

# Modeling synapses: Noise and dynamics

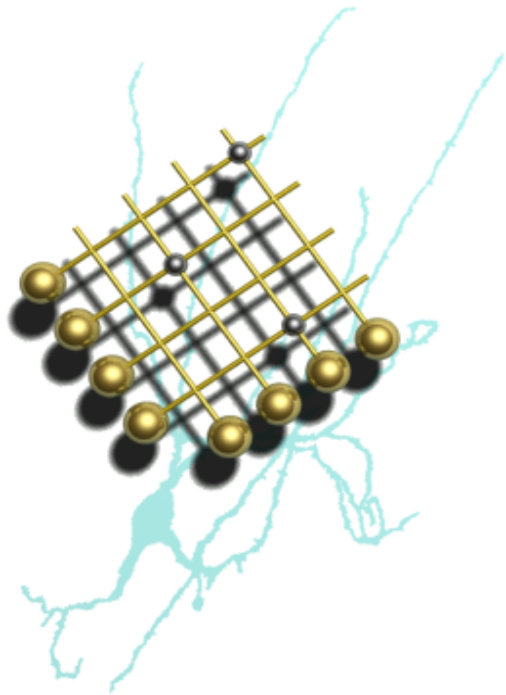
1

Mark van Rossum

ANC

University of Edinburgh

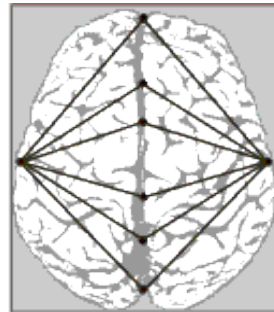
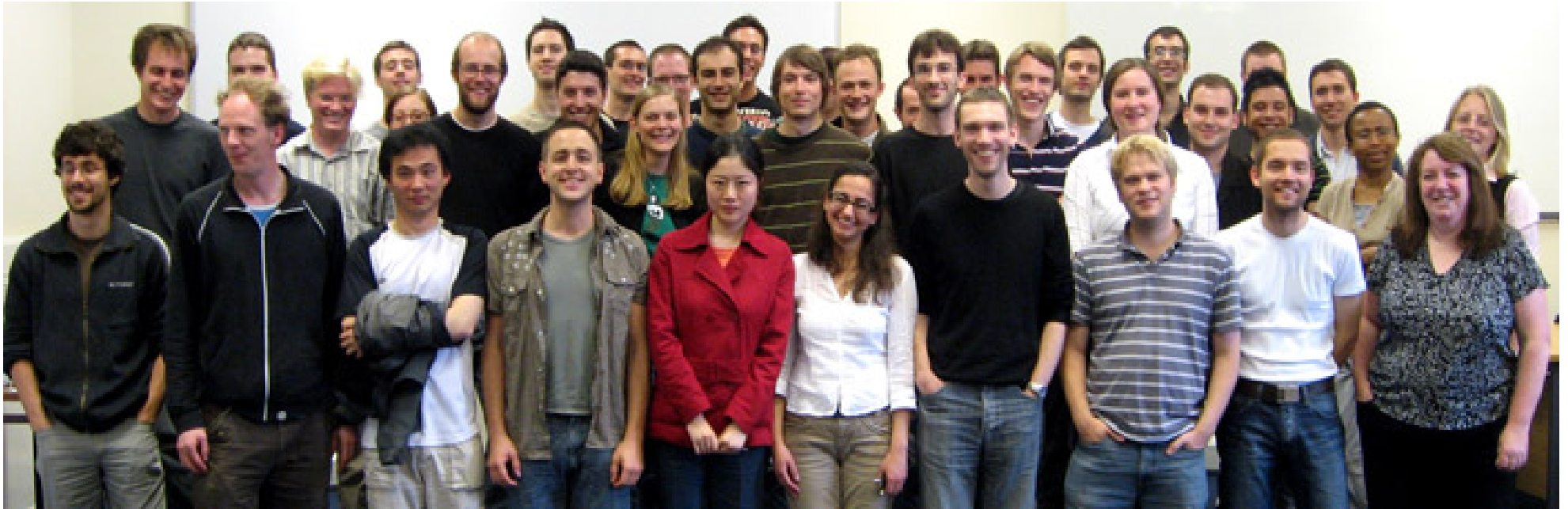
[mvanross@inf.ed.ac.uk](mailto:mvanross@inf.ed.ac.uk)



Lawrence York, Jesus Cortes, Cian O'Donnell,  
Paul Clark,  
Robert Smith (U Penn)

# Adverts

2

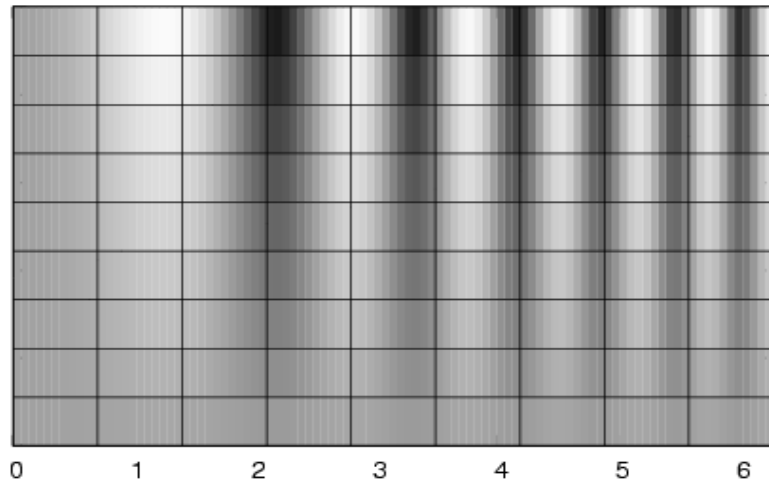


Neuroinformatics  
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# Links between synapses and perception in the visual system

3

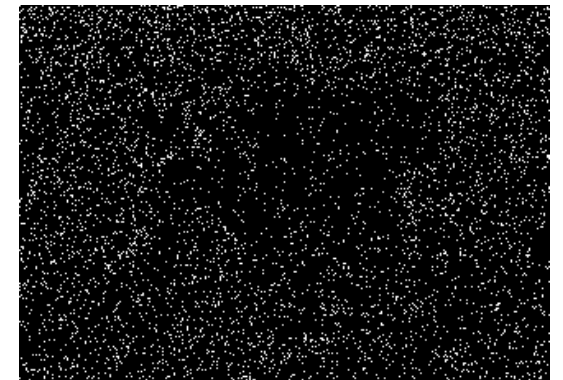
Contrast perception (TUESDAY)



Synaptic requirements for night vision

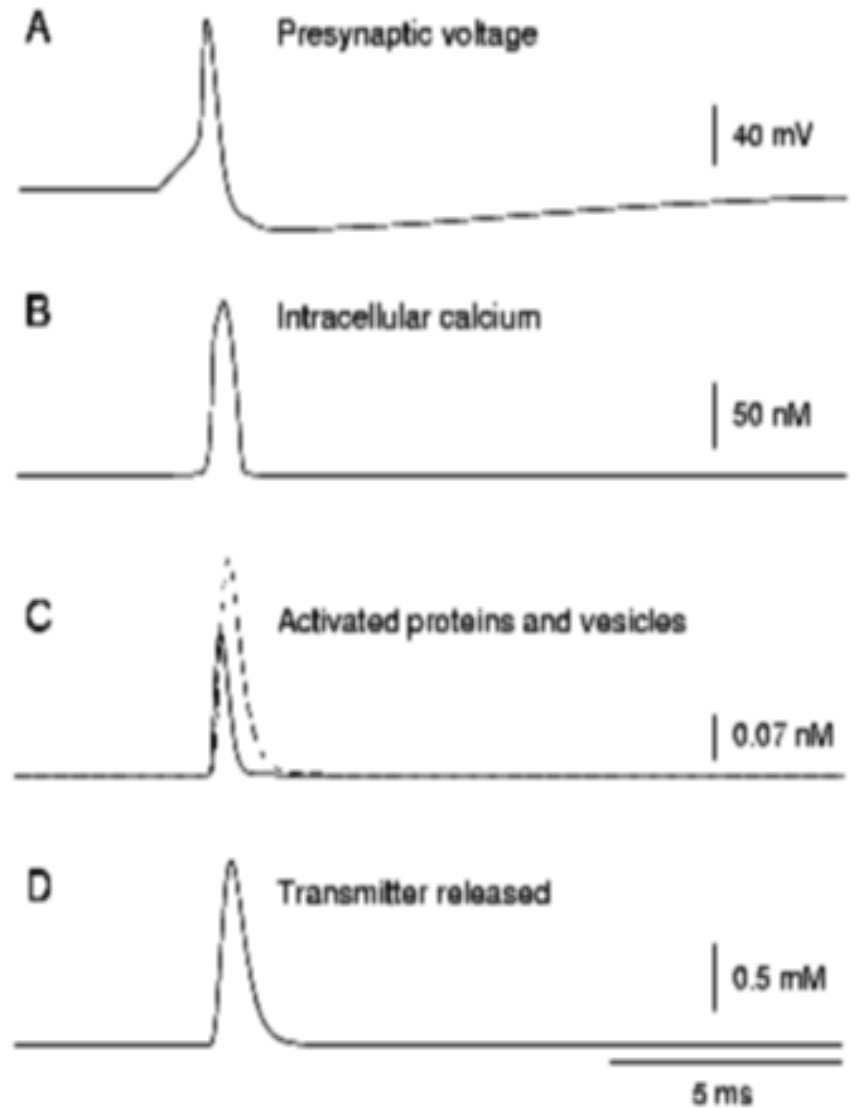
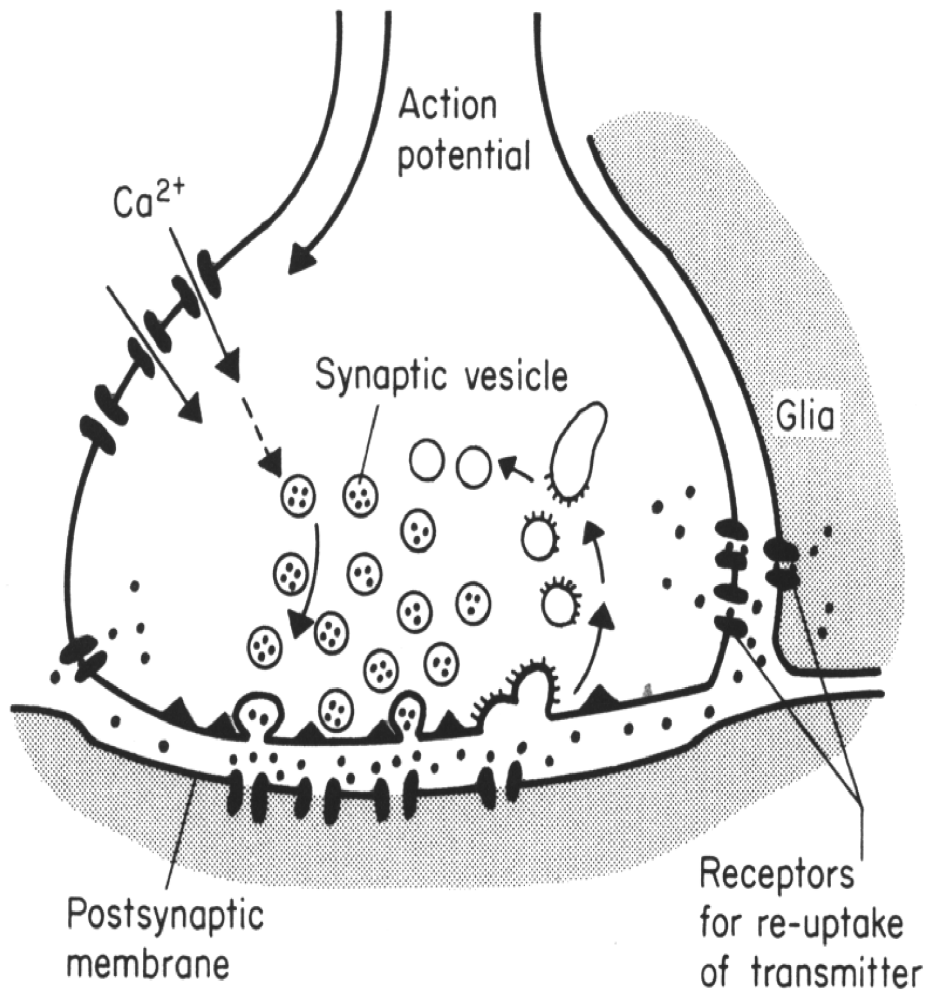


Day light

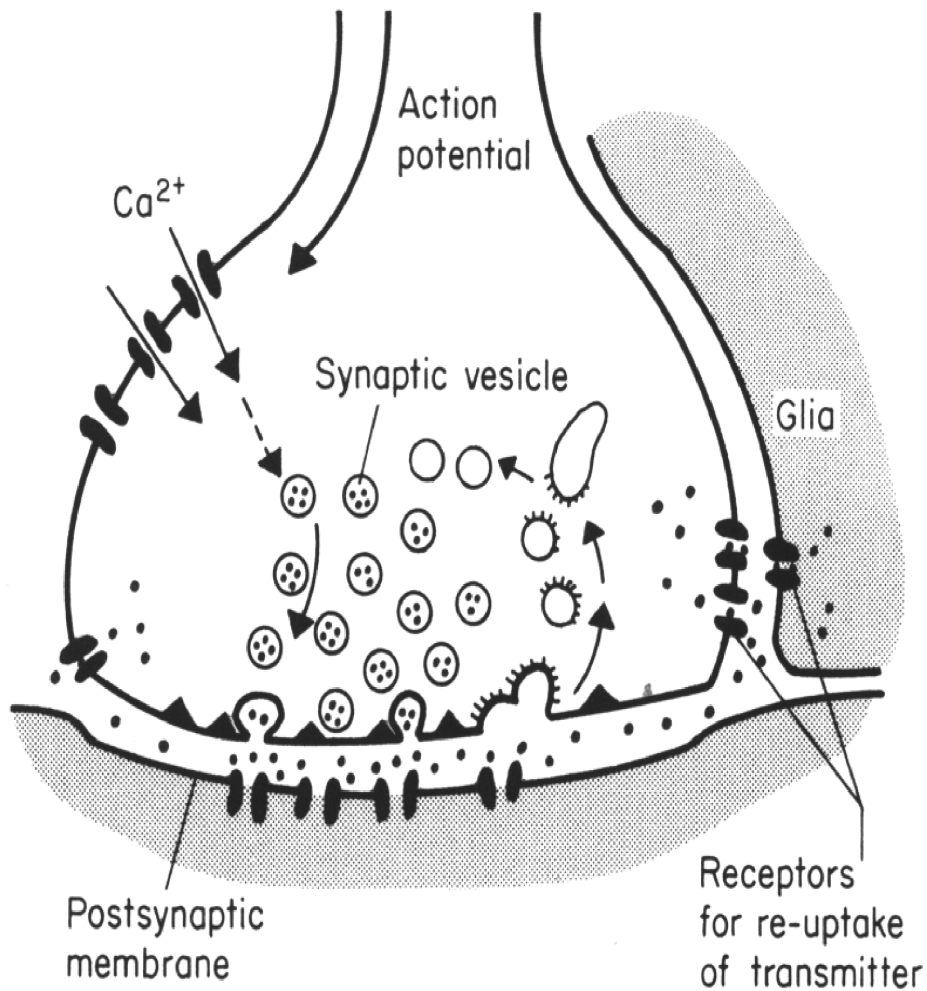


At night

# The synapse



# The synapse



## Table of contents:

- synapse models and synaptic noise
- synaptic non-linearities
- release probability

## Not discussed:

- pre-synaptic receptors
- LTP
- receptor insertion/diffusion
- release biophysics
- synaptic integration on dendrites
- spines

# PART 1: Synaptic channel models

# Synaptic currents

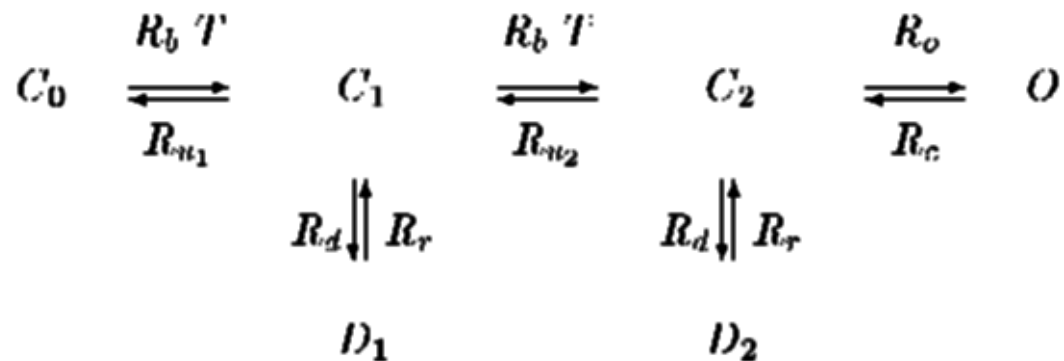
- Current based, e.g.

$$I_{syn} = I_0 \exp(-t/\tau_{syn})$$

- Conductance based, e.g.

$$I_{syn} = g_{syn} [V(t) - V_{syn}^{rev}] \exp(-t/\tau_{syn})$$

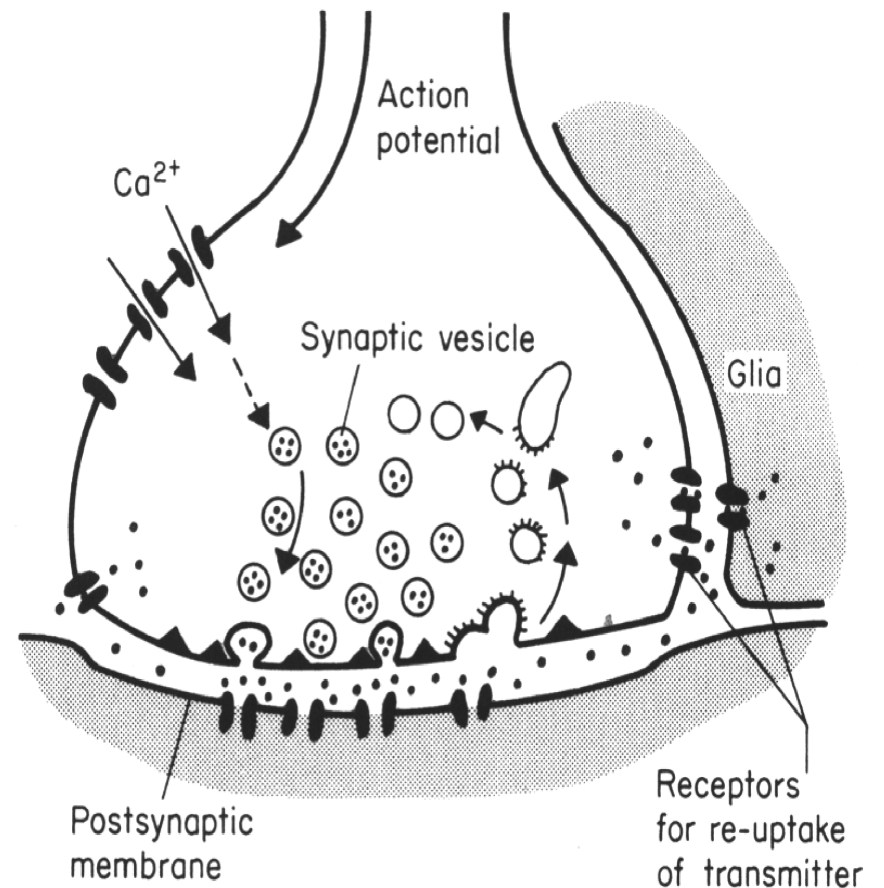
- Full state diagram: e.g. AMPA channel



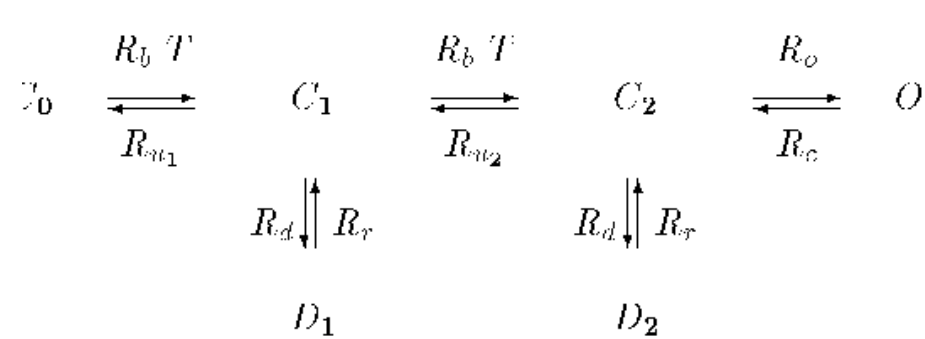
# Noise in synaptic input

Synaptic current is noisy because

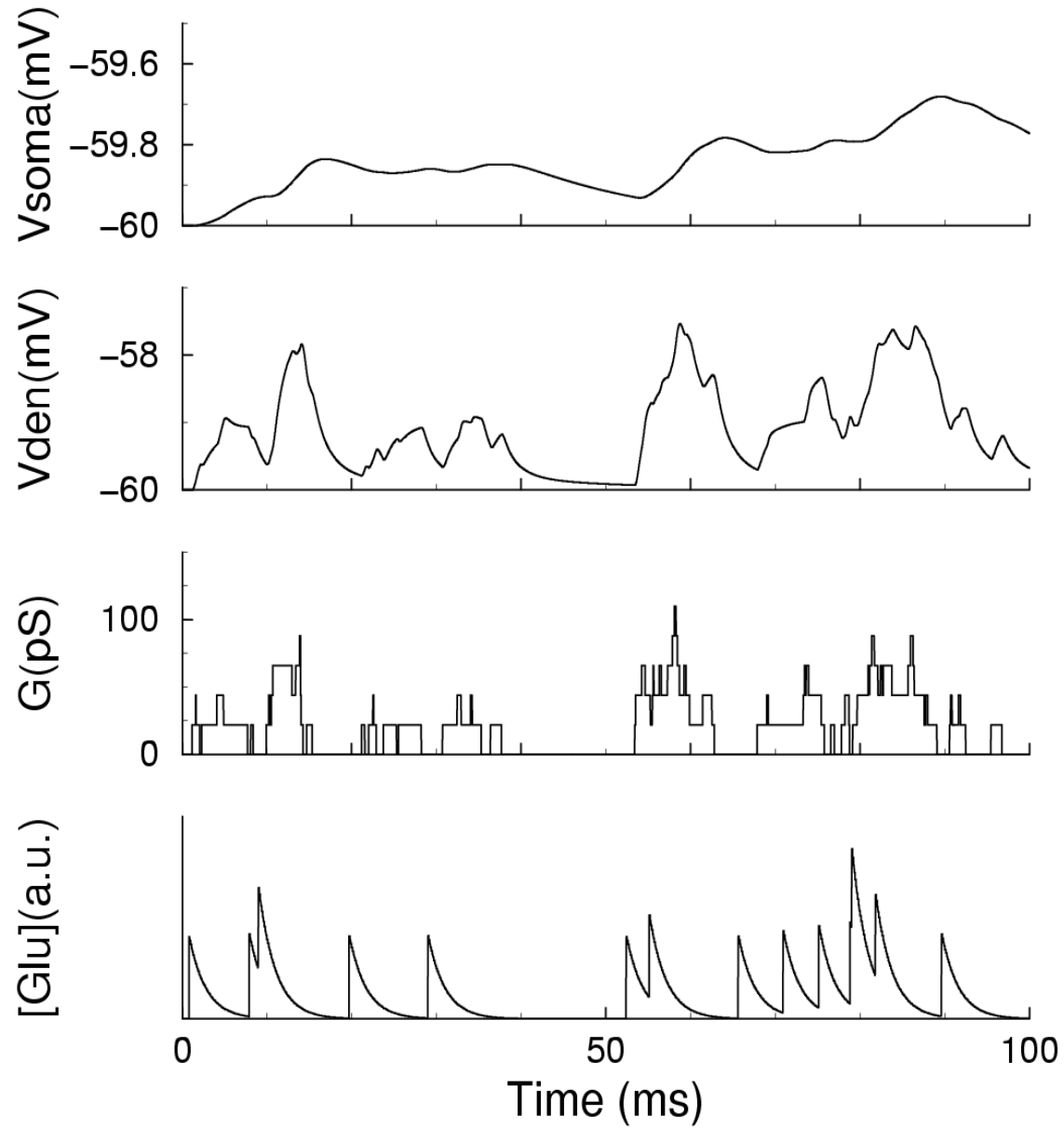
- vesicles vary in size
- vesicle release is stochastic
- variation in release location
- synaptic channel opening is stochastic



# Synaptic currents simulation



Markov diagrams can be simulated  
Stochastically  
or  
deterministically



# Markov formulation



State vector  $s$ , given occupation of each state  $\frac{d\vec{s}}{dt} = W \cdot \vec{s}$

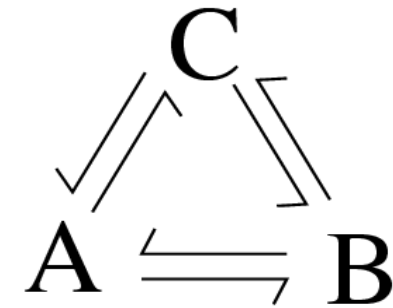
$$W = \begin{pmatrix} -kT & \alpha & 0 \\ kT & -k'T - \alpha & \alpha' \\ 0 & k'T & -\alpha' \end{pmatrix} \quad \lambda = \{0, \lambda_1, \lambda_2\}$$

$$\vec{s}(t) = \vec{s}_\infty + \vec{c}_1 \exp(-t/\lambda_1) + \vec{c}_2 \exp(-t/\lambda_2)$$

Preserve probability  $\frac{d \sum \vec{s}}{dt} = 0$

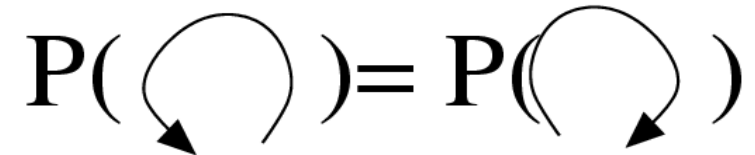
# Markov formulation

Detailed balance (statistical physics)



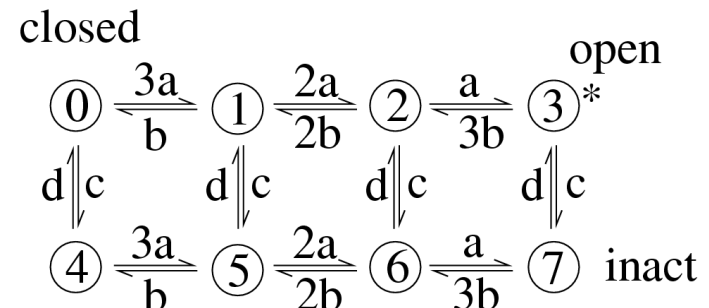
In equilibrium detailed balance should hold.

Without detailed balance, looping would be possible.



In diagrams without loops, detailed balance is trivial.

$$W_{ij} S_j^{equil} = W_{ji} S_i^{equil}$$



# Markov formulation

$$C \Leftrightarrow C' \Leftrightarrow O$$

Suppose equilibrium has been reached (e.g. transmitter in the bath). Autocorrelation function

$$c(t) = \int dt' s(t') s(t'+t)$$

Autocorrelation has two time-constants in this case:  $1/\lambda_1, 1/\lambda_2$

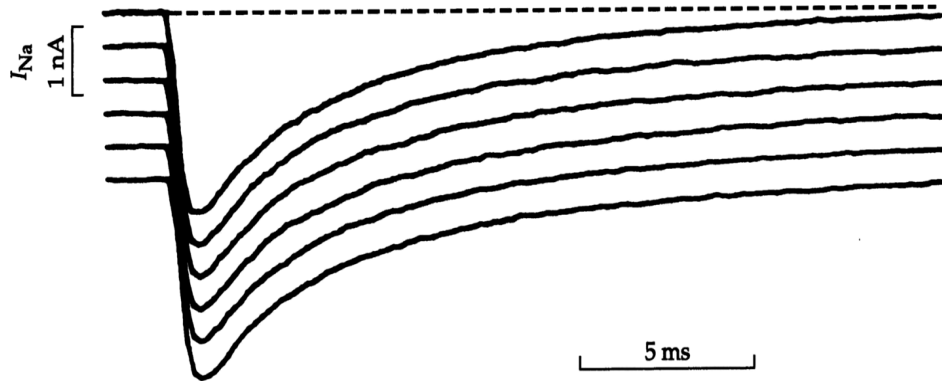
Wiener-Khinchin theorem:

$$PSD(f) = 4 \int_0^{\infty} c(t) \cos(2\pi f t) dt$$

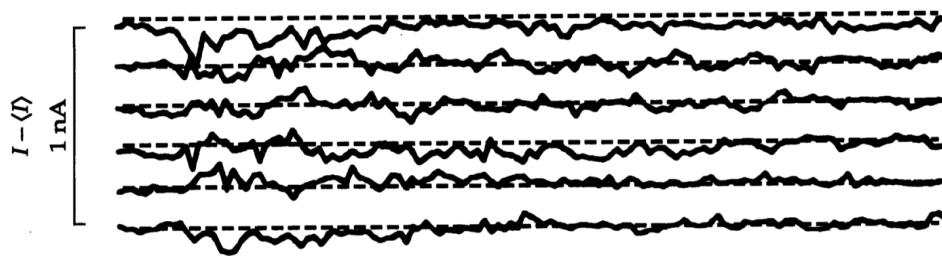
$$PSD(f) = \frac{w(0)}{\prod_i [1 + (2\pi f / \lambda_i)^2]}$$

The channel dynamics determine the noise spectrum

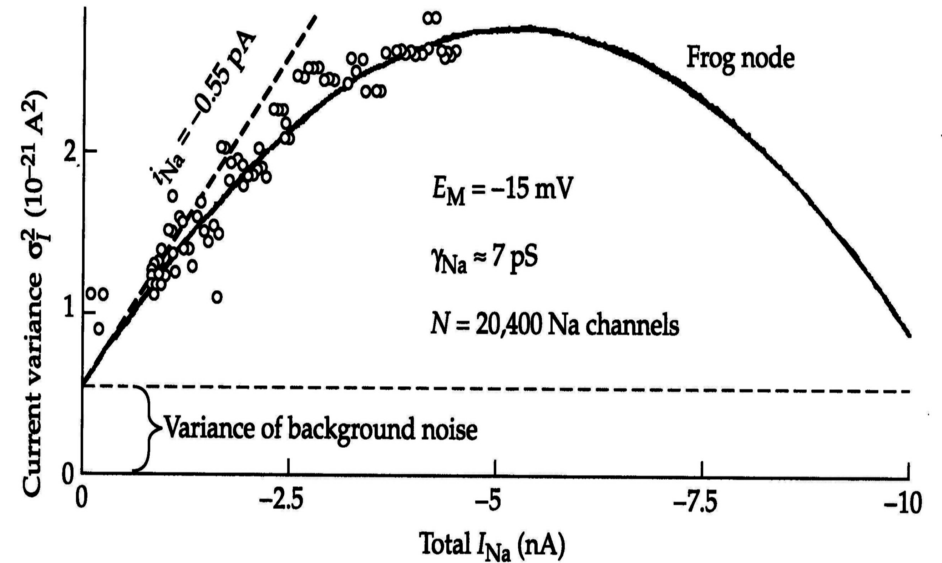
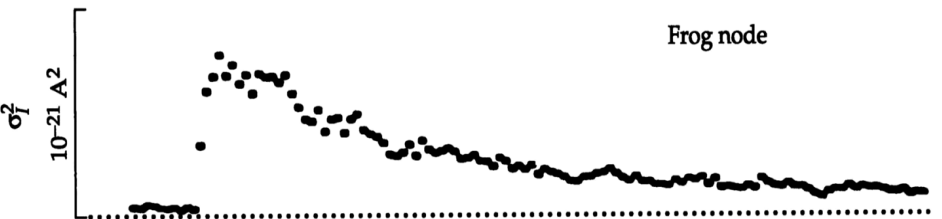
# Using the noise: Noise analysis



(B) DEVIATIONS FROM MEAN



(C) VARIANCE



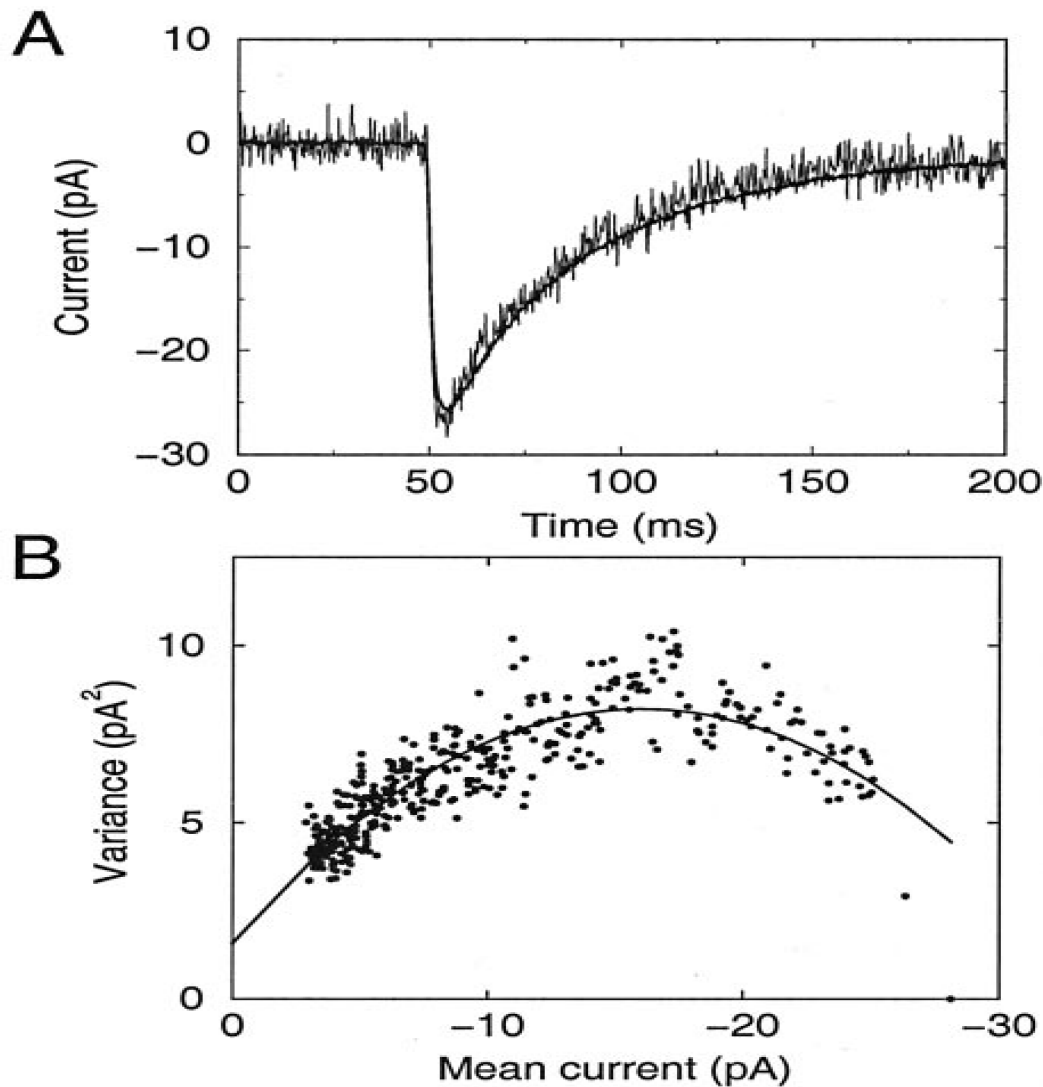
Binomial distribution:  $n$  out of  $N$  are open

$$\langle I(t) \rangle = i_0 N p(t)$$

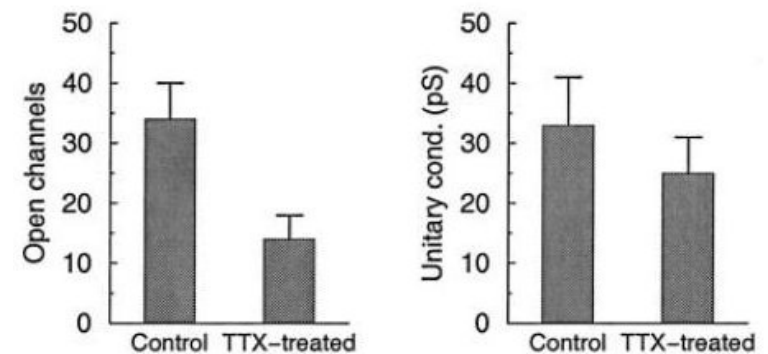
$$\langle \delta I^2(t) \rangle = i_0^2 N p(t) [1 - p(t)]$$

$$\langle \delta I^2(t) \rangle = i_0 \langle I(t) \rangle - \langle I(t) \rangle^2 / N$$

# Using the noise: Noise analysis



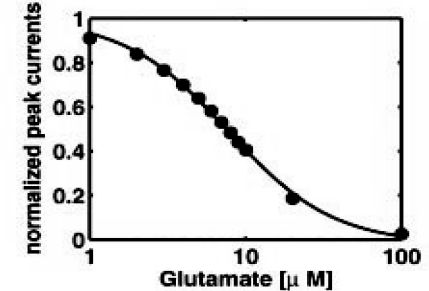
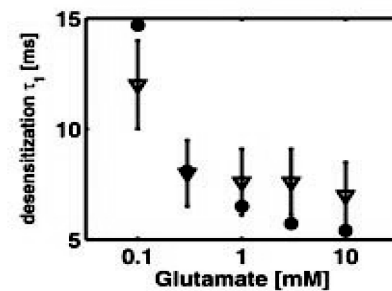
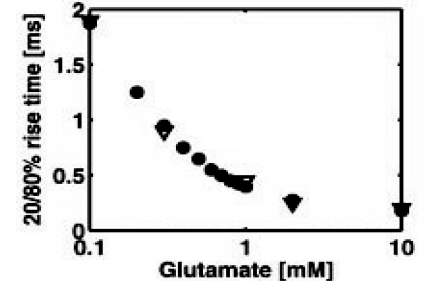
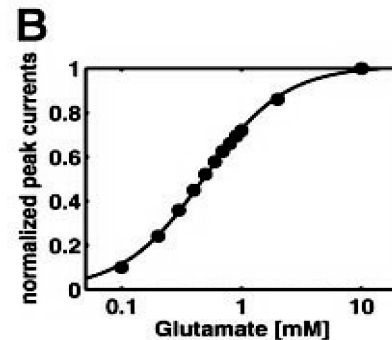
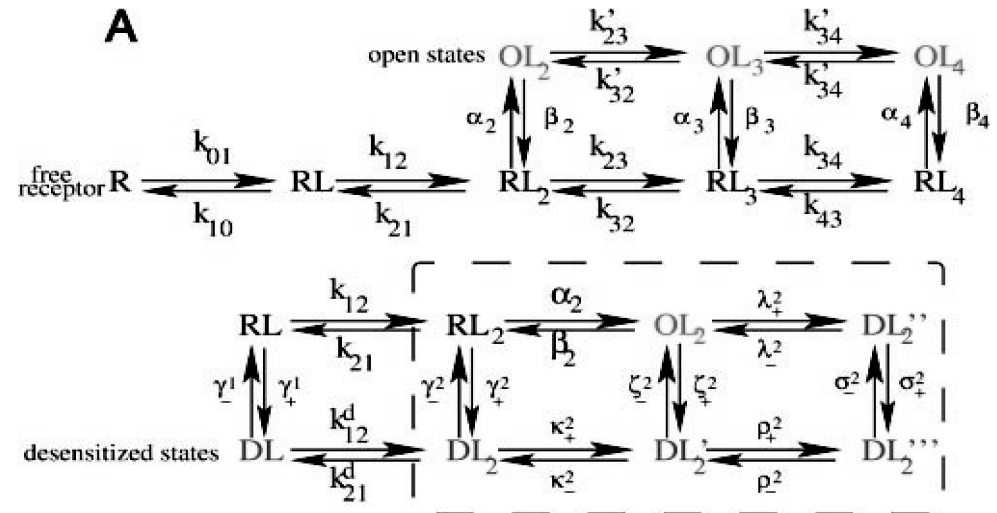
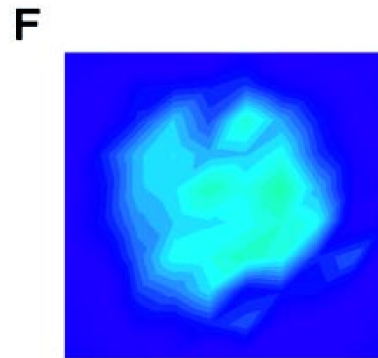
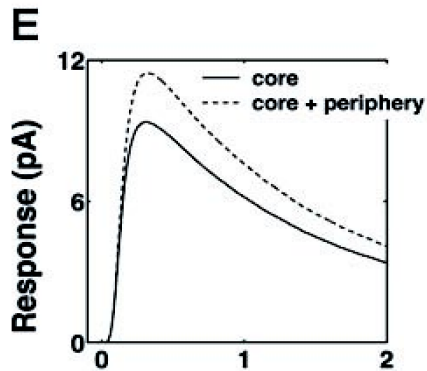
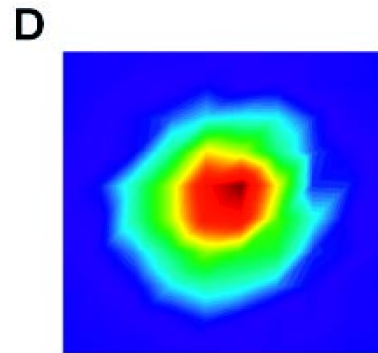
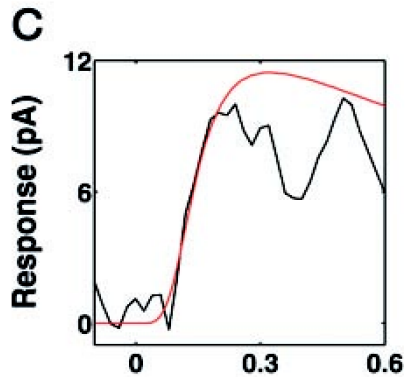
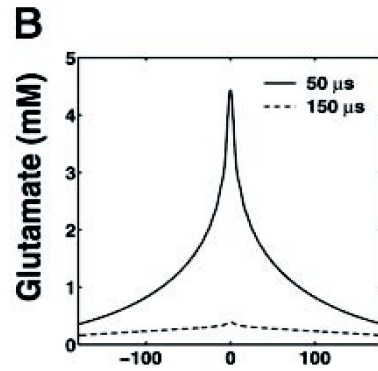
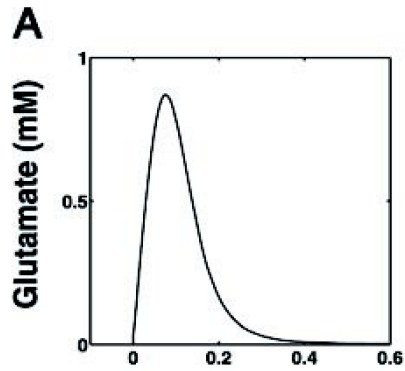
Noise analysis in practice:  
baseline subtraction,  
onset alignment,  
background noise, ...



These synapses are close to  
the soma (inhibitory).

# Model with geometry

## [Raghavachari & Lisman '04]



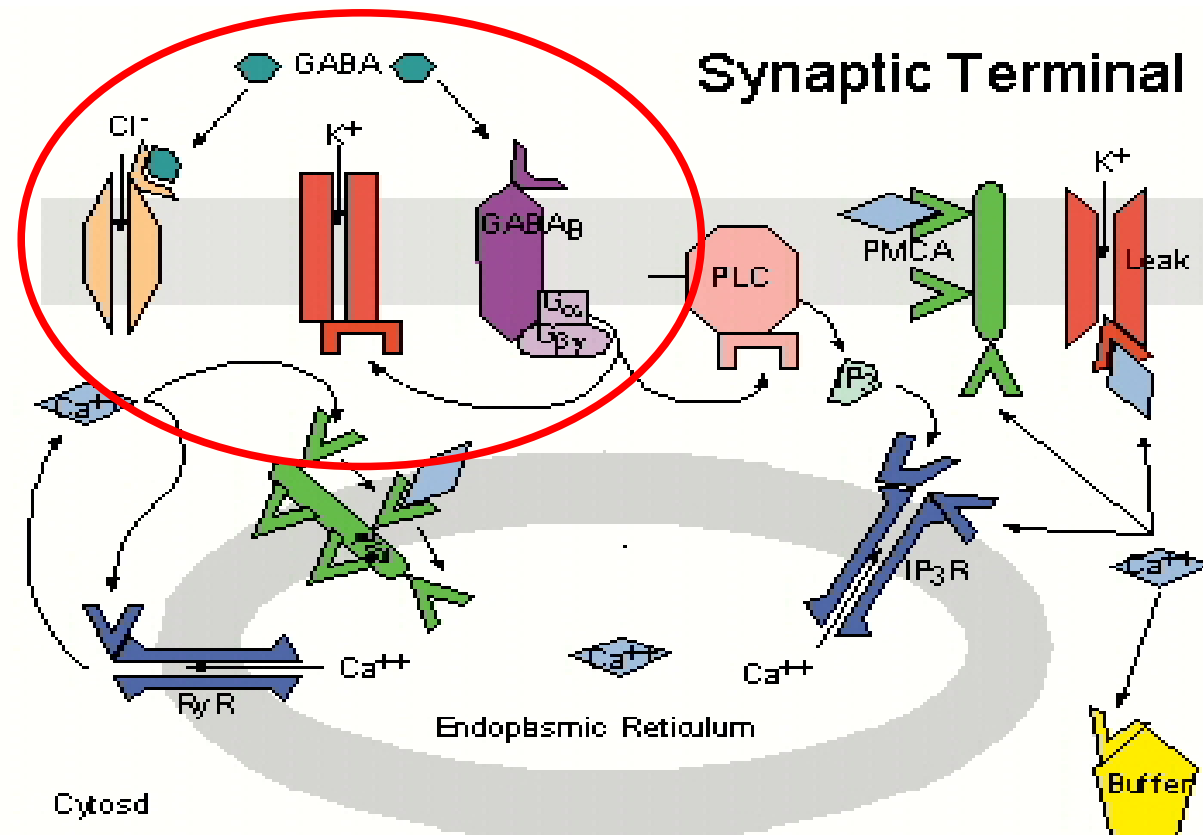
# PART 2: Post synaptic non-linearities

# Post-synaptic complications

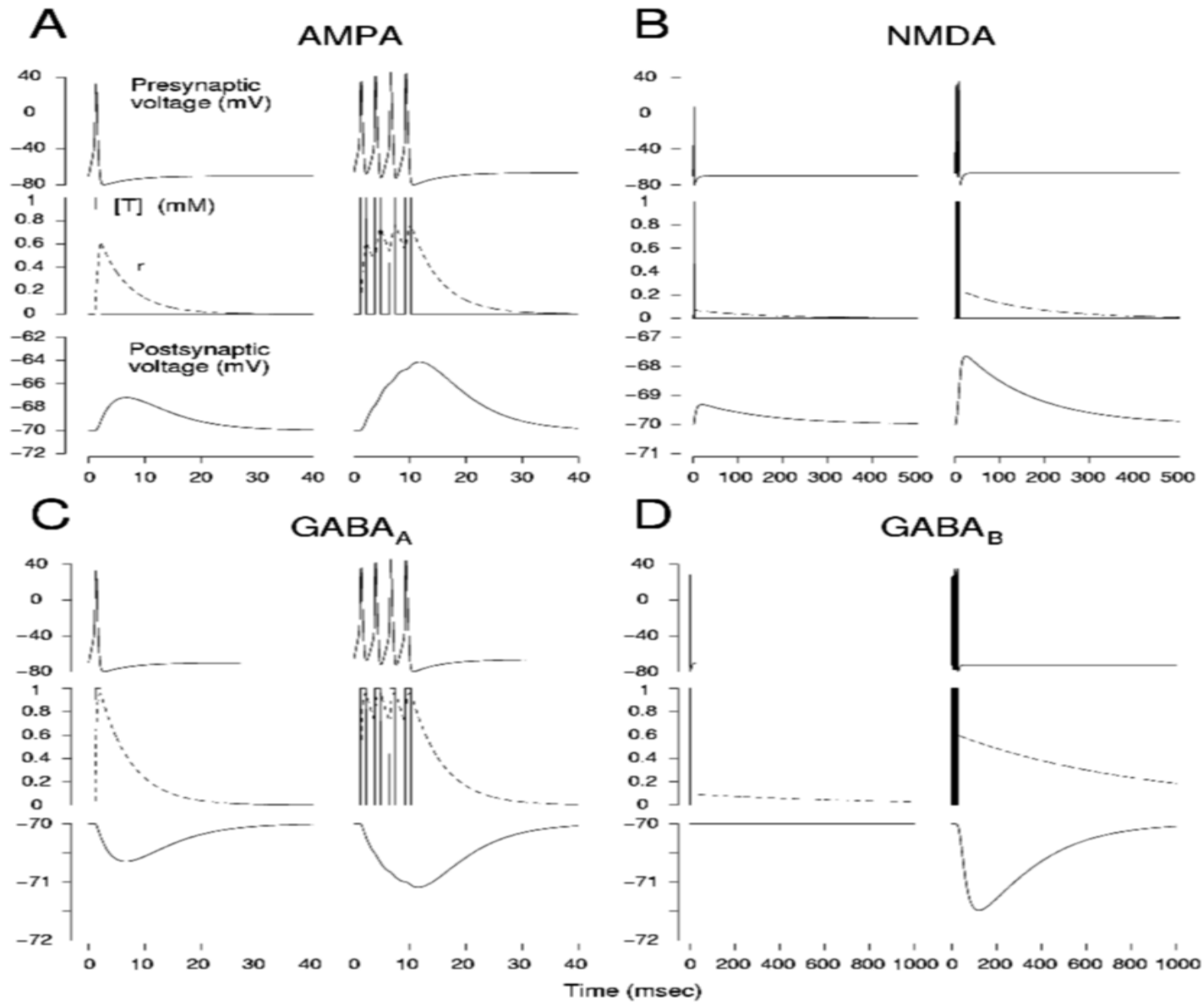
Second messenger synapses, channel opening is caused by activated membrane bound messenger.

Example in cortex: GABA<sub>B</sub>, mGluR

Retina: mGluR



# GABA<sub>b</sub>



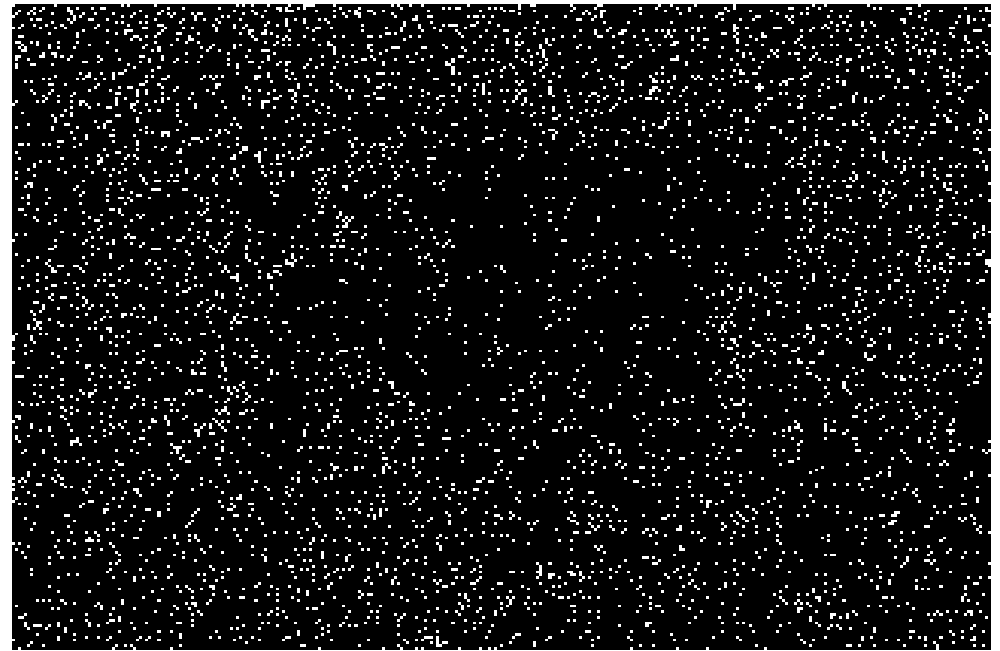
# Second messenger: noise removal at rod to rod-bipolar synapse

19

Day light



Scotopic : 1 photon / rod / 10 min



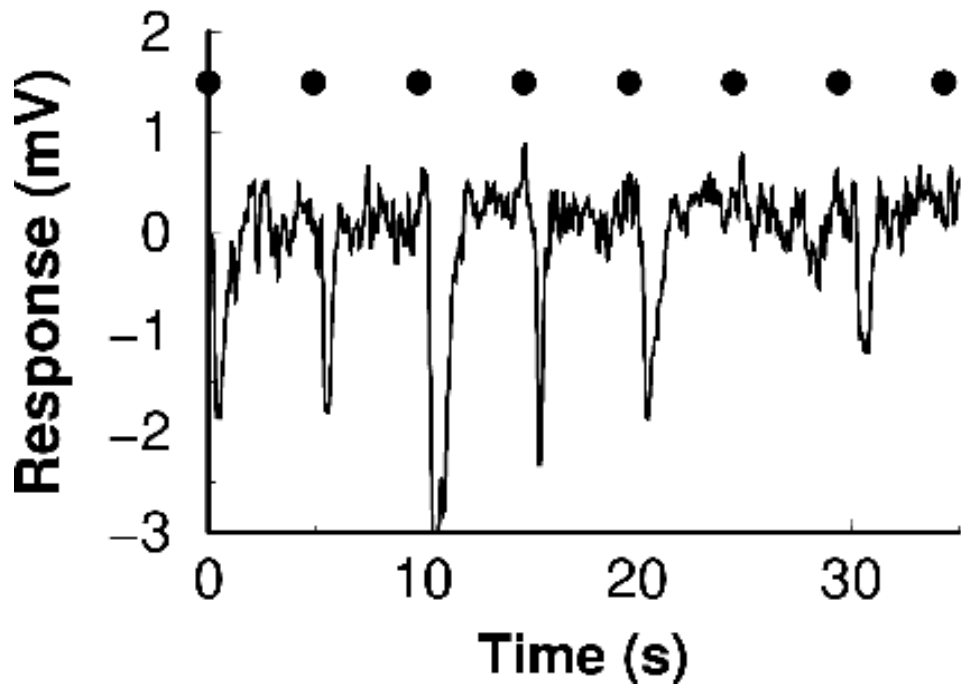
Retinal performance:

- Miss at most 75% of photons: *false negative* rate < 75 %
- *False positives*: less than dark light (0.01 /rod /sec)

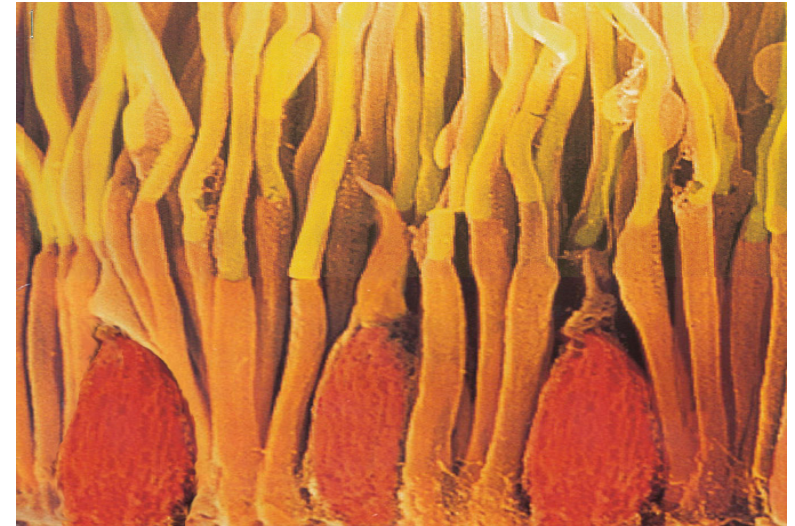
[van Rossum & Smith 1998, Clark & van Rossum 2006]

# Rod pathway circuitry

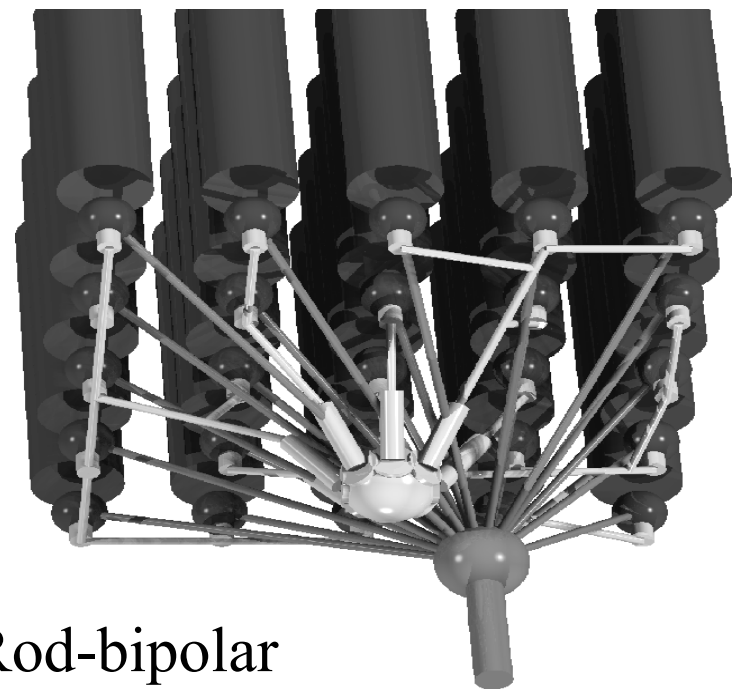
Rod response to dim flashes  
Schneeweis and Schnapf '95



Note: 20% intrinsic noise



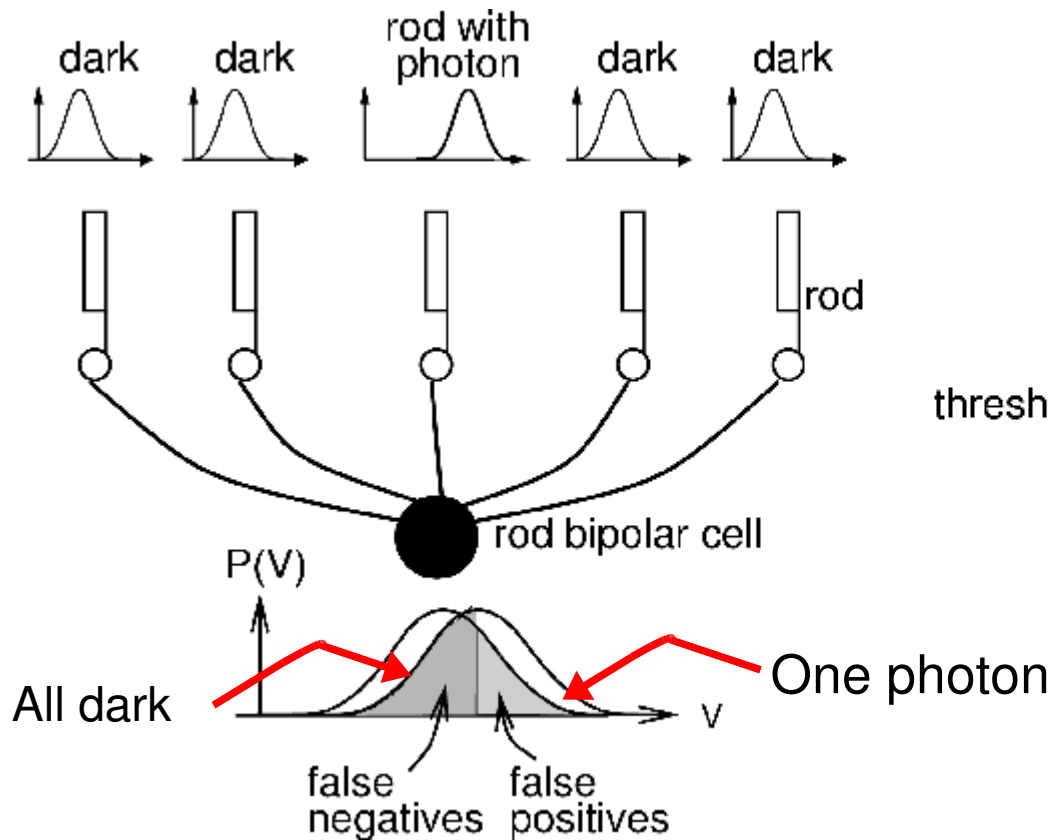
Rods



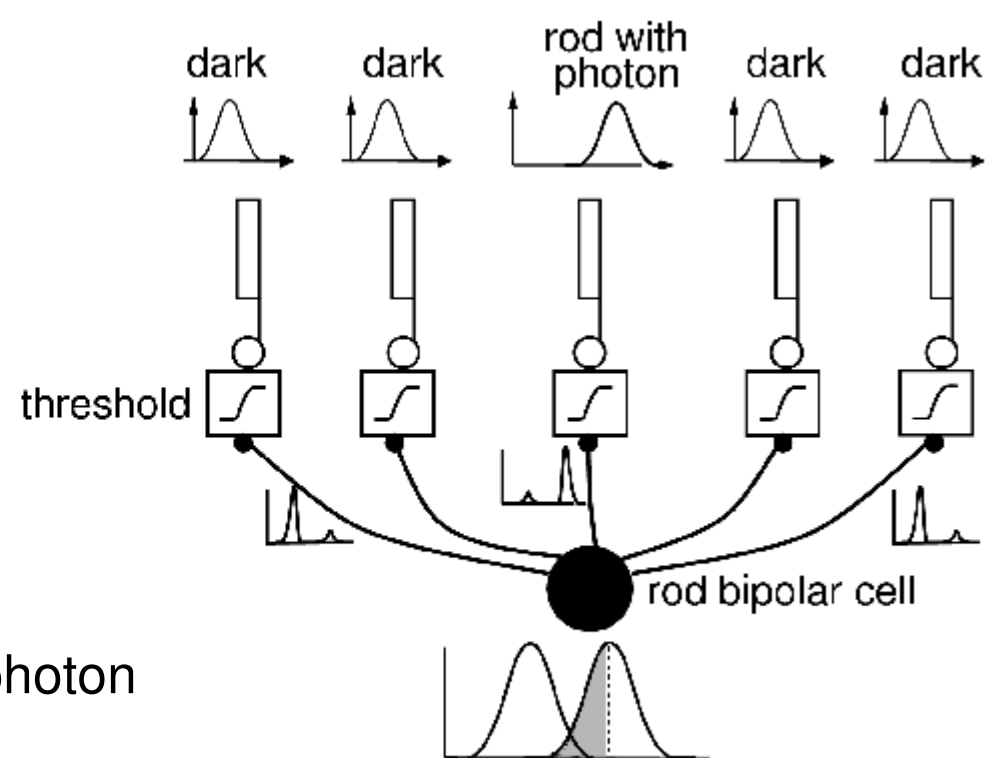
Rod-bipolar

# The need for a thresholding synapse

21



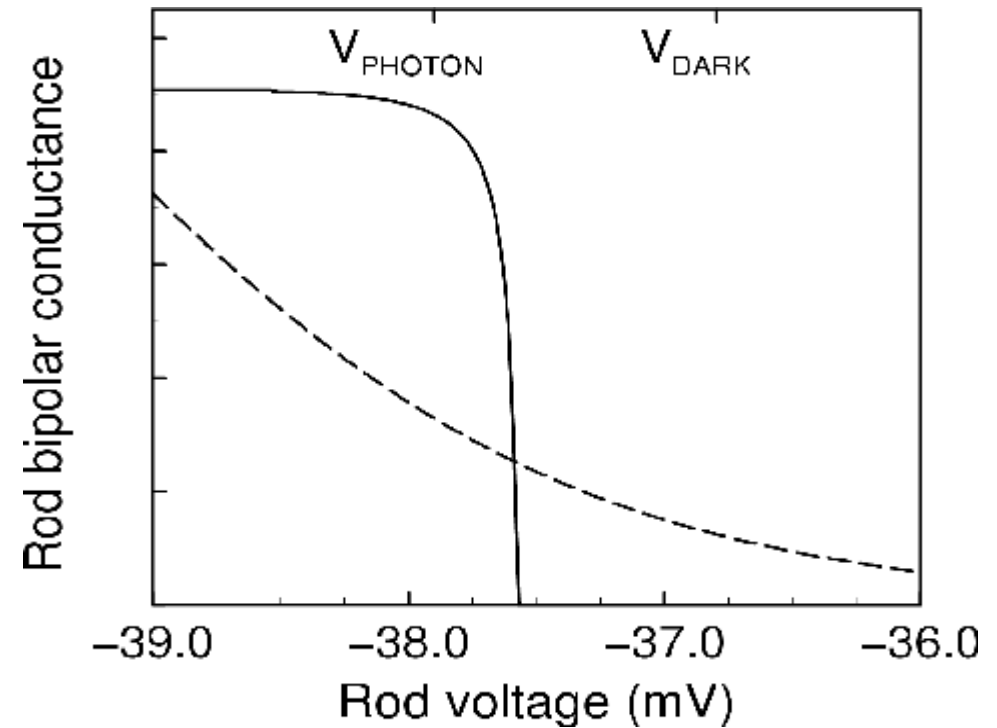
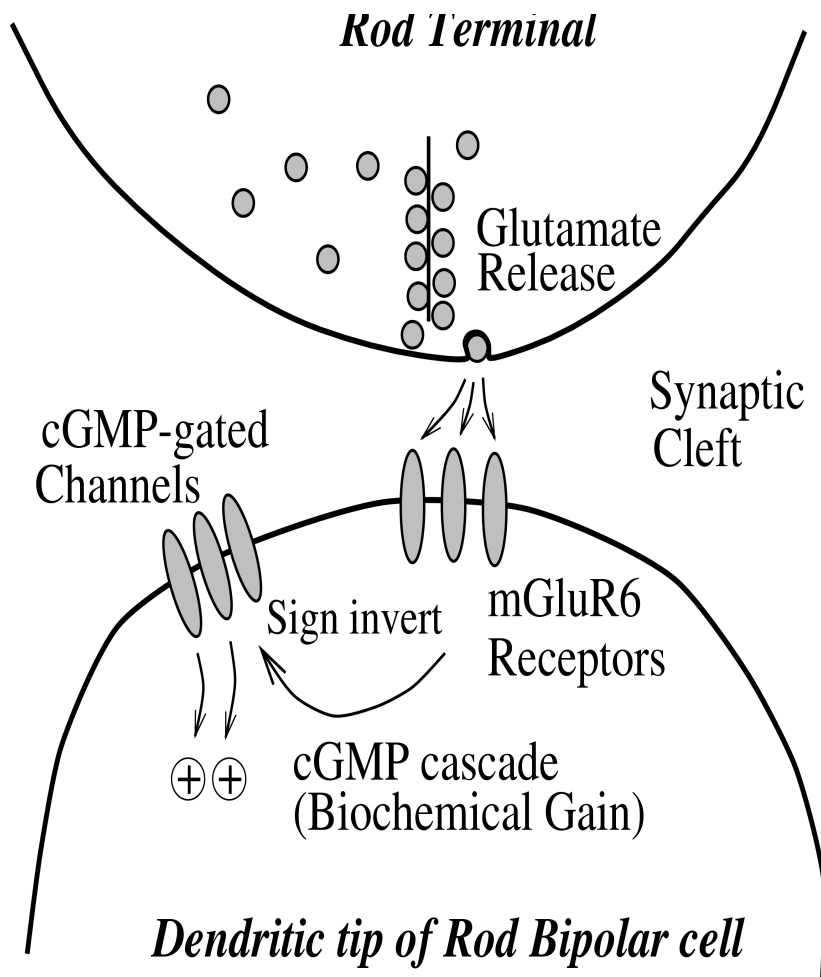
No threshold: many errors



Thresholding: keep errors small

- Simple pooling is bad [Baylor et al 1984].
- Threshold before summing
- Set threshold high (reject many events).

# Second messenger synapse



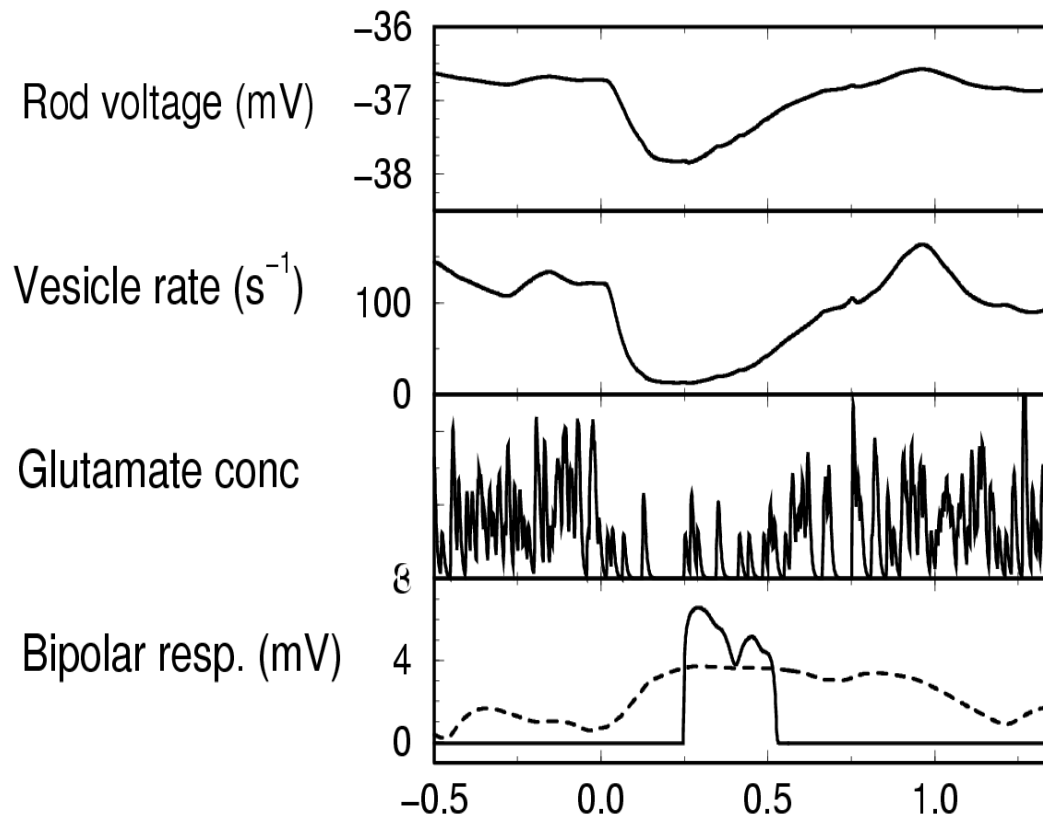
$$[cGMP] = 1 - c_{gain} [Glu]_{bound}$$

$$but [cGMP] > 0$$

$$g = \frac{[cGMP]}{k_2 + [cGMP]}$$

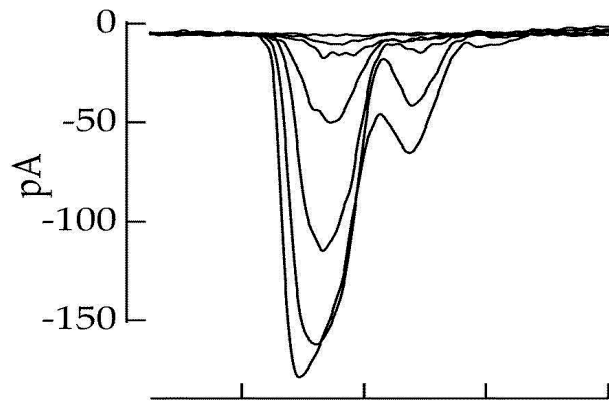
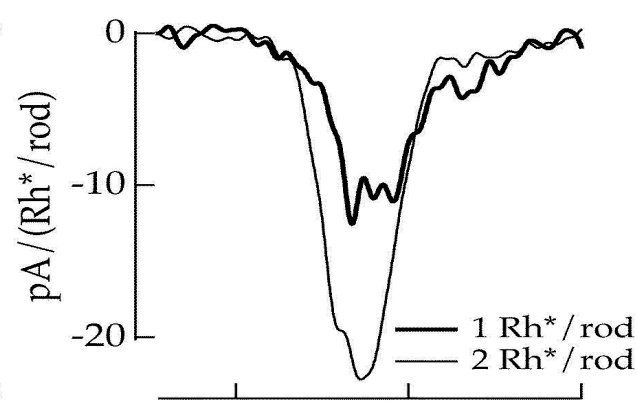
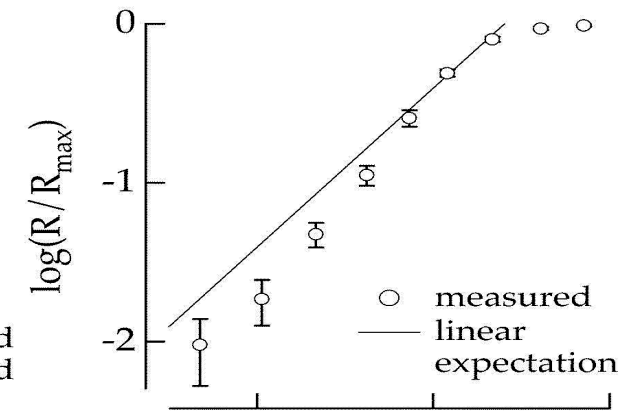
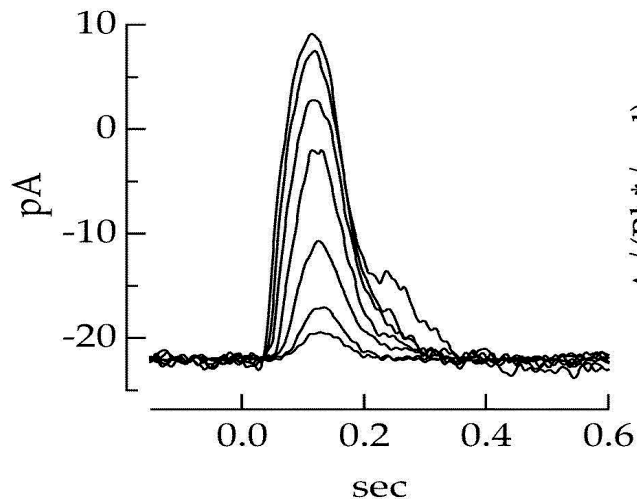
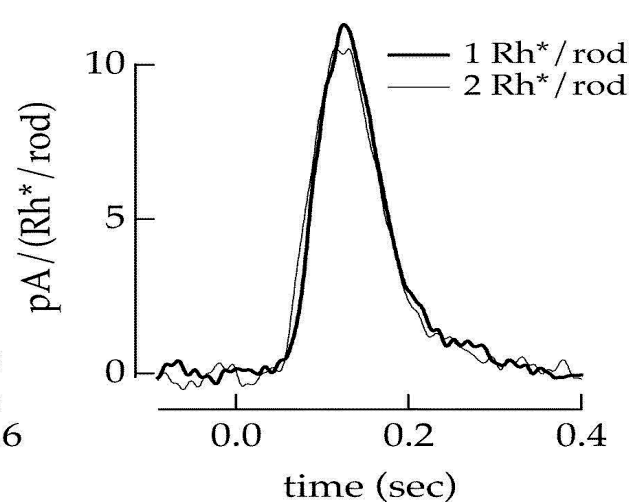
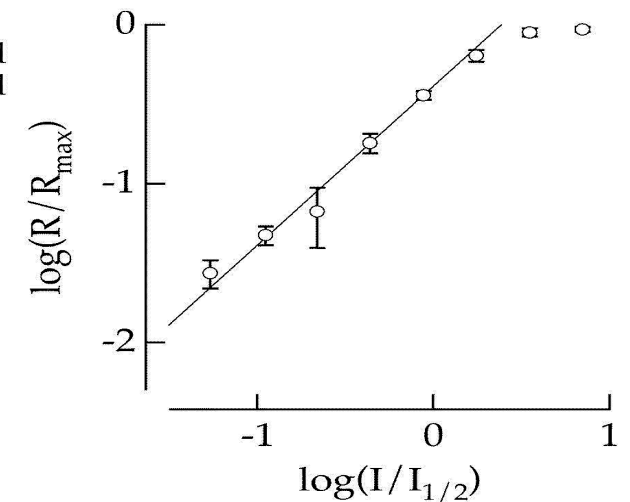
# Second messenger synapse

23



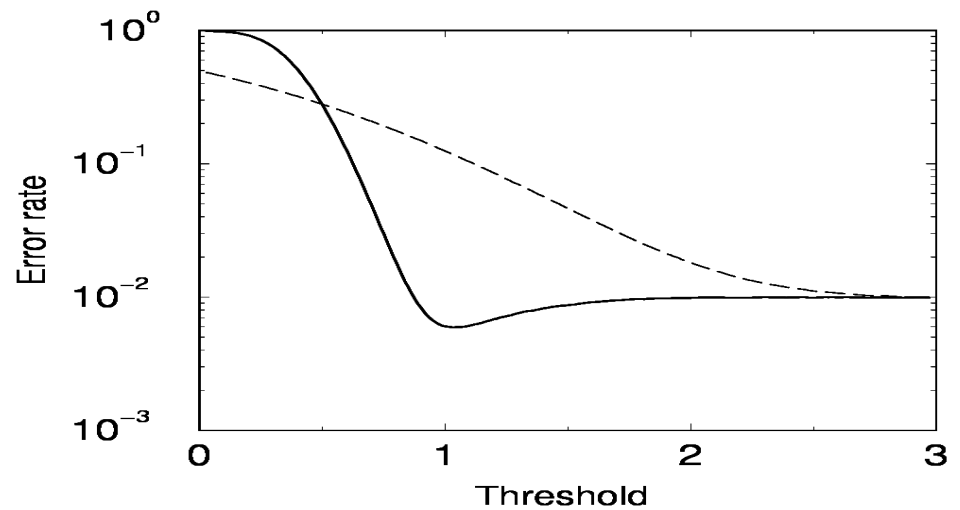
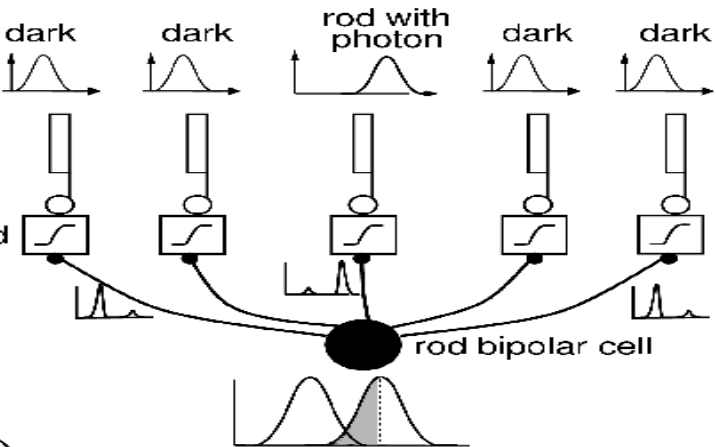
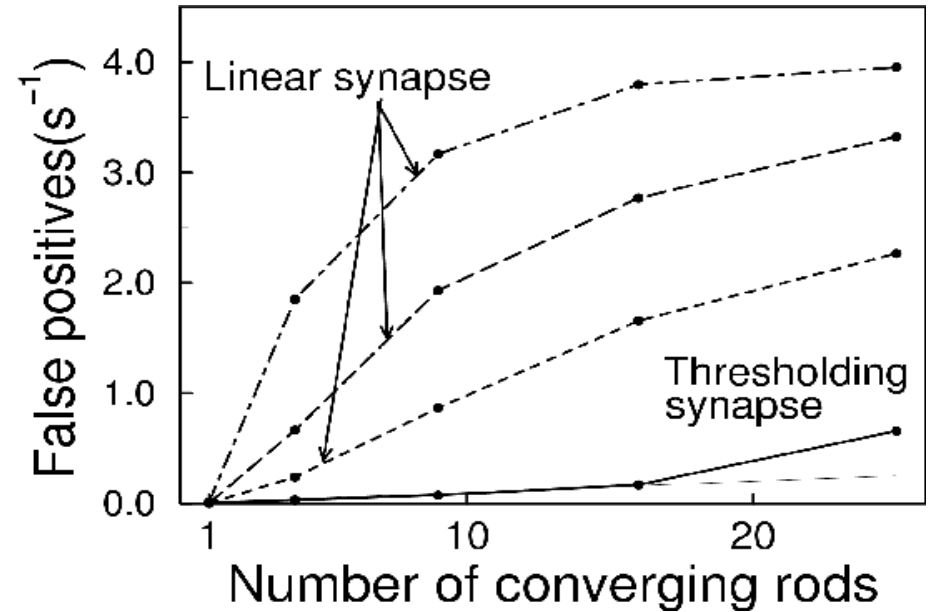
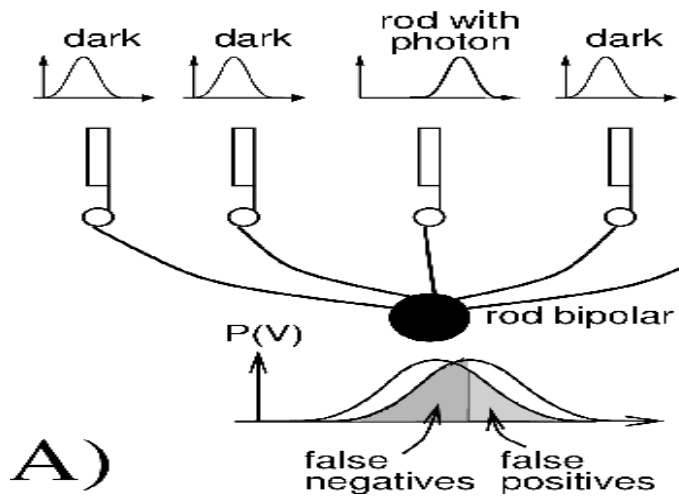
- Requires high vesicle release rate. About 80 ves/sec for Poisson release.
- Experiments confirm model [Sampath & Rieke, '04]
- Only 0-2 channels open in the dark, 30 open for single photon response.

# Experimental evidence

**A** Rod bipolar**B****C****D** OFF bipolar**E****F**

Data from Field & Rieke 2002.

# Performance with thresholding synapse



# Simulation of visual perception

26



Different synapses, different perception  
Low level modeling with high level consequences

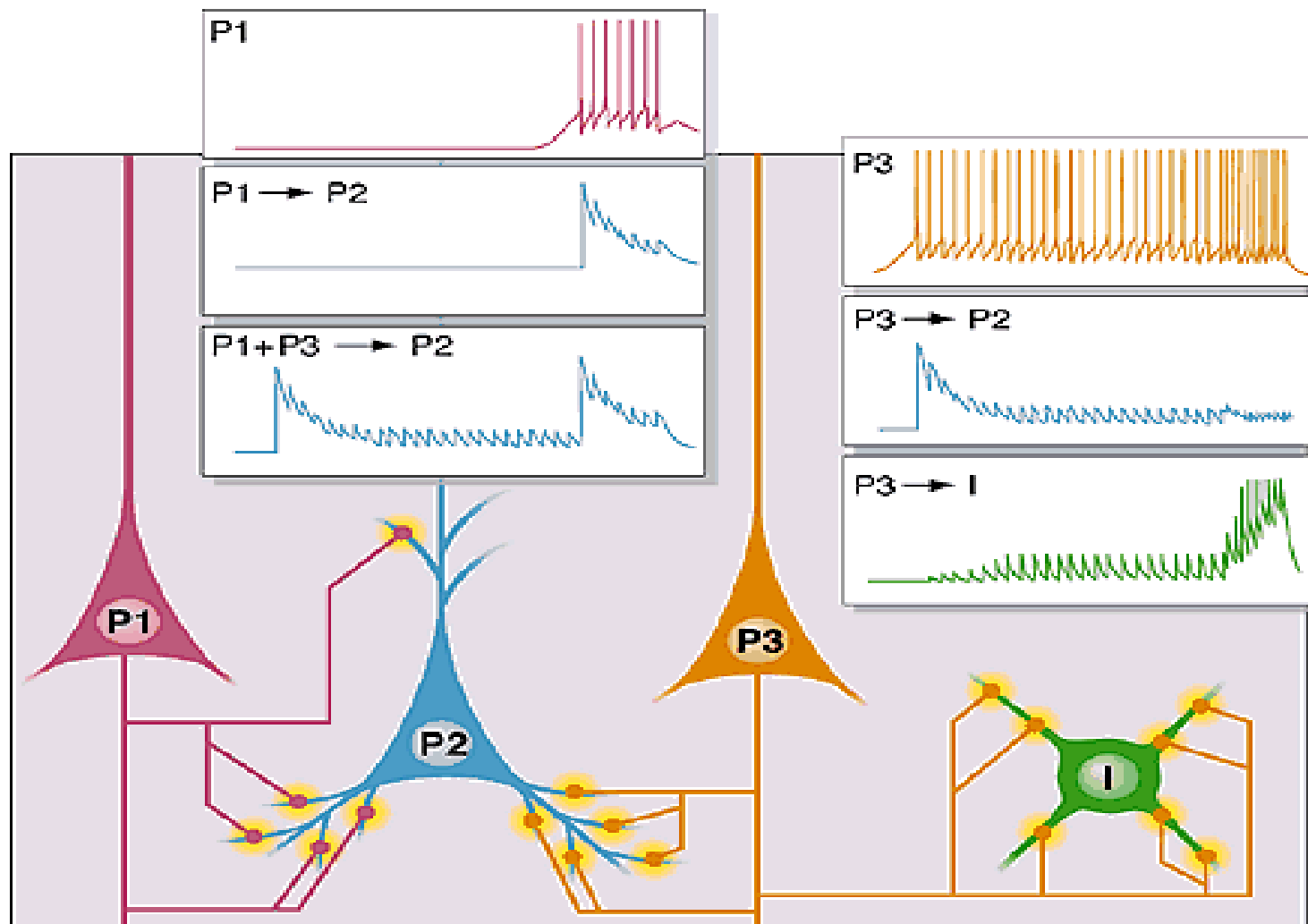
Other applications ?

Detection of single, binary signal amidst many, noisy inputs

# PART 3: Synaptic depression and network dynamics

# Short term synaptic depression

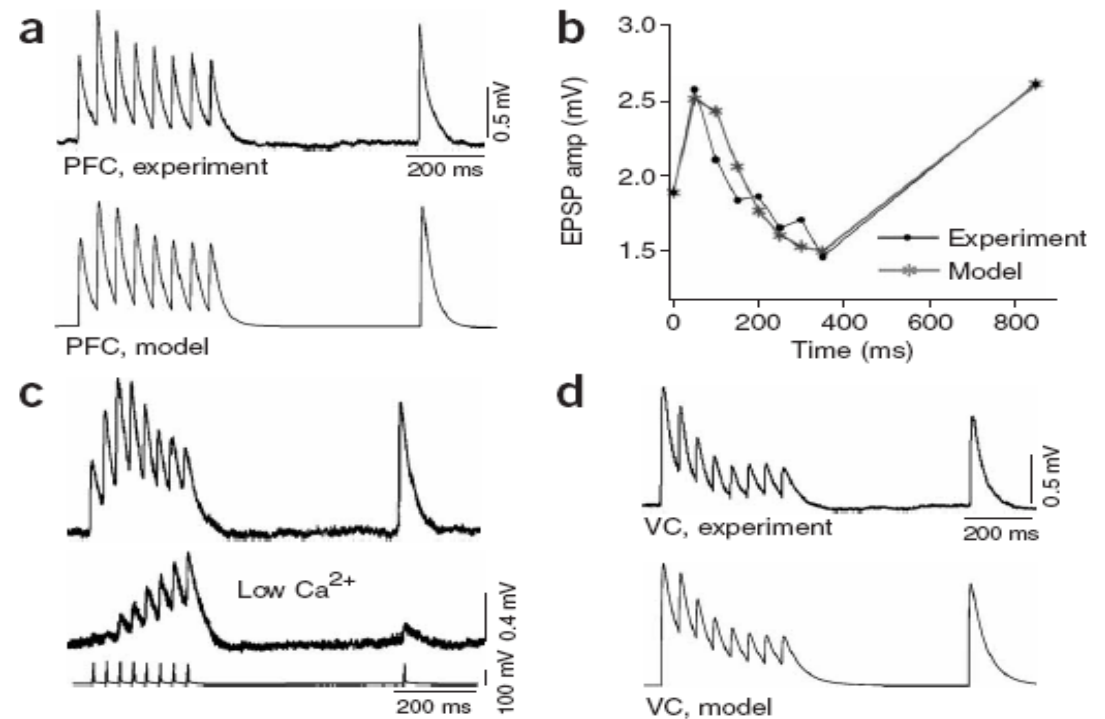
28



# Short term synaptic depression



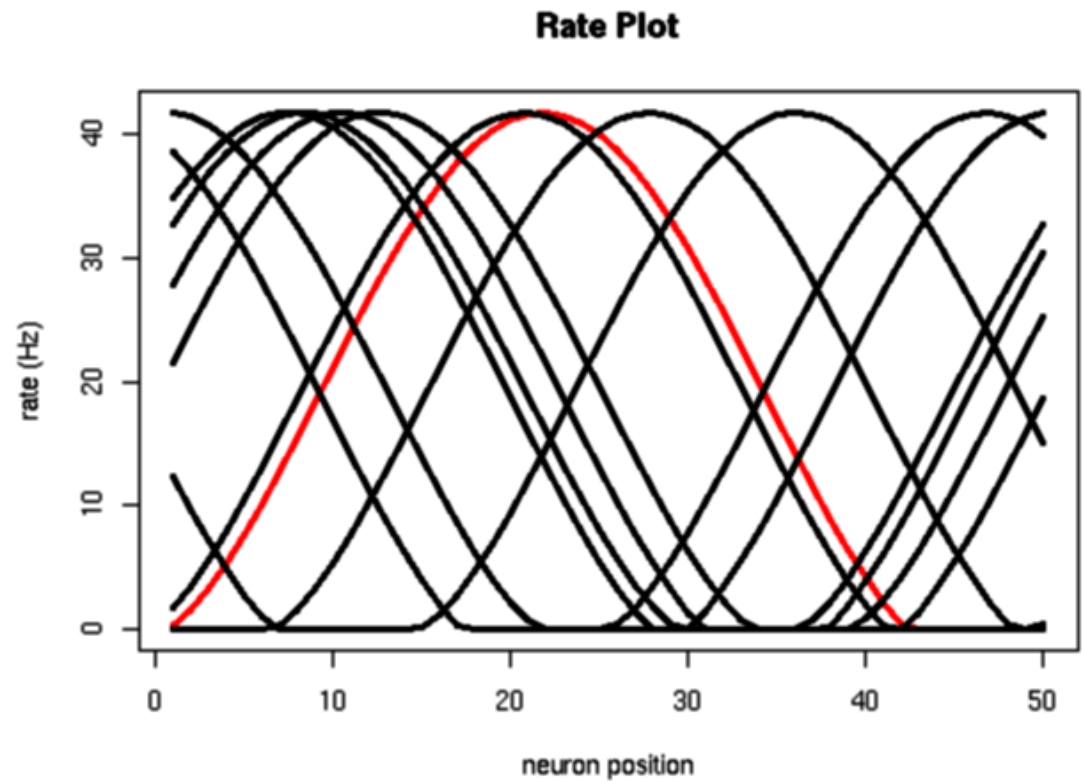
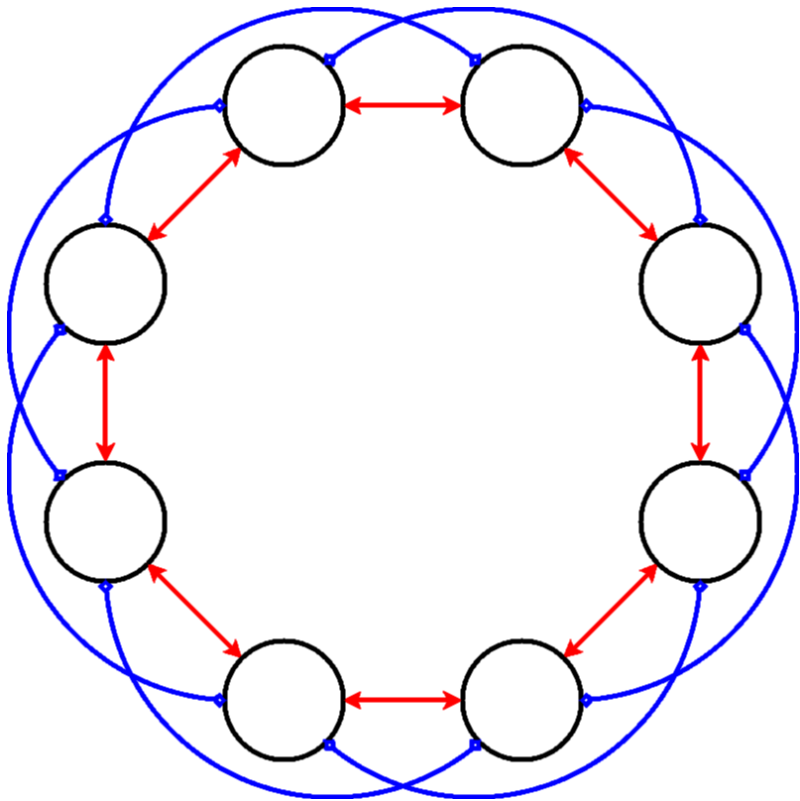
[Markram et al '98]



[Wang et al 2007]

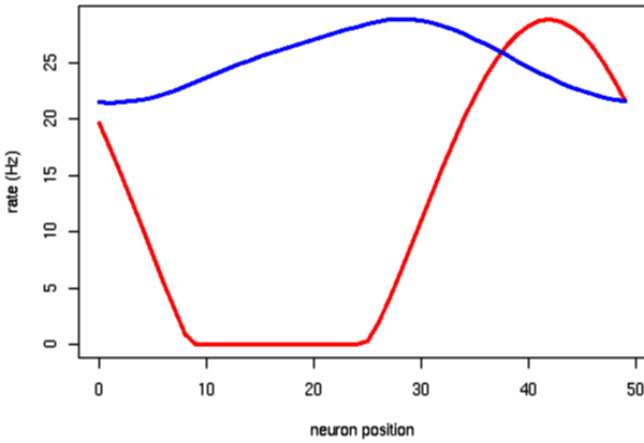
# Ring attractor network

- Ring network with centre-surround profile
- With strong enough recurrence: line-attractors

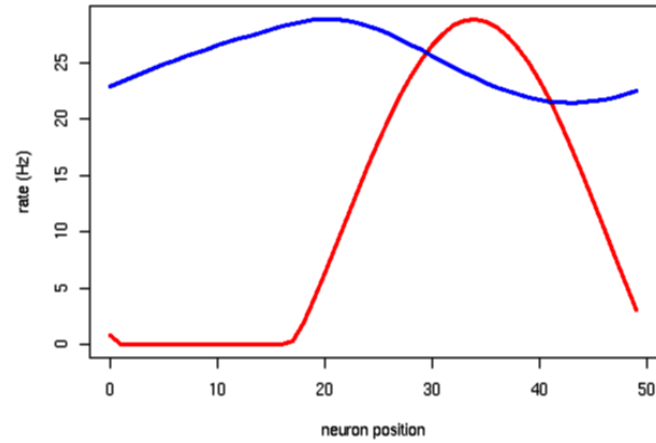


# Ring attractor network

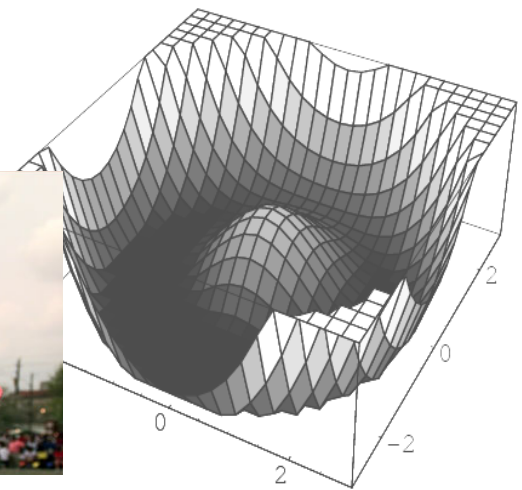
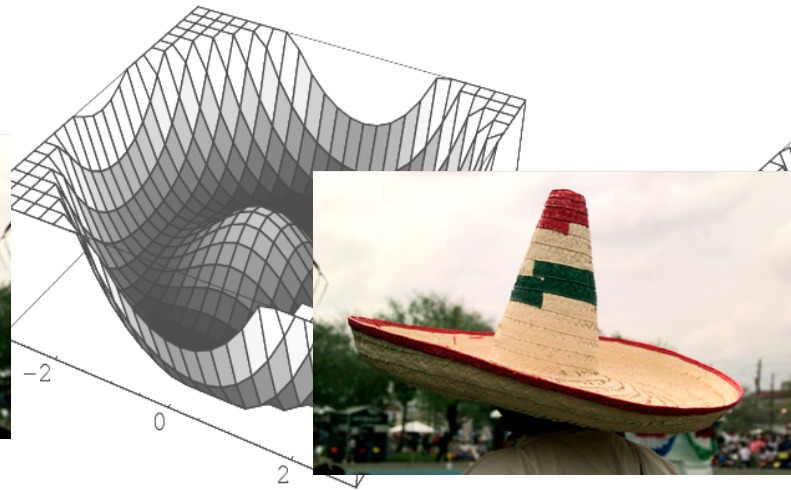
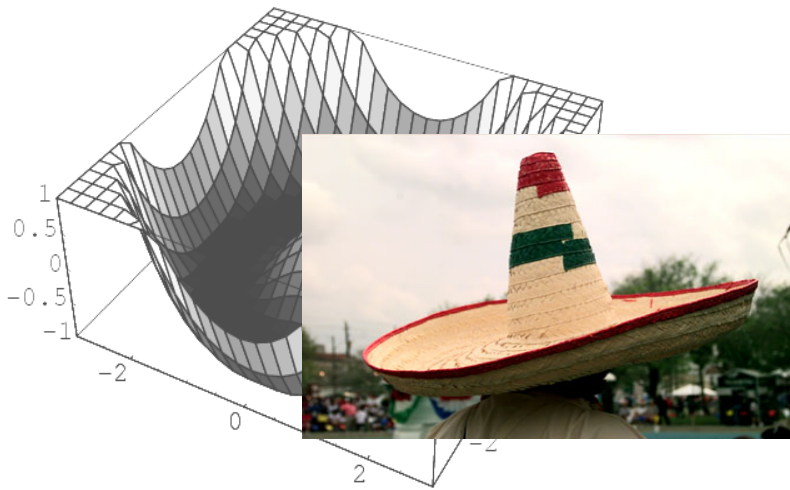
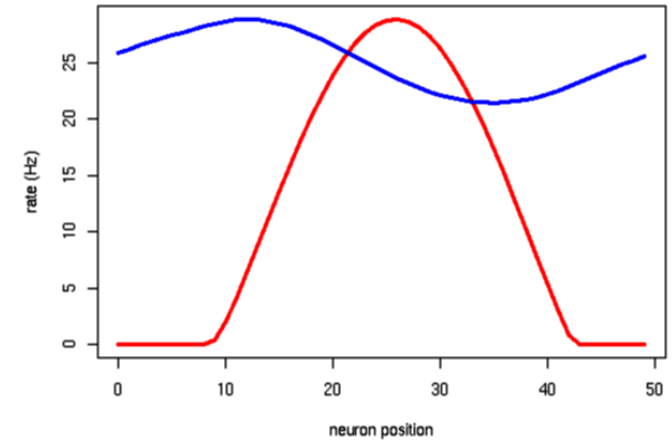
Rate Plot



Rate Plot



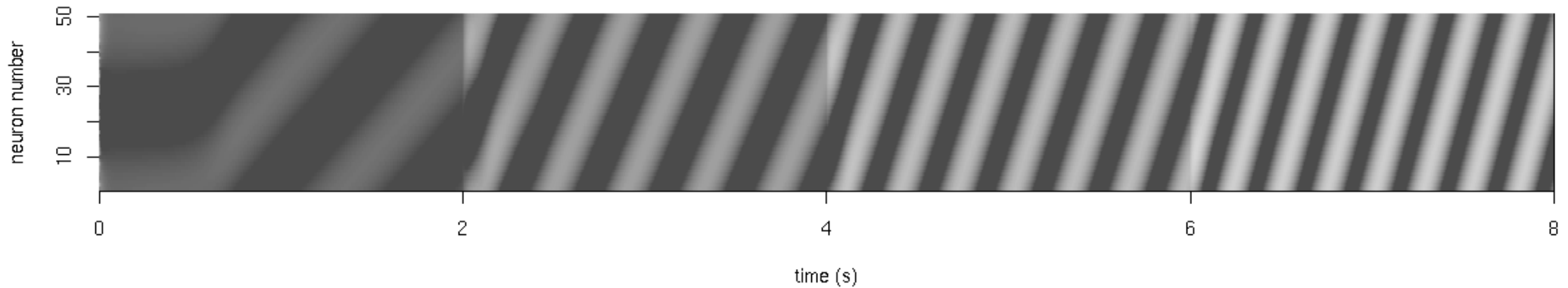
Rate Plot



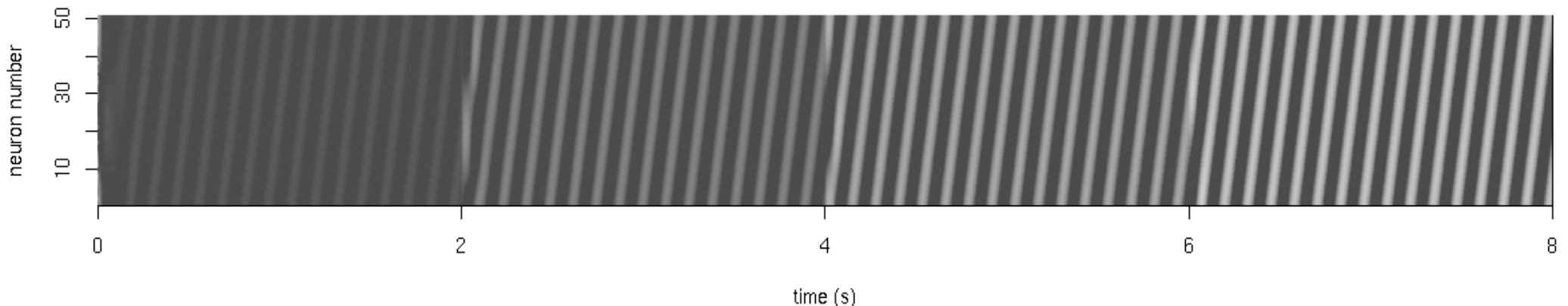
# Destabilization, rotating attractors

- With depressing synapses, rotating solution
- The rotation speed depends on uniform background current
- With spike rate adaptation, speed is fixed[Hansel and Sompolinsky]

Rate Image

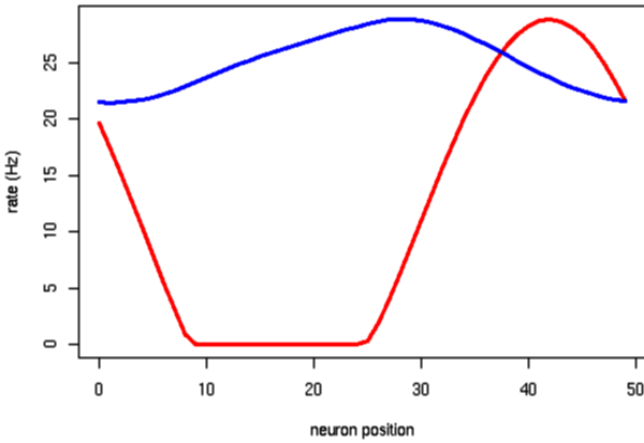


Rate Image

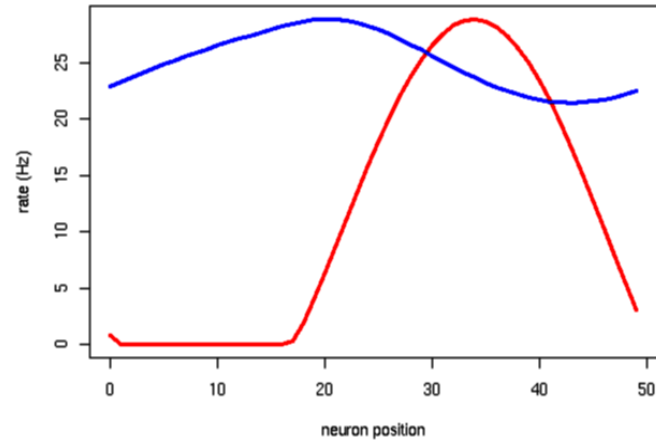


# Ring attractor network

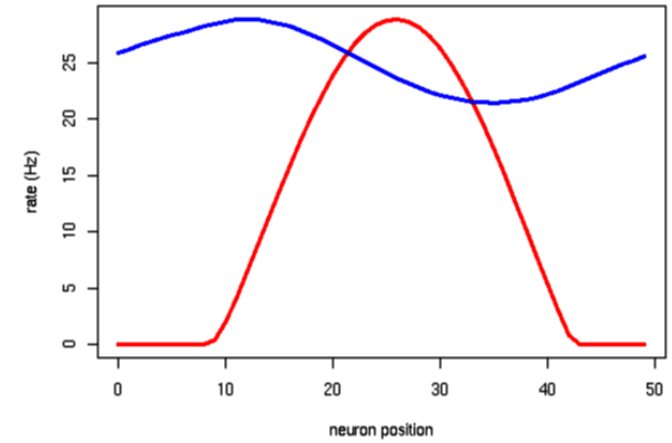
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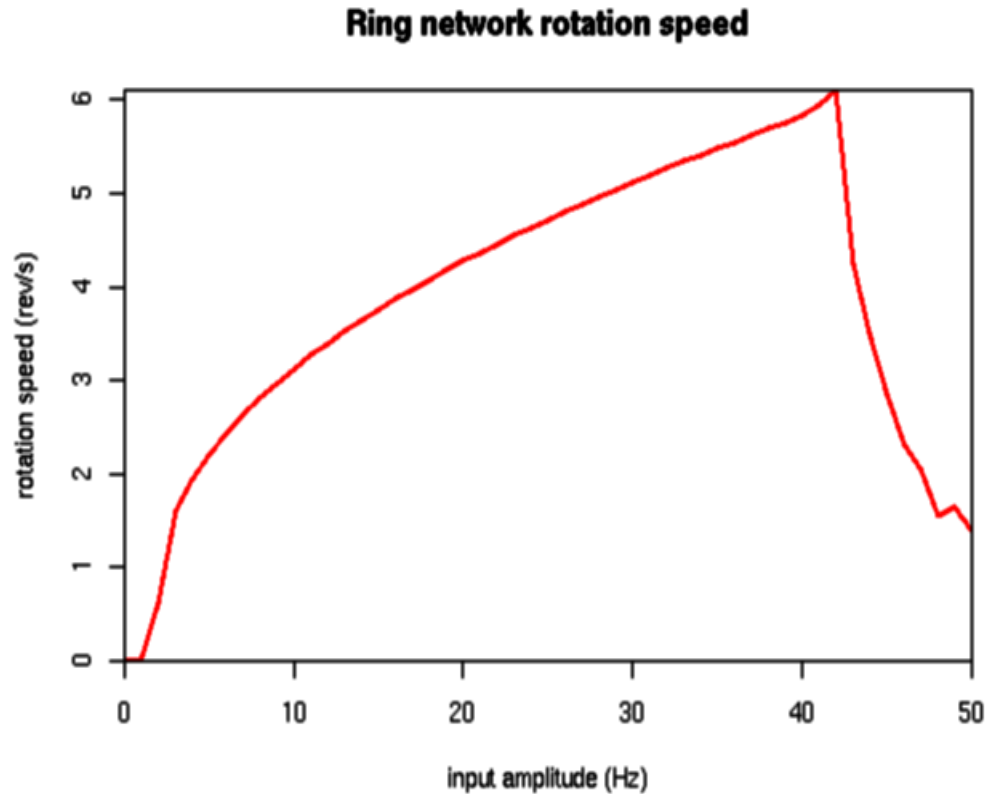
Rate Plot



Rate Plot



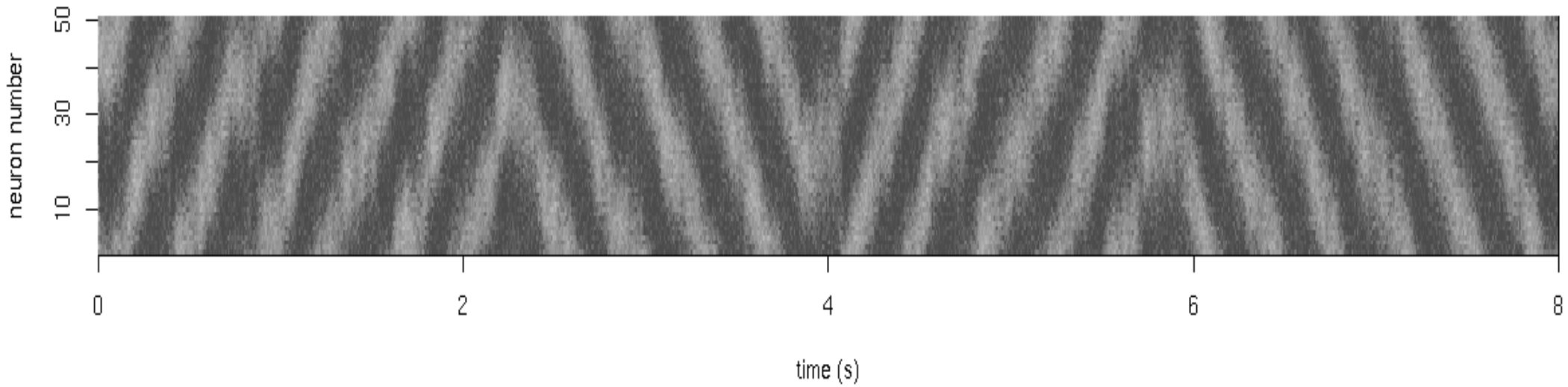
# Rotation speed



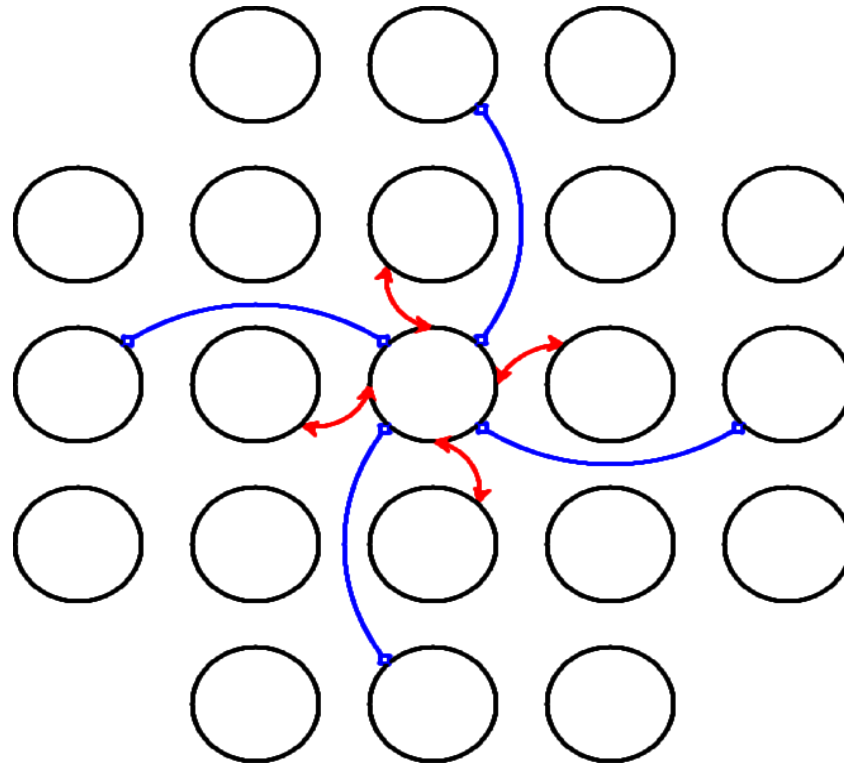
Possible application: pattern generation with variable frequency.

# Alternative solutions in the presence of noise

Rate Image



# 2 dimensional field



Movies...

# Conclusions

- Synapses are not only a way to do connect neurons
- They can have rich behavior: Non-linear, noisy, dynamical
- And this is reflected on a systems level