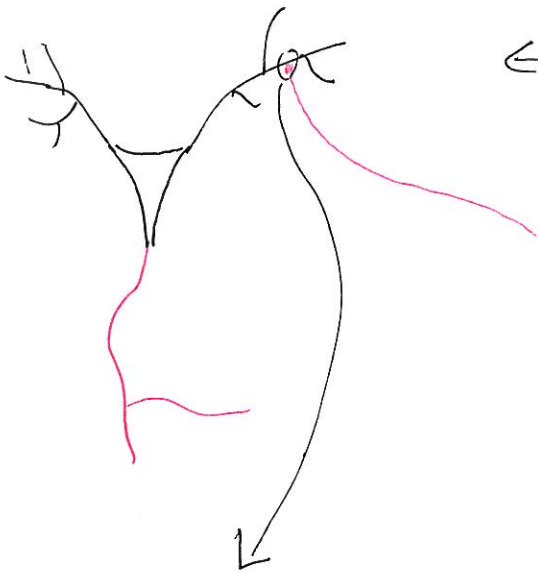


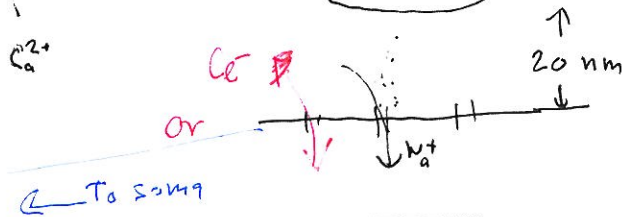
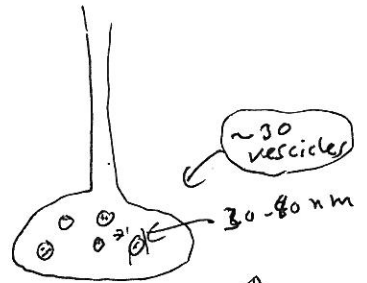
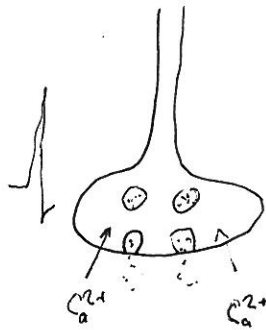
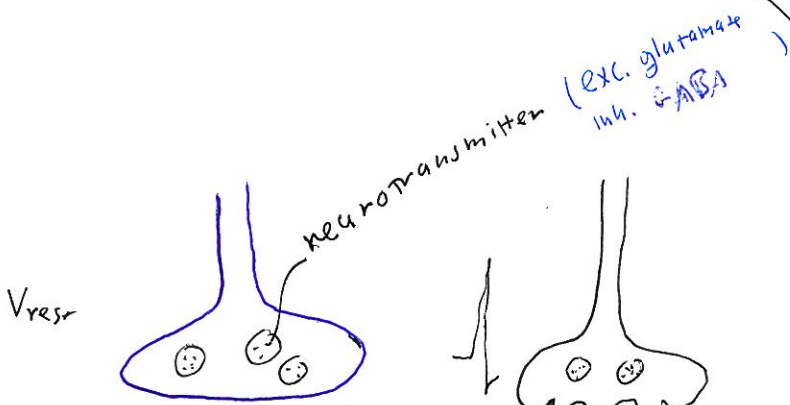
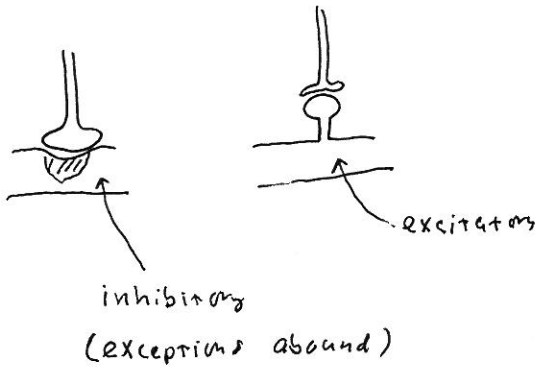
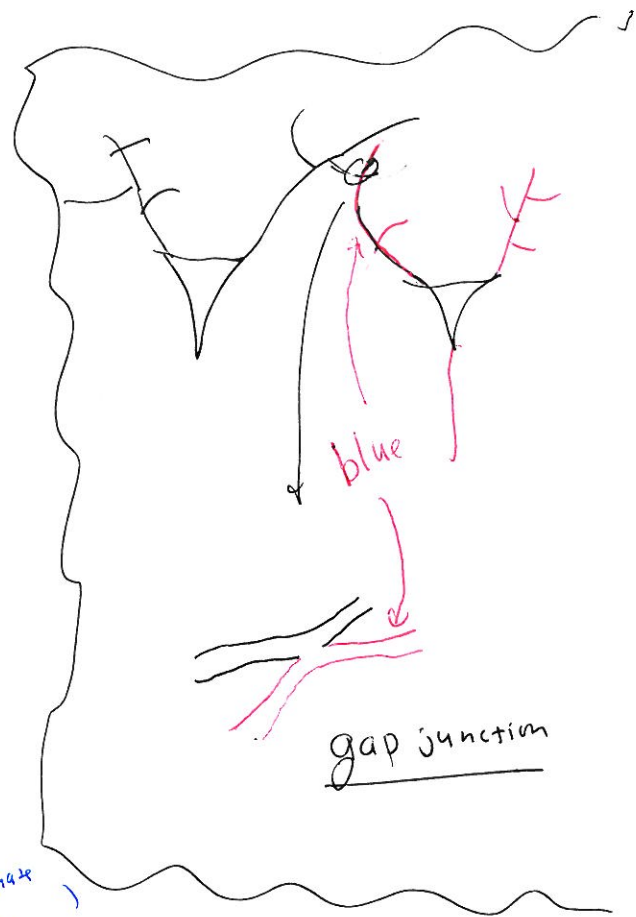
10/15/04

# Synapses

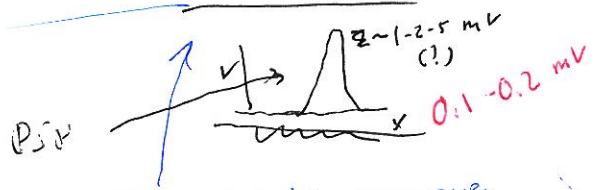
① replaced by 10-16!! (except for page 2)



~ 5-10,000 connections



mentim repackaging here!



Several mV, can even cause spikes (& co-operative effect). 0.1-0.2 mV at soma

Questions

1a) How many packets released / spike?

1b) How are they repackaged?

1c) What ~~kind~~ neurotransmitters are released?  
→ What channels are opened?

← glutamate

} presynaptic: facts

2a) ~~What~~ What is the time course of release?

2b) How does ~~#~~ <sup>release probability</sup> depend on history

} presynaptic: facts/math

3a) How does PSP size depend on parameters?

~~3b)~~ ~~How~~ How do we put synaptic release into models

3b) How " " " " " history?

} postsynaptic: facts/math



2006

What do we want to know! - write down full set of eqns.

~~1) What kind of~~

1) what happens when AP arrives at presynaptic terminal

a) quantum model - (AMPA, GABA, NMDA, ... ) pg 10

2) what determines <sup>synaptic strength</sup> ~~?~~ ?

a) # packets

b) prob of release

c) postsynaptic channel dynamics

} binomial pg 8, 9

3) history dependent pg 6

Short term: PS,  $\tau$

Long term: PS.

3

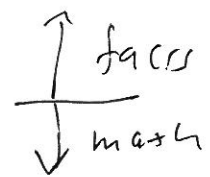
1a) quantal release

- n vesicles
- each released independently w/ probability p
- amount released =  $q_0$

$$p(k) = \frac{n!}{k!(n-k)!} p^k (1-p)^{n-k}$$

Average release =  $q_0 \langle k \rangle$

Standard deviation =  $q_0 [\langle k^2 \rangle - \langle k \rangle^2]$



////

$$\begin{aligned} \langle k^r \rangle &= \sum_{k=0}^n \frac{n!}{k!(n-k)!} k^r p^k (1-p)^{n-k} \\ &= \sum_{k=0}^n \frac{n!}{k!(n-k)!} k^r p^k q^{n-k} \Big|_{q=1-p} \end{aligned}$$

$$p \frac{d}{dp} p^k = k p^k$$

$$p \frac{d}{dp} \left[ p \frac{d}{dp} \right] p^k = k^2 p^k$$

$$k^r = \left( p \frac{d}{dp} \right)^r p^k$$

$$\Rightarrow \langle k^r \rangle = \left( p \frac{d}{dp} \right)^r \underbrace{\sum_{k=0}^n \frac{n!}{k!(n-k)!} p^k q^{n-k}}_{(p+q)^n} \Big|_{q=1-p}$$

$$\langle k^r \rangle = \left( p \frac{d}{dp} \right)^r (p+q)^n \Big|_{p+q=1}$$

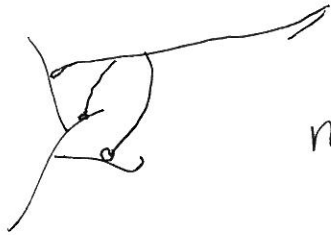
$$\left. \begin{aligned} \langle k \rangle &= np \\ \langle k^2 \rangle &= n(n-1)p^2 + np \end{aligned} \right\} \Rightarrow \langle k^2 \rangle - \langle k \rangle^2 = np(1-p)$$

④

NMJ = Katz

$n = 100 - 1000$

central synapses:  $n = 1$  per synapse (hypothesis)



$n = 3$

$p: 0.1 - 0.9$

total mystery!!!!

$\langle \text{release} \rangle = q_0 np$

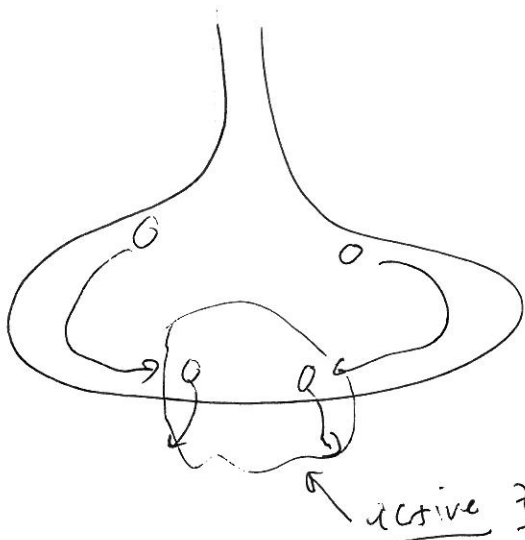
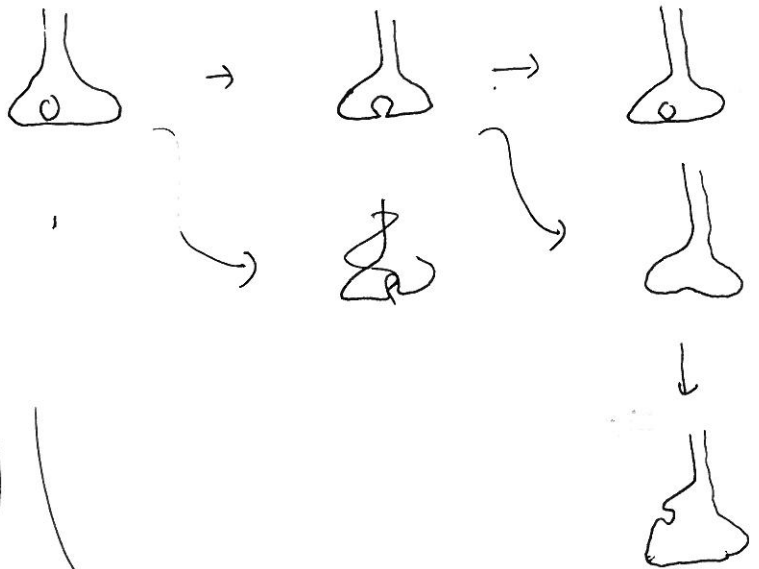
$\text{Var} [ ] = q_0^2 np(1-p)$

$\left[ \frac{\text{Std}}{\text{mean}} \right]^{var} = \frac{np}{\sqrt{np(1-p)}} = \sqrt{\frac{np}{1-p}}$   
 $\hookrightarrow \frac{\sqrt{np(1-p)}}{np} \cdot \sqrt{\frac{1-p}{np}} = \text{big!!!}$

//////

1b) - Kiss + run

- standard model



$\sim 20-30$  vesicles

active zone

5

Later

IC) Excitatory: ~~AMPA~~ AMPA  
NMDA → mix of  $Na^+$ ,  $K^+$ ,  $Ca^{++}$   
 $E \sim 0$  mV

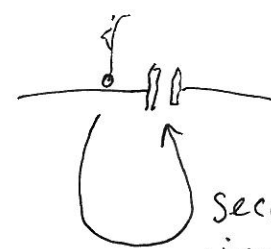
Inhibitory:  $GABA_A$   
 $GABA_B$  →  $Cl^-$   $E \sim -70$  mV  
 $-100$  mV

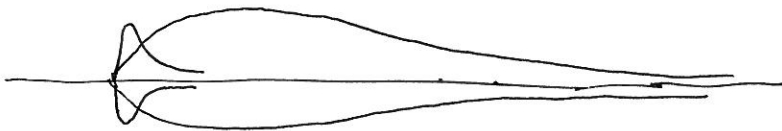
AMPA + ~~AMPA~~  $GABA_A$ : fast ionotropic (fast) 2-10 ms

NMDA +  $GABA_B$ : metabotropic (slow) 50-200 ms  
the only metabotropic one, according to Koch

fast  binding causes opening

in beginning

slow  second messenger cascade, G-protein mediated  
involving ~~the coupled~~



Start w/ new  
 Fig - tell them about  $Ca^{++}$  + active zones, mention  
 ionotropic vs metabotropic

2a) Working hypothesis: only Prelease changes as a function of history.

- Facilitation (A.K.A. augmentation or potentiation) probably result from a buildup of  $Ca^{++}$  remember,  $Ca^{++}$  triggers release
- Depression probably results from a depletion of resources (meaning everybody in the readily releasable pool is gone)
- We'll look at the two separately

$$\tau \frac{dP_{rel}}{dt} = P_0 - P_{rel} + \left[ \sum_i \delta(t-t_i) \right] \left\{ \begin{array}{l} -\tau(t-t_0) P_{rel} \\ +\tau f_F (1-P_{rel}) \end{array} \right.$$

between spikes:

do this first  
 $x = f(x) + g(x) \delta(x)$   
 $x(t+\Delta t) = x(t) + \Delta x$   
 process spikes:  
 $\int f(x) \delta(x) dx$   
 $\int g(x) \delta(x) dx = g(x)$   
 $x \rightarrow x + \Delta x$   
 $\Delta x \rightarrow P_{rel} + \left\{ \begin{array}{l} -P_{rel} + f_F P_{rel} \\ +f_F (1-P_{rel}) \end{array} \right.$

$$P_{rel}(t) = P_0 + [P_{rel}(0) - P_0] e^{-\frac{t}{\tau}}$$

[check this!]

$$+ f_0 P_{rel}$$

[check this]

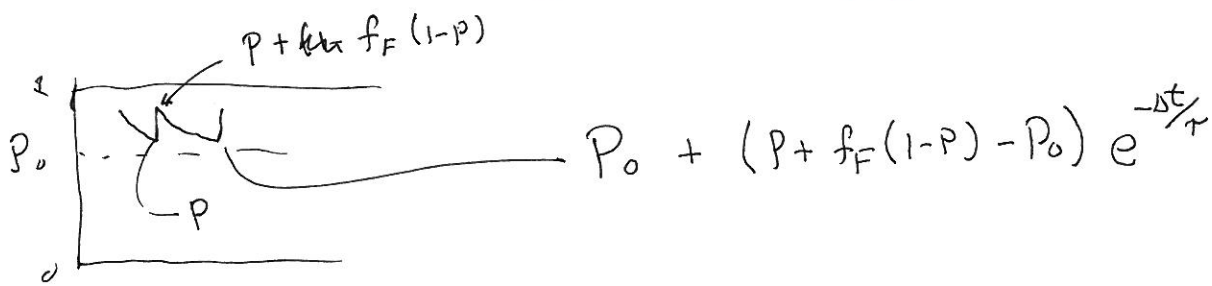
$$+ f_F (1 - P_{rel})$$



⑦

2b) Implications

SKIP ~~exercise~~; facilitate  
- HWK



$$P(t_{n+1}) = P_0 + (P(t_n)(1-f_F) + f_F - P_0) e^{-\frac{\Delta t}{\tau}}$$

$$\langle P(t_{n+1}) \rangle = P_0 + (f_F - P_0) \langle e^{-\frac{\Delta t}{\tau}} \rangle + (1-f_F) \langle P(t_n) e^{-\frac{\Delta t}{\tau}} \rangle$$

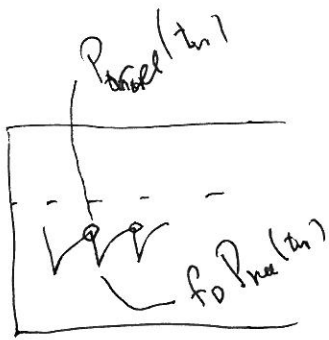
$$\bar{P} = P_0 + (f_F - P_0) \langle e^{-\frac{\Delta t}{\tau}} \rangle + (1-f_F) \underbrace{\langle P(t_n) \rangle \langle e^{-\frac{\Delta t}{\tau}} \rangle}_{\text{Approx!!!}}$$

Poisson at rate  $r$ :

$$p(\Delta t) = r e^{-\Delta t r}$$

$$\langle e^{-\frac{\Delta t}{\tau}} \rangle = \int_0^{\infty} d(\Delta t) r e^{-\Delta t (r + \frac{1}{\tau})}$$

$$= \frac{r}{r + 1/\tau} = \frac{r\tau}{1+r\tau}$$



8

depression:

$$P_{pre}(t_{n+1}) = P_0 + [P_{pre}(t_n) - P_0] e^{-\frac{\Delta t}{\tau}}$$

$$P(t_{n+1}) = P_0 + (f_0 P(t_n) - P_0) e^{-\frac{\Delta t}{\tau}}$$

$$\Rightarrow \langle P \rangle \approx \frac{P_0}{1 + (1 - f_0)r\tau}$$

← do averaging here because we skip facilitation

average drive to the cell:

$$r \langle P \rangle \approx \frac{P_0 r}{1 + (1 - f_0)r\tau} \rightarrow \frac{P_0}{\tau(1 - f_0)} \quad \text{as } r \rightarrow \infty$$

- independent of  $r$ !!
- synaptic depression democratizes ~~synaptic~~ presynaptic spikes
- why?? - change detection!!!



mention minus!!



9

3a) 4 main kinds of channels:

AMPA } glutamate receptors / excitatory  $E \sim 0$  mV  
NMDA }

GABA<sub>A</sub> } GABA receptors / inhibition,  $E = -70$  mV  
GABA<sub>B</sub> }  $-100$  mV

model:

$$C_m \frac{\partial v}{\partial t} + \sum \frac{\partial v}{\partial x} = \lambda^2 \frac{\partial^2 v}{\partial x^2} - (I_m + v_n i_e)$$

$$g(v - E)$$

this we know

~~$\bar{g} x$  (Transmitter)~~

~~$\dot{x} = \alpha(1-x) - \beta x$~~

~~$\tau \dot{x} = x_{\infty} - x$   $\tau = \frac{1}{\alpha + \beta}$  } AMPA~~

~~$x_{\infty} = \frac{\alpha}{\alpha + \beta}$  } GABA<sub>A</sub>~~

(10)

AMPA

GABA<sub>A</sub>

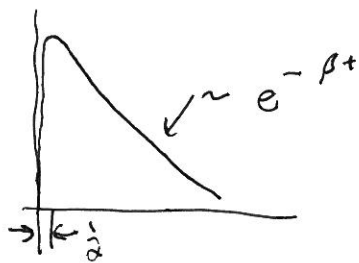
GABA<sub>B</sub>

$$I_m = \bar{g} x (V - E)$$

$\leftarrow 0, -70$

$$\dot{x} = \alpha (\text{transmitter}) (1-x) - \beta x$$

$\leftarrow \text{constant}$

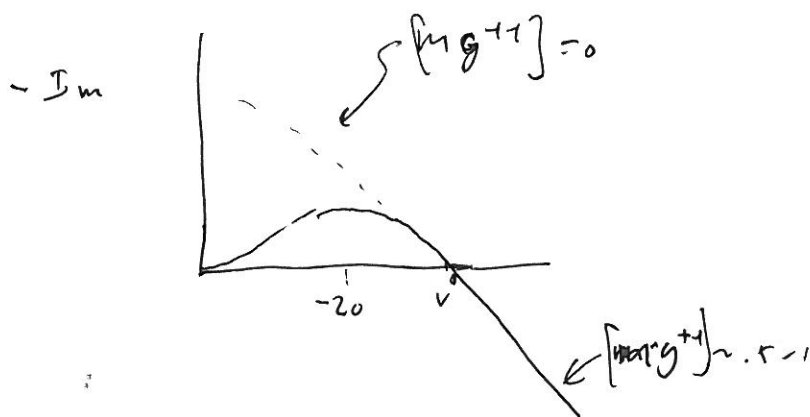


NMDA:

$$I_m = \frac{\bar{g} x (V - E)}{\left(1 + \frac{[Mg^{++}]}{2.57 \text{ mM}} \exp\left(-V/16.1 \text{ mV}\right)\right)}$$

$\leftarrow 0$

$\leftarrow 16.1$



NMDA only opens when postsynaptic cell is ~~activated~~ depolarized: coincidence detector.

11

~~Ampa~~

	$\tau_{rise} (1/\mu s)$	$\tau_{decay} (1/\mu s)$
AMPA	$\sim 0$	5
NMDA	1-5	-150
GABA <sub>A</sub>	0.3	5
GABA <sub>B</sub>	$\sim 10$	$\sim 200$



Full model - single comparm

$$C_m \frac{dv_i}{dt} = -\bar{g}_L (V_i - E_L) - \bar{g}_{Na} m^3 h (V_i - E_{Na}) - \bar{g}_K n^4 (V_i - E_K) - \sum_j \bar{g}_{ij} X_{ij} (V_i - E_x)$$

$$\dot{X}_{ij} = \alpha_x (1 - X_{ij}) \sum_e \delta(t - t_j^e) - \beta X_{ij}$$

↑ instantan rise

(can replace w/ a square pulse) ]

3b) ~~Hebb's rule~~

LTP

o  
h  
s  
e  
r  
m  
o  
t  
e  
n  
t  
i  
a  
t  
i  
o  
n  
s

- memory stored in synaptic strength
- learning changes strength (weight)

Give example



randomly connected network:

- output doesn't tell you much about input

after learning:

- output says "yes, that's a black dog in my visual field"

- What is synaptic strength?

- $\tau$  can increase
- $n$  can increase
- $g_0$  probably not
- $\bar{g}_{postsynaptic}$

img argument over which one changes  
- final answer: both.

- ~~how~~ <sup>what causes</sup> does it change?

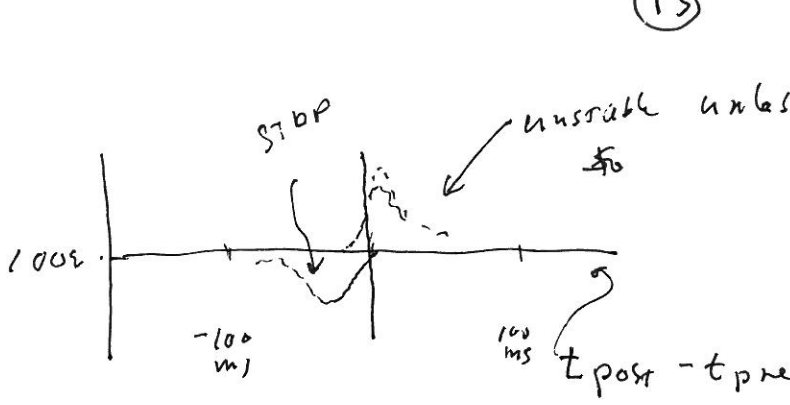
history dependent

↖ both pre-synaptic activity + post-synaptic activity.

- ~~how~~ does

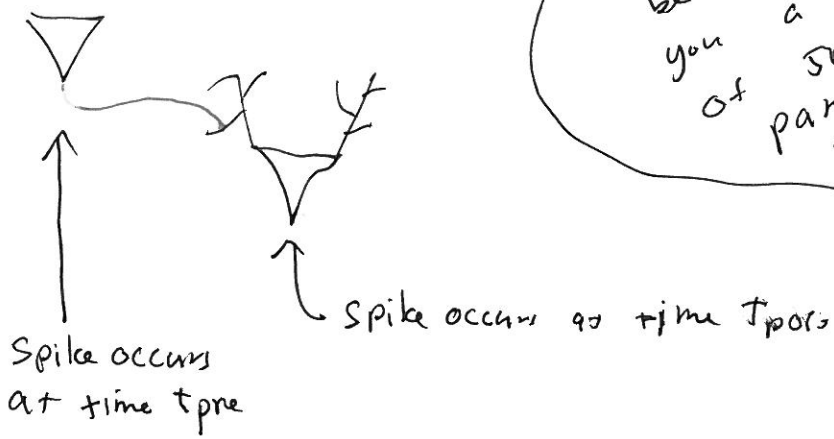
↘ near-simultaneous pre + post-synaptic spikes produce either a decrease or increase in strength.

(13)

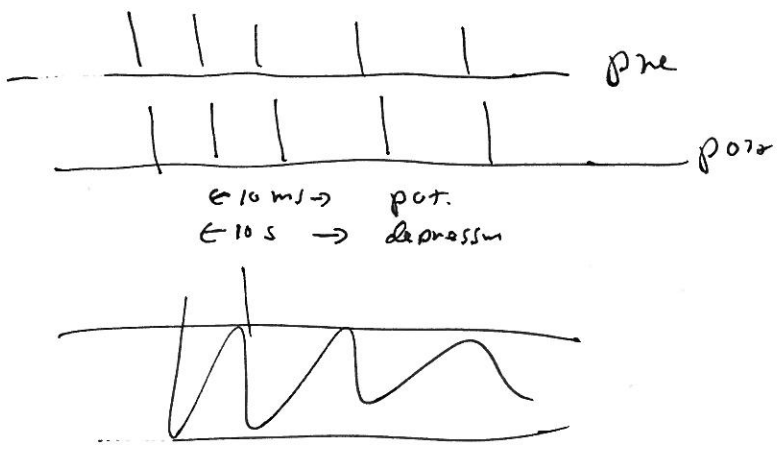


a normal rate is  
 $\sum \text{weights} = \text{const.}$   
 neuron

Important because it gives you a factor of 500 more parameters to play with.



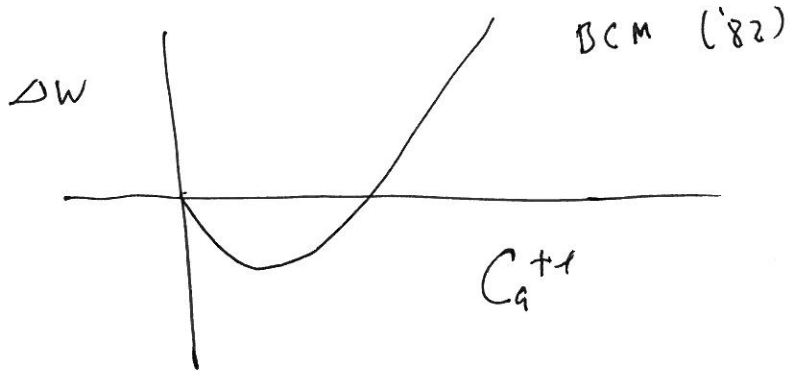
- Situation is more complicated:



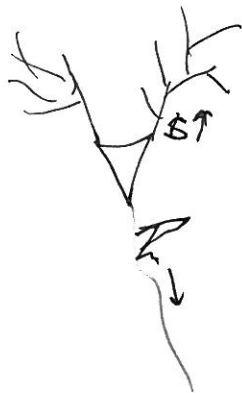
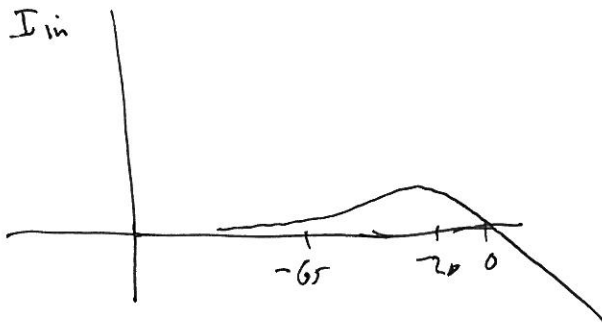
- nobody really know how all of this works in network
- Peter Dayan will provide more info.

# Mechanism

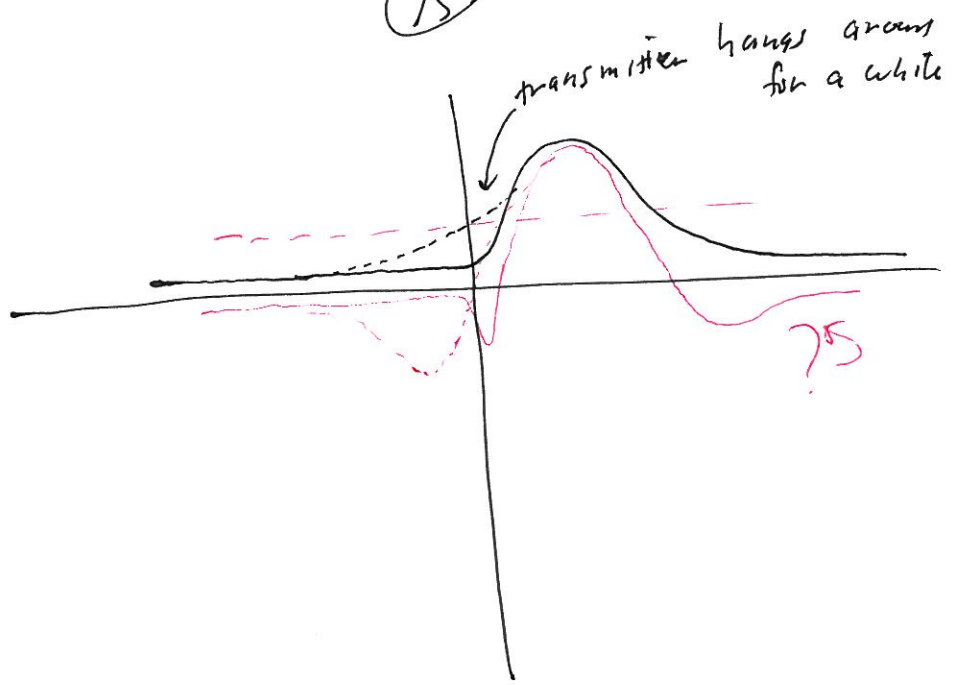
high  $Ca^{++}$  in pre



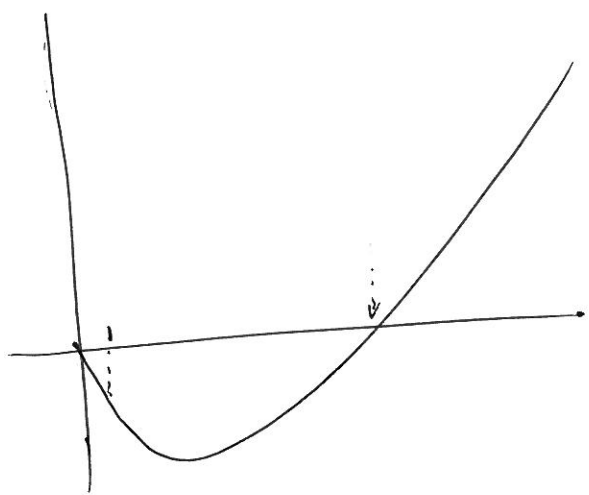
NMDA  
lets in  
 $Ca^{++}$ !



15



unsolved problem!



Be sure to do ML!