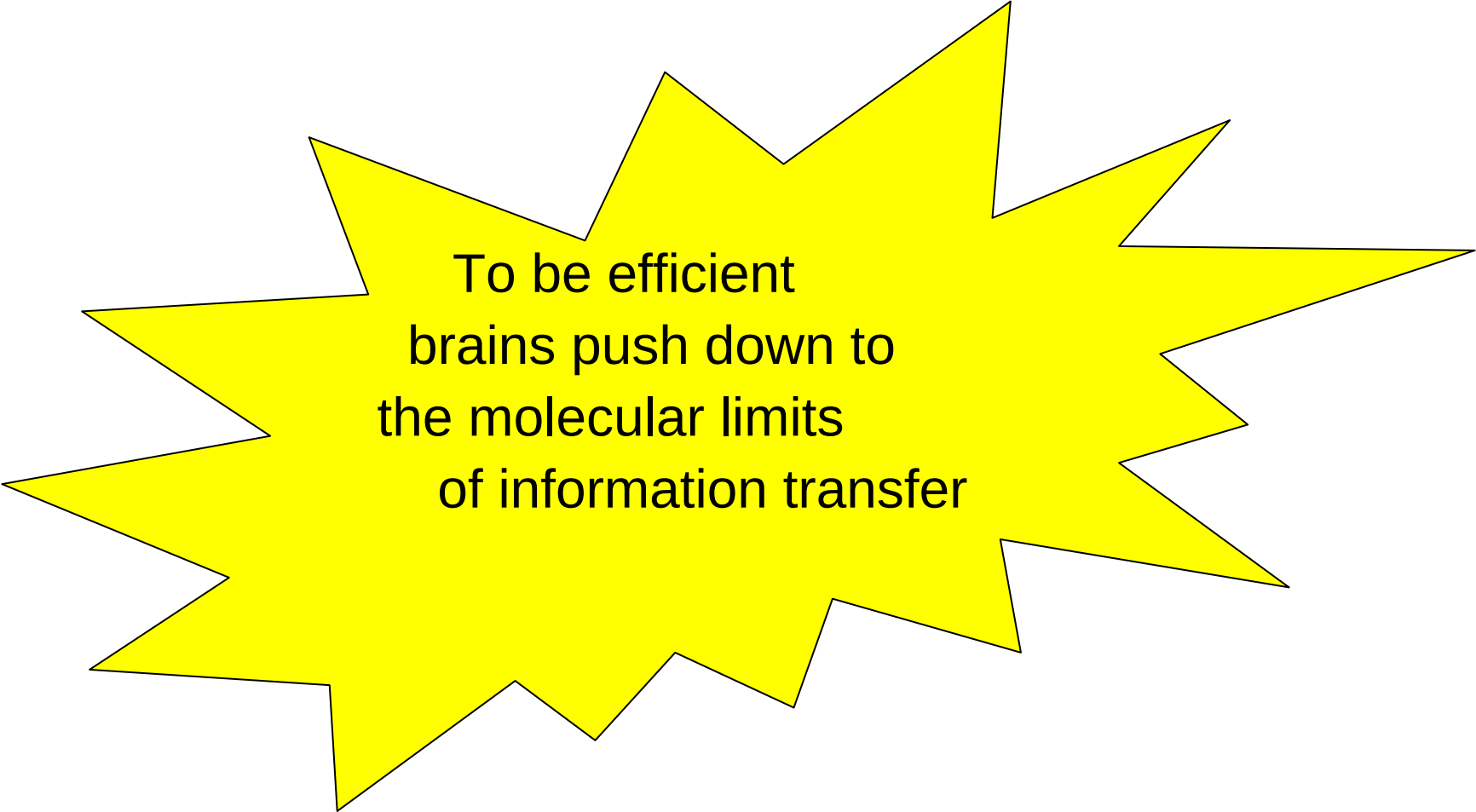
stochastic axon model





Spontaneous rate increases rapidly with decreasing diameter, below $0.2\ \mu\text{m}$





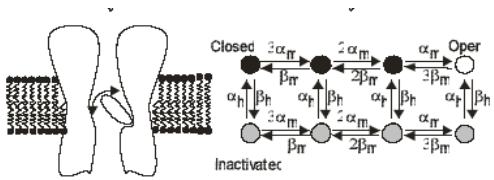
To be efficient
brains push down to
the molecular limits
of information transfer

Secret 10

Do it with chemistry
(because ye have flesh)

The voltage sensitive sodium channel

a molecular machine



Voltage
sensors



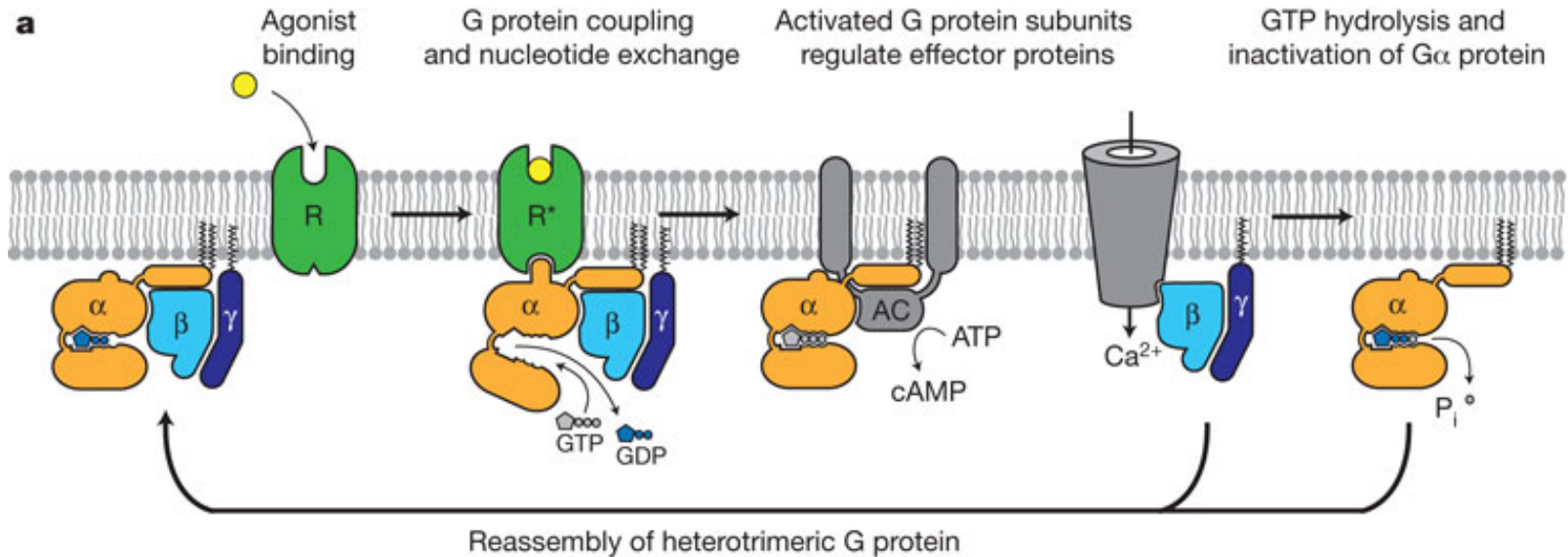
Voltage
sensors

From Horn, *Nature* 2011
summarising Payandeh *et al*, *Nature* 2011

Protein power

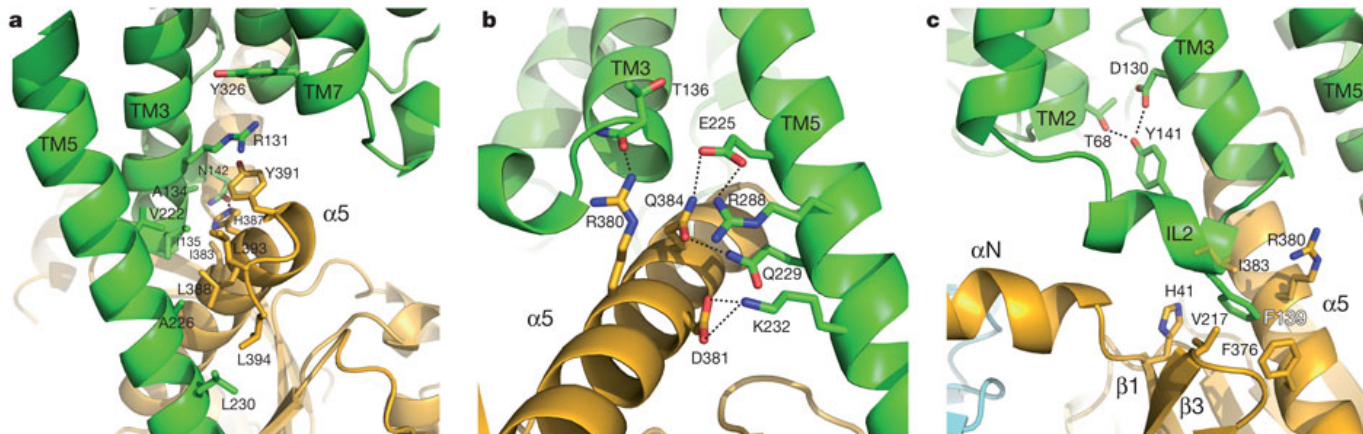
Nano (single molecule) finite state machines
switching by conformational changes

G protein cycle for the β_2 AR–Gs complex.



SGF Rasmussen *et al. Nature* **000**, 1-7 (2011) doi:10.1038/nature10361

Receptor-G protein interactions.

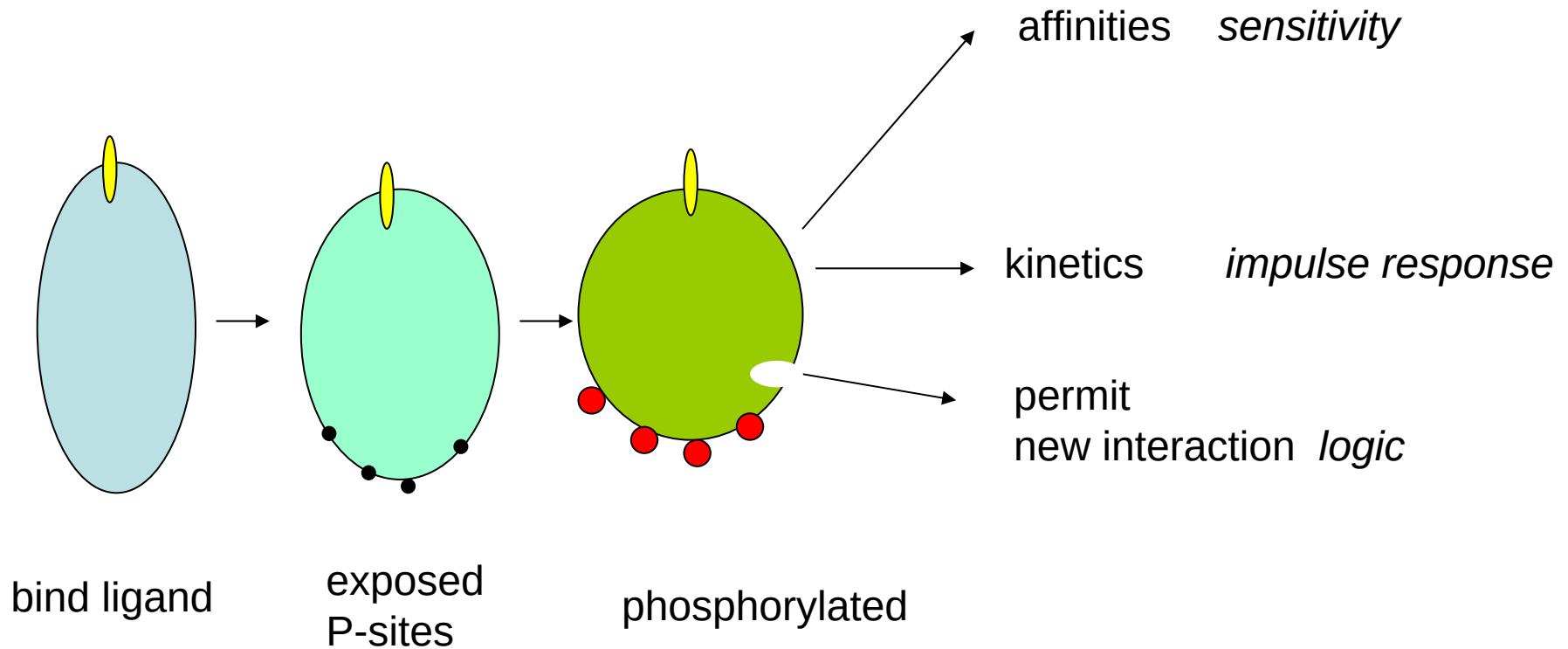


SGF Rasmussen *et al. Nature* **000**, 1-7 (2011) doi:10.1038/nature10361

Note: This figure is from a near-final version AOP and may change prior to final publication in print/online

nature

Changing protein nano machine states by changing their energy landscape

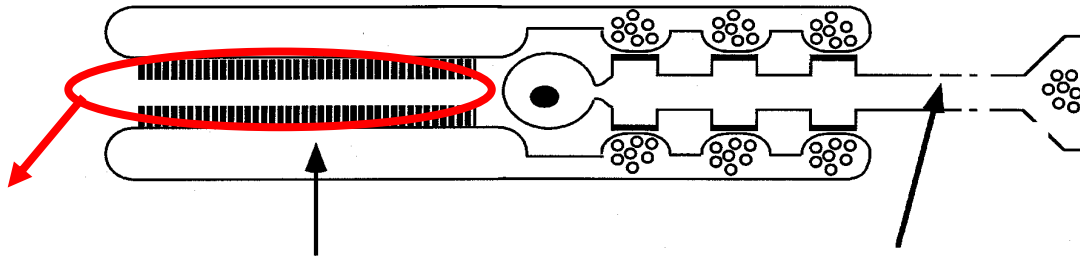


Protein molecules are flexible and versatile

- Easily fine tuned
 - variable gain amplifiers
 - temporal filters
- Process information
 - coincidence detectors (AND gates)
 - NAND gates
- Assembled into information processing complexes and networks

Fine tuning

photoreceptor protein network implements
van Hateren's optimum filter



*Tune
impulse
response
to SNR
(light level)*

R1-6

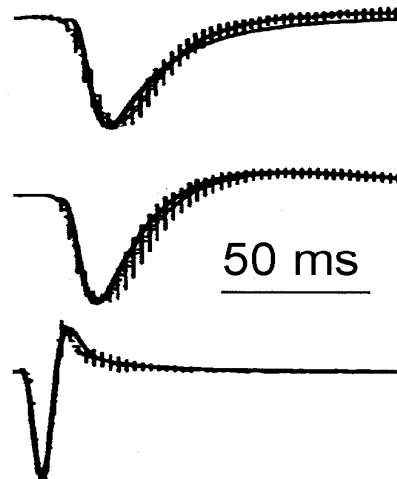
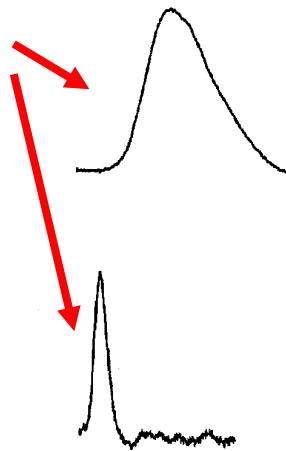
LMC

LMC Data |
Model ____

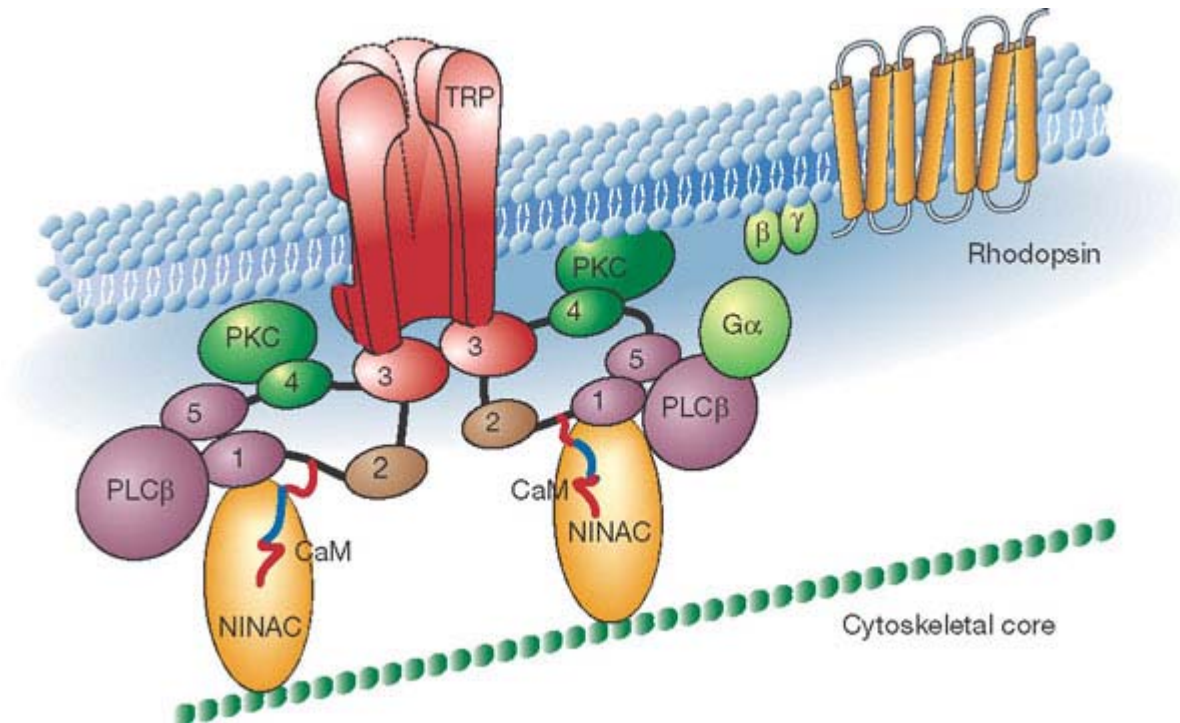
Very dim

Quite dim

daylight



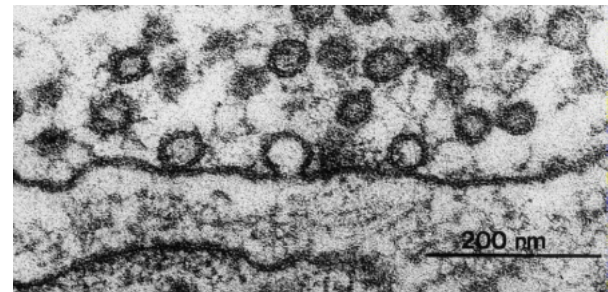
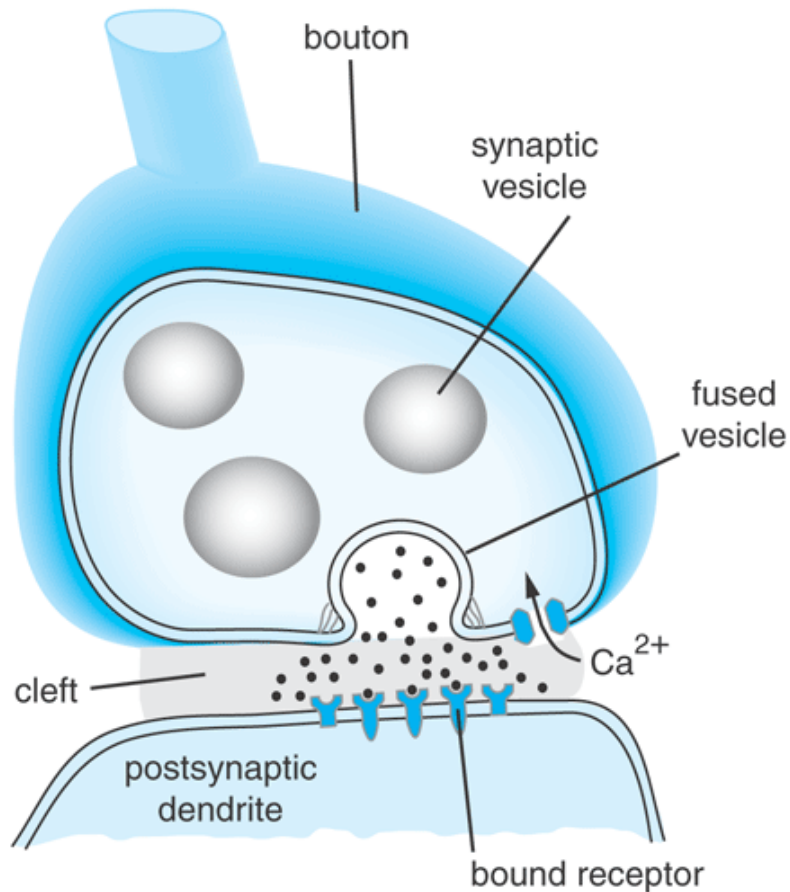
Molecular machinery for phototransduction in the fly photoreceptor



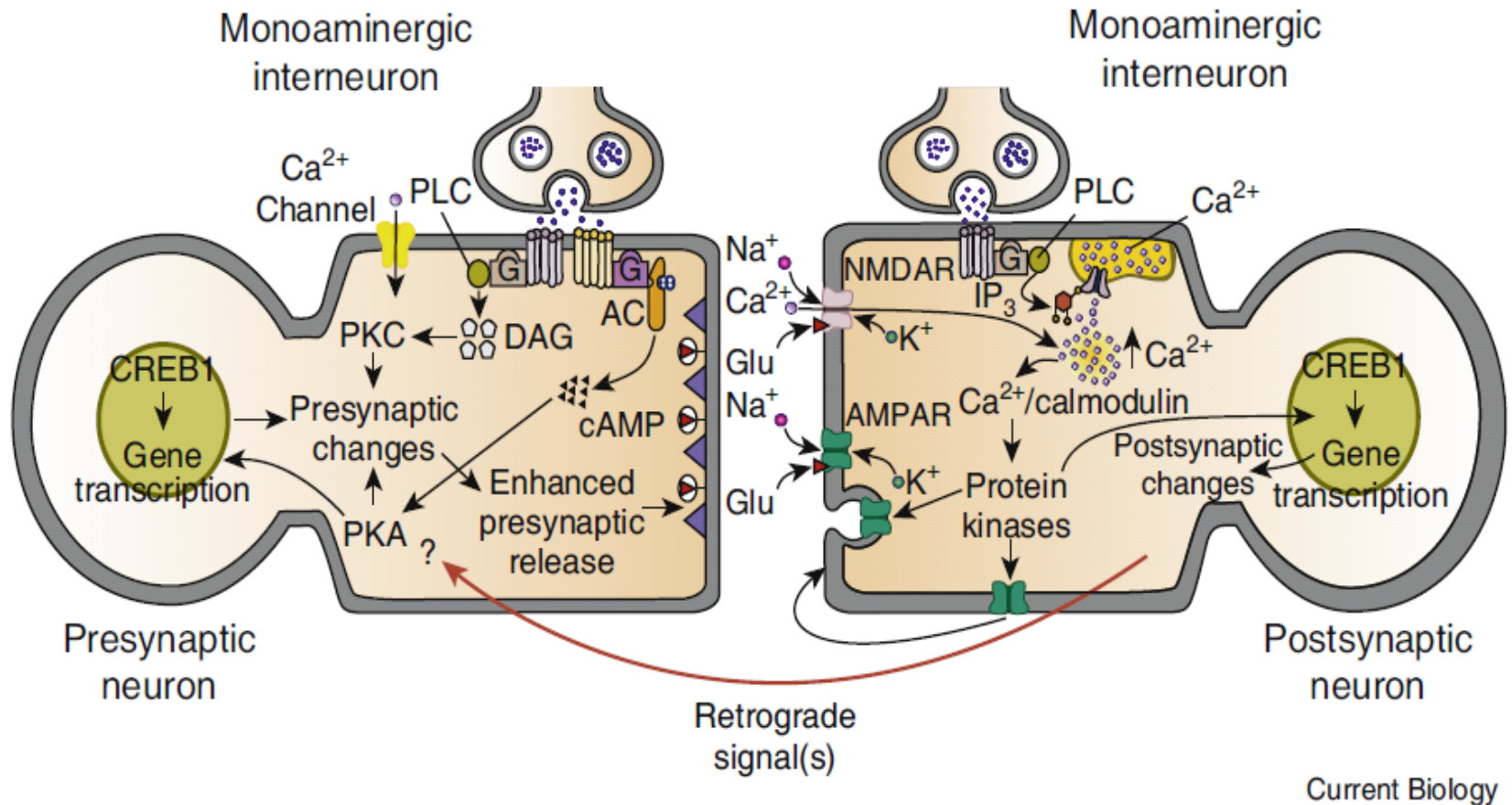
Hardie & Raghu, 2001

Networks revealed

*The chemical synapse before
the molecular neurobiology revolution*



The outcome of revolution - promiscuity



Protein potential

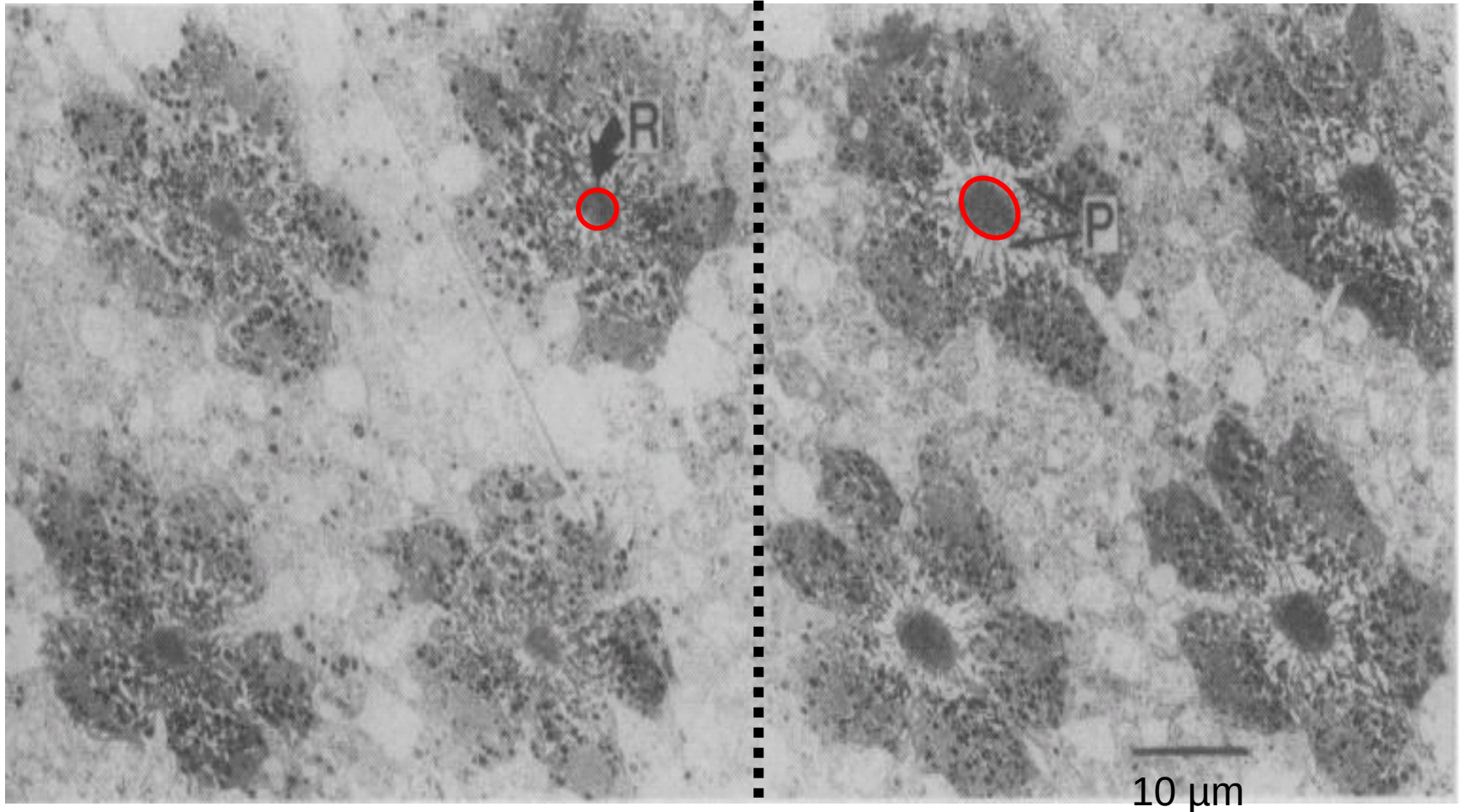
- Nano-scale devices
- Receive transduce and transmit signals
- Select, amplify and filter signals
- Rich repertoire of analogue and logic operations
- Form networks that integrate and process signals, *therefore*
- *Control and drive all cellular processes*
- *Reconfigure networks – demand, history*
- *Allocate resources – demand, history*

Changing sensor sensitivity – day to night

Locust

light

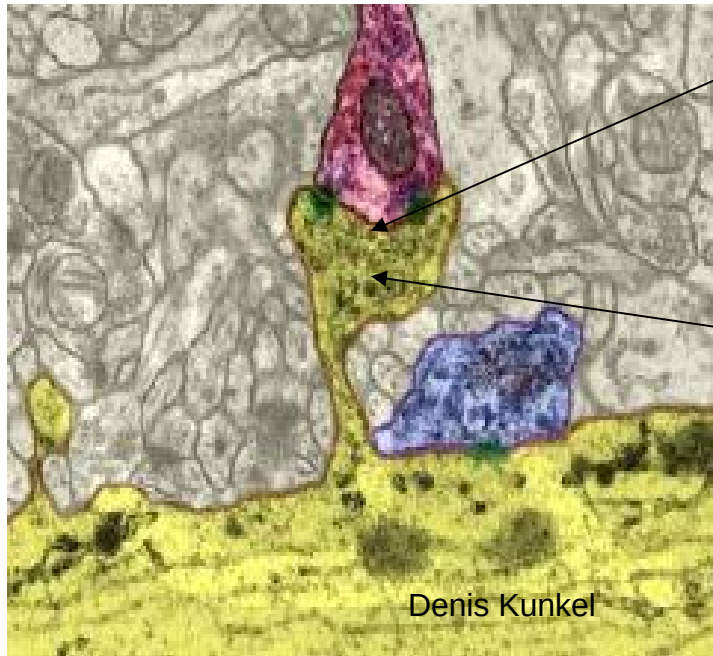
dark



David Williams, Science 1982

Chemistry is cheaper than
Electronics

Cost of signalling at single cortical spine



Synaptic current flowing through receptor ion channels

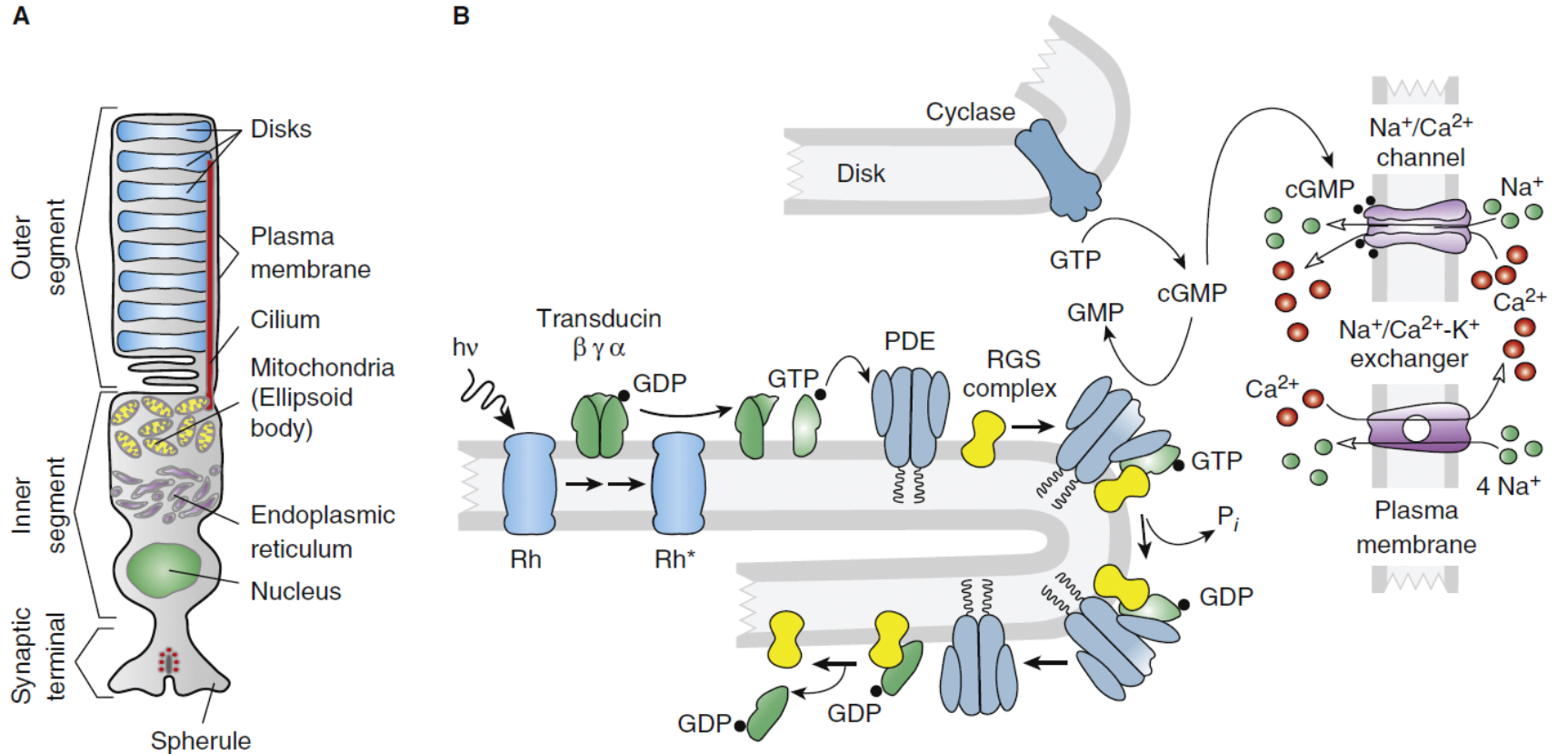
200,000 ions, 67,000 ATPs

Post-synaptic Ca transient triggered by G-protein coupled receptors

1000 ions 1000 ATP

Attwell & Laughlin, 2001

A heavy weight chemical signalling network rod photoreceptors

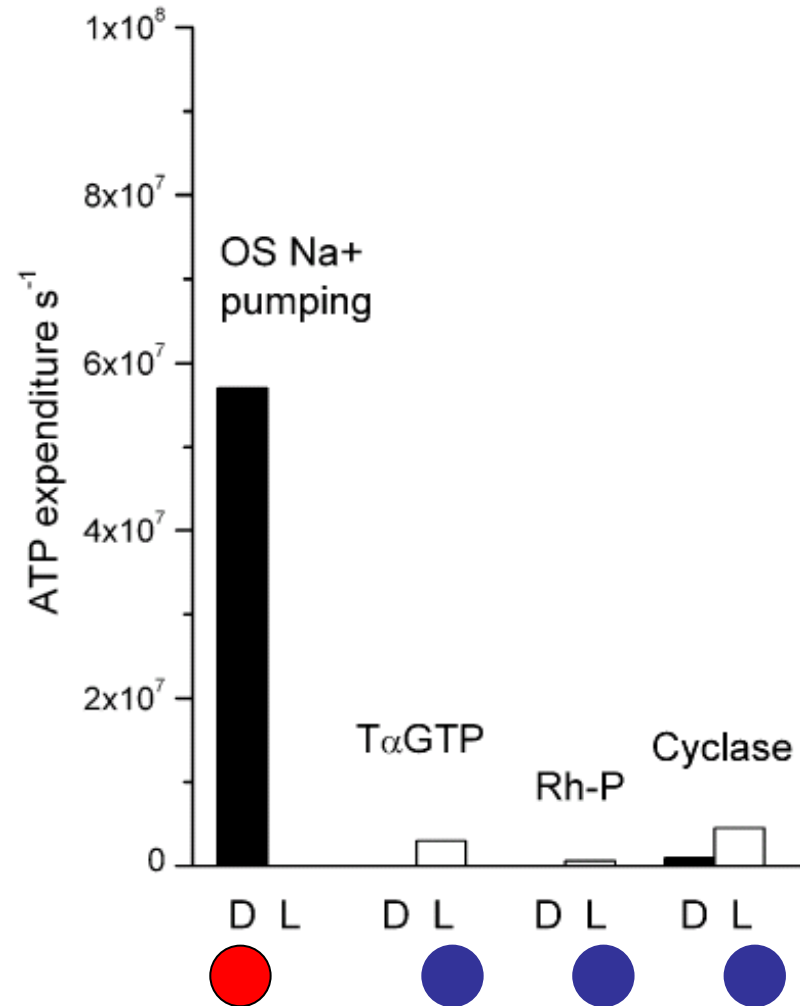


Current Biology

Fain, Hardie & Laughlin, Current Biology 2010

Energy costs of **electrical** and **chemical** signalling in a mouse rod photoreceptor

Okawa, Sampath, Laughlin & Fain, 2008



Secret 11

Mix electronics and chemistry

Getting the mix balance pro and cons

- **Chemistry**

- versatile
- computationally powerful
- economical
- links all levels
- slow
- no universal currency at a node

- **Electricity**

- less versatile
- computationally weak
- expensive
- isolated (uses chem)
- fast
- universal currency, "instantaneous" integration of many to one.