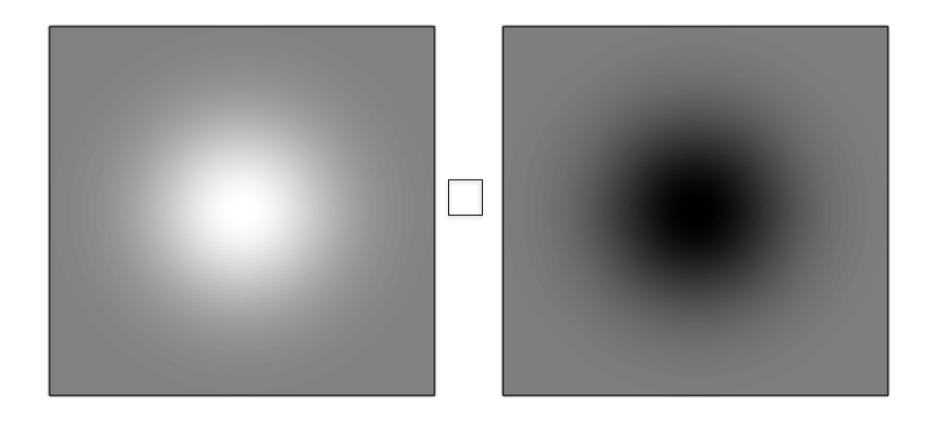
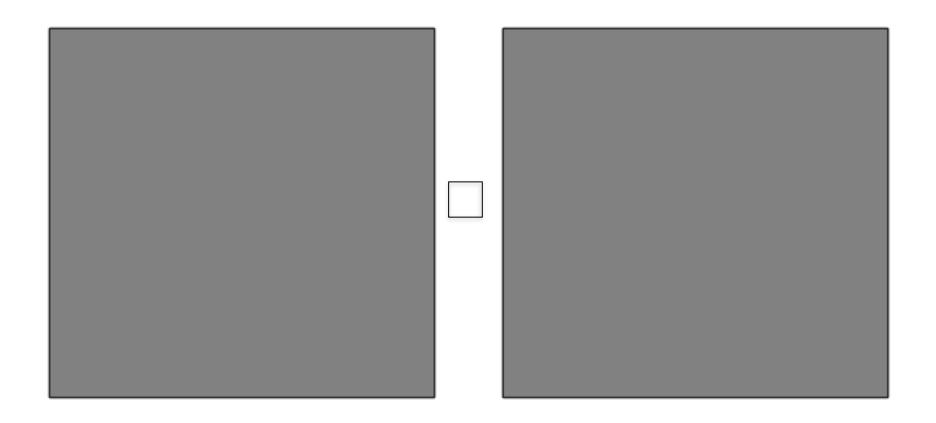
#### NEPR208 Synapses, plasticity and learning

- Synaptic depression and adaptation
- Synaptic facilitation and working memory
- Spike time dependent plasticity and long term memory
- Heterosynaptic long term depression and network stability
- Anti-hebbian plasticity and decorrelation

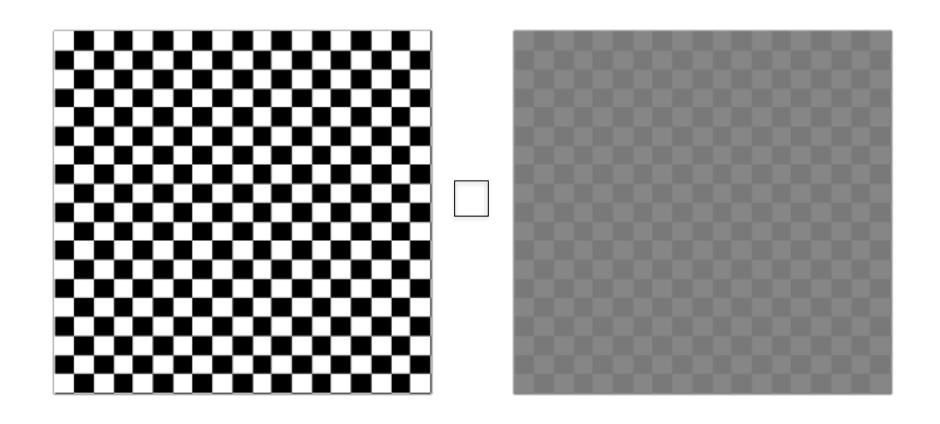
# The visual system adapts to the mean luminance



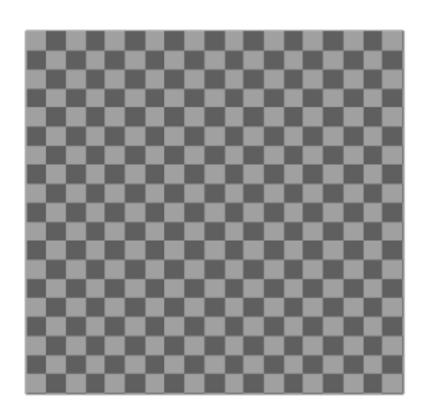
# The visual system adapts to the mean luminance

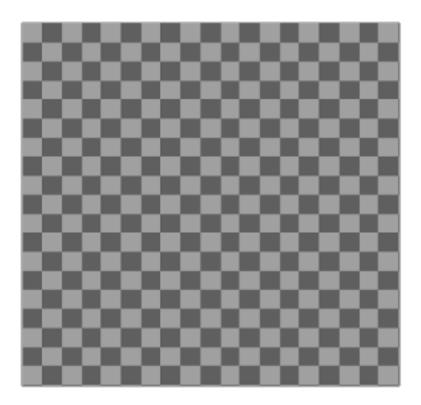


# The visual system adapts to contrast

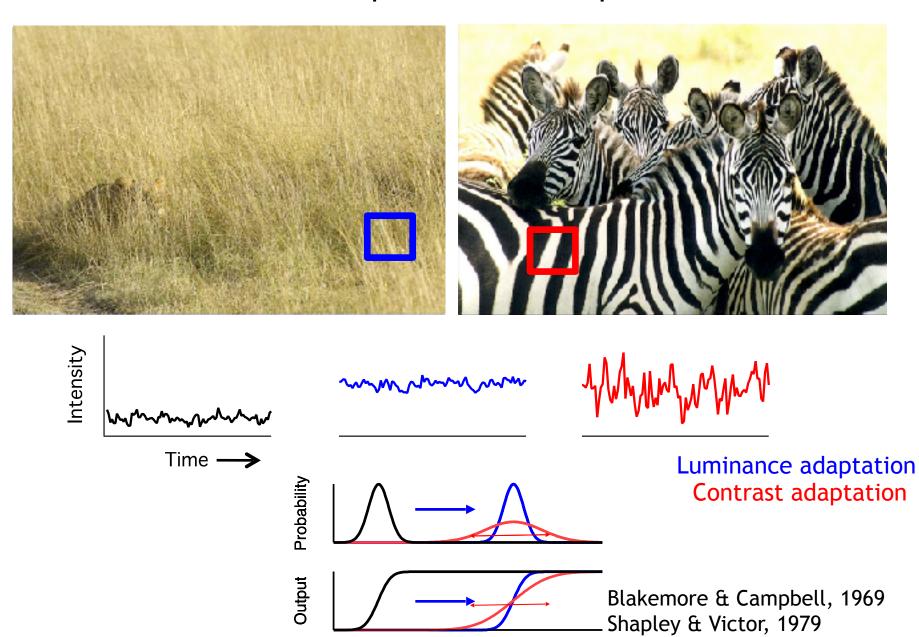


# The visual system adapts to contrast





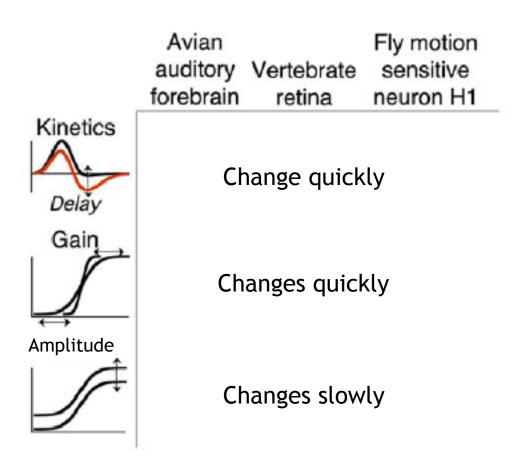
## Functional importance of adaptation



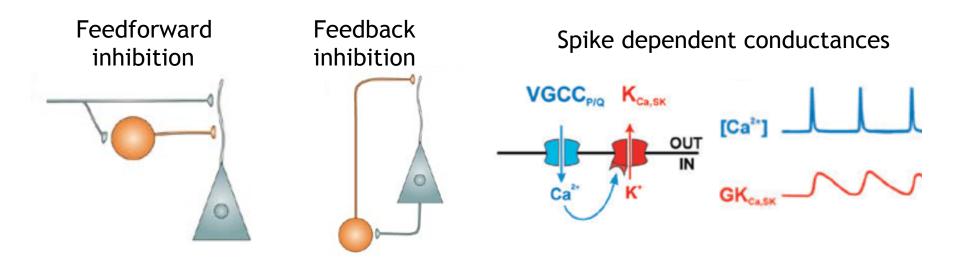
Input (Light intensity)

Smirnakis et al., 1997

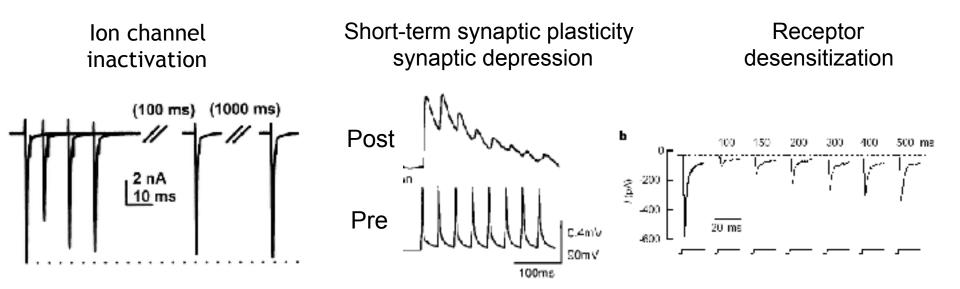
## Common properties of contrast adaptation



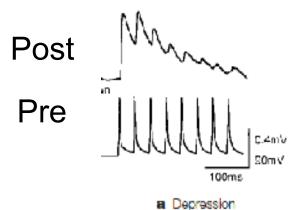
## Change in sensitivity by modulation



## Change in sensitivity by depletion



## Short-term synaptic plasticity – synaptic depression



n: Number of vesicle

p: Probability of vesicle release

Release = 
$$n \times p$$



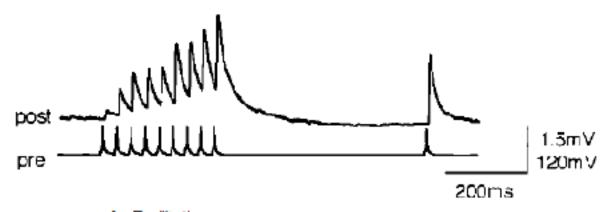
Depletion of available vesicles as a mechanism for depression

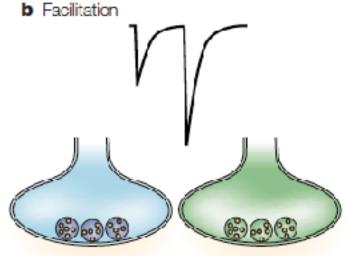
$$\frac{dn(t)}{dt} = \underbrace{\frac{1 - n(t)}{\tau_r}}_{\text{replenishment}} - \underbrace{\sum_{j} \delta(t - t_j) \cdot p \cdot n(t)}_{\text{release}}$$

Hennig, 2013. Theoretical models of synaptic short term plasticity

Chance FS, Nelson SB, Abbott LF. (1998) Ozuysal & Baccus (2012)

## Short-term synaptic plasticity – synaptic facilitation

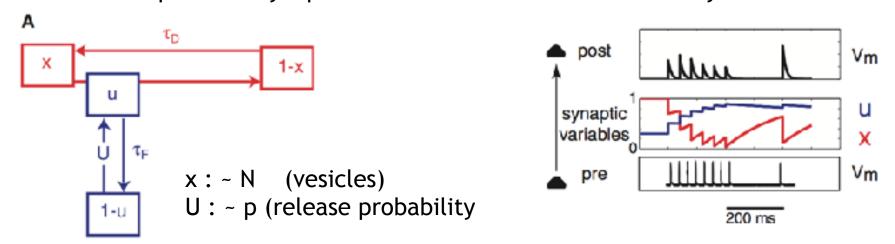




Residual calcium as a mechanism for increased release

$$\frac{dp(t)}{dt} = \frac{p_0 - p(t)}{\tau_f} + \sum_j \delta(t - t_j) \cdot a_f \cdot (1 - p(t))$$

#### Proposal for synaptic facilitation in short-term memory

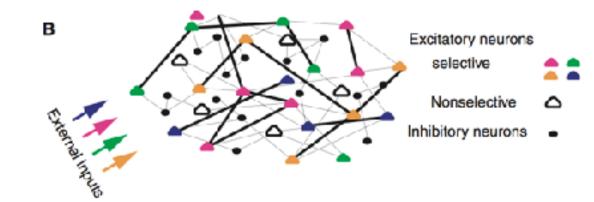


#### Release & Replenishment

$$\frac{dx}{dt} = \frac{1-x}{\tau_{\rm D}} - u x \delta(t-t_{\rm Sp})$$

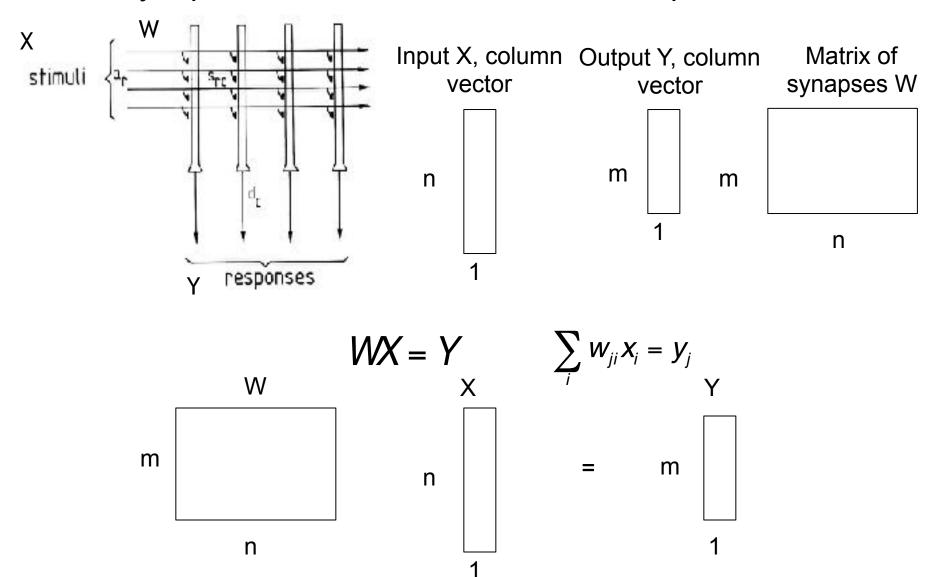
#### **Facilitation**

$$\frac{du}{dt} = \frac{U - u}{\tau_{\rm F}} + U (1 - u) \delta(t - t_{\rm Sp})$$

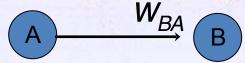


Synaptic theory of working memory Mongillo, Barak & Tsodyks, 2008

## Synaptic transmission as a matrix multiplication



## The Hebb rule for synaptic plasticity



When an axon of cell A is near enough to excite cell B or repeatedly or persistently takes part in firing it, some growth process or metabolic change takes place in one or both cells such that A's efficiency, as one of the cells firing B, is increased.

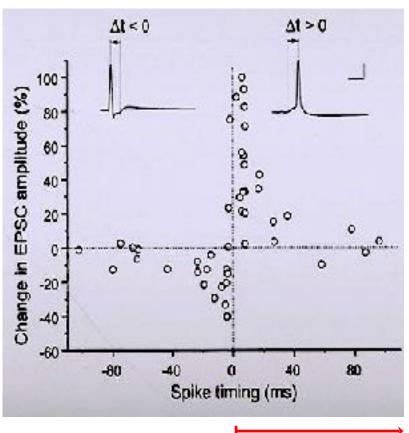
## Hebbian plasticity

A excites B: Coincidence leads to greater release

$$\frac{dW_{BA}}{dt} = \langle AB \rangle$$

## Hebbian spike timing plasticity (STDP)

#### hippocampal neurons



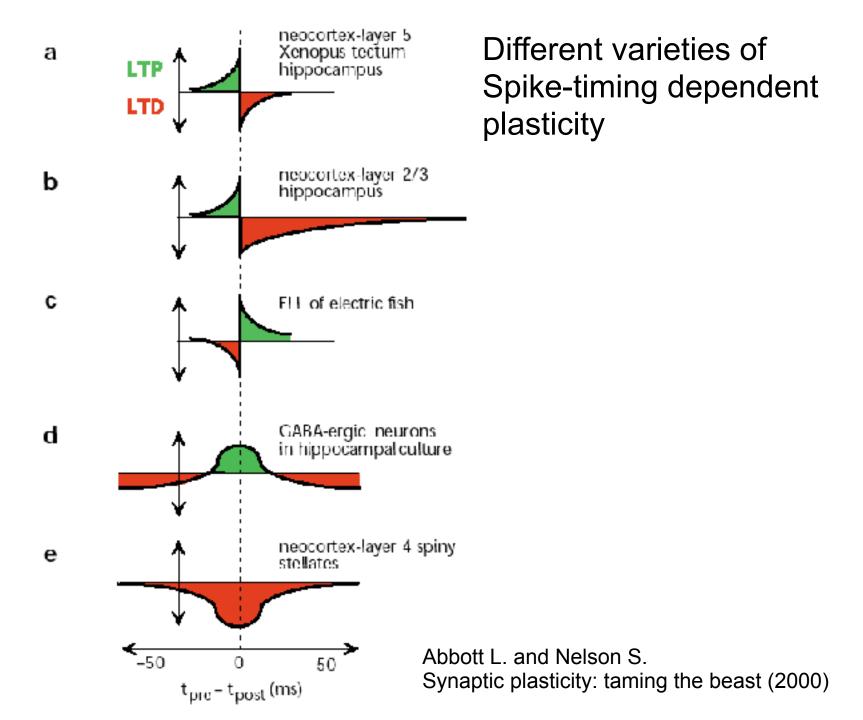
$$\frac{dW_{ba}}{dt}(t) = f(\tau)a(t)b(t-\tau)$$

Simplified:

 $\frac{dW_{ba}}{dt}(t) = a(t)b(t-1) - a(t-1)b(t)$ 

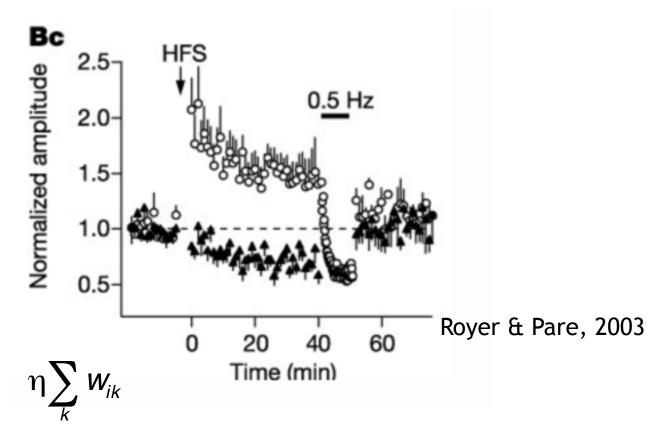
Pre before Post

Bi & Poo (1998)



#### How can Hebbian networks be stable?

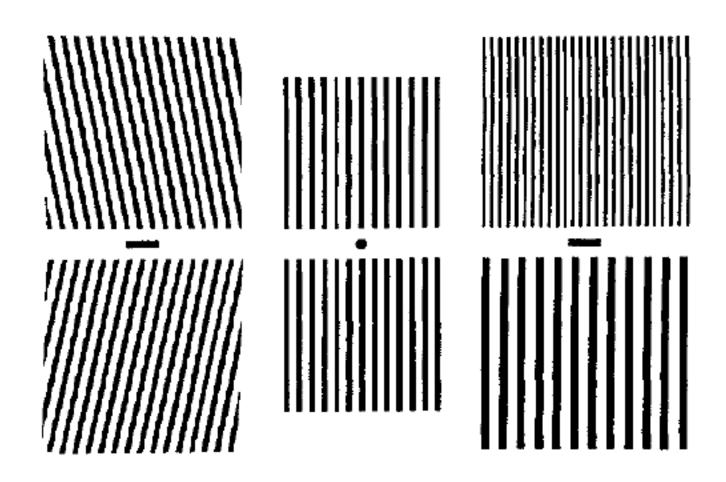
Heterosynaptic Long Term Depression



Decrease each  $\boldsymbol{w}_{ij}$  by

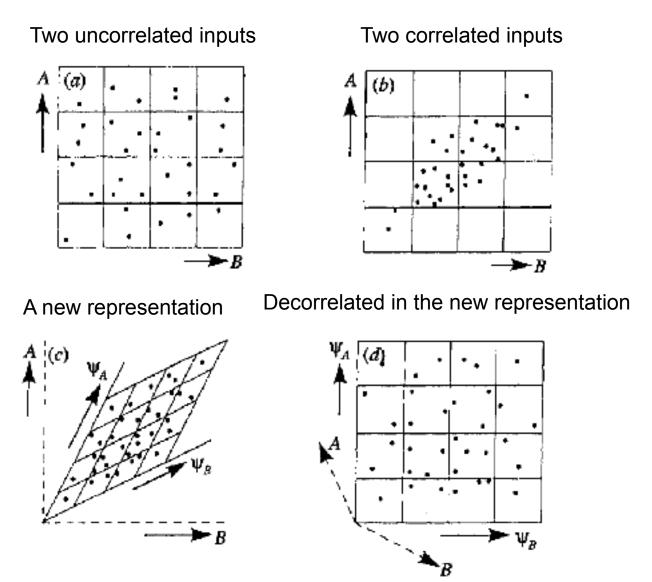
$$\frac{dW_{ji}}{dt} \propto X_i Y_j - Y_j^2 W_{ji}$$

# Predictive coding: removing predictable signals from the brain various aftereffects including tilt and motion



Barlow HB, The knowledge used in vision and where it comes from (1997) Hosoya et al., Dynamic predictive coding in the retina (2005)

### "Decorrelation" – increasing the difference between inputs



Barlow HB, The knowledge used in vision and where it comes from (1997)



## Hebbian plasticity

$$\frac{dW}{dt} = \langle AB \rangle$$

A excites B: Coincidence leads to greater release

A inhibits B: Coincidence leads to less release

"Anti-Hebbian" plasticity 
$$\frac{\partial W}{\partial t} = -\langle AB \rangle$$

A excites B: Coincidence leads to less release

A inhibits B: Coincidence leads to greater release