

# BNN Course: Synaptic Plasticity

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# Reminder:

- ▶ git pull (To access new changes to the repository)
- ▶ Handouts ([www.nest-simulator.org/introduction-to-pynest/](http://www.nest-simulator.org/introduction-to-pynest/))

# Synaptic plasticity

- ▶ considered to be the biological substrate of learning and memory

Synaptic changes can be induced by specific stimulation conditions:

- ▶ presynaptic firing rates
- ▶ spike timing

Detailed biophysical models

- ▶ crucial to understand underlying biological mechanisms

Phenomenological models

- ▶ describe synaptic changes without reference to mechanism
- ▶ generally more tractable and less computationally expensive

# Outline

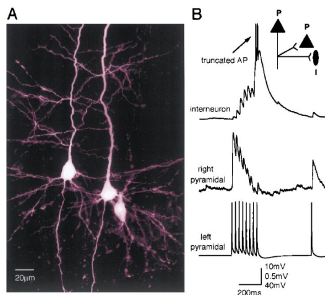
## Basic experimental findings

- Short-term plasticity
- Long-term plasticity
- Spike-timing dependent plasticity
- Homeostasis

## Phenomenological models of synaptic plasticity

- Short-term plasticity (very briefly)
- Spike-timing dependent plasticity (in detail)

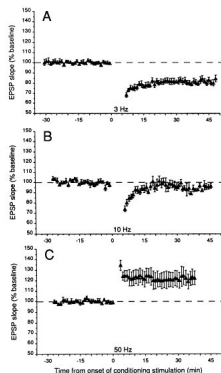
# Short-term plasticity



Markram et al. (1998)  
*Proc Natl Acad Sci USA*. **95**(9):5323–5328

- ▶ sequence of eight presynaptic spikes at 20Hz evokes successively smaller (depression) or successively larger (facilitation) responses in postsynaptic cell
  - ▶ same presynaptic neuron makes connections to different types of target neurons with different plasticity properties
- 
- ▶ enables the synaptic efficacy to represent the history of presynaptic activity (e.g higher activity - faster depletion of resources - STD)
  - ▶ change persists only for a few hundred milliseconds (amplitude of the postsynaptic response recovers to close-to-normal values within less than a second)

# Long-term plasticity

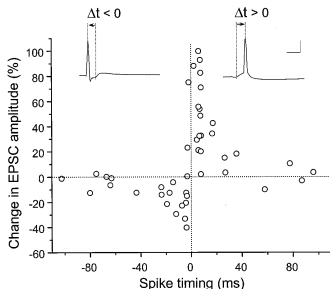


- ▶ 900 presynaptic stimulation pulses yield either persistent depression (LTD) or potentiation (LTP) depending on rate

Dudek & Bear (1992)  
*Proc Natl Acad Sci USA.* **89**:4363–4367

- ▶ sensitive to the presynaptic firing rate over a time scale of tens or hundreds of seconds
- ▶ change can persist for more than one hour
- ▶ final stabilization on a time scale of hours (e.g. late phase of LTP reported by Frey & Morris, 1997)

# Long-term plasticity

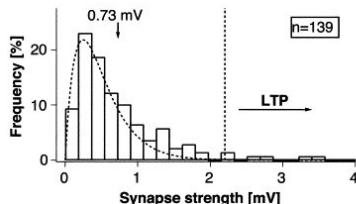


- ▶ LTP if presynaptic spike precedes postsynaptic spike by 10 ms
- ▶ LTD if order of spikes reversed
- ▶ repetitions of 60 pairs of spikes (single pair has no effect)

Bi & Poo (1998)  
*J Neurosci* **18**:10464–10472

- ▶ depends on exact timing of pre- and postsynaptic spikes on time scale of milliseconds
- ▶ so called spike-timing dependent plasticity (STDP)

# Spike-timing dependent plasticity (STDP)

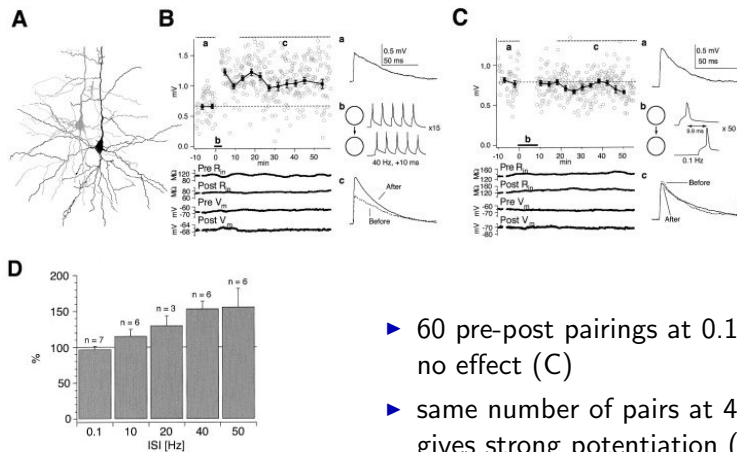


Sjostrom et al. (2001) *Neuron*. 32(6):1149–64

- ▶ synaptic strength (e.g. EPSP amplitudes) in data collected across several pairs of neurons reported to be unimodal



# Spike-timing dependent plasticity (STDP)

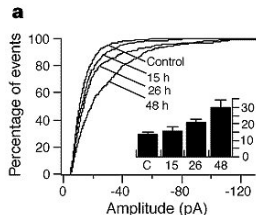


- ▶ 60 pre-post pairings at 0.1 Hz have no effect (C)
- ▶ same number of pairs at 40 Hz gives strong potentiation (B)

Sjostrom et al. (2001)  
*Neuron*. 32(6):1149–64

- ▶ depends on repetition frequency of pre-post spike-pairings

# Homeostasis of synaptic efficacies

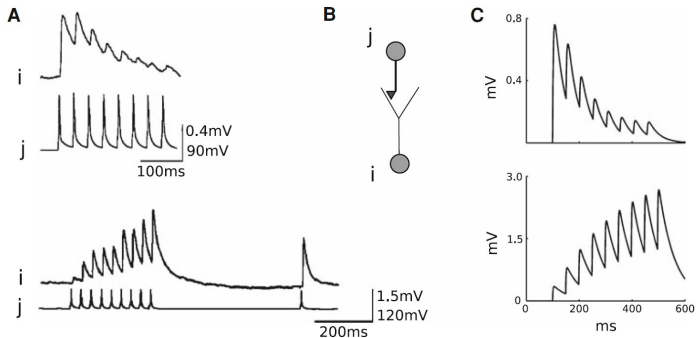


- ▶ chronic blockade of cortical culture activity increases amplitude of mEPSCs without changing their kinetics

Turrigiano et al. (1998)  
*Nature*. **391**:892–896

- ▶ on a time scale of hours, rescaling of synaptic response amplitudes may occur
- ▶ useful to stabilize neuronal firing rates

# Phenomenological models of short-term plasticity



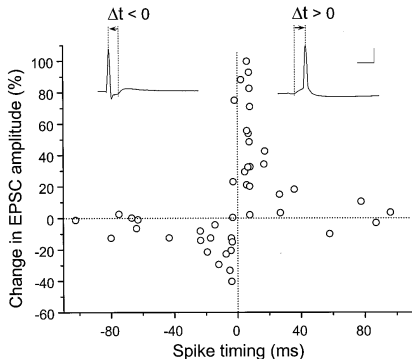
Left: experimental results adapted from Markram et al. (1998)

Right: simulation results using Markram-Tsodyks model

- ▶ fast synaptic dynamics firmly established in biological literature (Markram et al. 1998; Gupta et al. 2000)
- ▶ well-accepted models exist (Abbott et al. 1997; Tsodyks et al. 1998)

# Phenomenological models of spike-timing dependent plasticity

What do we need to specify a pair-based model of STDP?



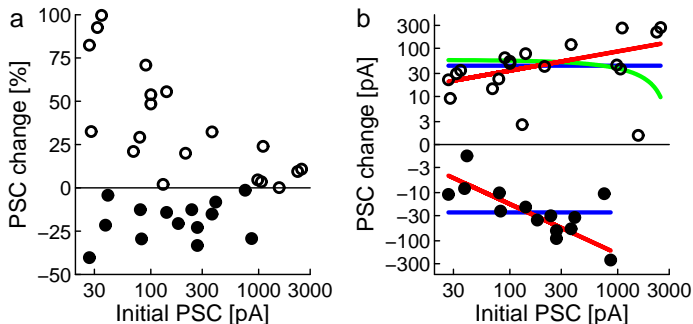
Bi & Poo (1998)  
*J Neurosci* 18:10464–10472

$$\Delta w_- = -F_-(w) \exp(-|\Delta t|/\tau_-)$$
$$\Delta w_+ = -F_+(w) \exp(-|\Delta t|/\tau_+)$$

- ▶ weight dependence i.e.  $F_{\pm}(w)$
- ▶ spike pairing scheme
- ▶ decomposition of synaptic delay into axonal and dendritic components
- ▶ asymmetry of STDP kernel ( $\alpha = \frac{F_-}{F_+}$ )

all of these aspects are important  
and should not be chosen arbitrarily

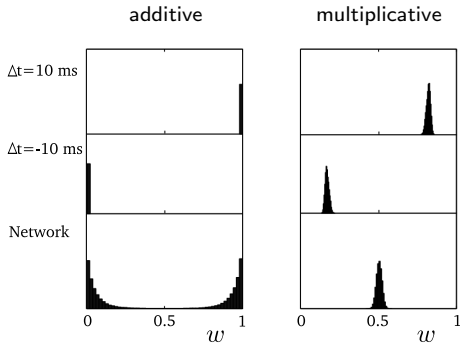
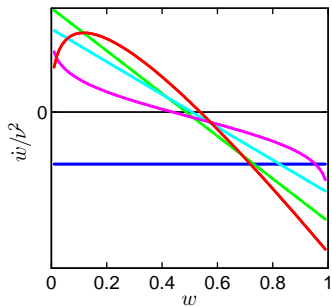
# Models of STDP: Weight Dependence



- ▶ Reanalysis of Bi & Poo (1998) data shows that the commonly used **additive** model ( $F_- = A_-$ ,  $F_+ = A_+$ ) is not a good fit for the data.
- ▶ Depression data are best fit by a **multiplicative/power law** model ( $F_- = A_- w^{-1}$ )
- ▶ Potentiation data are best fit by a **power law** model ( $F_+ = A_+ w^\mu$ ,  $\mu = 0.4$ )

# Models of STDP:

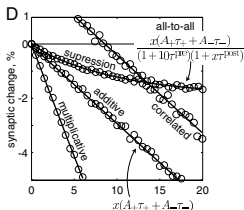
## Distribution of synaptic weights



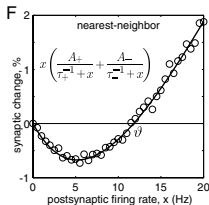
- ▶ all weight dependent models exhibit a stable fixed point and generate a unimodal distribution (**multiplicative**, **power law**, **Gütig**, **Van Rossum**)
- ▶ the **additive** model does not have a stable fixed point and produces a bimodal distribution

# Models of STDP: Spike Pairing Scheme

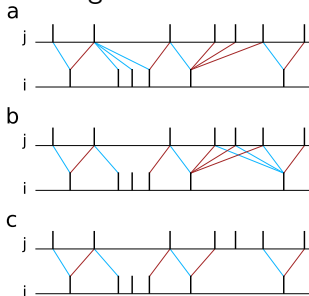
all-to-all...



... or nearest neighbour?



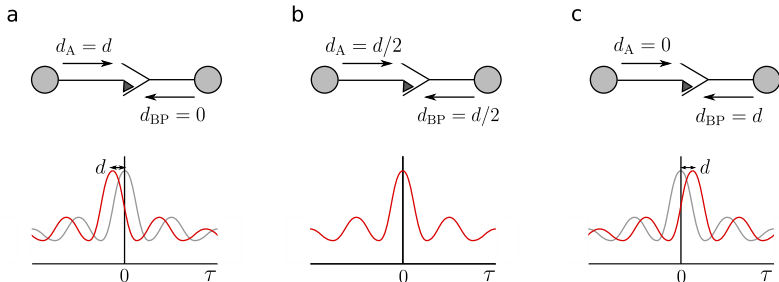
...but which nearest neighbour?



Izhikevich & Desai (2003) *Neural Comput* **15**:1511–1523

- ▶ there are many possible spike pairing schemes
- ▶ they can give qualitatively different results (e.g. Kempter et al. 2001; Morrison et al. 2007)
- ▶ see Burkitt et al (2004) for copious analysis

# Models of STDP: Synaptic Delays



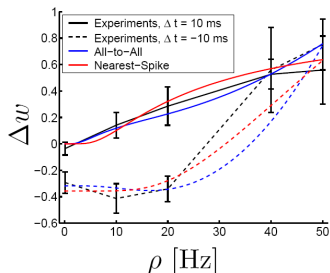
- ▶ The synaptic drift depends on the **cross-correlation with respect to the synapse**
- ▶ This is shifted from to the left or right of the (measurable) cross-correlation with respect to the soma depending on whether the synaptic delay is largely axonal (a) or dendritic (c)
- ▶ Therefore the same STDP rule coupled with the same network dynamics can give rise to either net depression, no change, or net potentiation



# Models of STDP: Beyond Pair Effects

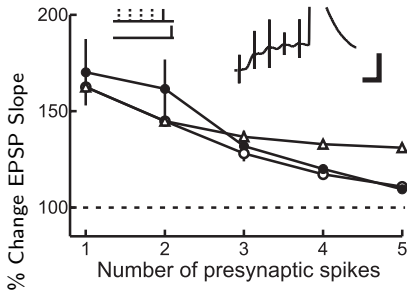
Many features of STDP cannot be explained by pair-based rules:

## Frequency dependence



Data: Sjostrom et al. (2001) *Neuron* **32**:1149–1164  
Modelling: Pfister et al. (2006) *J Neurosci* **26**:9673–9682  
Modelling: Also see Clopath et. al (2010) *Front. Synap Neurosci.*, Shouval et. al (2010) *Front. Comp. Neurosci.*, Costa et al. (2015) *eLife*

## Burst dependence



Data and modelling: Froemke et al. (2006) *J Neurophysiol* **95**:1620–1629

There is generally no cross-compatibility between models accounting for these features

# Conclusions

- ▶ A modeller needs to specify a complete model despite lack of clear experimental evidence
- ▶ These choices can have profound consequences
- ▶ Almost all network modelling papers fail to consider whether the effects shown are artifacts of their specific choices
- ▶ There is so far no phenomenological model which accounts for all the nonlinearities exhibited by STDP
- ▶ For those who want to know more...

Biol Cybern (2008) 98:459–478  
DOI 10.1007/s00422-008-0233-1

REVIEW

Biological  
Cybernetics

## Phenomenological models of synaptic plasticity based on spike timing

Abigail Morrison · Markus Diesmann ·  
Wulfram Gerstner