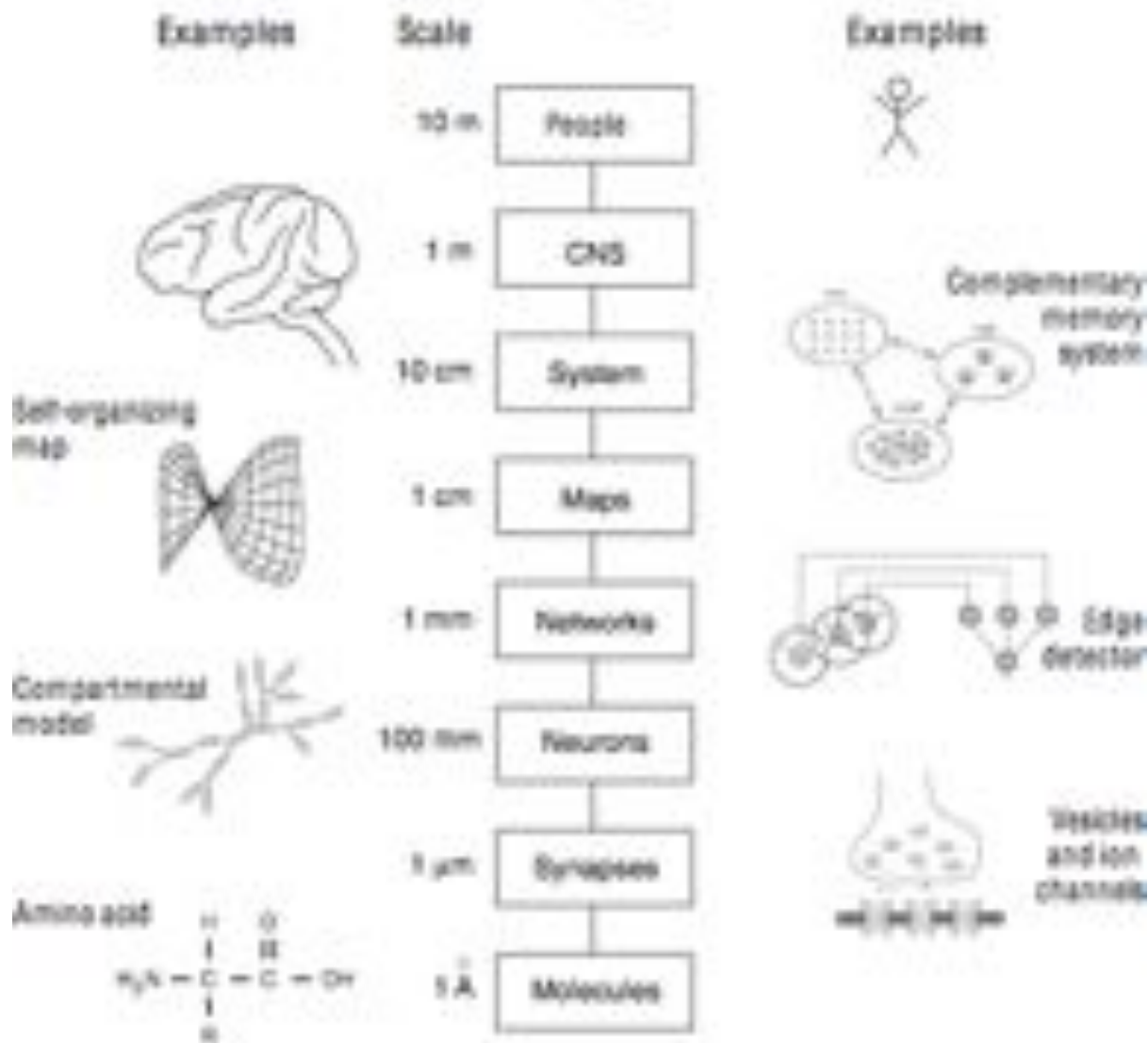


Memory and Learning

Levels of organization in the nervous system



Marr' s approach

1. Computational theory: What is the goal of the computation, why is it appropriate, and what is the logic of the strategy by which it can be carried out?
2. Representation and algorithm: How can this computational theory be implemented? In particular, what is the representation for the input and output, and what is the algorithm for the transformation?
3. Hardware implementation: How can the representation and algorithm be realized physically?

Marr puts great importance to the first level:

”To phrase the matter in another way, an algorithm is likely to be understood more readily by understanding the nature of the problem being solved than by examining the mechanism (and hardware) in which it is embodied.”

Non-associative learning:

- Habituation
- Dishabituation
- Sensitization

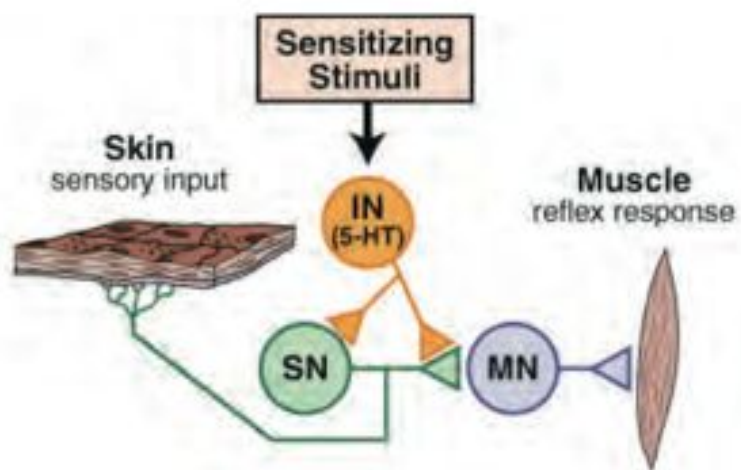
Associative learning:

- Classical conditioning
- Instrumental (operant) conditioning
- Recognition Memory

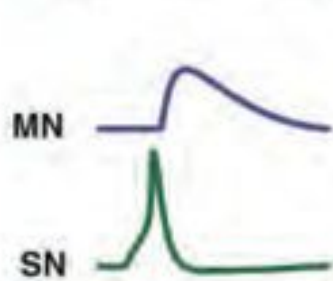
Synaptic Plasticity

Induction, Expression, Maintenance

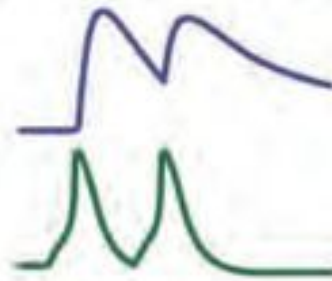
A1



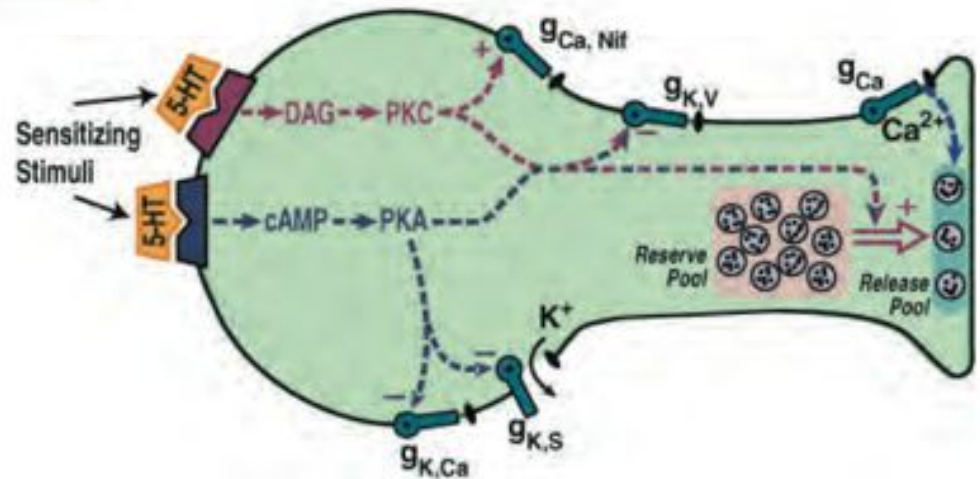
A2 Before activation of IN



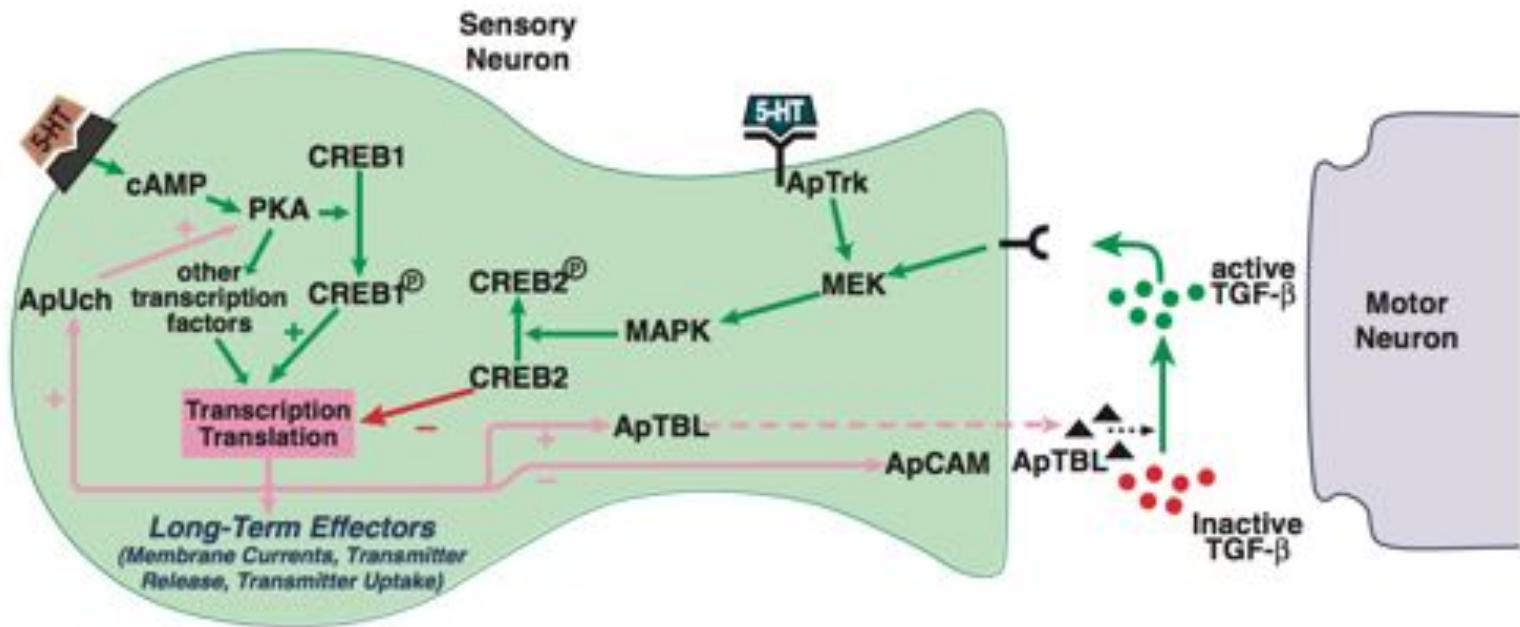
A3 After activation of IN



Short Term

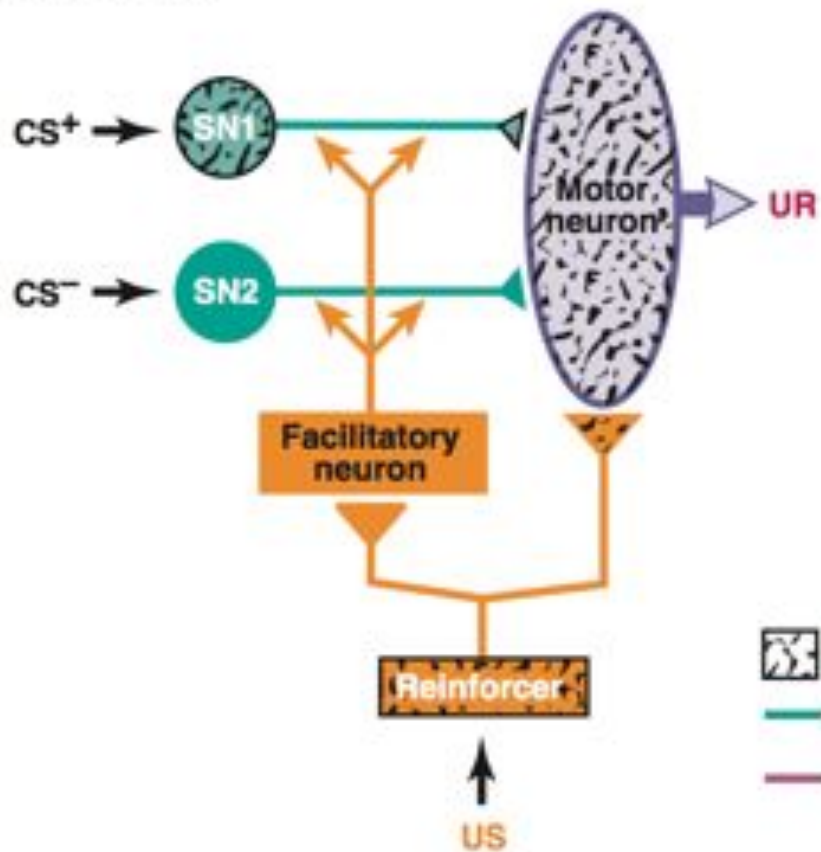


Long Term Sensitization

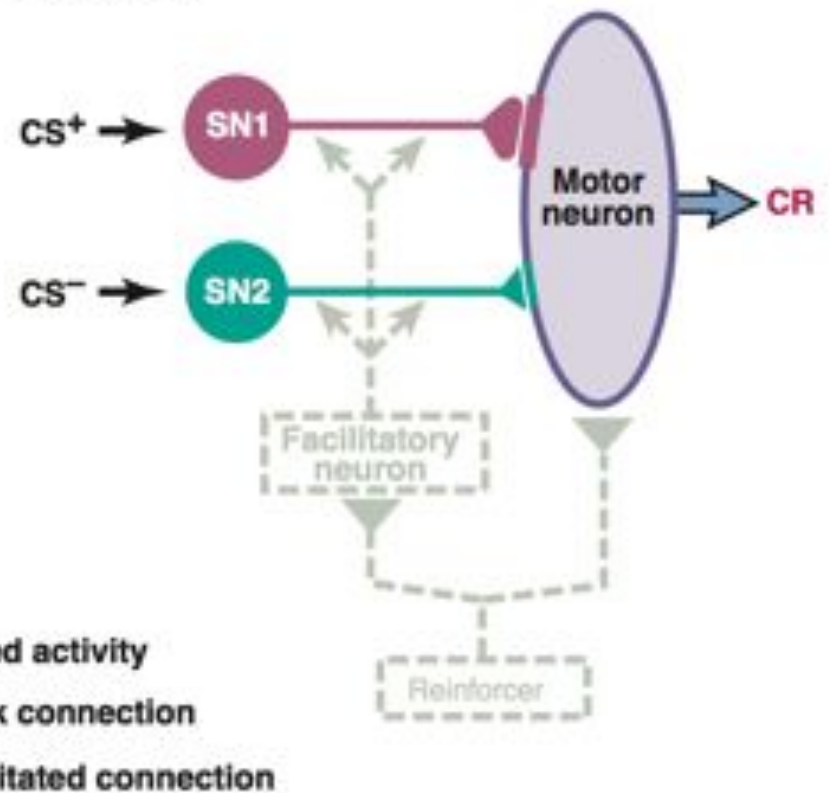


Classical condition of withdraw reflex in Aplysia

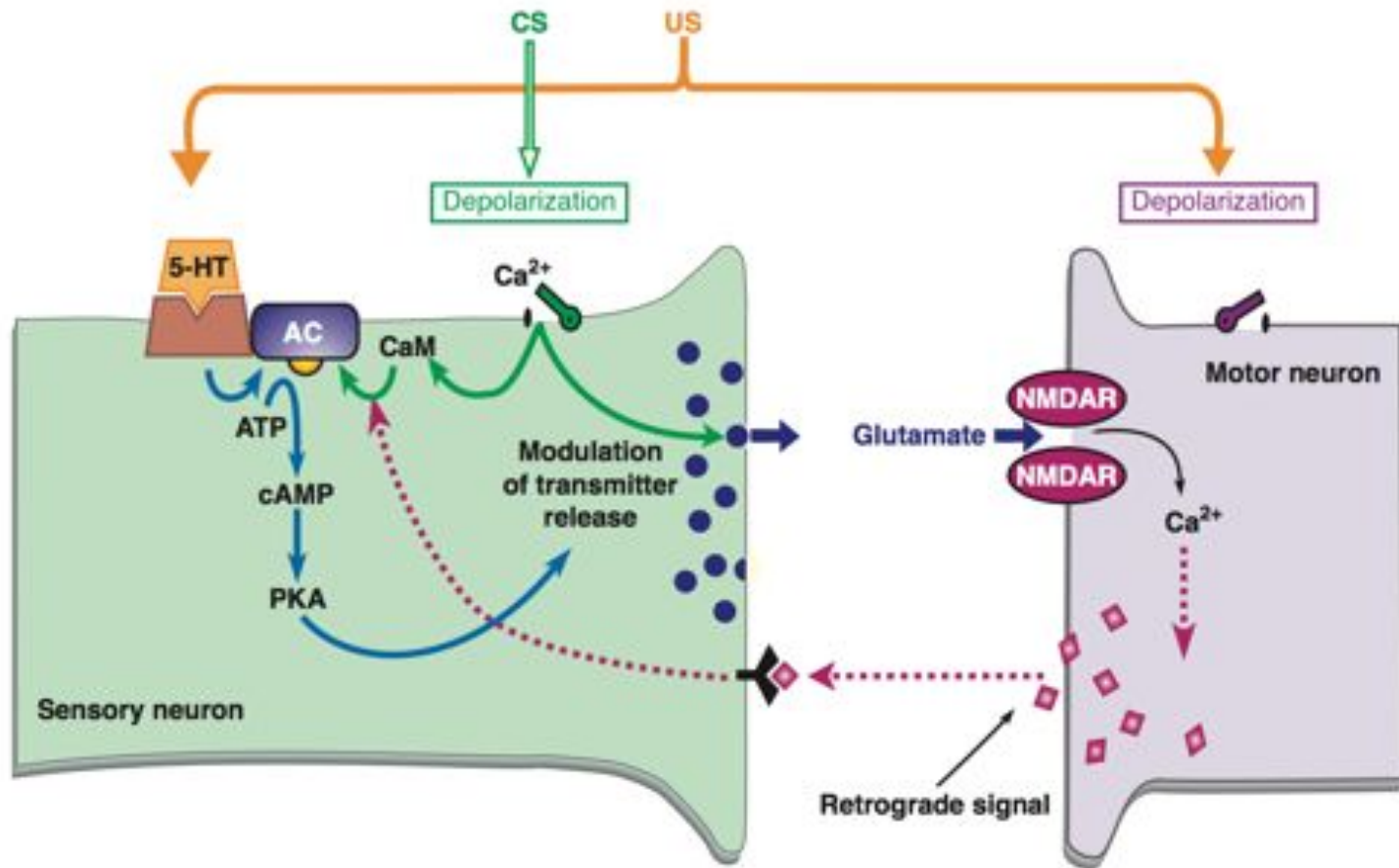
A. Learning

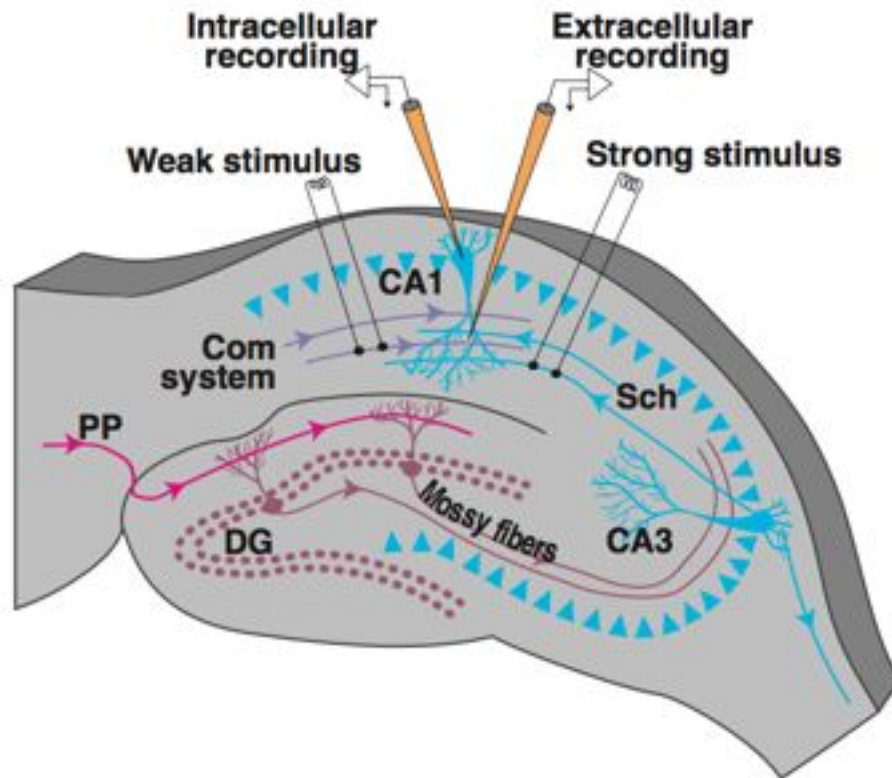


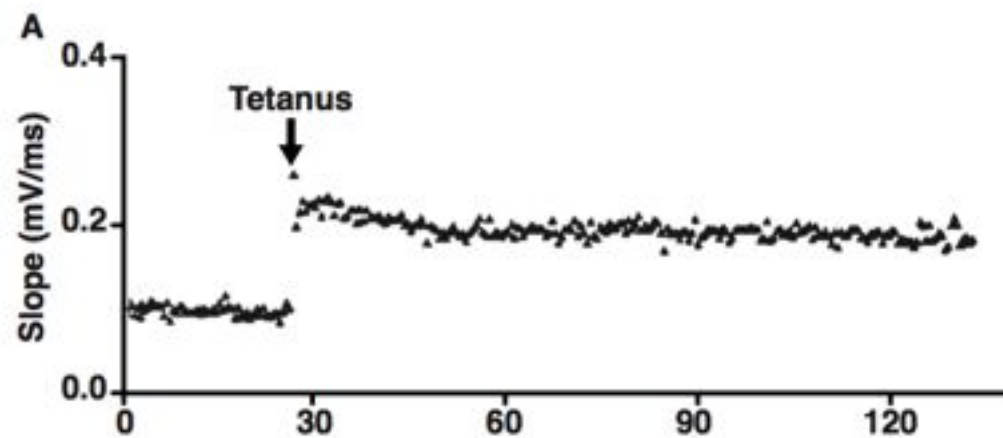
B. Memory



Coincidence detection and retrograde signaling







B1. Control

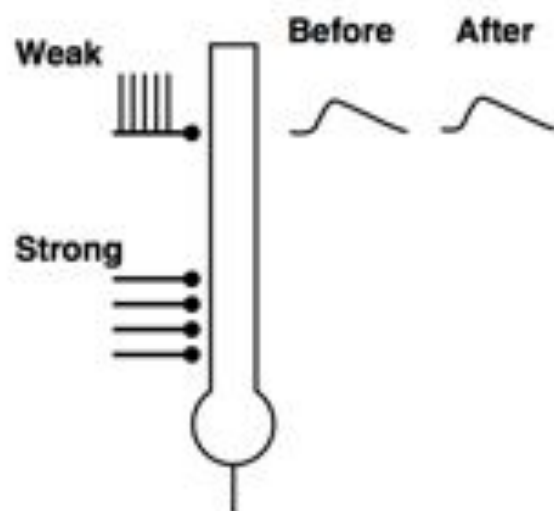
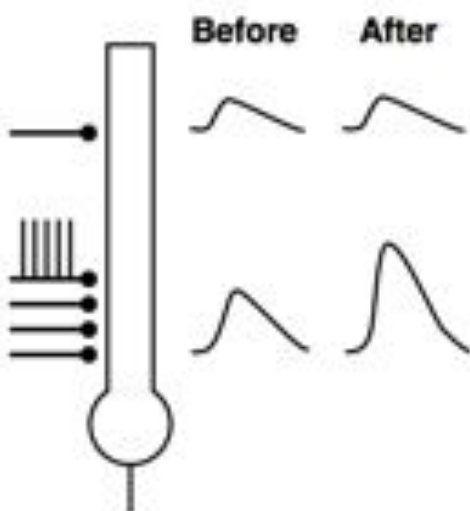
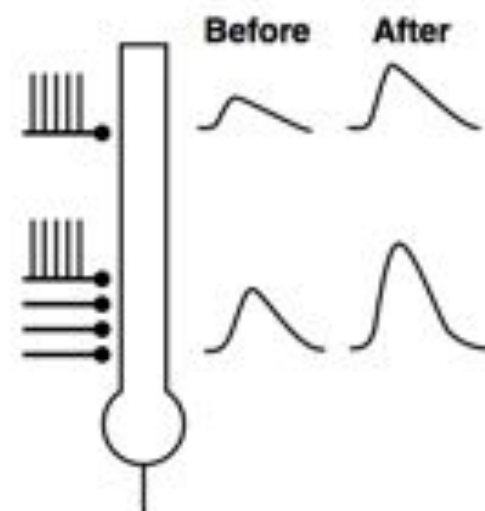


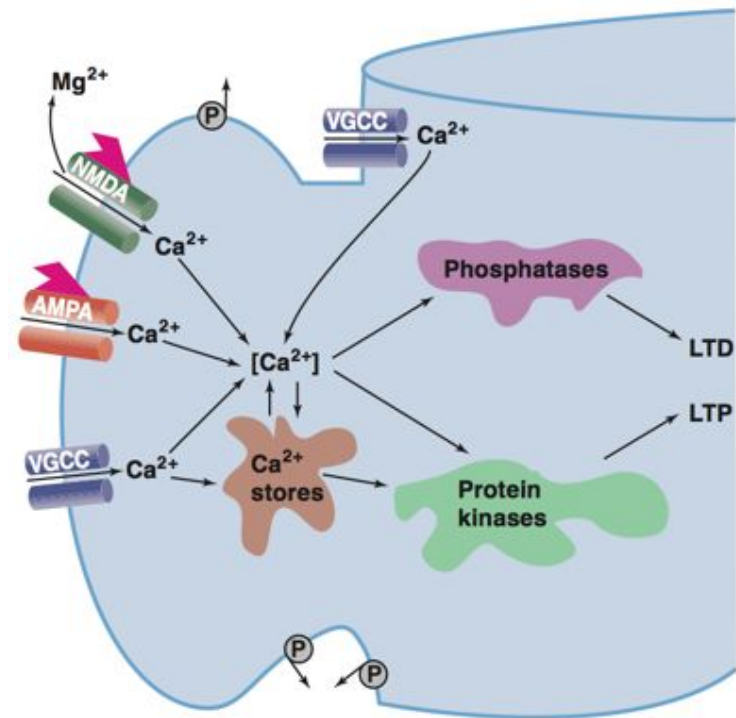
B2. 90 min.

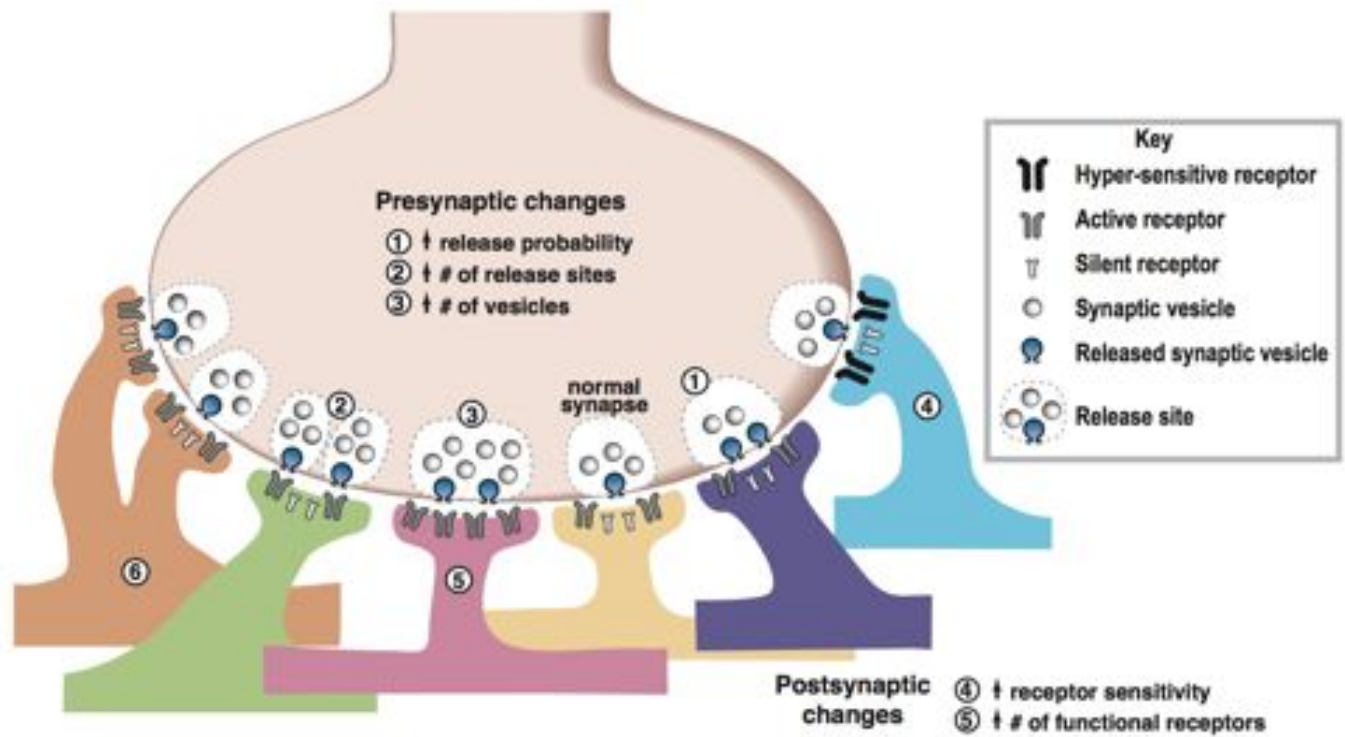


B3. Superimposed

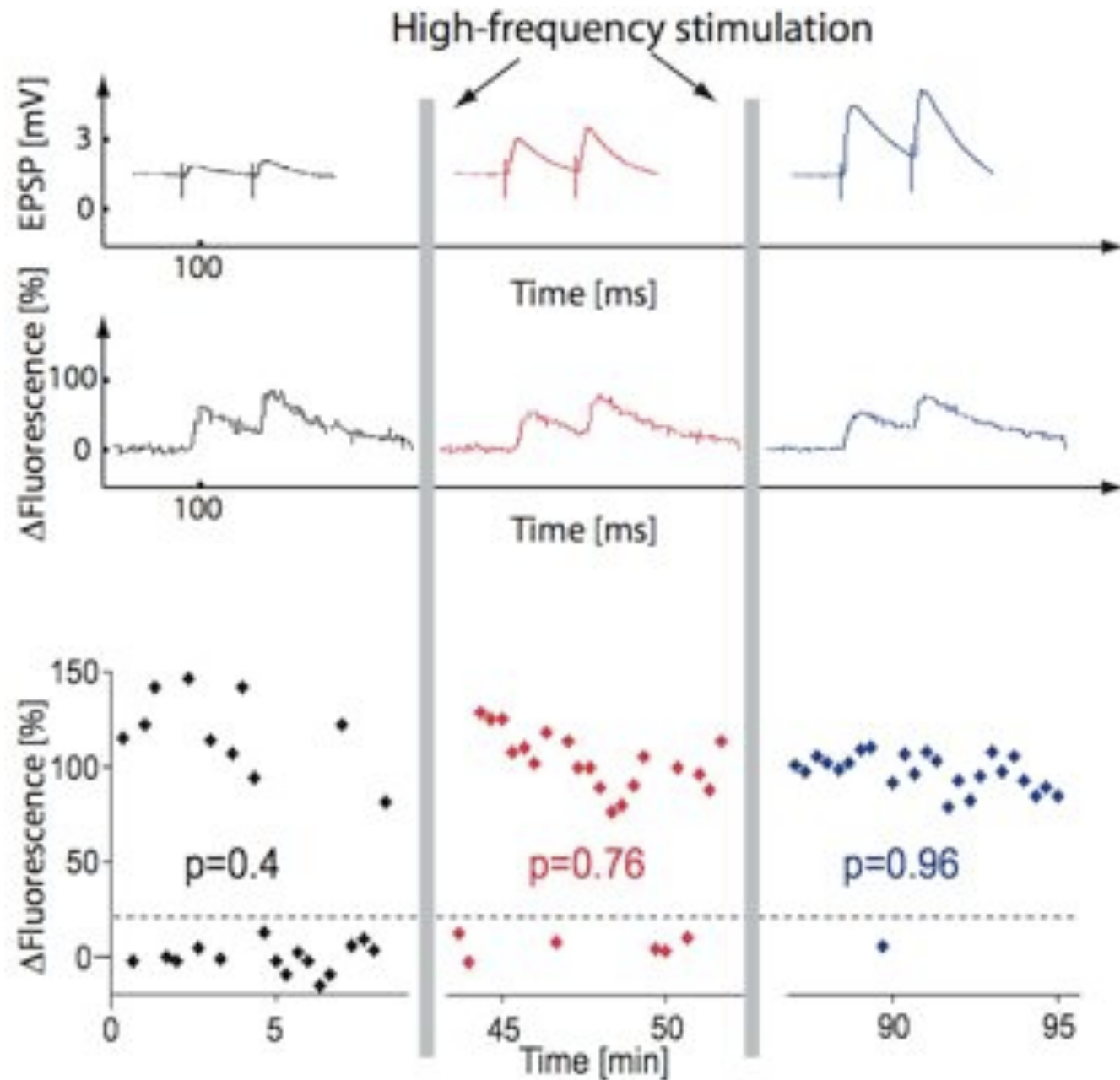


A**B****C**

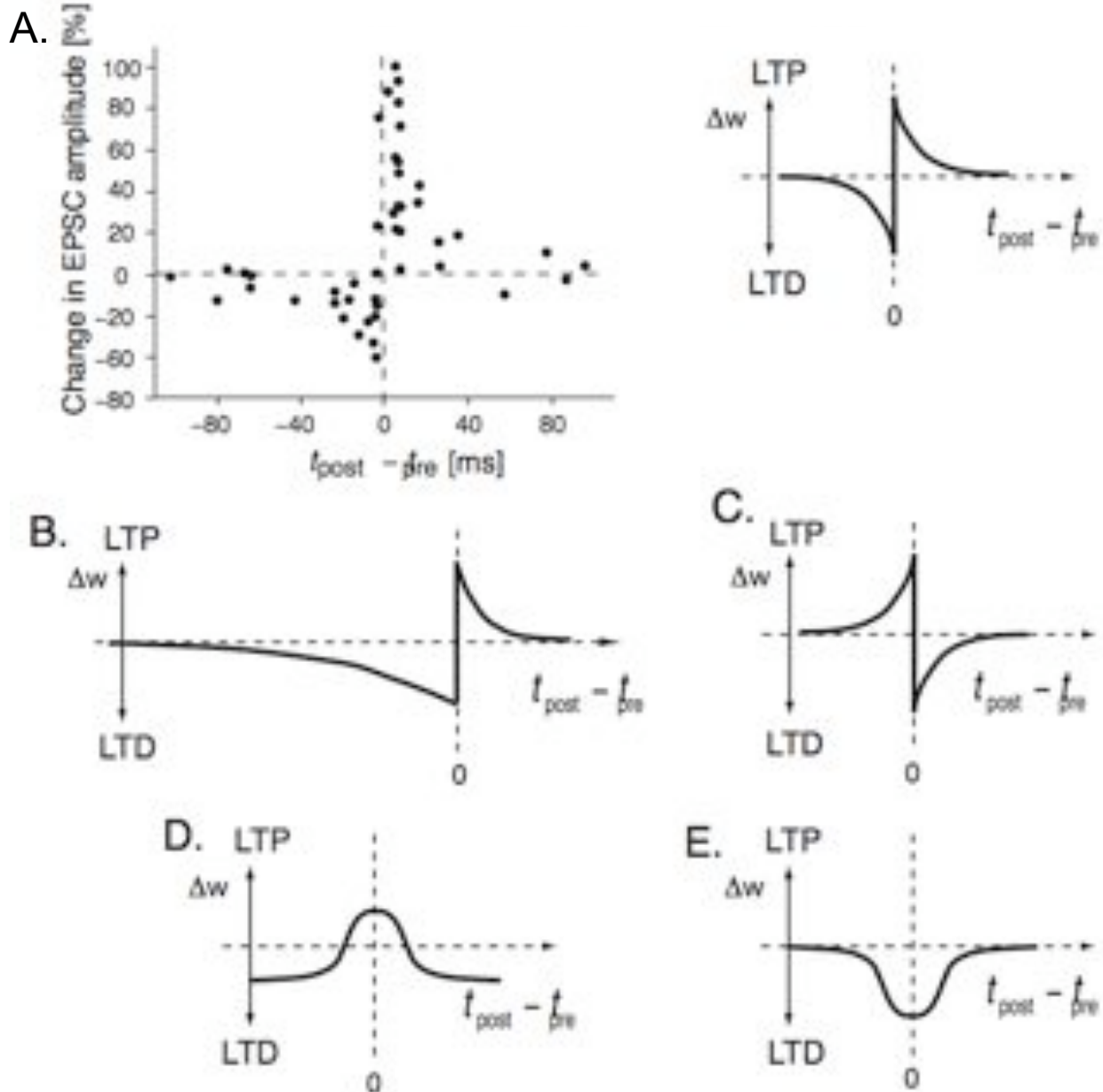




Synaptic neurotransmitter release probability



Spike timing dependent plasticity (STDP)



Hebbian learning in population and rate models

General: $\Delta W_{ij} = \epsilon(t, W)[f_{\text{post}}(r_i)f_{\text{pre}}(r_j) - f(r_i, r_j, W)]$

Mnemonic equation (Caianiello): $\Delta W_{ij} = \epsilon(W)[r_i r_j - f(W)]$

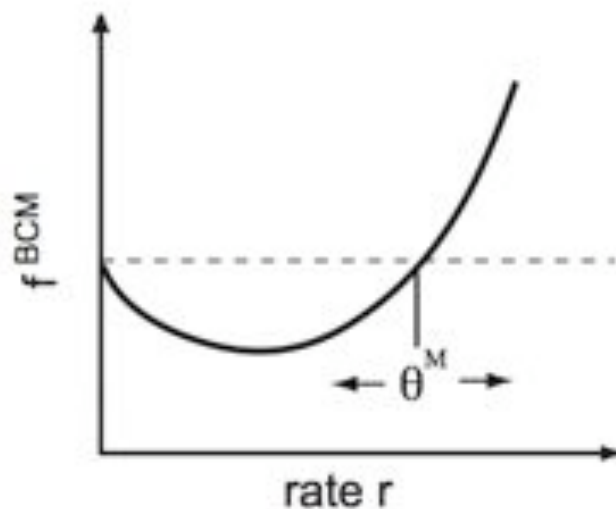
Basic Hebb: $\Delta W_{ij} = \epsilon r_i r_j$

Covariance rule: $\Delta W_{ij} = \epsilon(r_i - \langle r_i \rangle)(r_j - \langle r_j \rangle)$

BCM theory: $\Delta W_{ij} = \epsilon(f^{\text{BCM}}(r_i; \theta^M)(r_j) - f(W))$

ABS rule: $\Delta W_{ij} = \epsilon(f_{\text{ABS}}(r_i; \theta^-, \theta^+) \text{sign}(r_j - \theta^{\text{pre}}))$

Function used in BCM rule



Function used in basic ABS rule

