

Systems Neuroscience NOTES

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INTRO

Readings / Tests / Course Outline

Predispositions / folk theory vs sci see how things are

10%

cell doctrine

smell brain

sex diffs

L/R

Taking notes

attn / multi-task analogy

type vs write

drawing

Separate and

400 parts ...

repeat

3 planes / nissl-myelin

Golgi

Evolution

comparative function

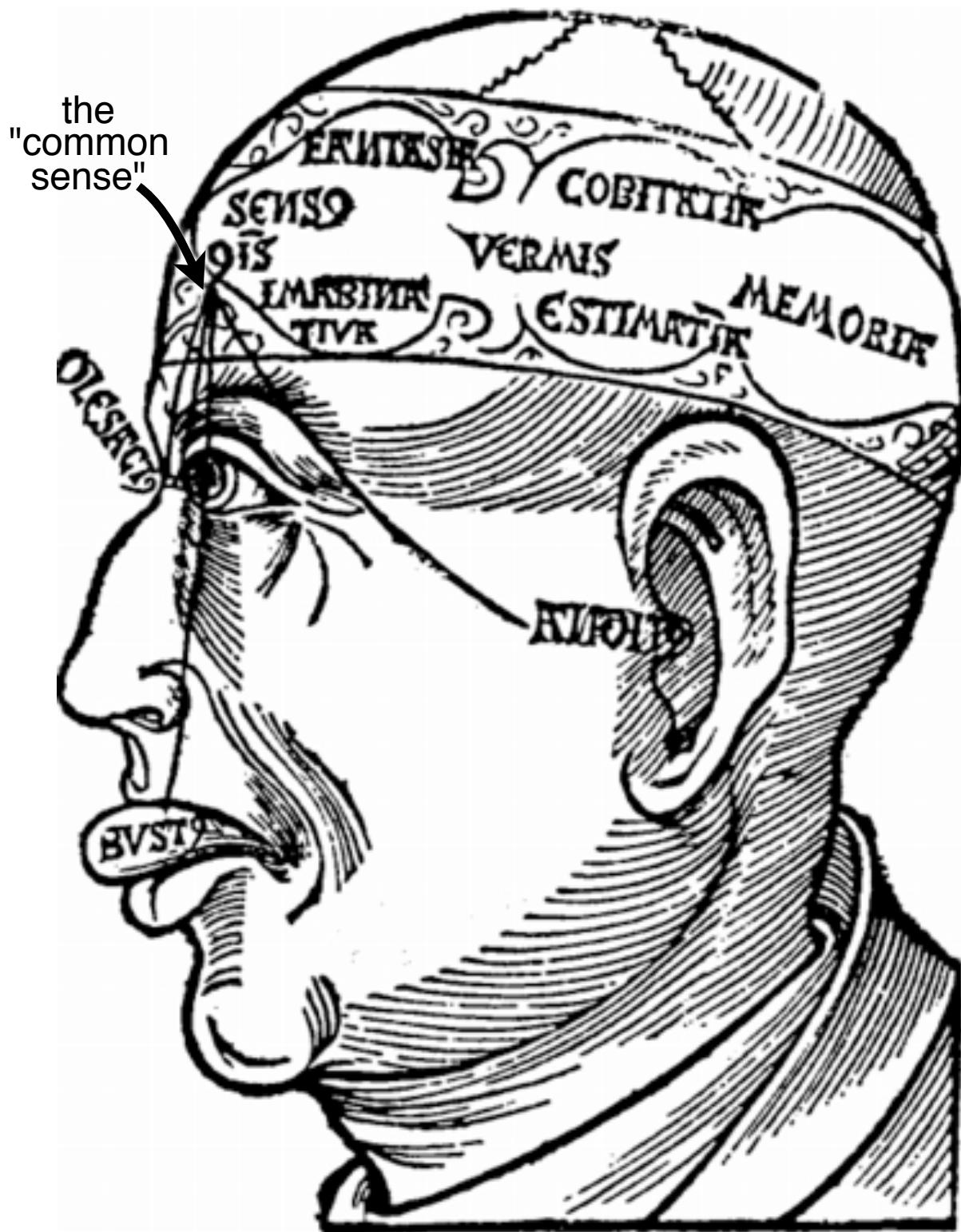
Machines ↗ not directly for machine learning

for engineer given biological raw material

- class summary
- signal / syst approach
machine learning (cf. multitask!)
vs.
engineer using bio material!
- folk theory
 - Aristotle (steam eng.)
 - Descartes (pineal cross bar)
 - *Cell Doctrine (cog sci)
 - Phrenology / Barnum
- modern folk theory
 - 10%
 - smell brain
 - sex diffs]
 - L/R diffs]

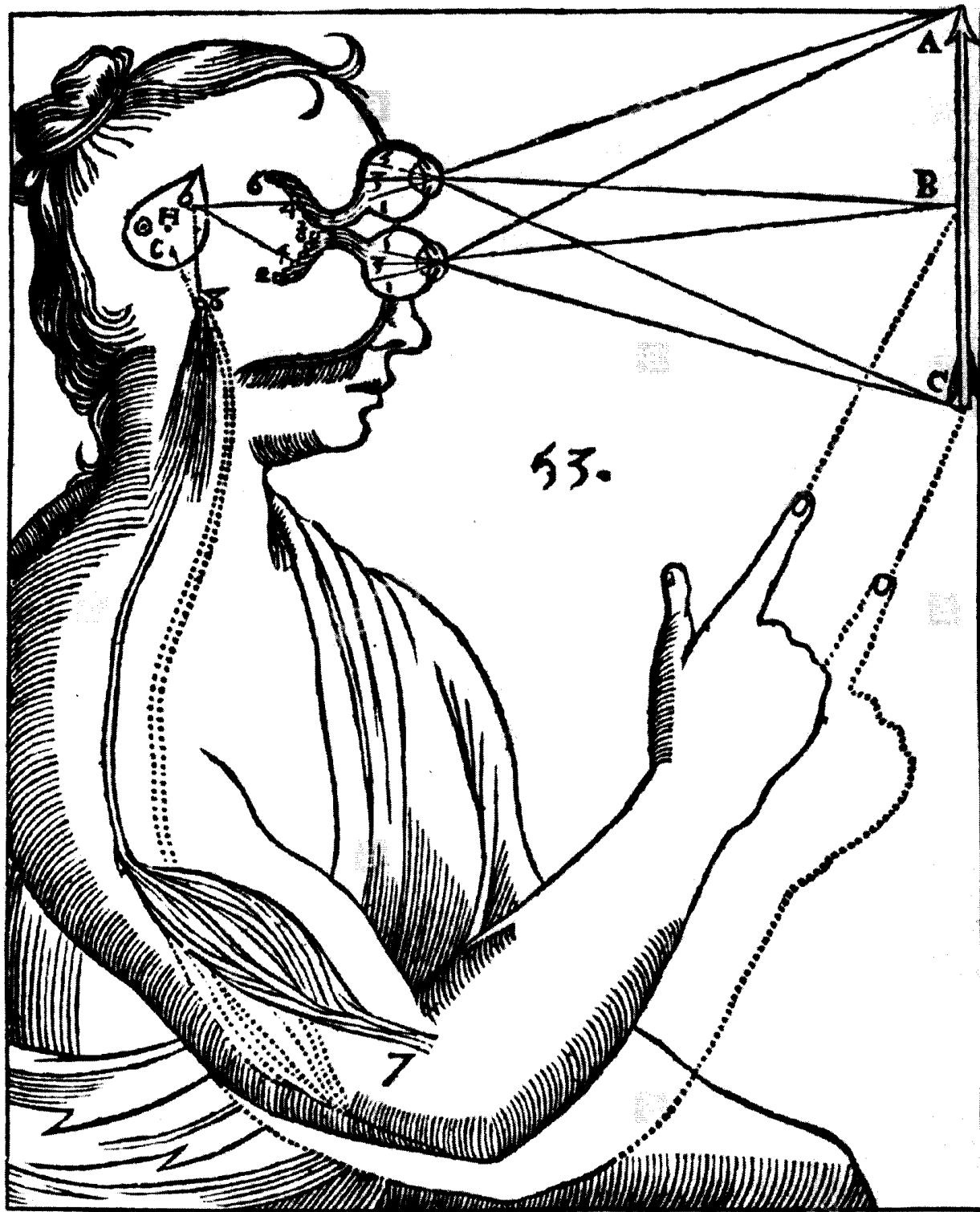
The "Cell Doctrine"

(default theory of human cognition in absence of any information about brain organization)



from Descartes

(a pneumatic crossbar switch in
the pineal controlled by the mind)

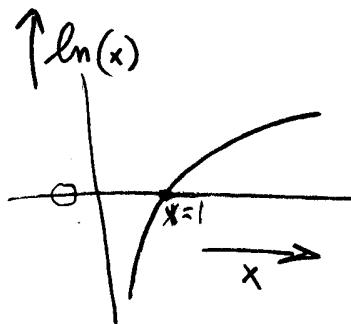


Nernst Potential

$$E_{\text{ion}} = \frac{RT}{F} \cdot \frac{1}{z} \cdot \ln \left[\frac{[\text{Ion}]_{\text{out}}}{[\text{Ion}]_{\text{in}}} \right]$$

→ Nernst eq: → reversal potential

natural log



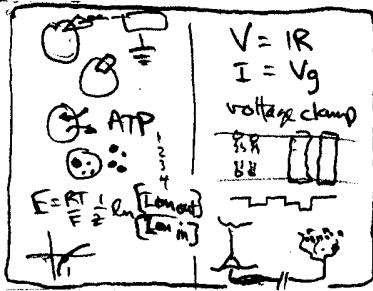
or
equilibrium potential

or
the membrane potential at which no current flows when channels for this ion are opened

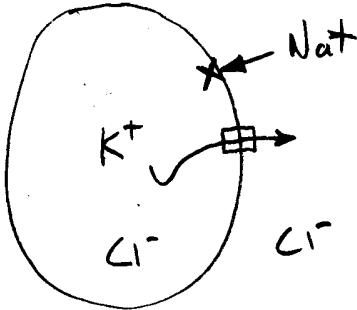
$\frac{RT}{F}$ gas const temp Kelvin
 $\approx 26 \text{ mV}$

Faraday const

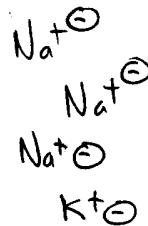
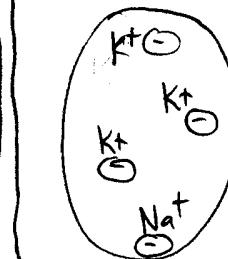
Summary



A



B



- start at 0, balanced
- a few K+'s leak out
- inside is negative

1 - differential concentration

2 - semi permeable membrane

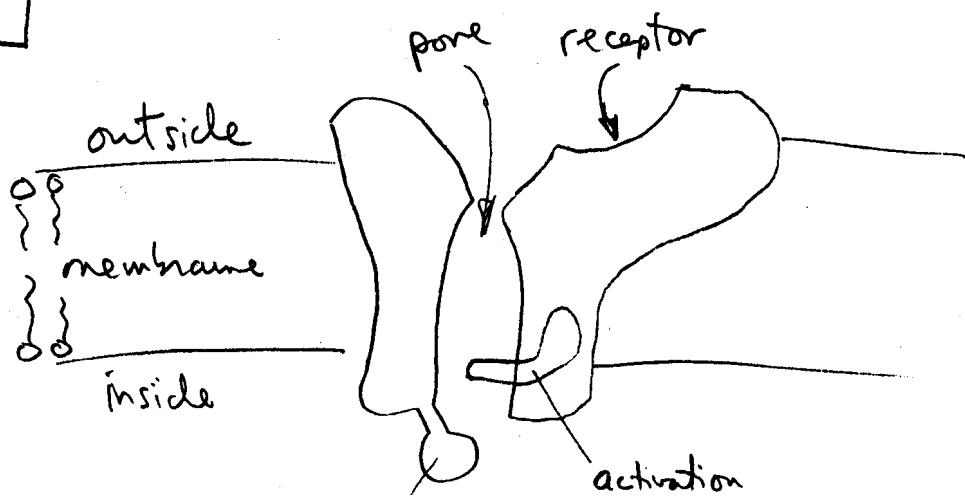
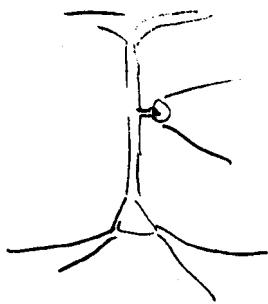
3 - resting potential is balance between diffusion & potential diff

4 - requires membrane channels but no voltage or neurotransmitter gating

Important Point:

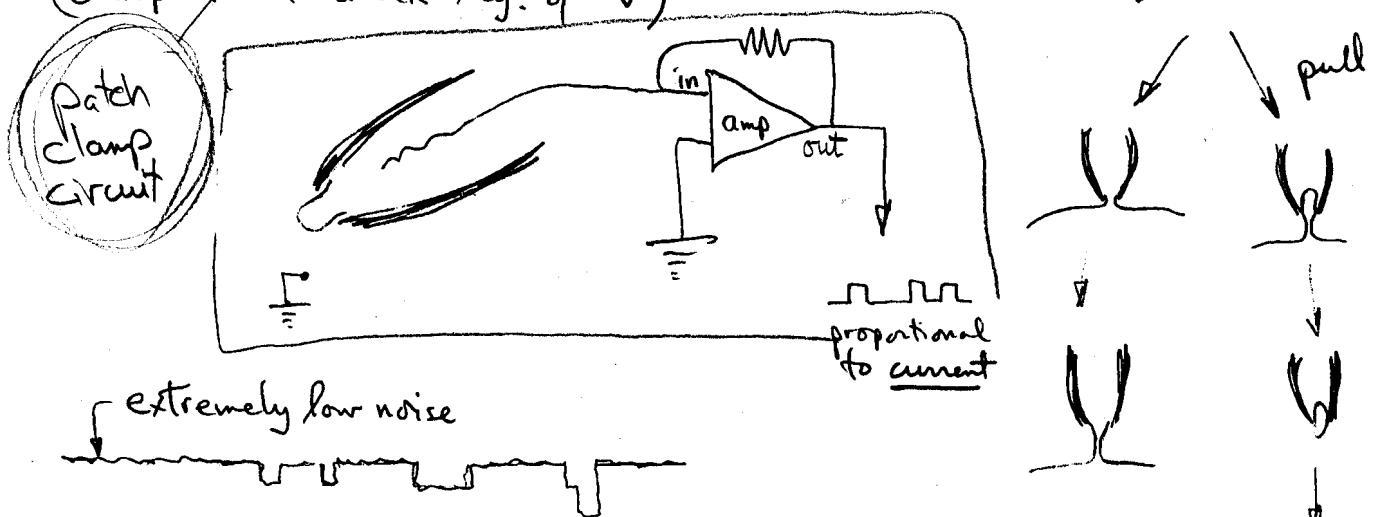
- Nernst potential describes a target voltage for an ion
- the present membrane potential is affected by other things!

CHANNELS



Patch clamp

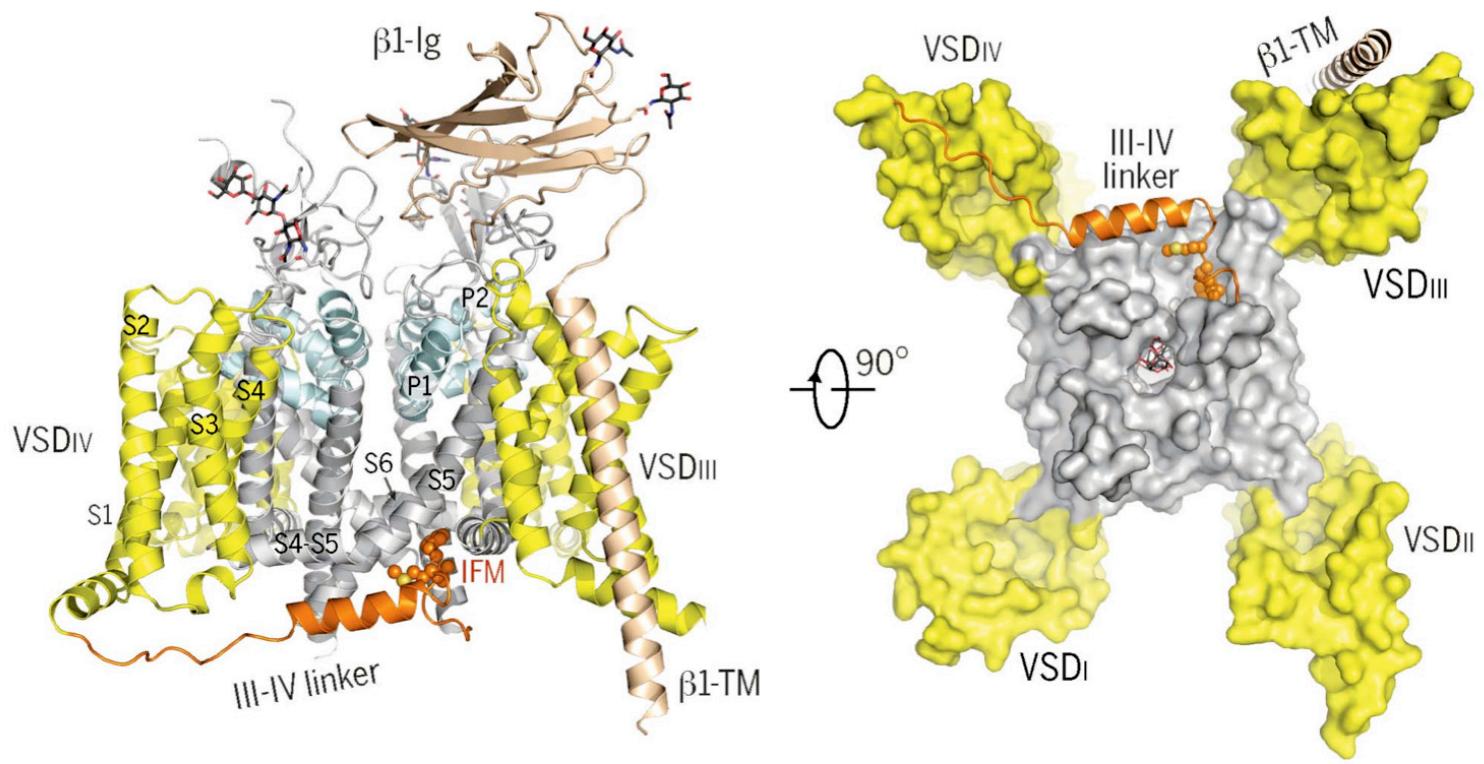
- electric current associated w/ single channel openings
- (clamp for feedback neg. sp. V)



- individual channel openings visualized
- 7000 ions/event ($10^7/\text{sec}$)
- conductance: $\sim 20 \text{ pS}$
- density: $100 - 1000 \mu\text{m}^{-2}$

S is Siemens
conductance (Siemens) = $\frac{\text{current (amps)}}{\text{voltage (volts)}}$

outside-out inside-out



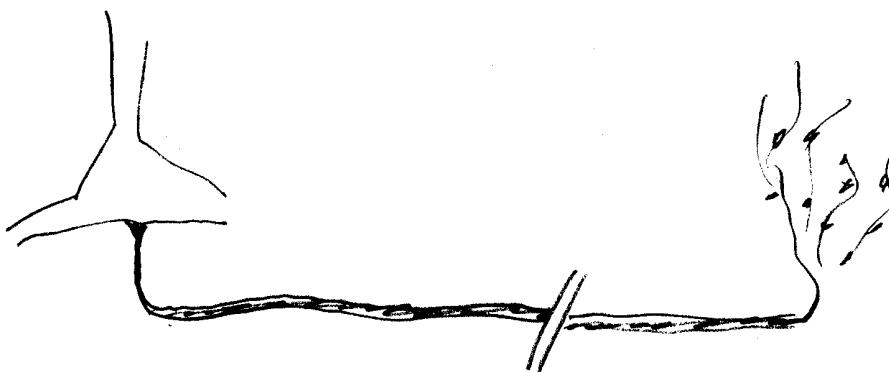
Structure of the human Na_v1.4-β1 complex. Two perpendicular views are shown. Left: Side view in ribbon cartoon. The VSDs are colored yellow, and the selectivity filter and supporting helices P1 and P2 are colored light cyan. The IFM motif is shown as spheres, and the III-IV linker is colored orange. The transmembrane segments in repeat IV are labeled. Right: Surface presentation for the bottom view to highlight the intracellular gate and the cavity that accommodates the IFM motif. The GDN molecule that penetrates the intracellular gate is shown as thin sticks.

Action Potential

- reason for action potential \Rightarrow passive flow dies out too quickly
- requires voltage-gated channels

[voltage-gated Na^+ (I_{Na})
voltage-gated K^+ (I_{K})]

Ohm's law
 $V = IR$
 $V = I/g$
 $I = Vg$

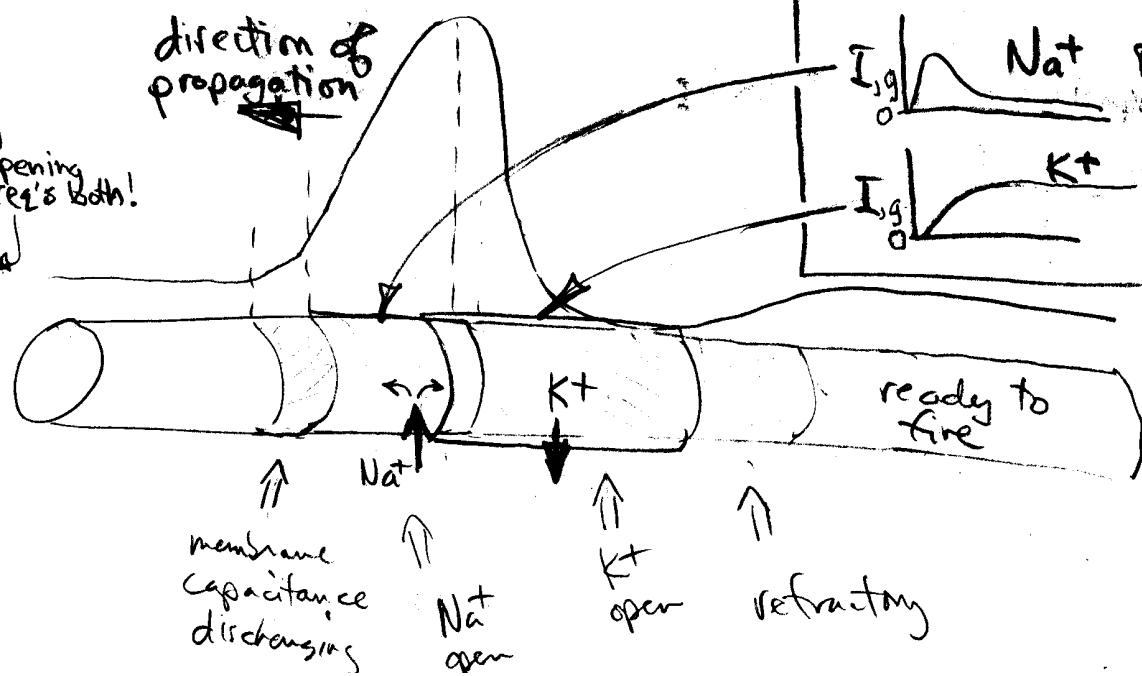
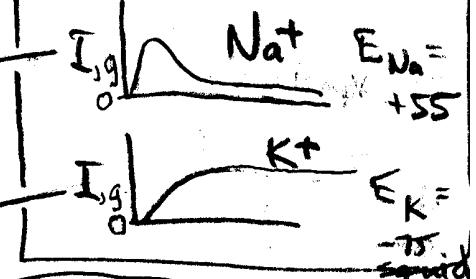
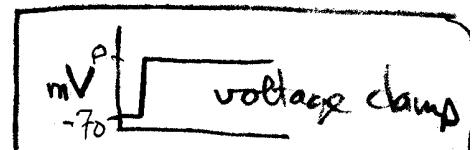


- cf. retina
- local pot. can cause NT release

[actual]

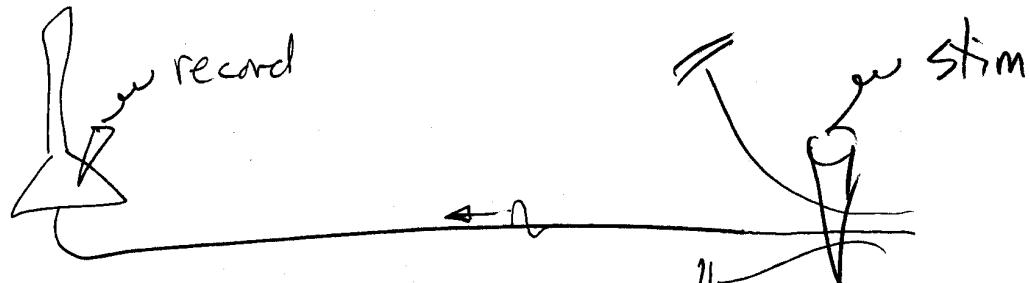
activation
deactivation
opening req'd both!
inactivation
deinactivation

direction of propagation



Collision test to verify antidromic

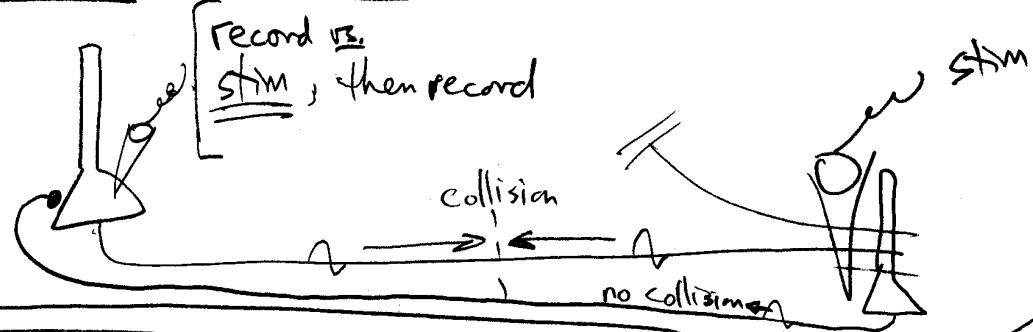
antidromic



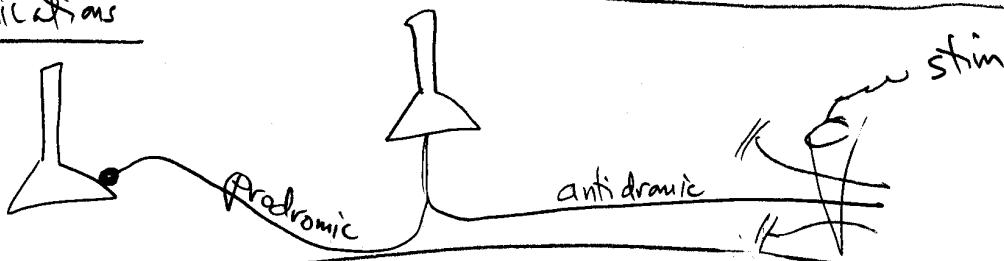
but could also be an afferent



Collision test

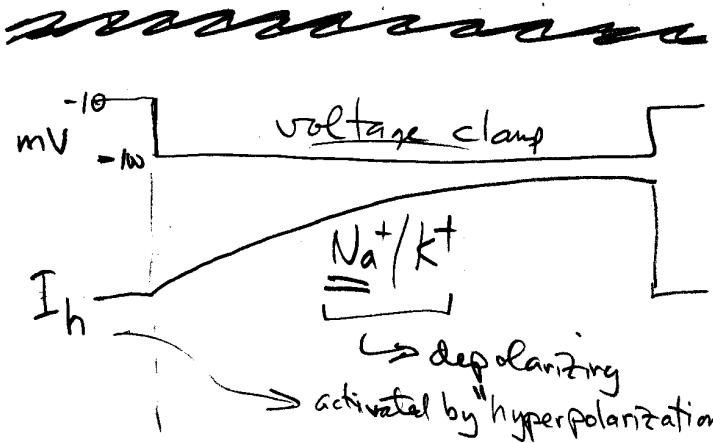
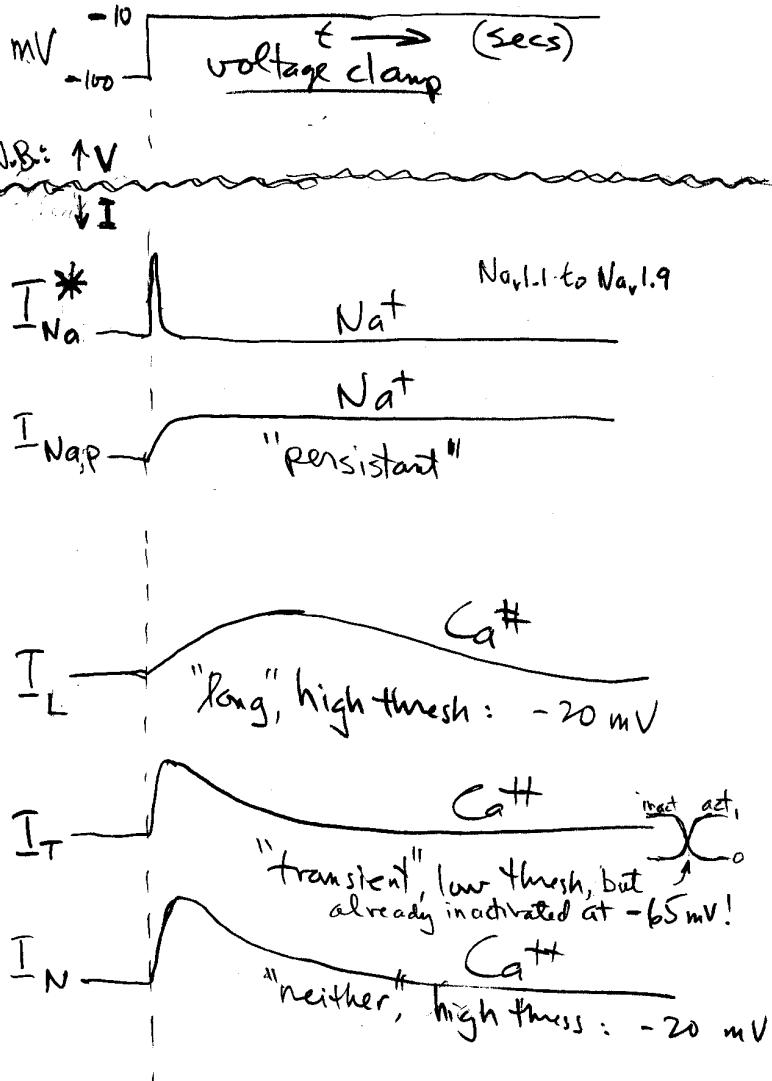


more complications

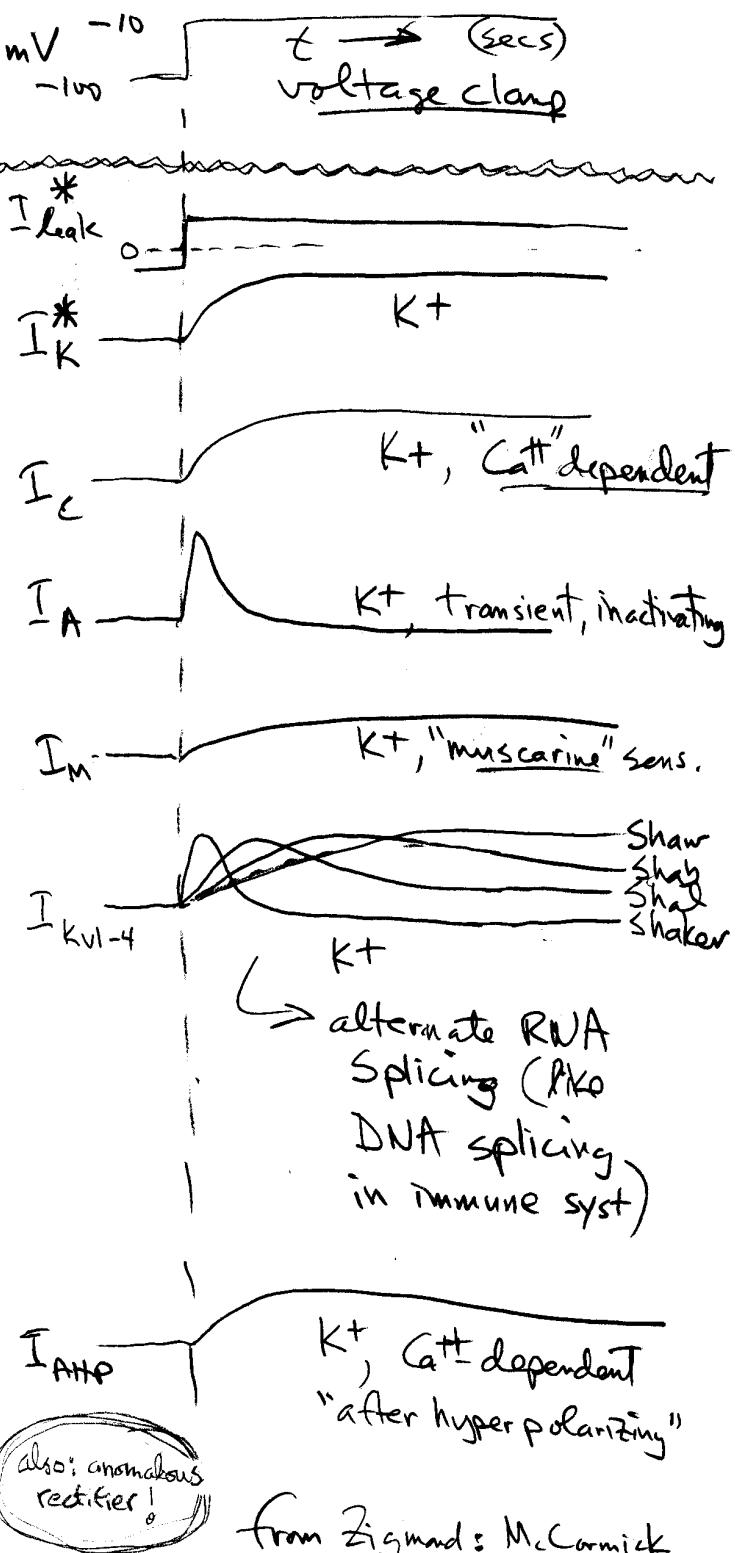


Voltage-Sens Other, Neuronal Ionic Currents

Excitatory ("inward") Na^+ currents



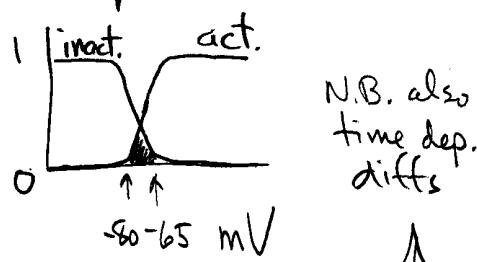
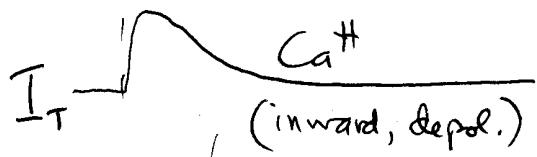
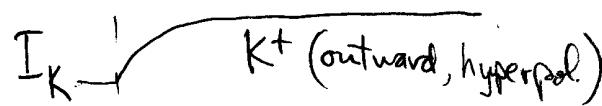
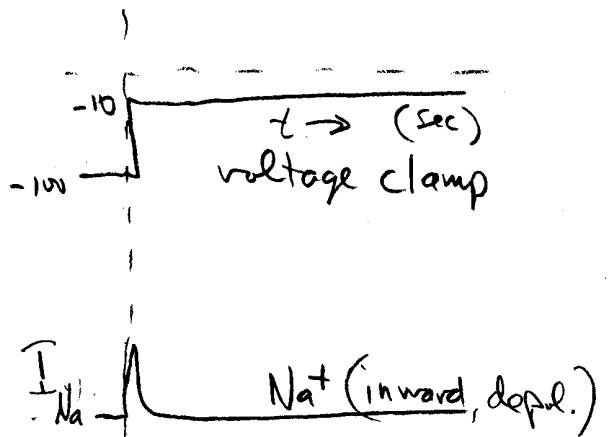
Inhibitory ("outward") K^+ currents



from Zigmund & McCormick

How I_T ("transient") leads to bursting

voltage clamp

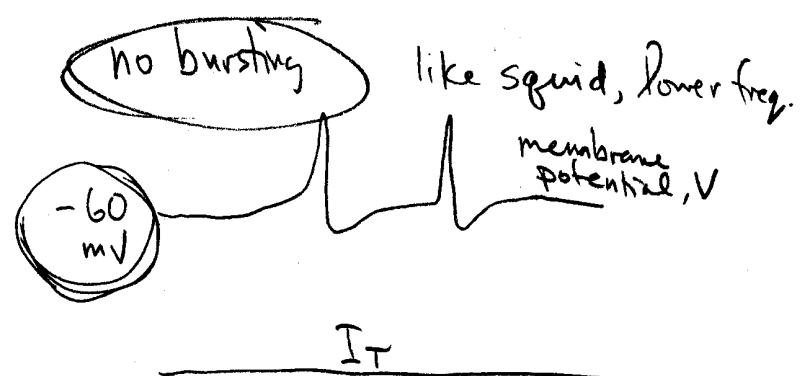
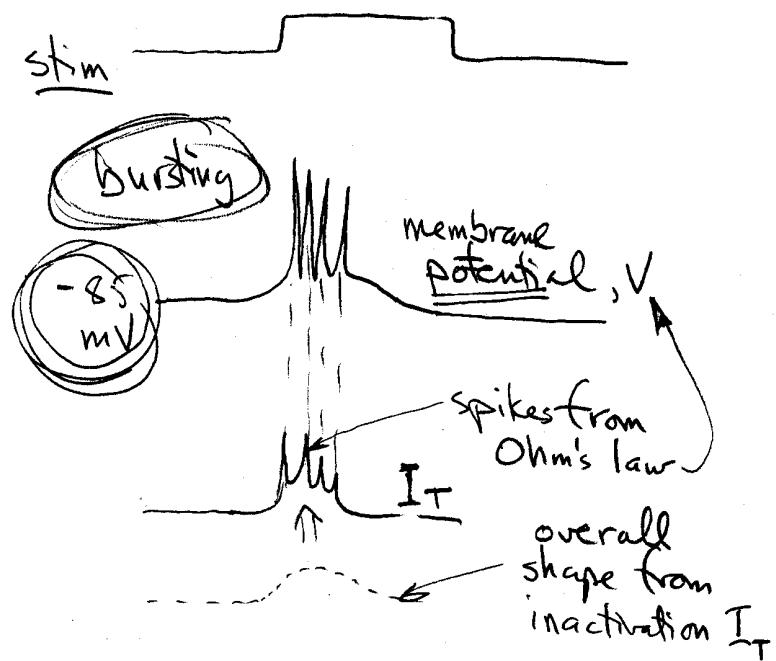


H-H

V.S. I_{Na} , higher volt before inact.

no voltage clamp

simple stim + 2 diff bias

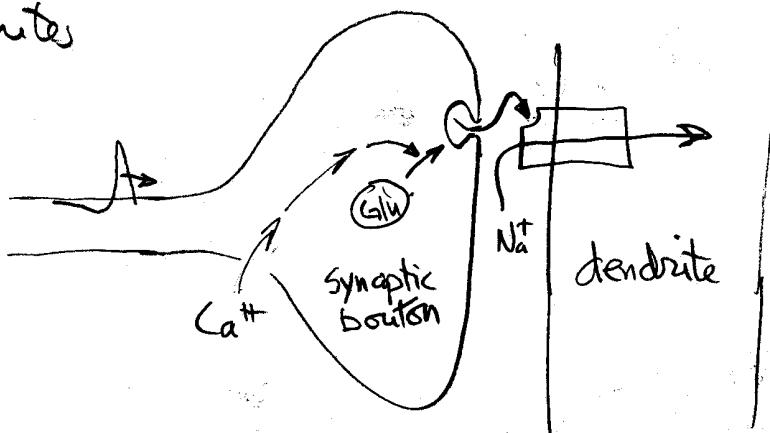
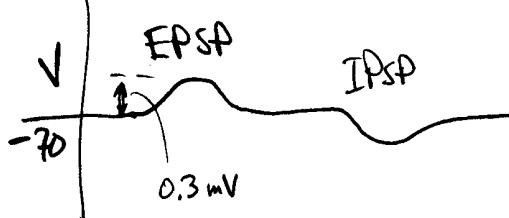


Another example:

I_A (outward) can cause delayed firing (firing starts after I_T inactivates)

Synaptic Potentials

- caused by neurotransmitter-gated channels
- mostly on dendrites



AMPA glutamate-gated Na^+

- "excitatory sodium"
- passes K^+ as well
- $E_{\text{Na}} \approx 0 \text{ mV}$
- not voltage-sensitive

NMDA glutamate-gated $\text{Na}^+, \text{Ca}^{++}$

- excitatory
- also passes Ca^{++}
- voltage-dependent
- blocked at rest

GABA_B K^+

↳ note: if resting pot. due entirely to K^+ , then these would not be hyperpolarizing

- hyperpolarizing inhibitory
- slower than AMPA

$$E_K \approx -100 \text{ (mammalian)}$$

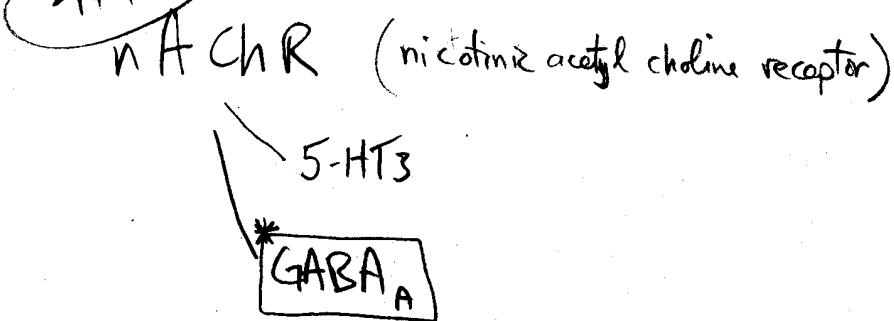
GABA_A Cl^-

- shunting inhibitory
- no effect at rest

$$E_{\text{Cl}} \approx -75 = \text{rest}$$

Channel Families

(4+1)



(4) Gi/o/o/o/o

non-NMDA =

* NMDA =

(7)

GPCRs - G-protein coupled receptors

m AChR

adrenergic

dopamine → D₁, D₅D₂, D₃, D₄

SHT - 1, 3, 4

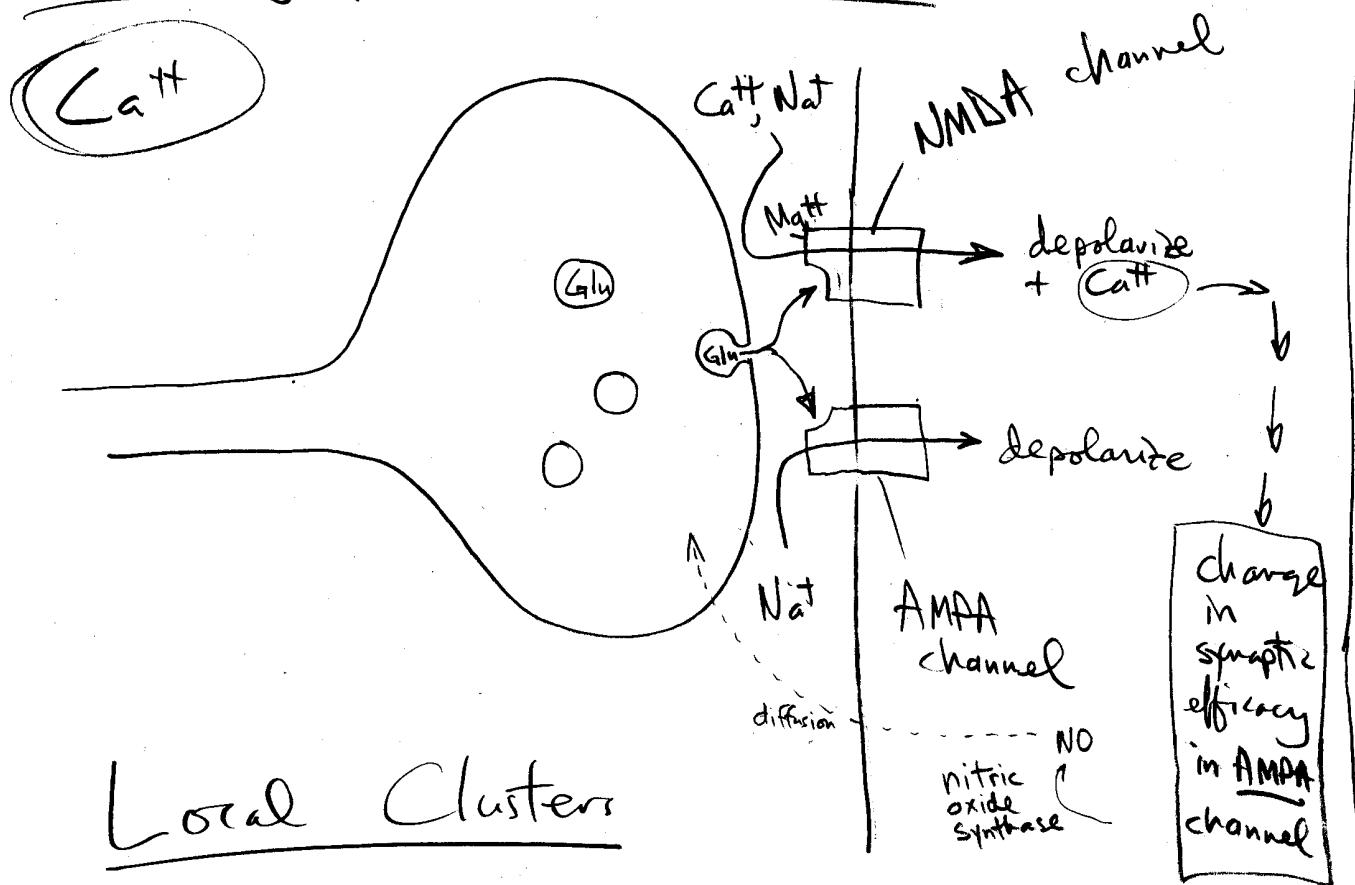
* glutamate

* GABA_B

neuropeptide

NMDA channels

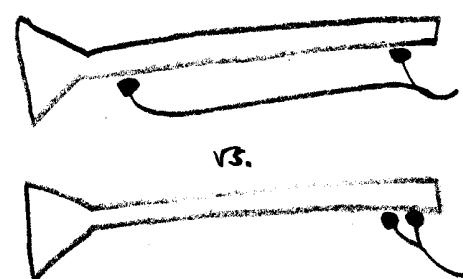
Detecting pre-post correlation



Local Clusters

$\text{Ca}^{++}, \text{Na}^{+} \Rightarrow$ voltage change
in cortex but not hippocampus \Rightarrow

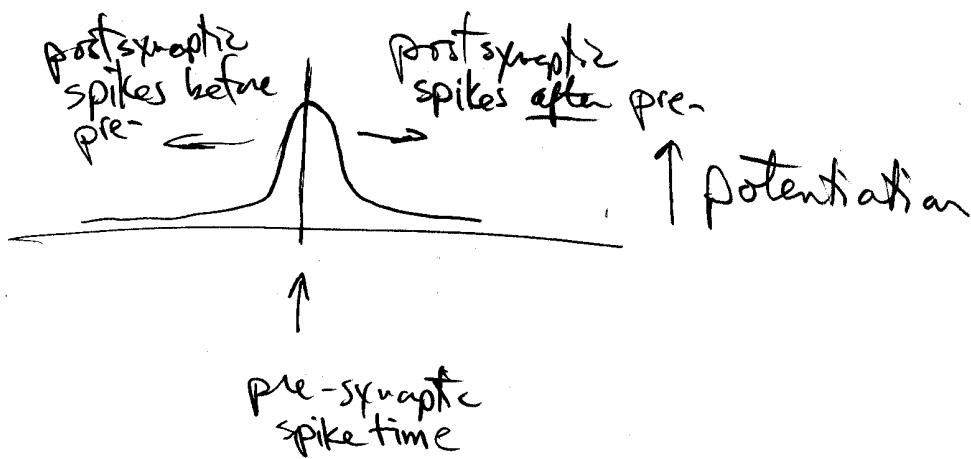
↓
synaptic strength change



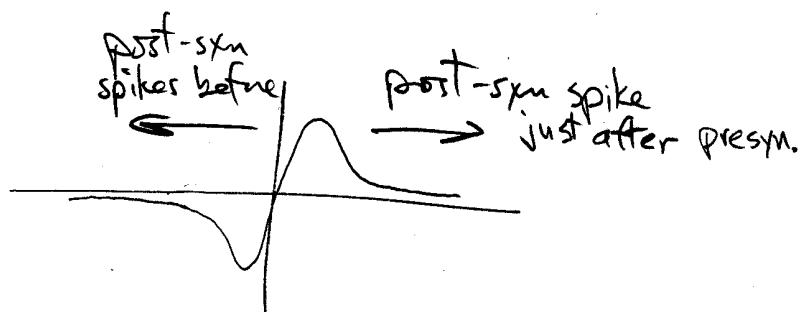
- (e.g. this one!)
- N.B. one synapse PSP not enough depol. for NMDA
- detect Corr. across synapses

Spike-timing Dependent Plasticity

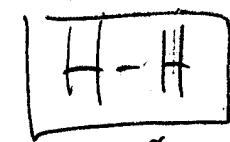
Old Idea



STDP



- idea: that synapses detect causality: if input caused cell to spike ↑
- and if cell spiked but input just missed helping out, ↓
- diff by +/- 100 msec → no effect



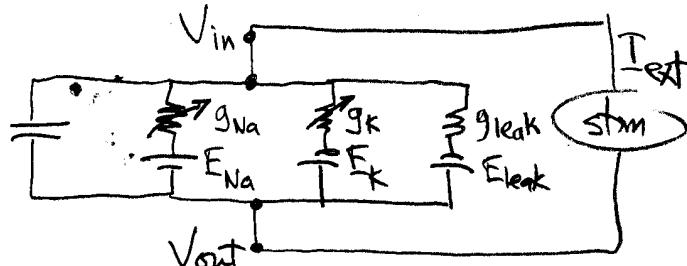
$$\frac{d\alpha_m}{dt} = \frac{\alpha_m}{T_m} (1 - \alpha_m) - \beta_m \alpha_m$$

$$\frac{d\beta_m}{dt} = \beta_m \alpha_m - \frac{\alpha_m}{T_m}$$

$$\frac{d\alpha_h}{dt} = \frac{\alpha_h}{T_h} (1 - \alpha_h) - \beta_h \alpha_h$$

$$\frac{d\beta_h}{dt} = \beta_h \alpha_h - \frac{\alpha_h}{T_h}$$

$\alpha_m, \beta_m, \alpha_h, \beta_h \approx 0 \Rightarrow \text{open!}$



$$C_m \frac{dV_{in}}{dt} + I_{ions} = I_{ext}$$

channels gating vars Ohm's law: $I = gV$

$$I_{ions} = \max(g_{ion}) \cdot m^2 \cdot h \cdot (V_{in} - E_{ion}) + \dots$$

other ions

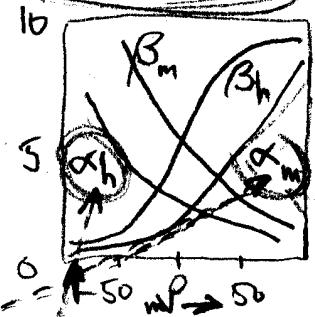
first order kinetics

$$\begin{aligned} \frac{dm}{dt}(t, V) &= \alpha_m(V)(1-m) - \beta_m(V)m \\ \frac{dh}{dt} &= \alpha_h(V)(1-h) - \beta_h(V)h \end{aligned}$$

rate const $(\frac{1}{msec})$

voltage dep
of fwd, backward
time consts
for m, h

* fit V -clamp data
get these curves



alternate way of writing

$$\frac{dm}{dt}(t, V) = \frac{m_\infty(V) - m}{T_m(V)}$$

steady state at partic. V

fit to data

$$m_\infty(V) = \frac{\alpha_m(V)}{\alpha_m(V) + \beta_m(V)}$$

time const

state of gating variables for partic. V at $t=0$

$$T_m(V) = \frac{1}{\alpha_m(V) + \beta_m(V)}$$

$= msec$

Solve simultaneously numerically integrate

$$\frac{dh}{dt}(t, V) = \frac{h_\infty(V) - h}{T_h(V)}$$

steady state rate const

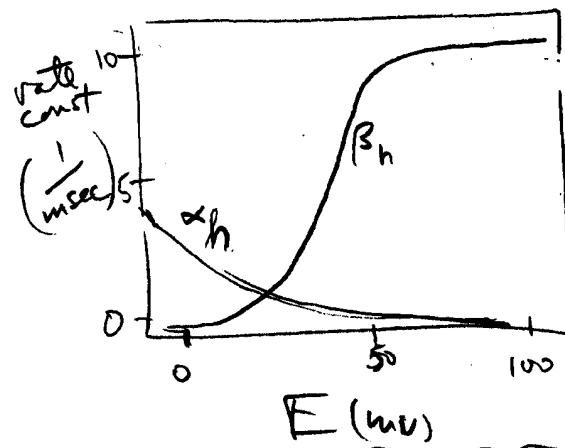
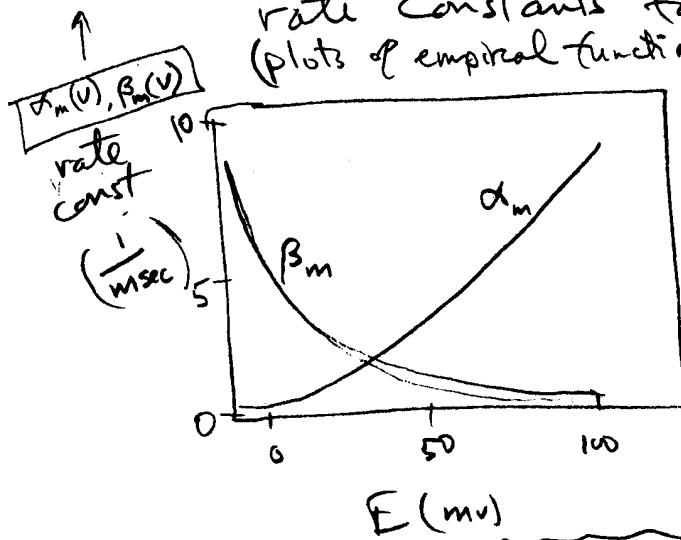
$$h_\infty(V) = \frac{\alpha_h(V)}{\alpha_h(V) + \beta_h(V)}$$

time const

$$T_h(V) = \frac{1}{\alpha_h(V) + \beta_h(V)}$$

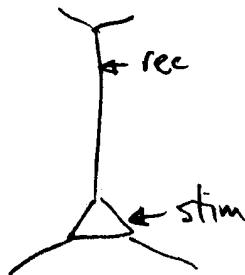
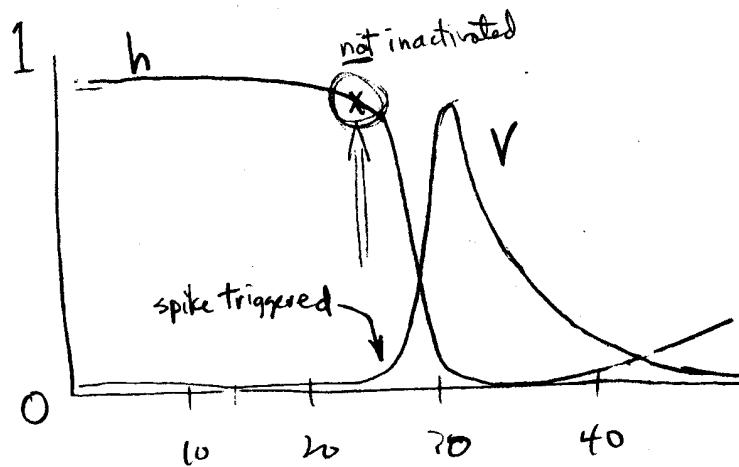
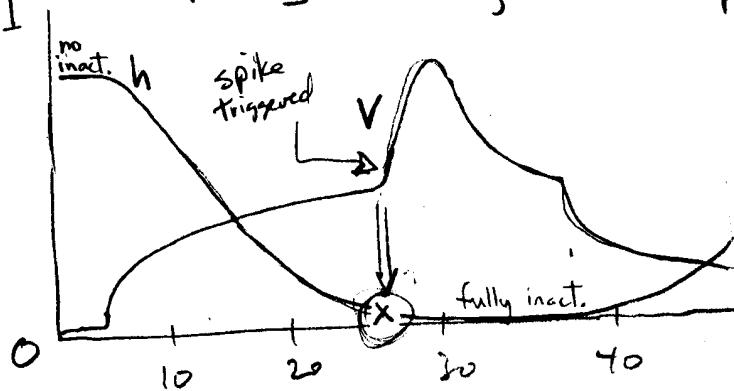
Add $\frac{dV}{dt}$
more eq's to
left space for graphs
redraft one more
time to separate
data fit m
in h

voltage dependence of forward & backward rate constants for Na activation & inactivation
(plots of empirical functions)



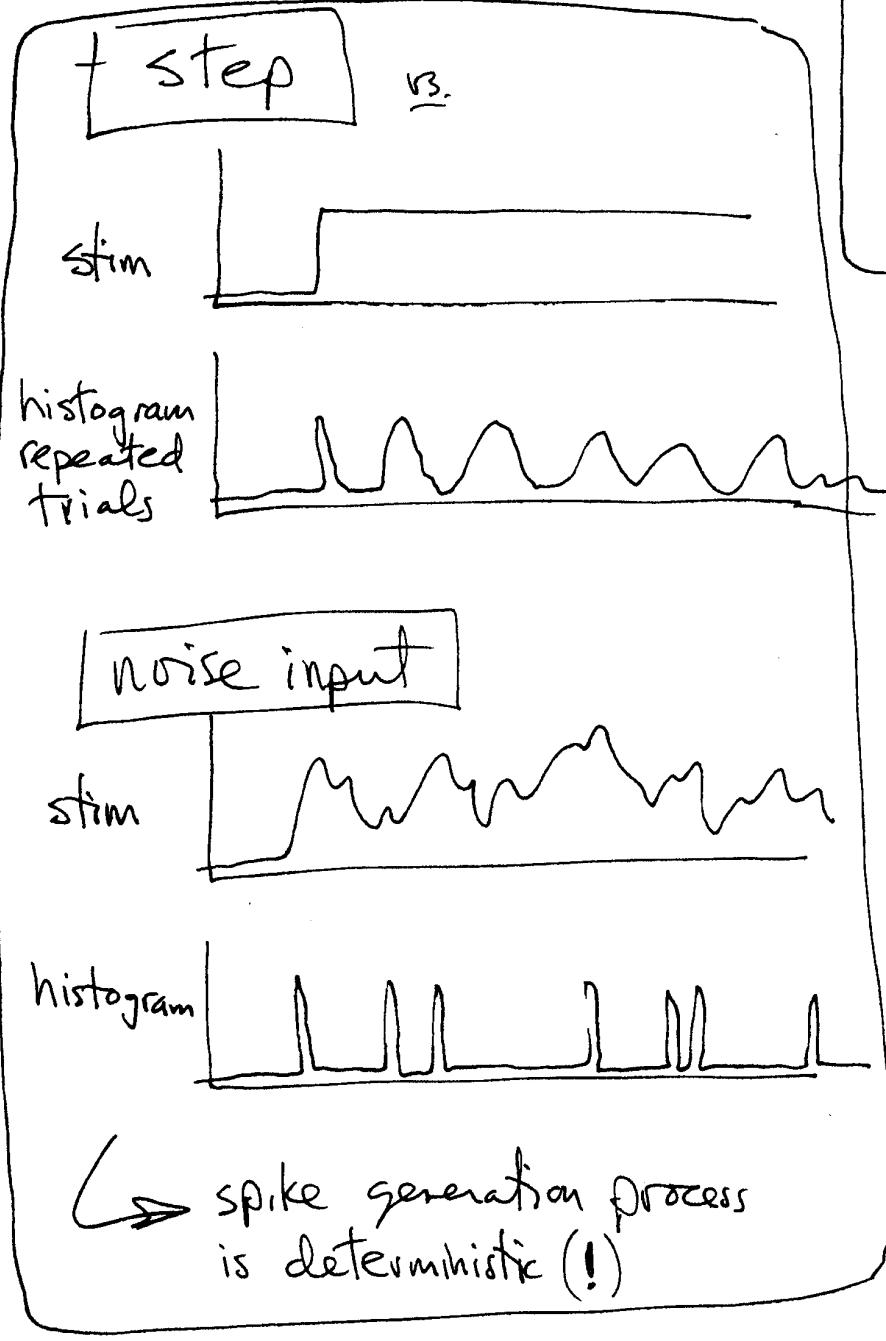
Why are spikes propagated backwards but not forwards in dendrites?

1 Nat inactivation, membrane pot.



Noise, Stochastic Processes, Firing Patterns

- cortical slices
- silent $\underline{\text{vs.}}$ background



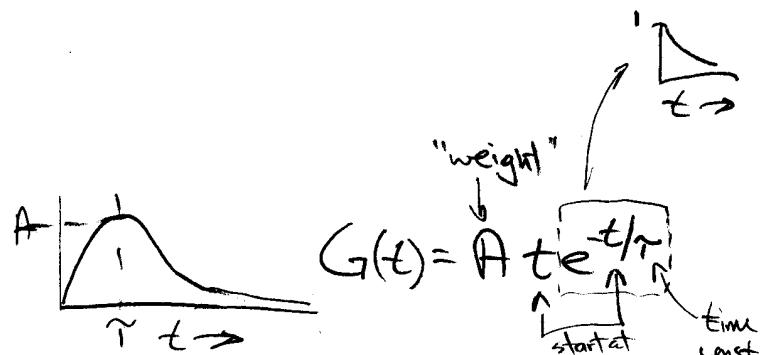
- Hodgkin-Huxley neuron
- 10,000 Poisson distributed inputs
- very regular firing

- real neurons have highly variable firing patterns in the cortex

- to simulate cortical variability in firing of stimulated neuron in slice, must inject bursty non-Poisson distributed spikes - filtered through PSP

Integrate & Fire

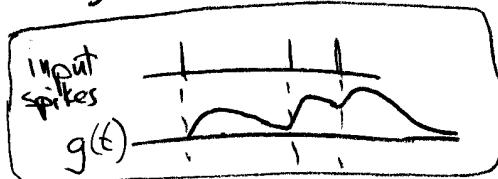
1) P.S.P. conductance waveform



2) Spike function (v3, H.H.)

$$S(t) = \begin{cases} 1 & \text{if } V(t) > \text{thresh} \\ & \text{and not refractory} \\ 0 & \end{cases}$$

3) Linearly add up conductance waveforms at one connection



$$g(t) = \int_0^{t-\text{delay}} G(\lambda) S(t - \lambda - \text{delay}) d\lambda$$

4) add up for all conductances like this $\hat{g}(t) = \sum_{\text{fibers}} g(t)$

$$5) \Delta V = \frac{1}{C_m} \sum_{\text{channels}} \left(V(t) - E_{\text{channel}} \right) \hat{g}_{\text{channel}}(t)$$

↑
next time step

↑
const

↑
current membrane pot

↑
reversal pot

↑
from above

(reversed in Wilson & Bower)

from:

- 1) Ohm's law $I = Vg$
- 2) $\Delta V = \frac{\sum I}{C}$

$\rightarrow \Delta V = \frac{1}{C} Vg$

Cable Theory \rightarrow Compartmental Models

$$V = \lambda^2 \left(\frac{\partial^2 V}{\partial x^2} \right) - \tau \left(\frac{\partial V}{\partial t} \right)$$

↑ space const. ↓ time const.

$$\lambda = \sqrt{\frac{r_m}{r_i}} = \sqrt{\frac{R_m/2\pi r}{R_i/\pi r^2}}$$

radius ↓ radius

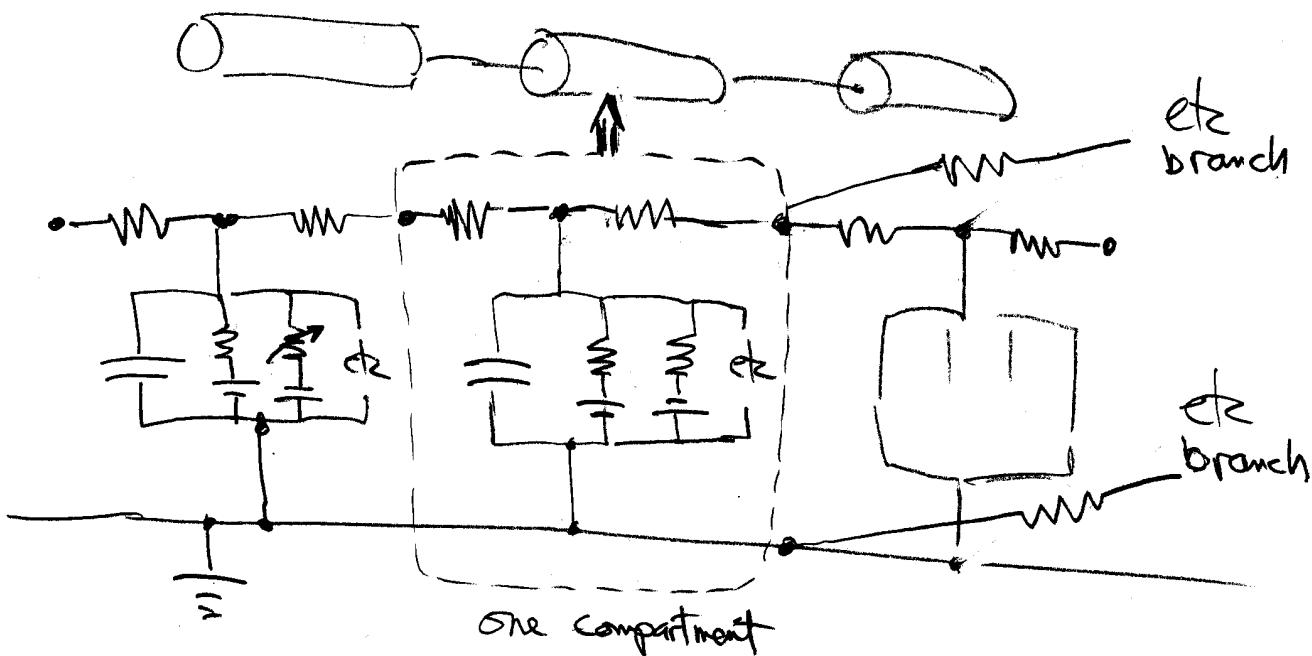
$$r_m C_m = (R_m/2\pi r)(C_m \cdot 2\pi r)$$

Cond vel:

$$\frac{dx}{dt} \propto \frac{\lambda}{\tau} \propto \sqrt{\frac{r}{R_i R_m C_m^2}}$$

Compartmental models

replace cont. partial D.E. eq w/ ordinary
 each compartment isopotential
 non-uniformities between compartments
 assume extracellular ground



circuits - 1

Definitions

Intro to Circuits

I - current ("amount of flow")

R - resistance

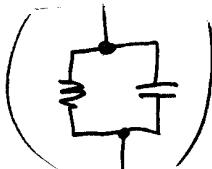
g - conductance ($= \frac{1}{R}$ = "diameter of pipe")

V - voltage, potential ("pressure")

C - capacitance ("spring loaded storage jar"; "balloon")

Q - charge ("what's stored")

Rules

V across parallel equal 

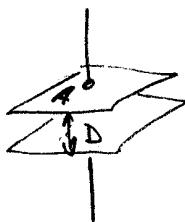
I across series equal 

It takes work (and finite time!) to change capacitors

$$V = IR \quad (\text{Ohm's law})$$

$$Q = CV$$

$$C \propto \frac{\text{Area}}{\text{Distance}}$$



Do this for each

a) hold one value constant

b) vary second

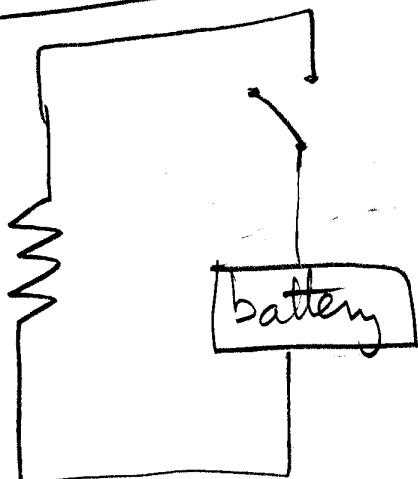
c) see how this affects third

$$\frac{dQ}{dt} = C_m \frac{dV}{dt}$$

circuits-2

Simplest Circuits

Resistor only



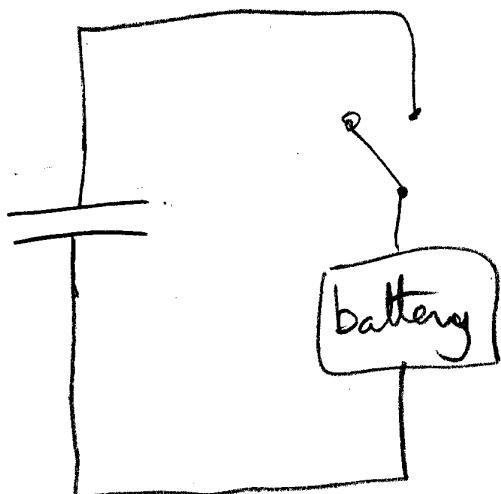
switch closing

output current
of source

I_R

V_R

Capacitor only



output current
of battery

I_C

V_C

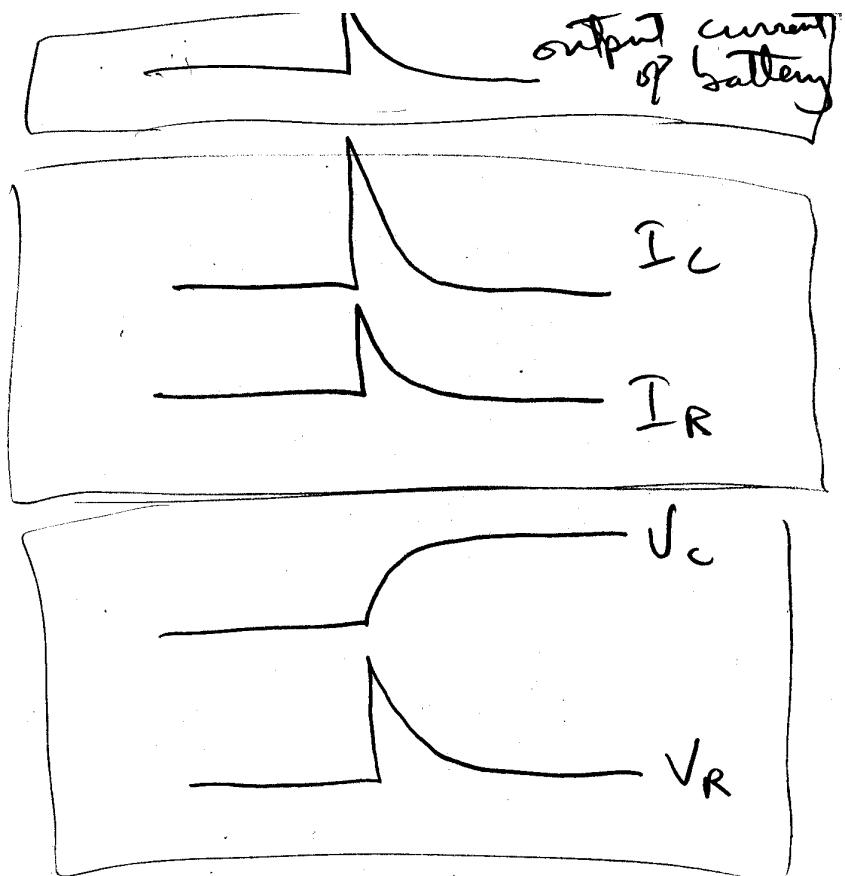
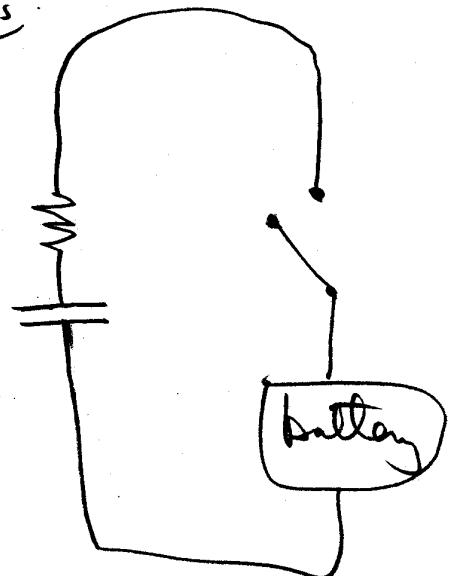
N.B.

- [] no resistance so current very high
- [] capacitor charges almost instantly
- [] a partly-charged capacitor still non-infinite effective resistance

Circuits-3

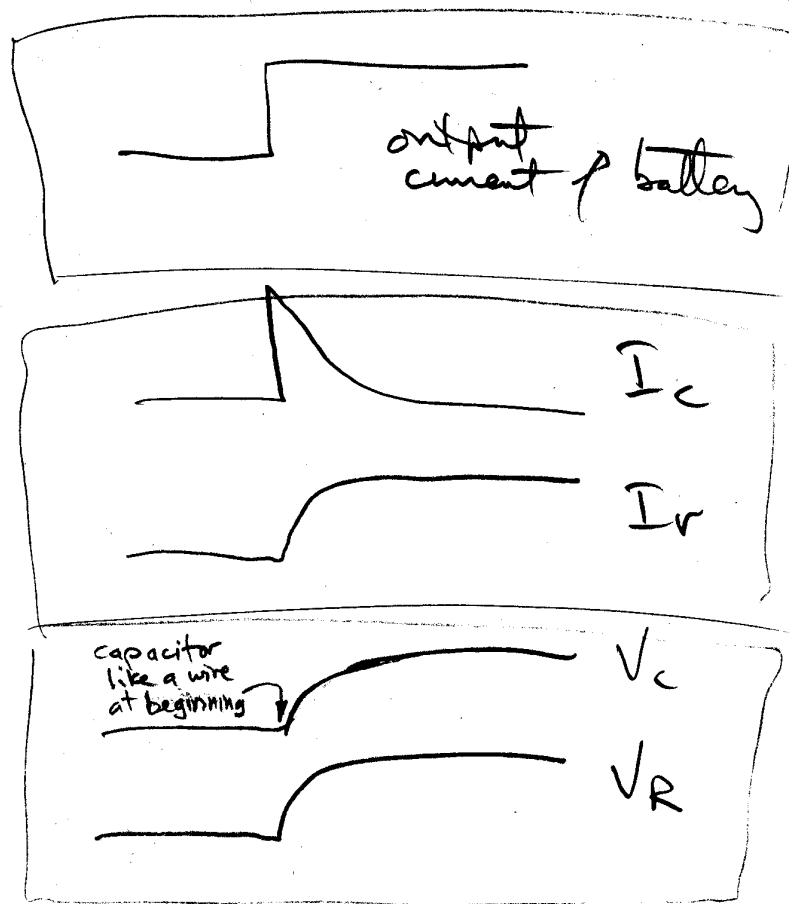
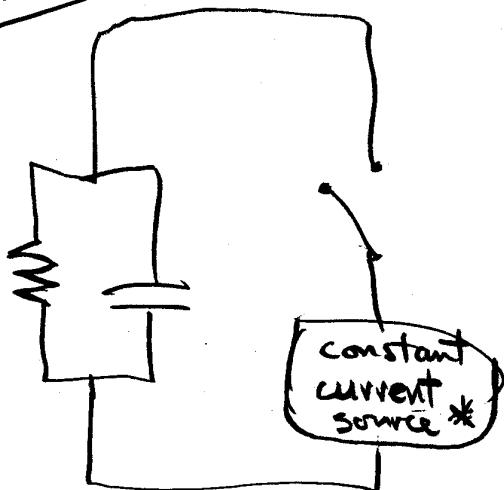
Simple Circuits

Series



- C charges slower because max voltage limited by R. (Voltage controls charging rate)

Parallel **



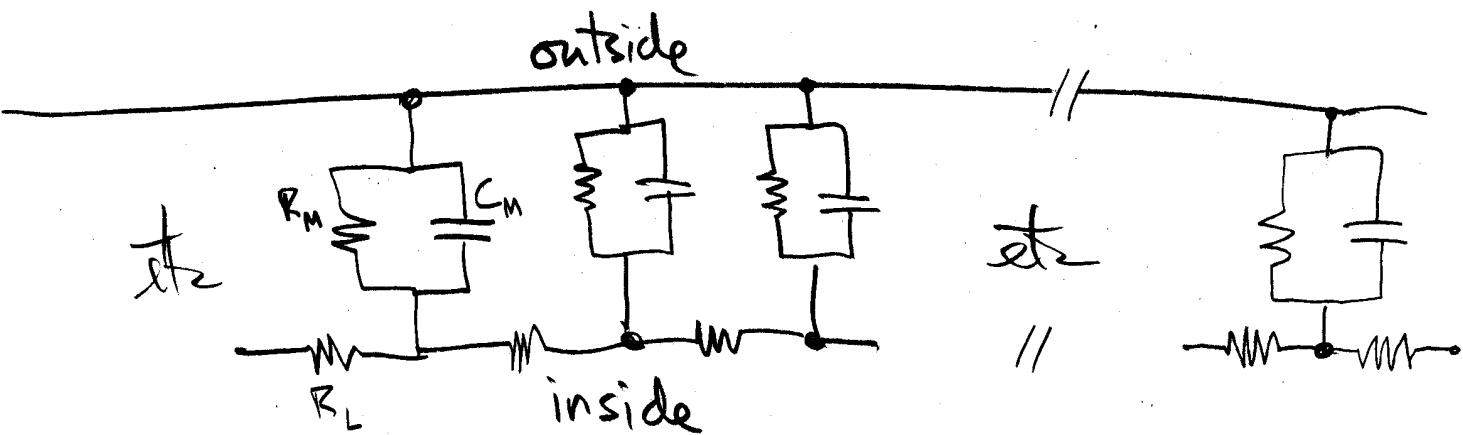
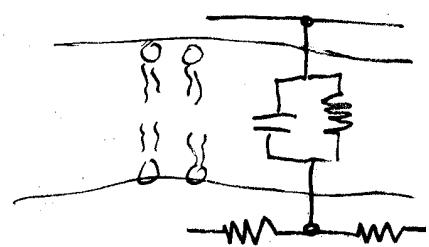
- ditto (w/ finite current source)
* if battery really big, would look instantaneous

circuits-4

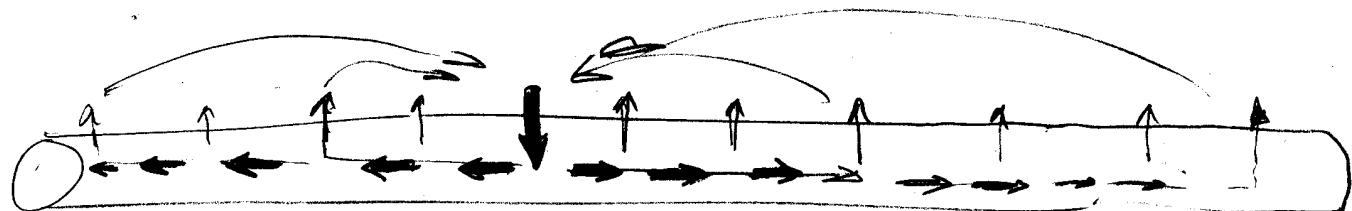
#2

Electrotomography / Dendrite Current Flow

- where do delays come from?
- cable theory (differential eq.)
- assume resting potential $\underline{0}$
- cable theory \rightarrow compartmental model

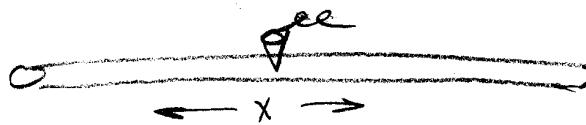


- why it's called cable theory $R_m \gg R_L$



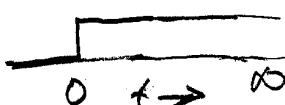
circuits-S

ON-STEP, OFF-STEP, SHORT PULSE IN SPACE & TIME

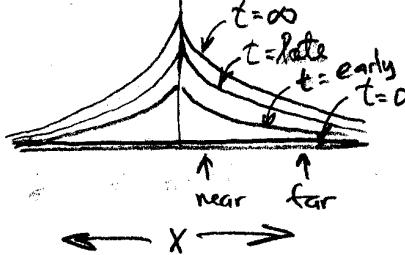


plot $\frac{V}{I}$ (conductance fixed)

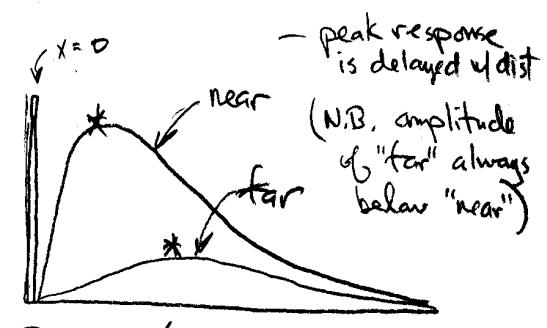
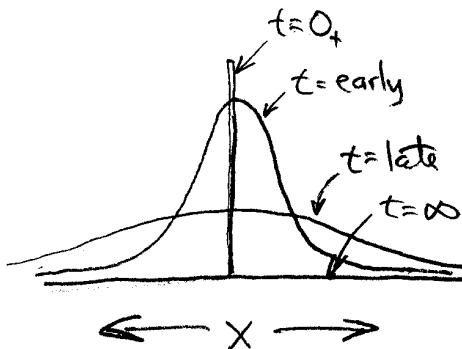
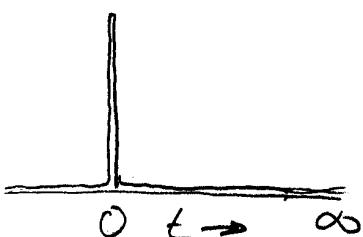
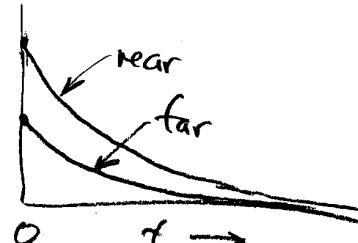
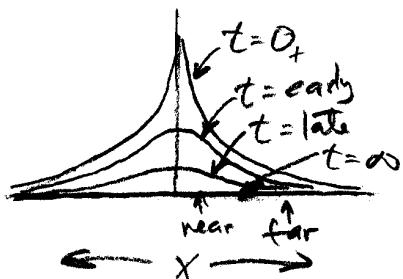
Input



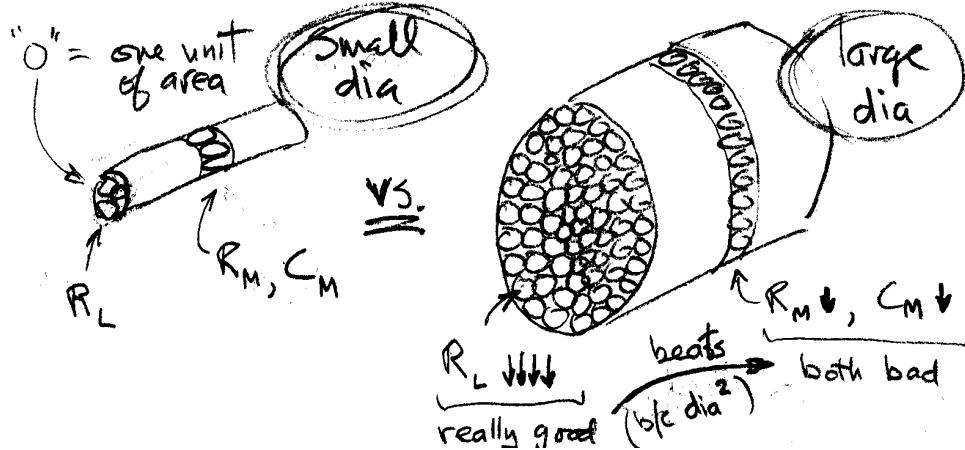
Space



Time



WHY DOES INCREASED DIA INCREASE COND. VELC.?



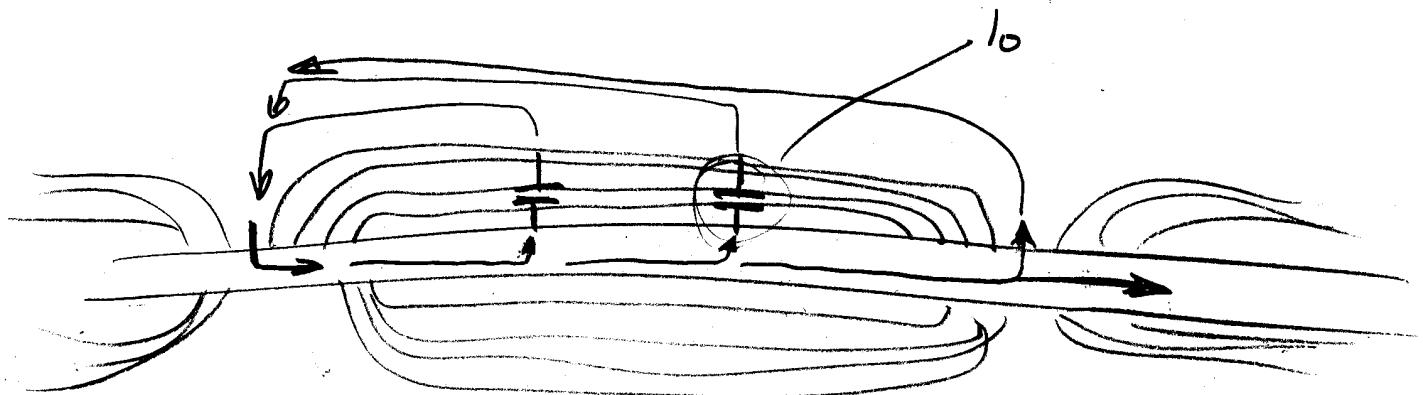
effects of expand area



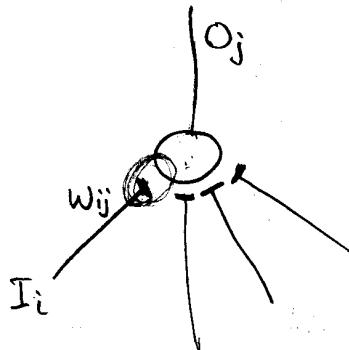
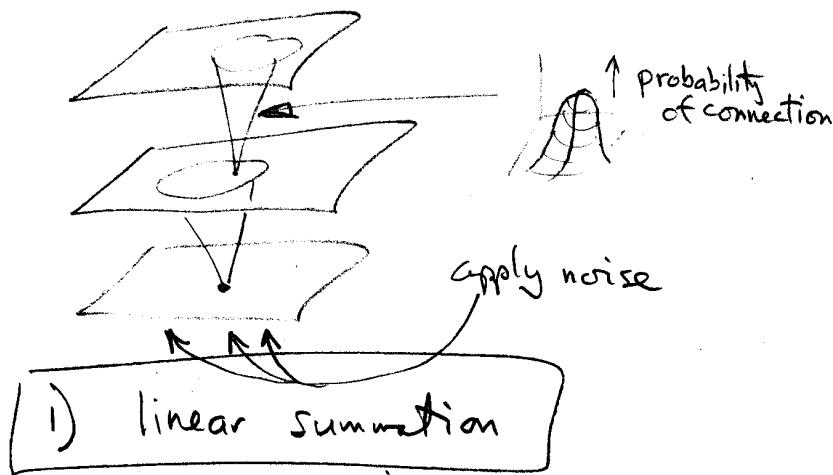
Circuits-6

Myelination

- myelin has low C
- conduction velocity of electronic pulse faster as result



Linsker (1) Basic Idea



2) simple Hebb

$$\Delta w_{ij} = \text{rate} * Q_{ij}$$

$$Q_{ij} = I_i * O_j$$

$$\Delta w_{ij} = \text{rate} * I_i * O_j$$

3) put together into one equation

$$\Delta w_{ij} = I_i \left(\sum_{\text{other}}^j I_i w_{ij} \right)$$

$$Q_{i,\text{other}} = I_i I_{\text{other}}$$

||

$$\Delta w_{ij} = \sum_{\text{other}}^j Q_{i,\text{other}} w_{\text{other},j}$$

||

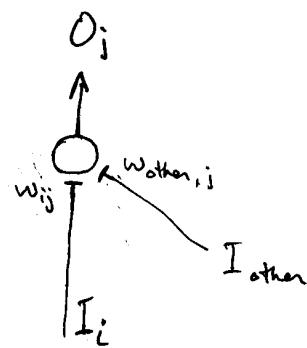
$$\Delta w_{ij} = K_1 + \sum_{\text{other}}^j (Q_{i,\text{other}} + K_2) w_{\text{other},j}$$

set no-change point

oneweight = $\underline{\text{Q vector}} \cdot \underline{\text{weight vector}}$

weight vector = $\underline{\text{Q matrix}} \cdot \underline{\text{weight vector}}$

("other" ranges over all the other connections to the unit but also includes current connection i)



Q examples

$$\begin{array}{r} 1 \times 2 \\ -2 \times -5 \\ 3 \times 1 \\ -4 \times 2 \\ \hline \text{Avg pos} \end{array}$$

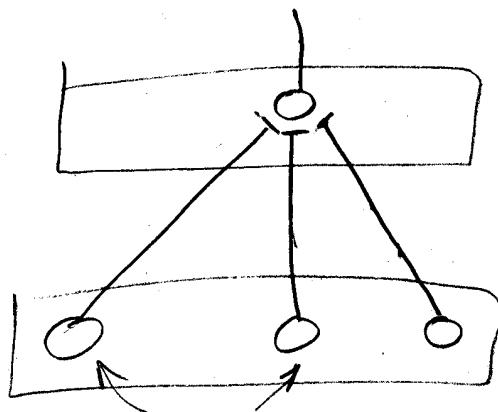


thus, Hebb synapses detect and amplify 2-point correlations in input lines

Linsker (2)

What happens in layer 2

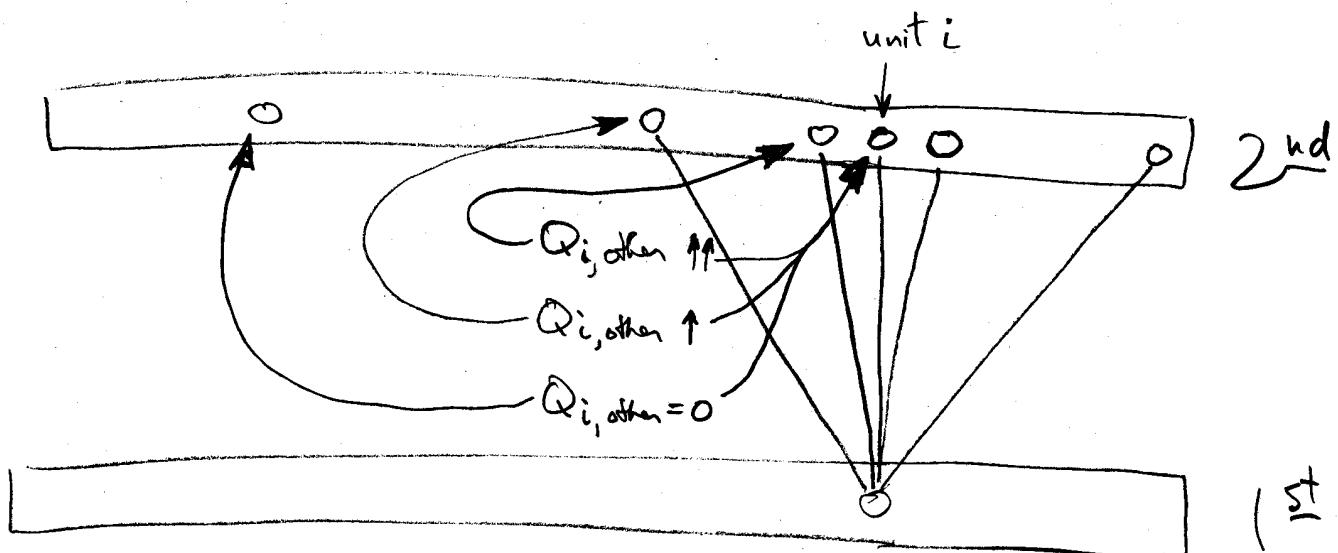
2nd layer input go all-neg or all-positive



$$\text{average } Q_{i,\text{other}} = 0$$

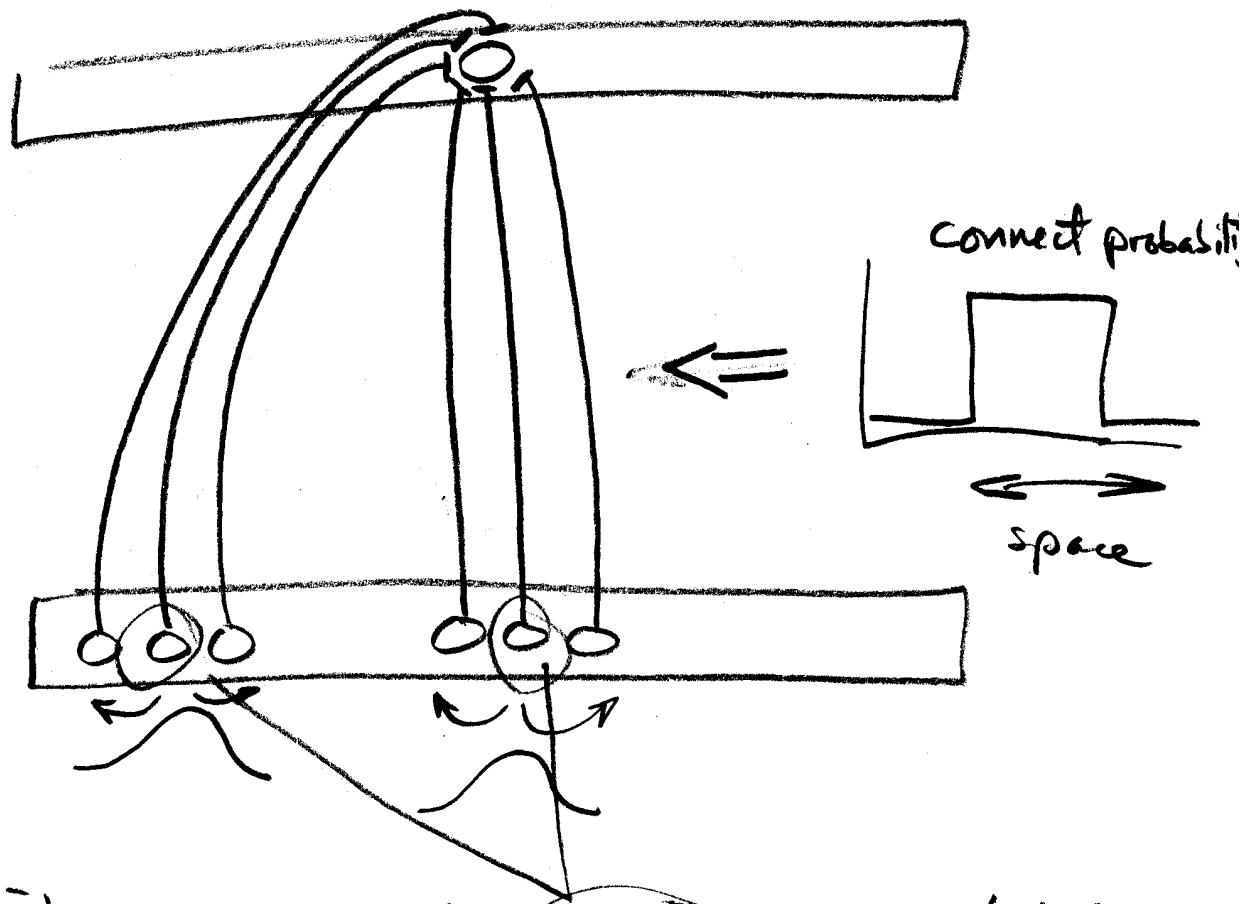
- as soon as average weight to a unit is slightly pos (or neg), Δw for current weight goes pos (or neg)
- the next cycle, tho $Q_{i,\text{other}}$ will still average zero, $\Delta w = \sum Q \cdot w_{\text{other}}$ will be more pos (or neg), since the other weights are more pos (or neg)

What endpoint $Q_{i,\text{other}}$ looks like for layer 2



Why the RF center weights go up

Counter example



- despite the fact that these two uncorrelated peripheral weights, will go up just as fast

Linsker(3)

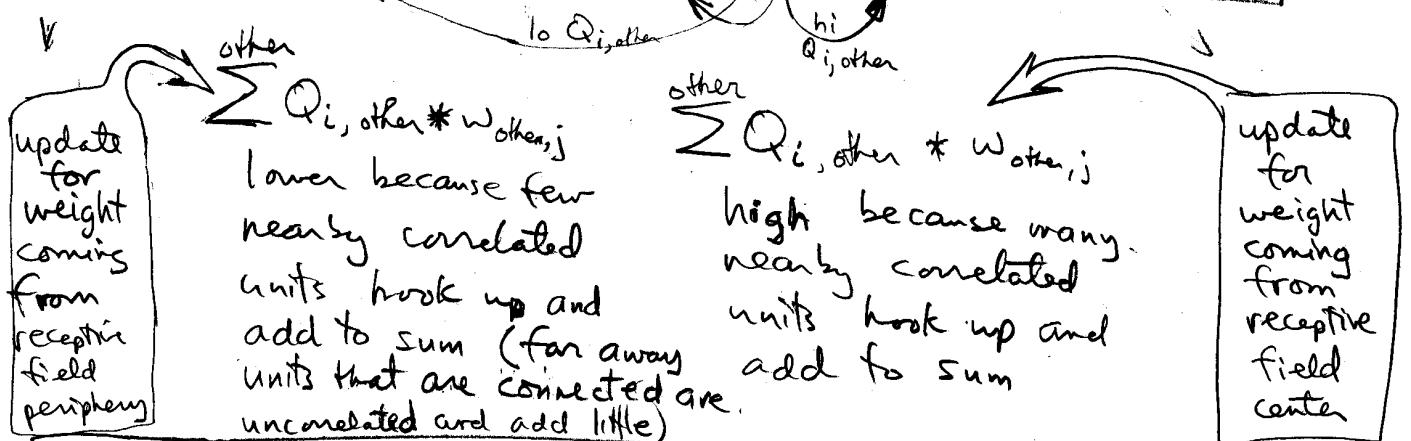
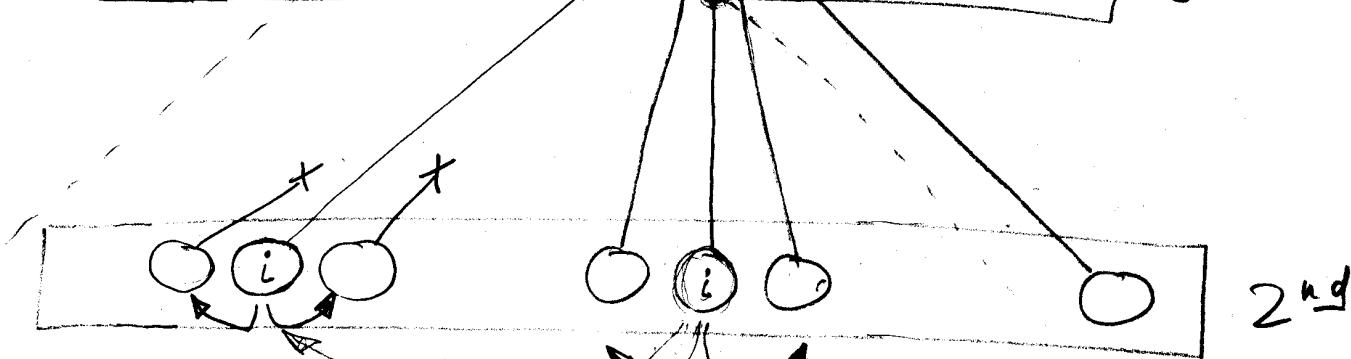
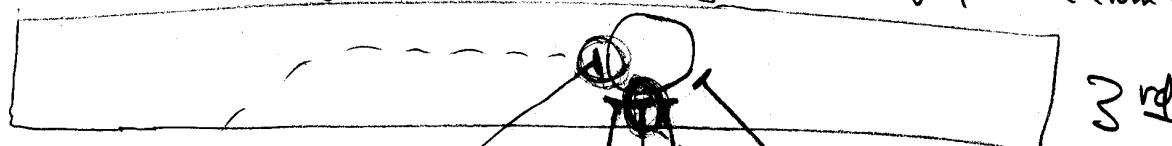
What happens in layer 3

1) 3rd layer inputs go center-surround

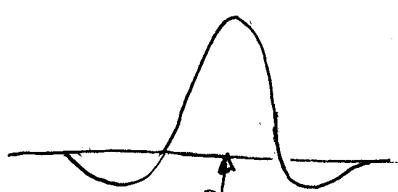
a) k_1 causes increase in weights

b) because [nearly inputs from layer 2 correlated] → center weights go higher
 [and, there are more units in center]

c) k_2 drags down everybody; center stays positive from b), surround goes neg.

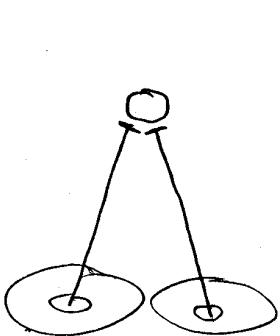
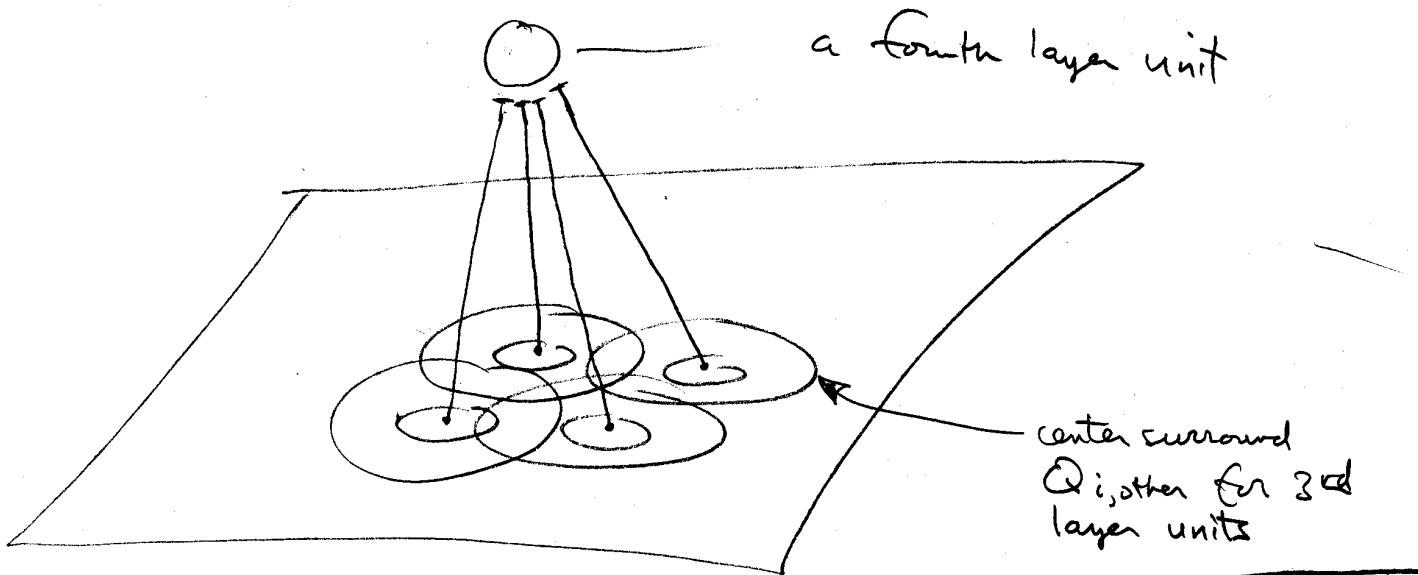


2) What the $Q_{i, \text{other}}$ looks like for layer 3

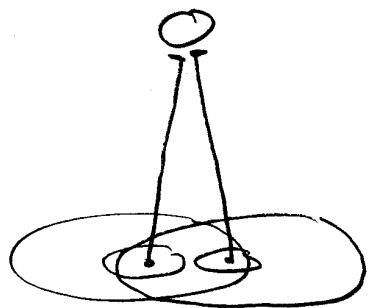


Linsker(4)

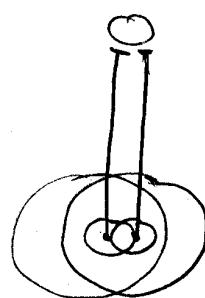
What happens in layer 4
(symmetry-breaking)



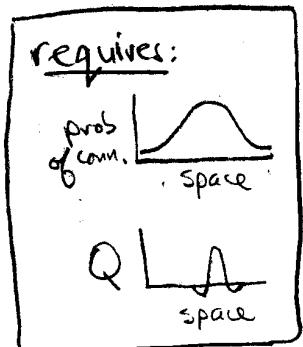
$\sum Q_{i, \text{other}}$ will be zero for these



$\sum Q_{i, \text{other}}$ will be negative for these



$\sum Q_{i, \text{other}}$ will be positive for these



Correlation of nearby centers $\Delta w \uparrow$

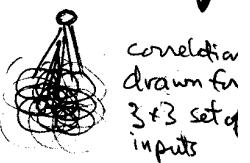
Anti-correlation of center and midrange surround $\Delta w \downarrow$
far away cells uncorrelated $\Delta w = 0$

- it is possible to calculate a Hebb energy

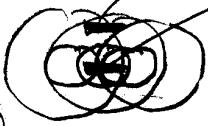
- each Δw will always change to lower Hebb energy

- a receptive field with elongated excitatory & inhibitory regions has lowest energy

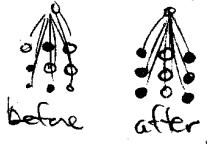
good here $\Delta w \downarrow \rightarrow$



good here $\Delta w \uparrow$

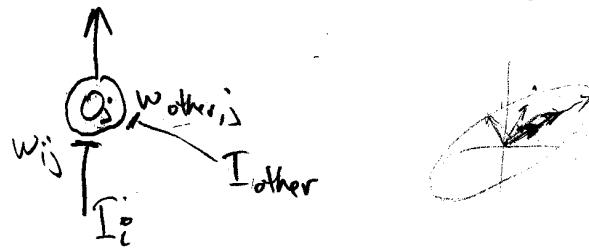


i.e., lines tend to emerge from circularly symmetric starting dist



Why Eigenvectors?

Linker



$\Delta \vec{w} = \vec{Q} \vec{w}$ is
a (system of) differential
equations

$$\Delta w_{ij} = k_1 + \frac{1}{\# \text{inputs}} \sum_{\text{other}} (Q_{i,\text{other}} + k_2) w_{\text{other},j}$$

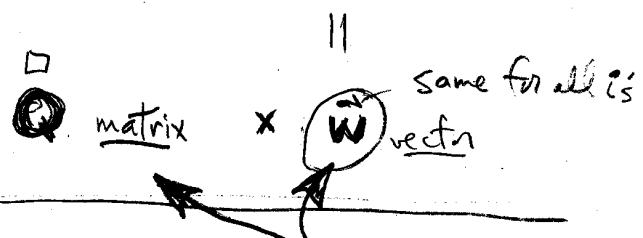
(ignore j b/c only one output unit)

$$\text{change in one weight } \Delta w_i =$$

dot prod ↑ if $\vec{Q} \vec{w}$ similar

$$\text{change in weight vector } \Delta \vec{W}_{\text{vectn}} =$$

(different i's)



Linear Transformations

$$Ax = b$$

matrix A [rotates scales]

vector x = output vector b

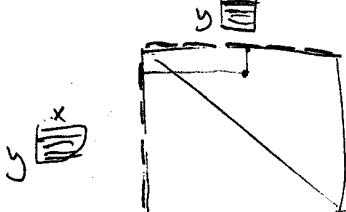
Eigenvalue problem

$$Ax = \lambda x$$

scalar eigenvalue

set of vectors x_n that when operated on by matrix A just grow (no rotation)
[- λ_n is how much each grows]
[- vectors not in direction of x_n rotate most toward nearest x_n]

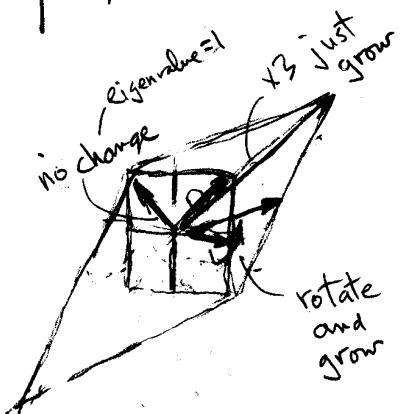
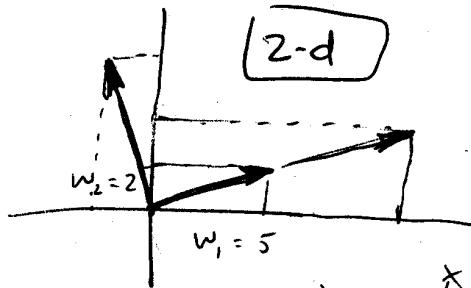
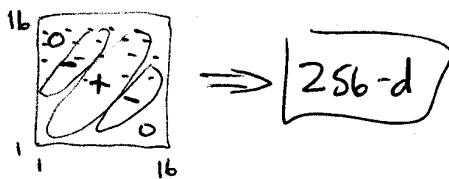
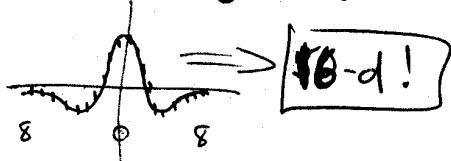
Covariance Matrix



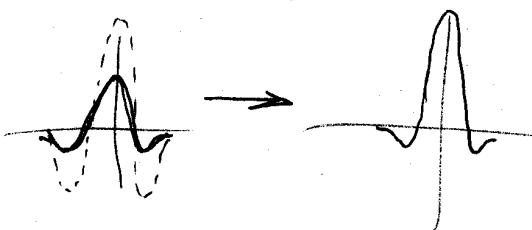
eigen vectors of stem covariance matrix tell which directions weight vectors will grow when operated on by covariance matrix during Hebb update

a vector in direction of eigenvector will just grow (no rot!)

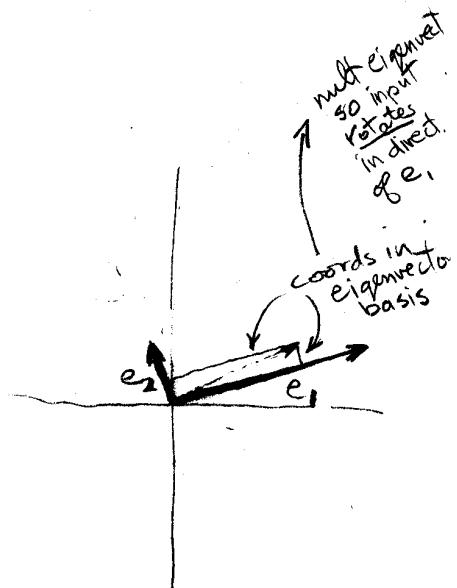
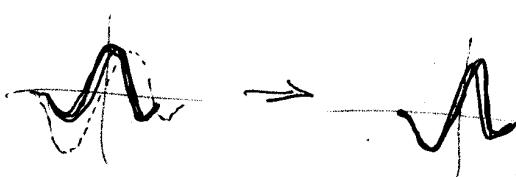
- vector is pattern of weights



- Scaling a pattern (multiplying by factor) doesn't change its "shape"

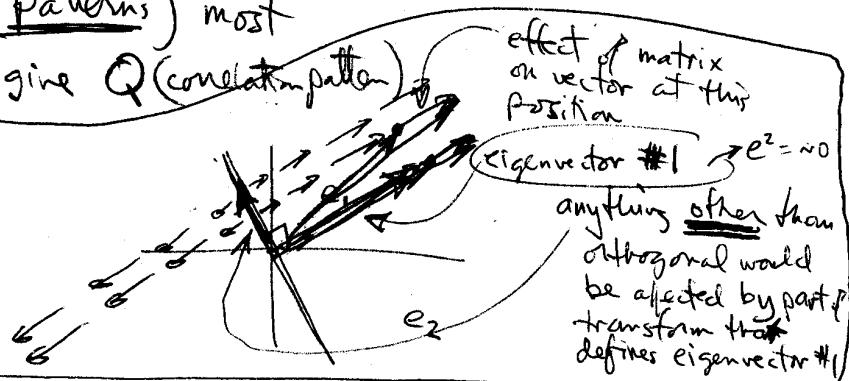


- rotating a pattern does change shape



- eigenvalue/eigenvector solution tells which directions (i.e. patterns) most likely to arise from given Q (correlation pattern)

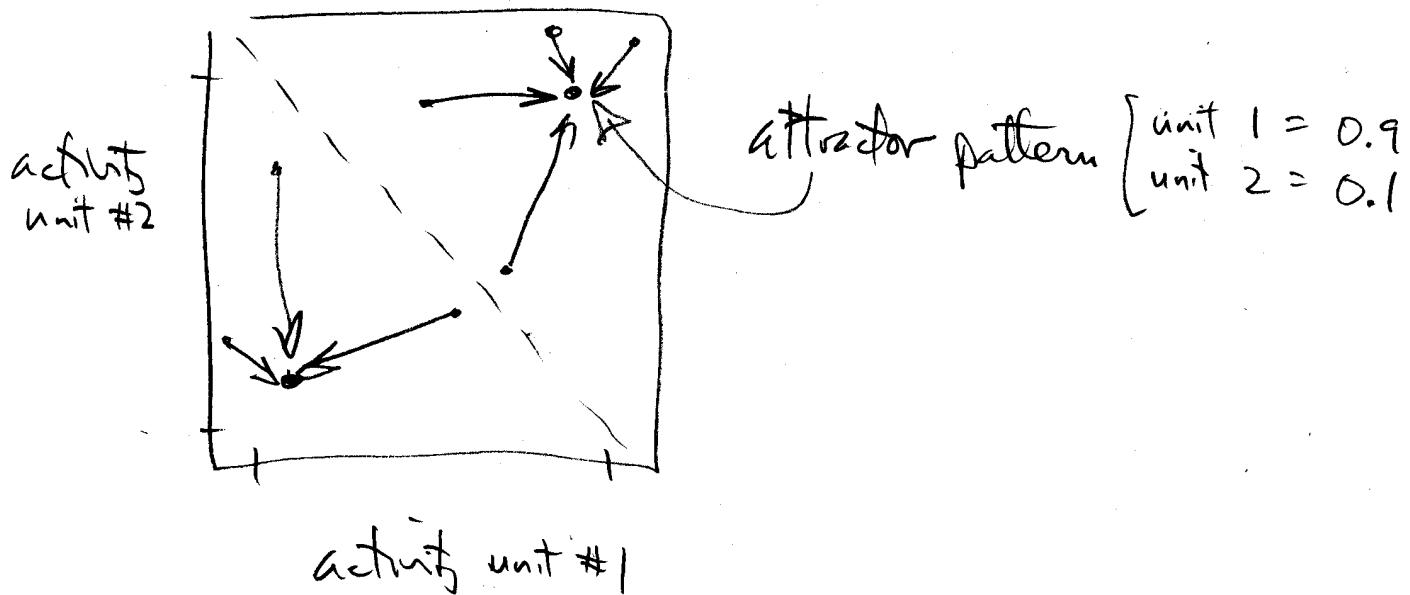
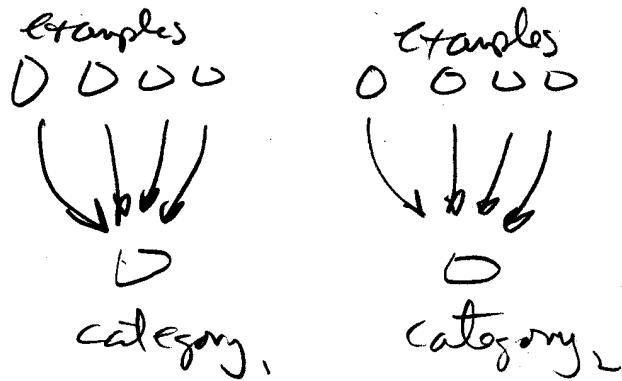
eigenvectors are orthogonal:



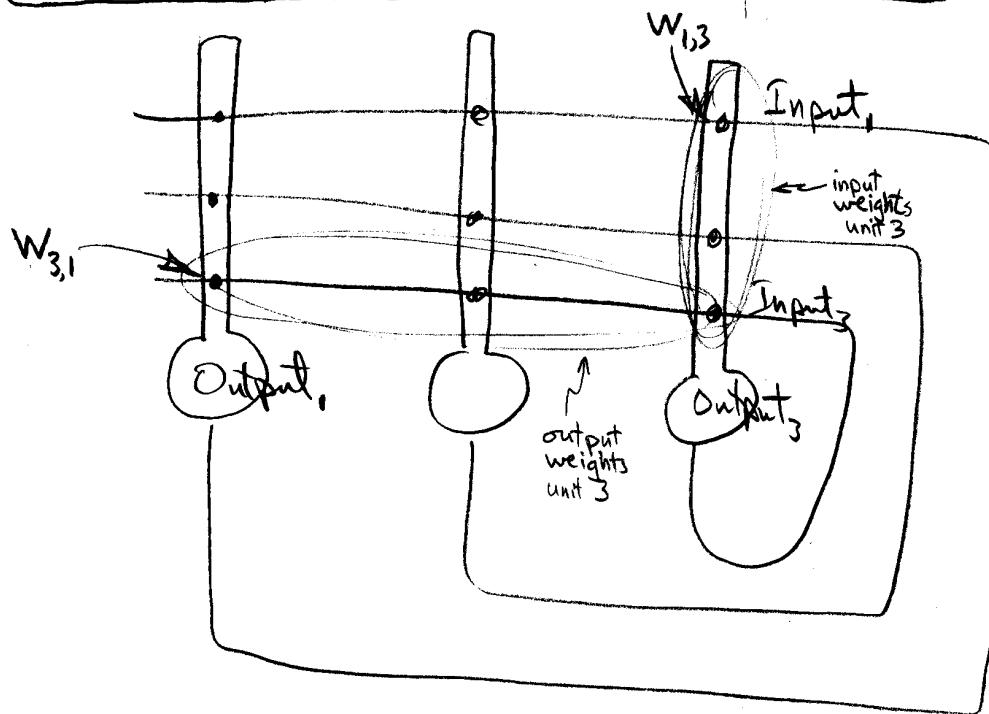
Recurrent Intro (1)

- 1) feedforward, learning (=weight change)
- 2) now: recurrent, dynamics w/o weight change

Attractor Networks



Recurrent - Why ΔE always \downarrow or ϕ (2)



Update

$$\text{Output}_j = \begin{cases} 1 & \text{if } \sum_i \text{Input}_i w_{ij} > 0 \\ -1 & \text{if } \sum_i \text{Input}_i w_{ij} \leq 0 \end{cases}$$

Energy

$$\text{Energy} = \underset{\substack{(\text{one} \\ \text{number})}}{- \sum_i \sum_j \text{Input}_i w_{ij} \text{Output}_j}$$

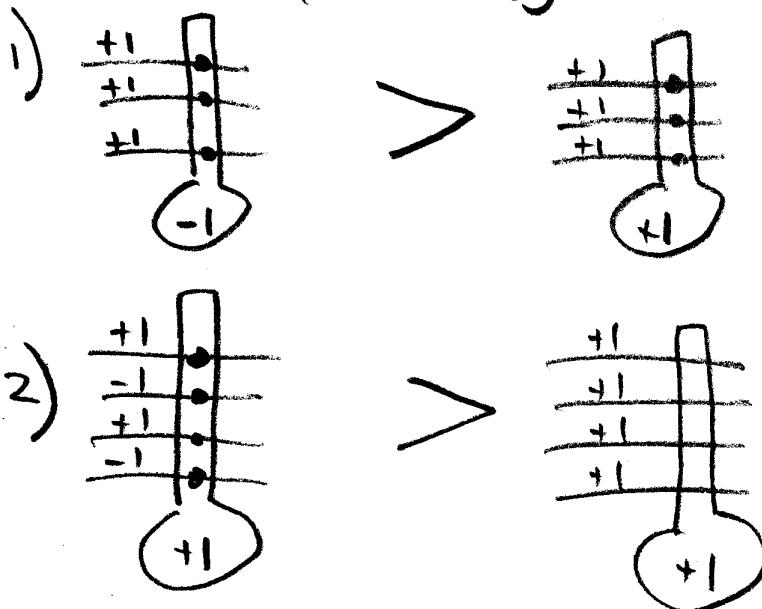
(where $\text{Input}_i = \text{Output}_i$)

Lyapunov funct

Show that update of any unit can only [reduce or leave same] energy

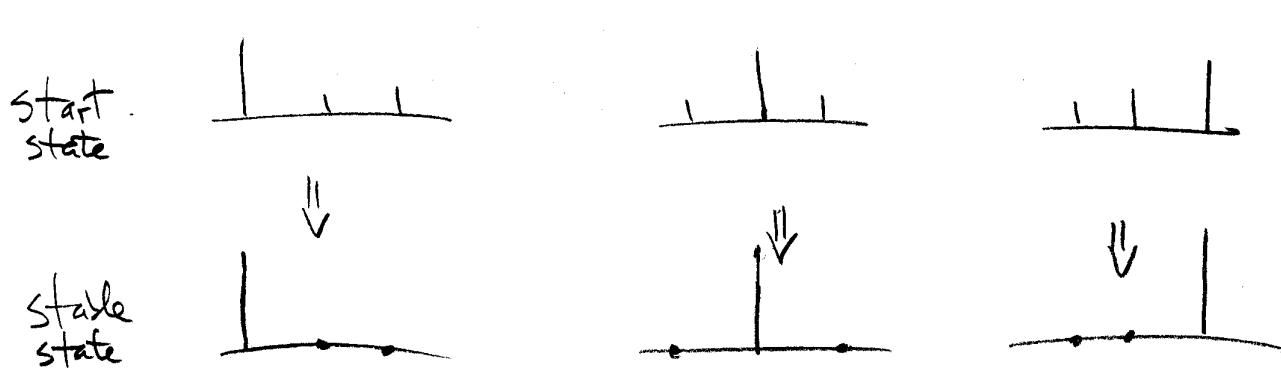
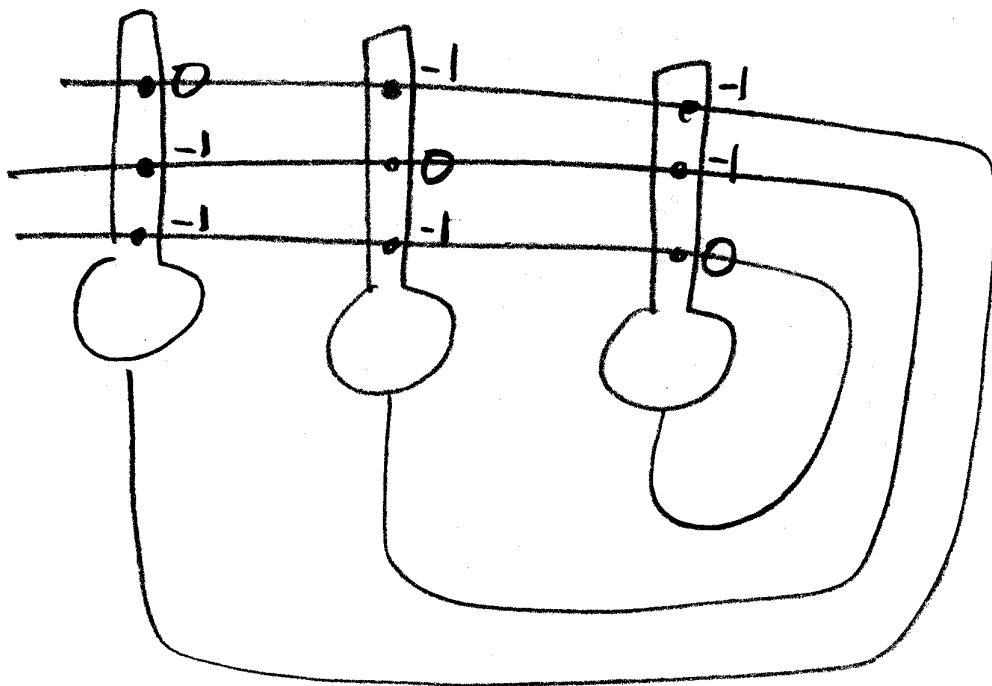
Recurrent - symmetry (3)

- intuitive idea of energy or "frustration"



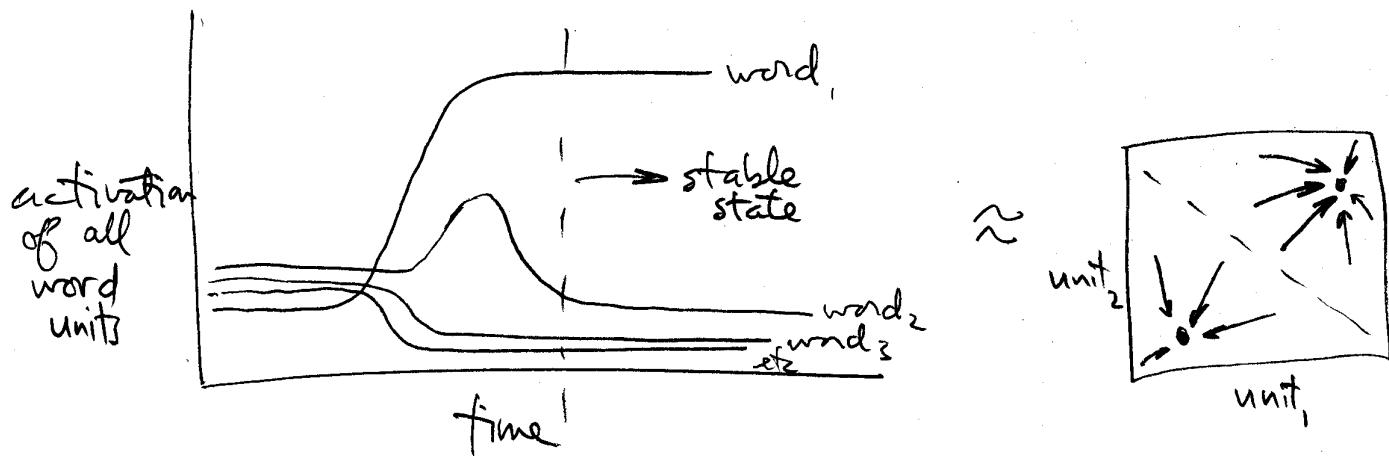
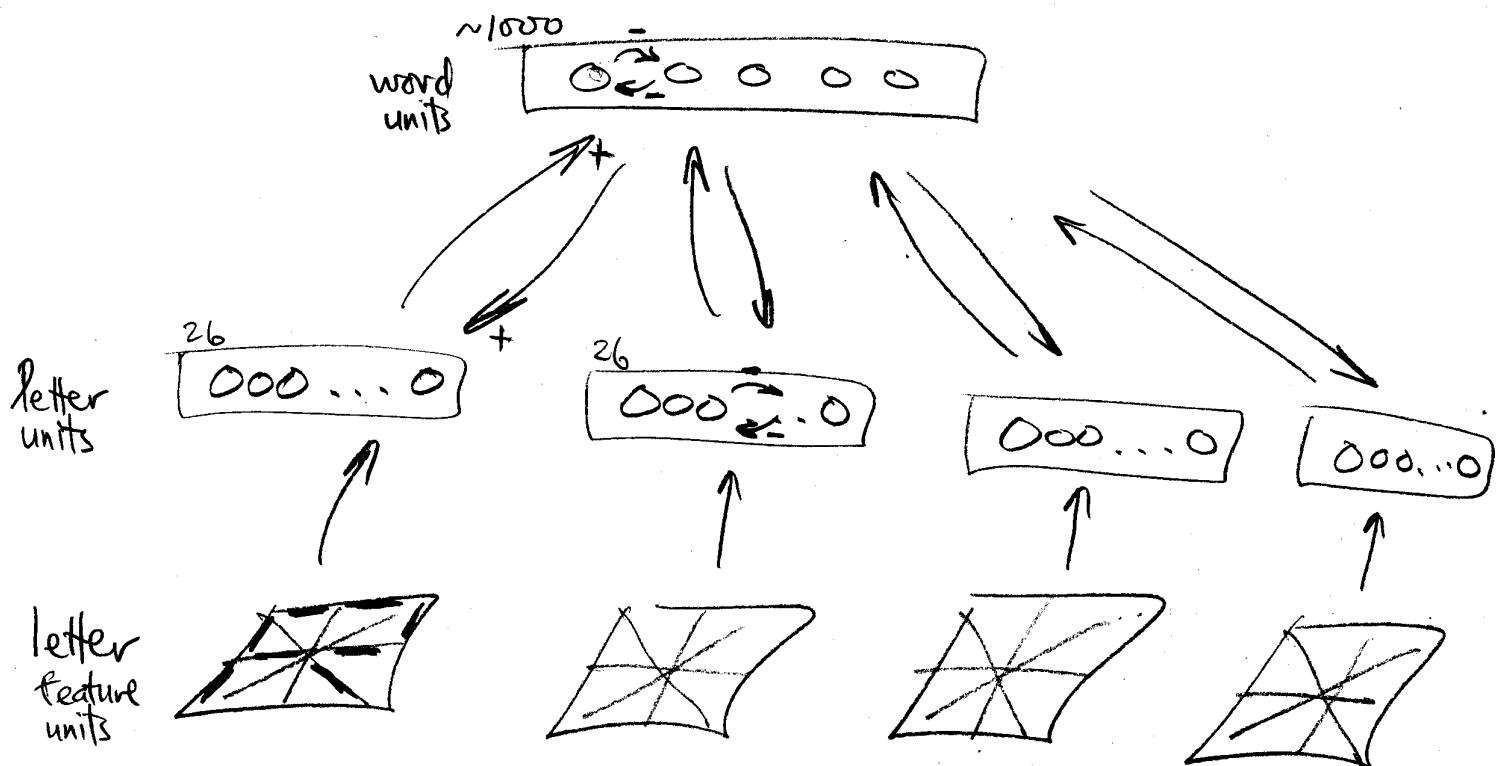
- update of one unit reduces (or keeps same) energy terms from its input connections
 \hookrightarrow (definition & update rule)
- output connection weights \rightarrow unknown effect
- if symmetry, output connection terms term-to-term same as input connection terms
- \therefore if symmetry \rightarrow stable states

Winner-take-All



- 3 stable states

Word-Rec model as Hopfield



- word-like non-words speed letter rec.
- deal w/partial occlusion/noise

#5

Back Prop Summary

$$\text{linear_output}_j = \sum_i \text{input}_i * w_{ij}$$

$$\text{squashed_output} = f(\text{linear_output})$$

change weight to reduce error:

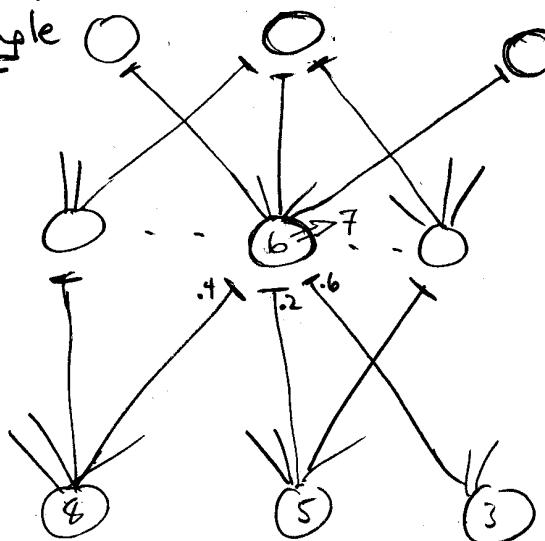
$$\Delta w = \text{learning rate} * \delta * \text{input}$$

where:

$$\delta_{\text{output}} = (\text{target} - \text{actual}) * f'(\text{linear_output})$$

$$\delta_{\text{hidden}} = \left(\sum_{\text{output}} \delta_{\text{output}} w_{\text{output output}} \right) * f'(\text{linear_output})$$

Feed forward pass example



gradient descent

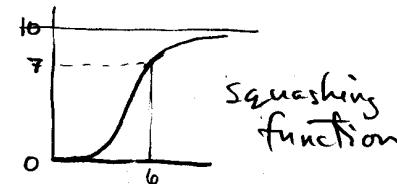
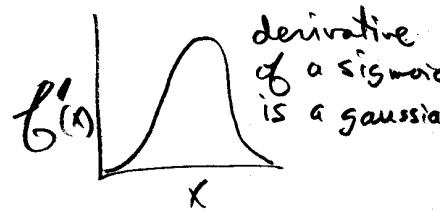
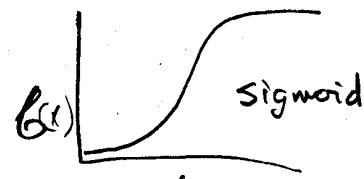
what $\Delta w = \text{des.-act.}$ does!

derivative $f'(x)$
get slope, take neg \rightarrow go downhill

gradient \rightarrow deriv. of $f(x)$
get grad ... (generalizes to vect arg)
vector, take neg \rightarrow go downhill



($f(x)$ e.g. a sigmoid)

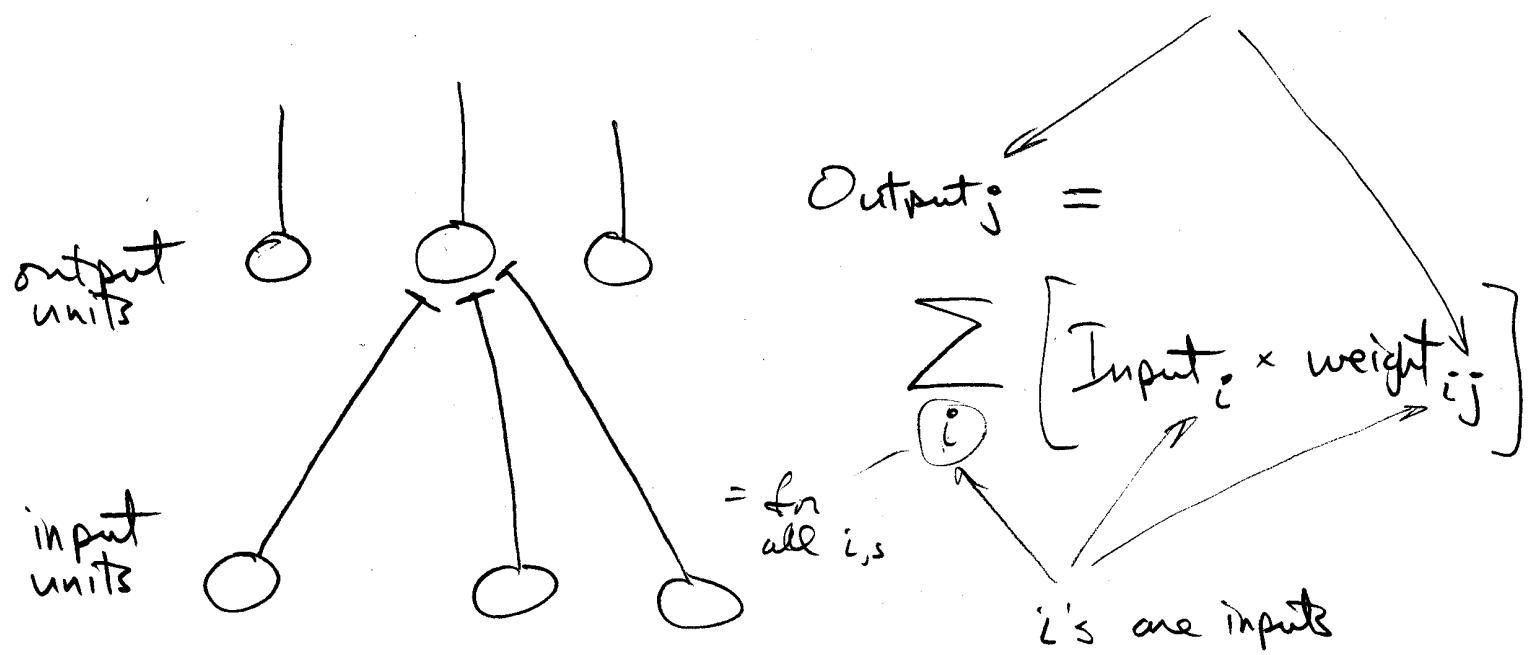


$$\begin{aligned} \text{linear_output}_j &= \sum_i \text{input}_i w_{ij} \\ &= (8 \times .4) + (5 \times .2) + (3 \times .6) \\ &= 6 \end{aligned}$$

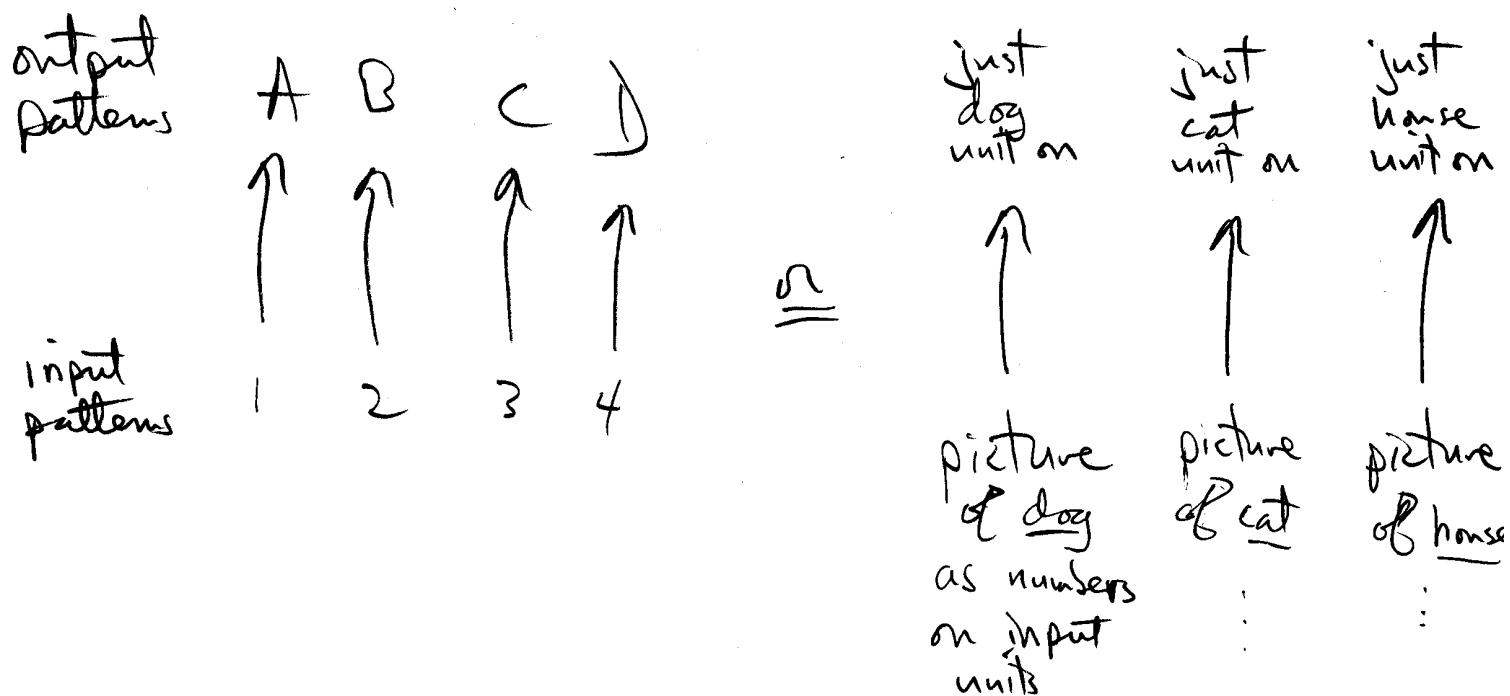
$$\begin{aligned} \text{squashed_output}_j &= f(\text{linear_output}_j) \\ &= f(6) \\ &= 7 \end{aligned}$$

Feedforward

Weighted Sum (1)



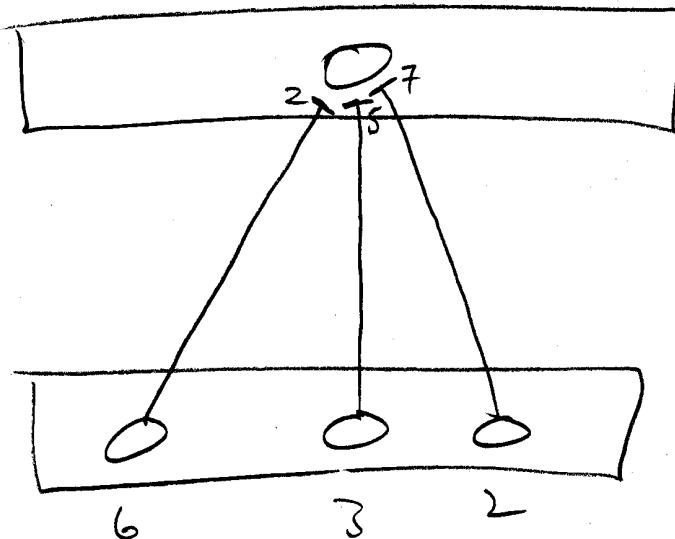
Example of learning problem



How to get output error (2)

desired = 55

actual = 41

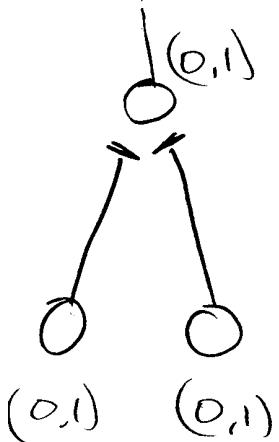


$$\text{error} = \frac{\text{desired}}{\text{activation}} - \frac{\text{actual}}{\text{activation}}$$

$$\begin{aligned}\text{output} &= \sum(\text{inputs} \cdot \text{weights}) = \\ &= 6 \times 2 + 3 \times 5 + 7 \times 2 \\ &= 12 + 15 + 14 \\ &= \boxed{41}\end{aligned}$$

Problems w/ 2-layer

Network



Problem

<u>Input Patterns</u>	<u>Output Patterns</u>
-----------------------	------------------------

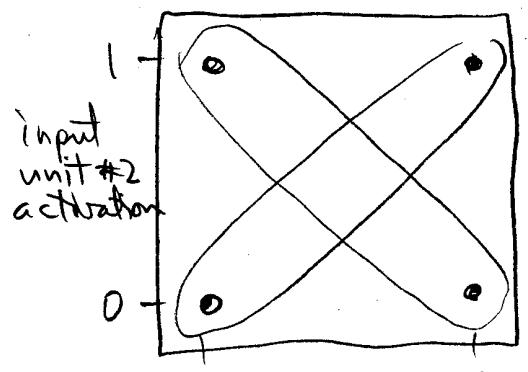
$$\begin{matrix} 0 & 1 \end{matrix} \rightarrow \begin{matrix} 0 \end{matrix}$$

$$\begin{matrix} 1 & 0 \end{matrix} \rightarrow \begin{matrix} 0 \end{matrix}$$

$$\begin{matrix} 1 & 1 \end{matrix} \rightarrow \begin{matrix} 1 \end{matrix}$$

$$\begin{matrix} 0 & 0 \end{matrix} \rightarrow \begin{matrix} 1 \end{matrix}$$

View a scatter plot of input space

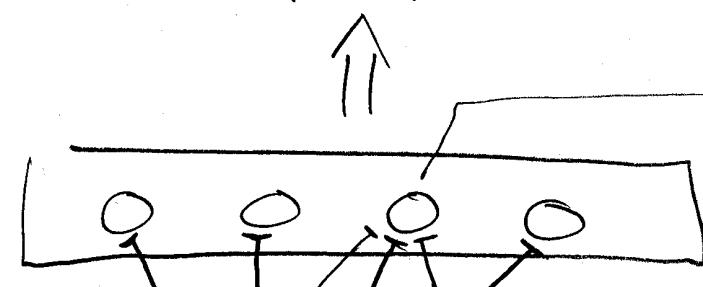


input unit #1 activation

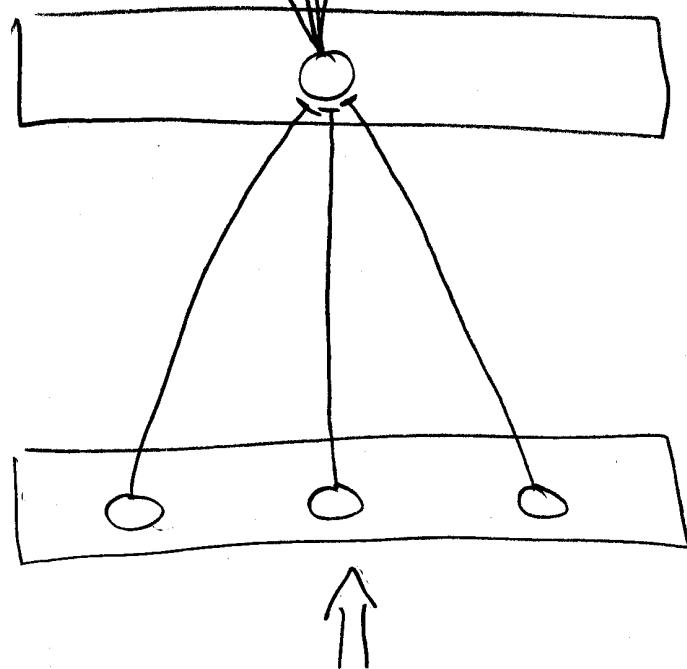
How to get hidden error (3)

output pattern calculated by

1) input \rightarrow hidden Σ
 2) hidden \rightarrow output Σ



$$\text{error}_{\text{output}} = \text{desired} - \text{actual}$$

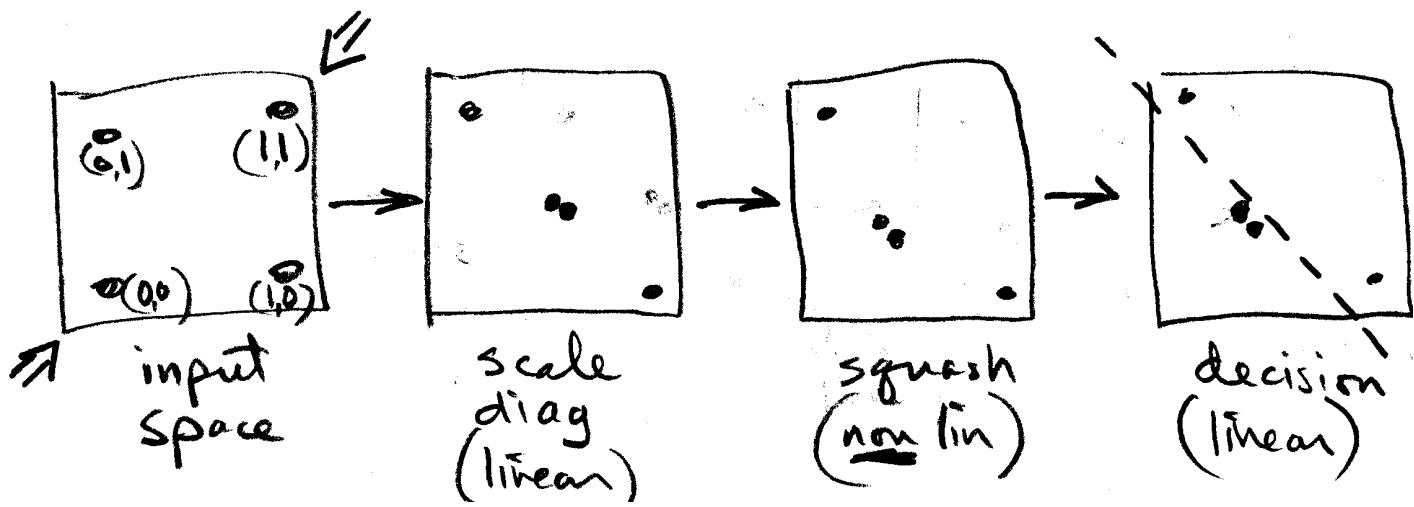


$$\text{error}_{\text{hidden}} = \sum_{\text{outputs}} [\text{error}_{\text{output}} \times \text{weight}_{\text{to output}}]$$

input pattern applied
to input units

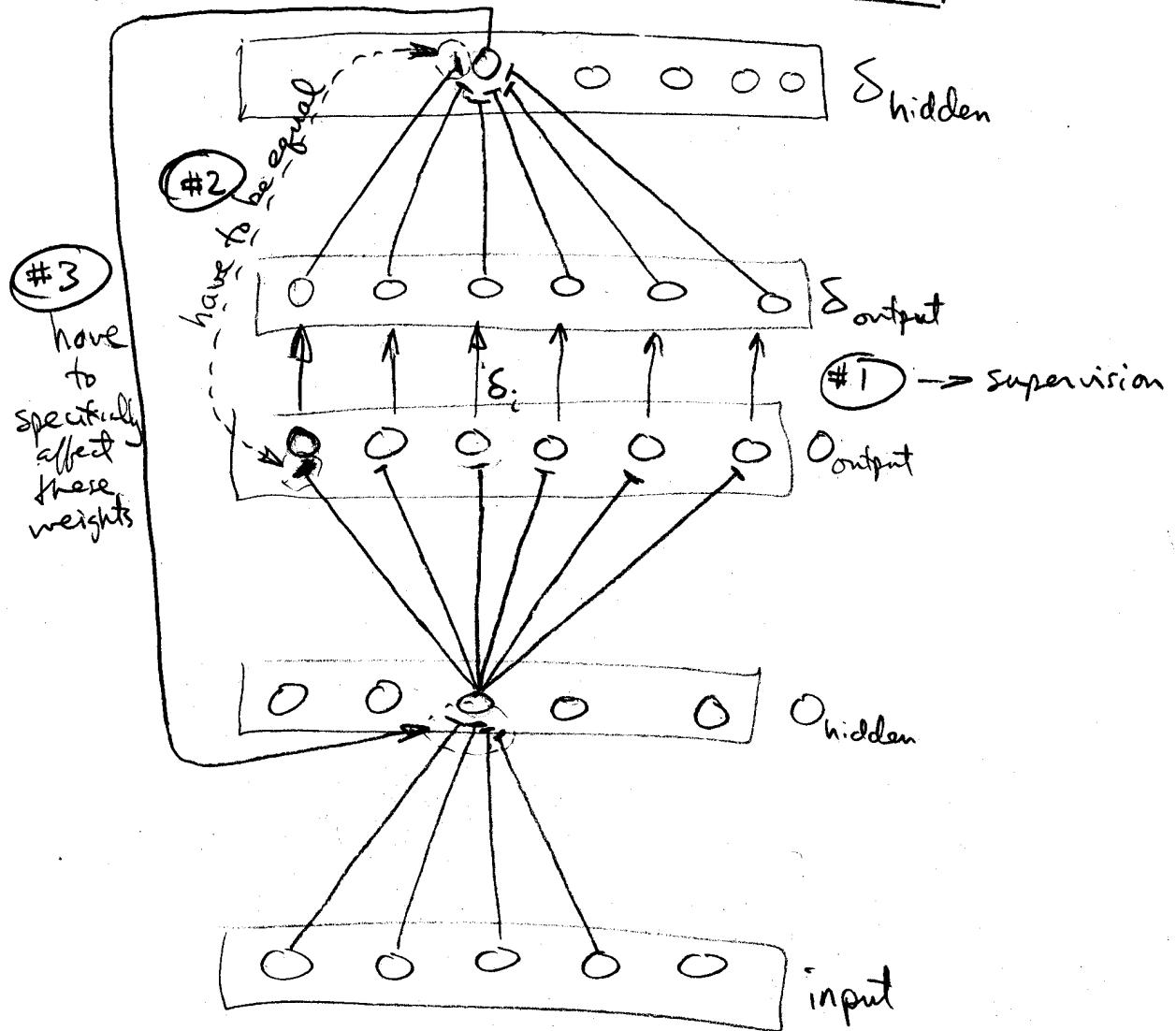
multiple layers
non-linearity
parity-like prob
(non-linearly separable)

- if net is linear, the operations performed by n layers can be done all at once by one layer
- non-linearly separable require
 - 1) 3 layers
 - 2) non-linear in at least one
- how parity solved:

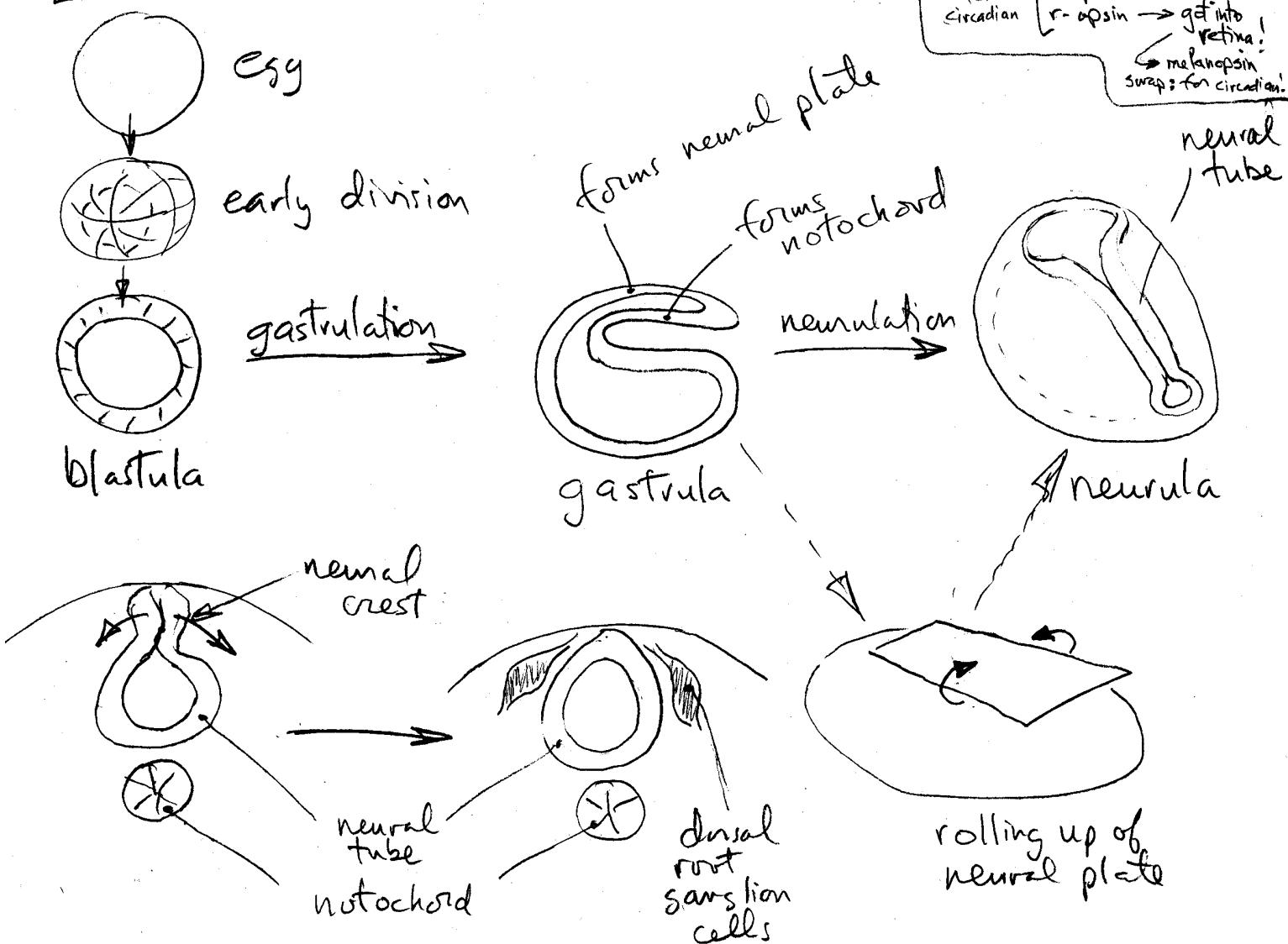


What's "bad" about backprop

("bad" \rightarrow bio-imausible)

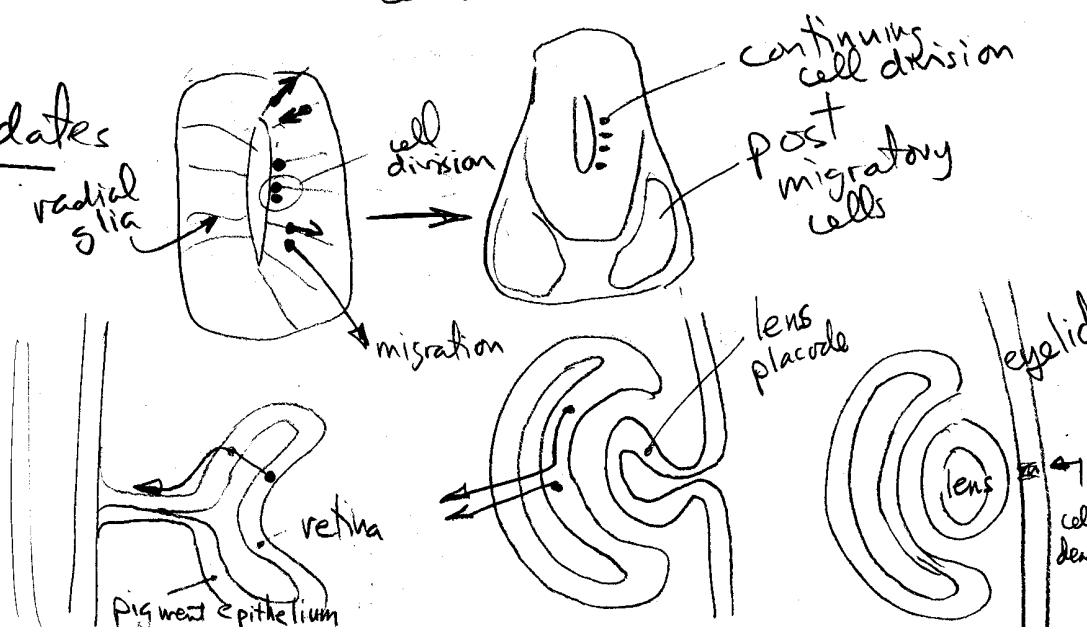
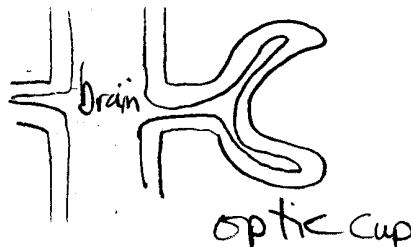


EARLY DEVELOPMENT



Migration/Birthdates

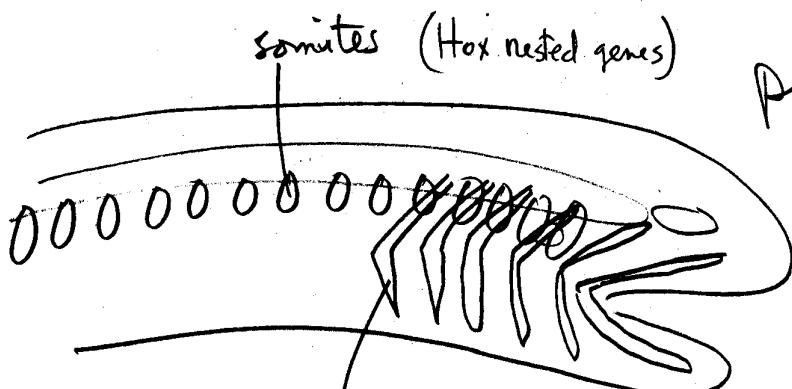
Eye Development



eyes Park's eyesless, vitamin A, opsin
trans. fact → insects
[rhabdomeric ciliary] → vertebrates
Polyphyletic w/ both!
for \leftarrow C-opsin
circadian r-opsin → get into retina!
melanopsin
Surp: for circadian!

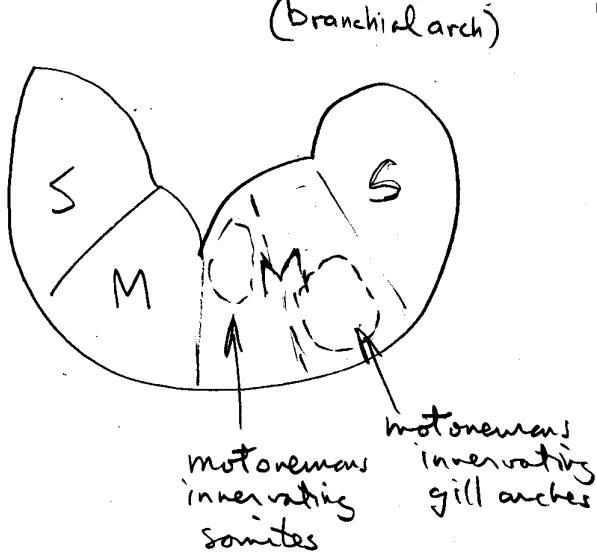
neural tube

Somites & Gill Arches



primitive
vertebrate

- antenepedia
- Hox DNA-binding proteins
- duplication
- tandem, like body order
- beware optimization

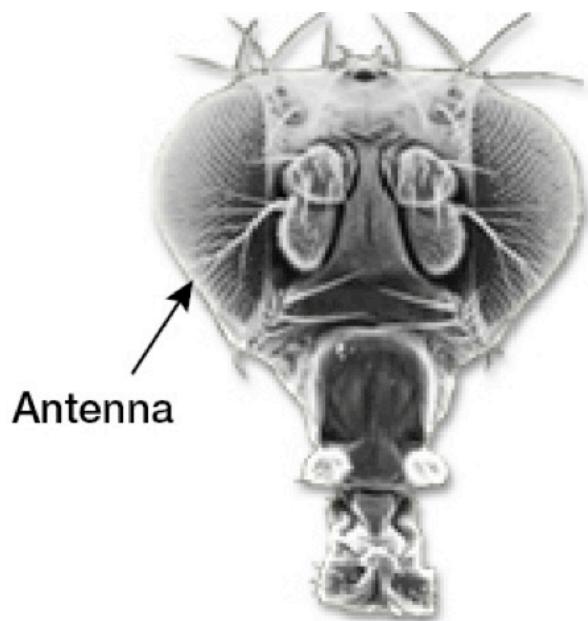


Somites

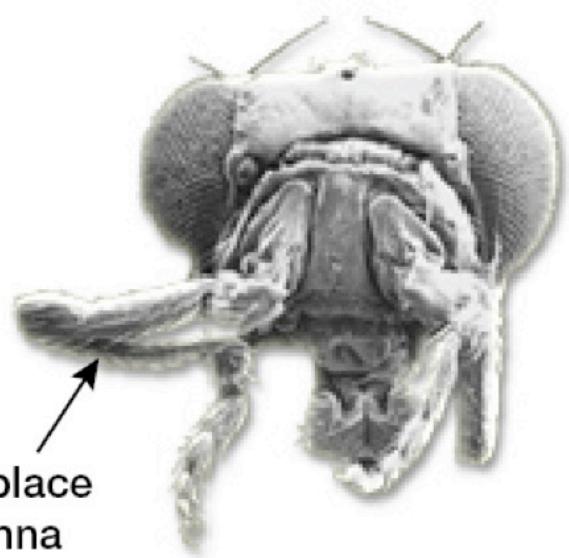
- | | | |
|---|-----------------------------------|--------------------------------------|
| 1 | | location of
motor nuclei in brain |
| 2 | \Rightarrow 6 eye
muscles | |
| 3 | | (4) III oculomotor |
| 4 | | (1) II trochlear |
| 5 | \Rightarrow tongue
muscles | (1) VI abducens |
| 6 | | XII hypoglossal |
| 7 | | |
| 8 | \Rightarrow body
musculature | |
| 9 | | spinal motor
neurons |

Gill Arches

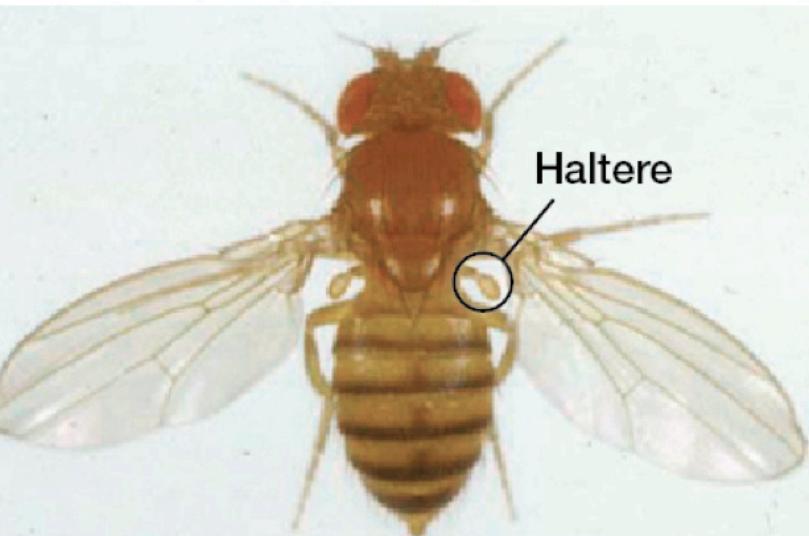
- | | | |
|---|--|---|
| 1 | \Rightarrow jaw
muscles | location of
motor nuclei in brain |
| 2 | \Rightarrow ear muscles | |
| 3 | | principle motor
nuc. of trigeminal (V) |
| 4 | | facial motor
nucleus (VII) |
| 5 | | |
| 6 | | |
| 7 | \Rightarrow larynx,
throat
muscles | nucleus
ambiguus |



Antenna

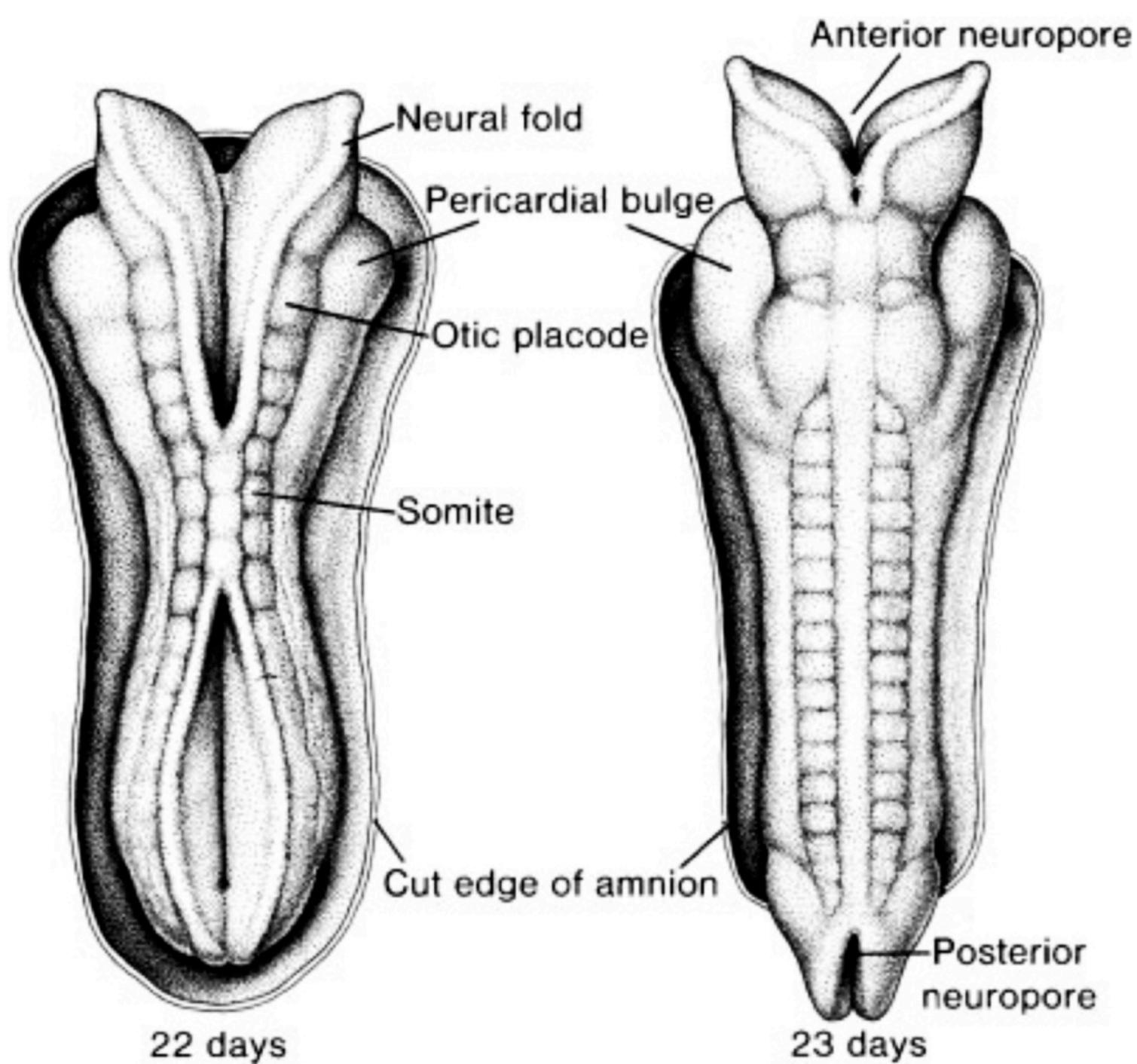


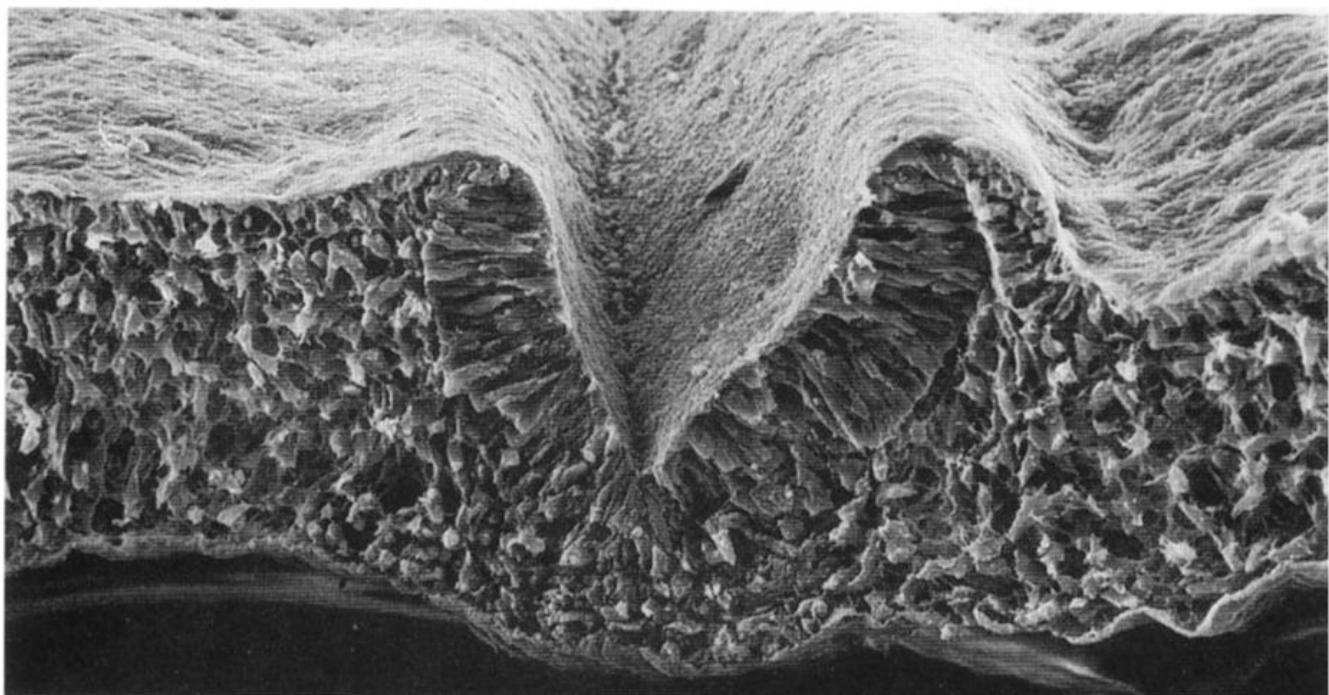
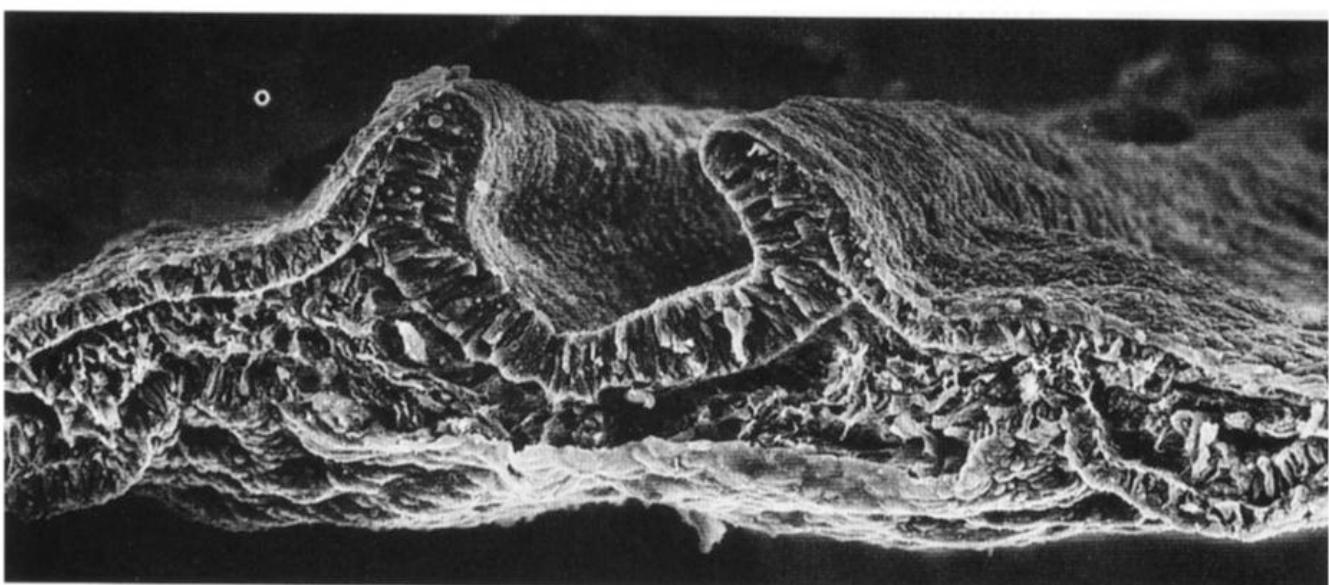
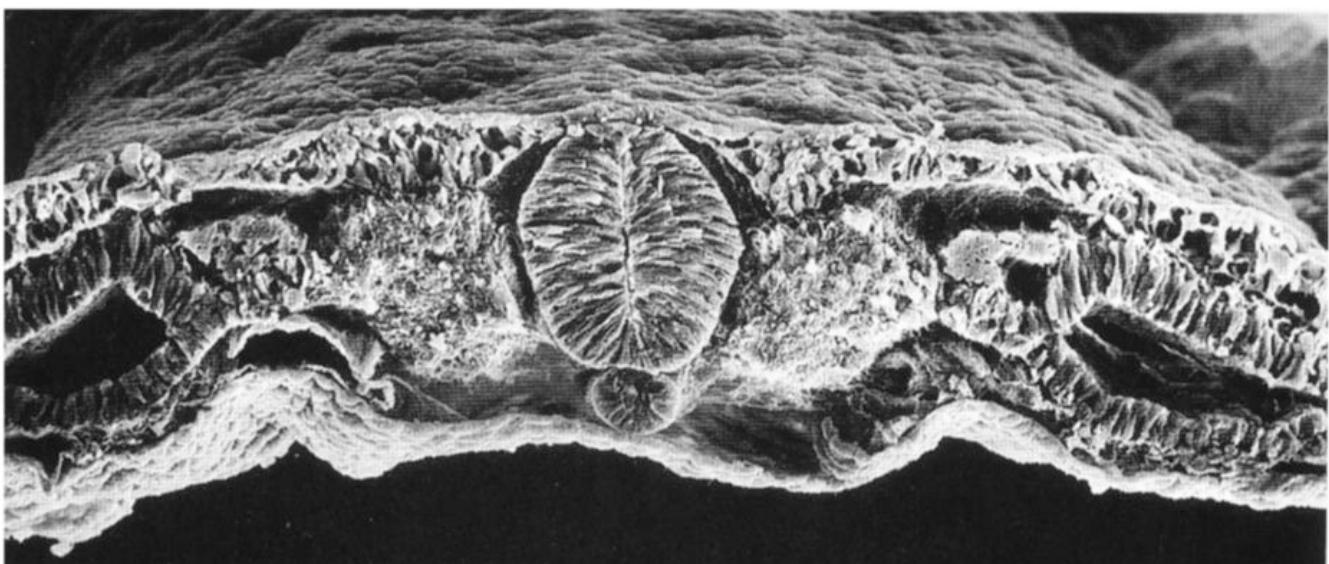
Leg in place
of antenna

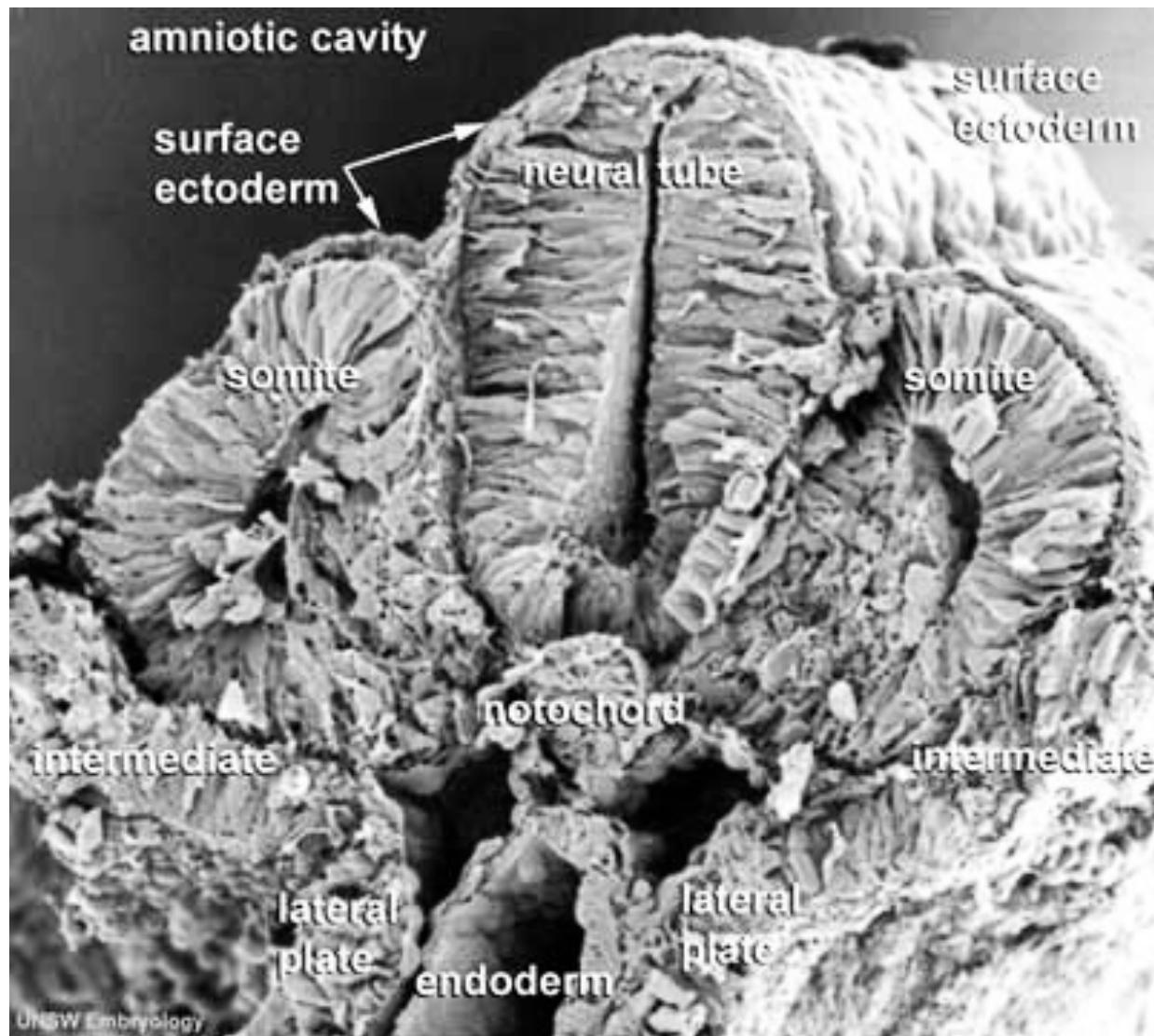


Haltere



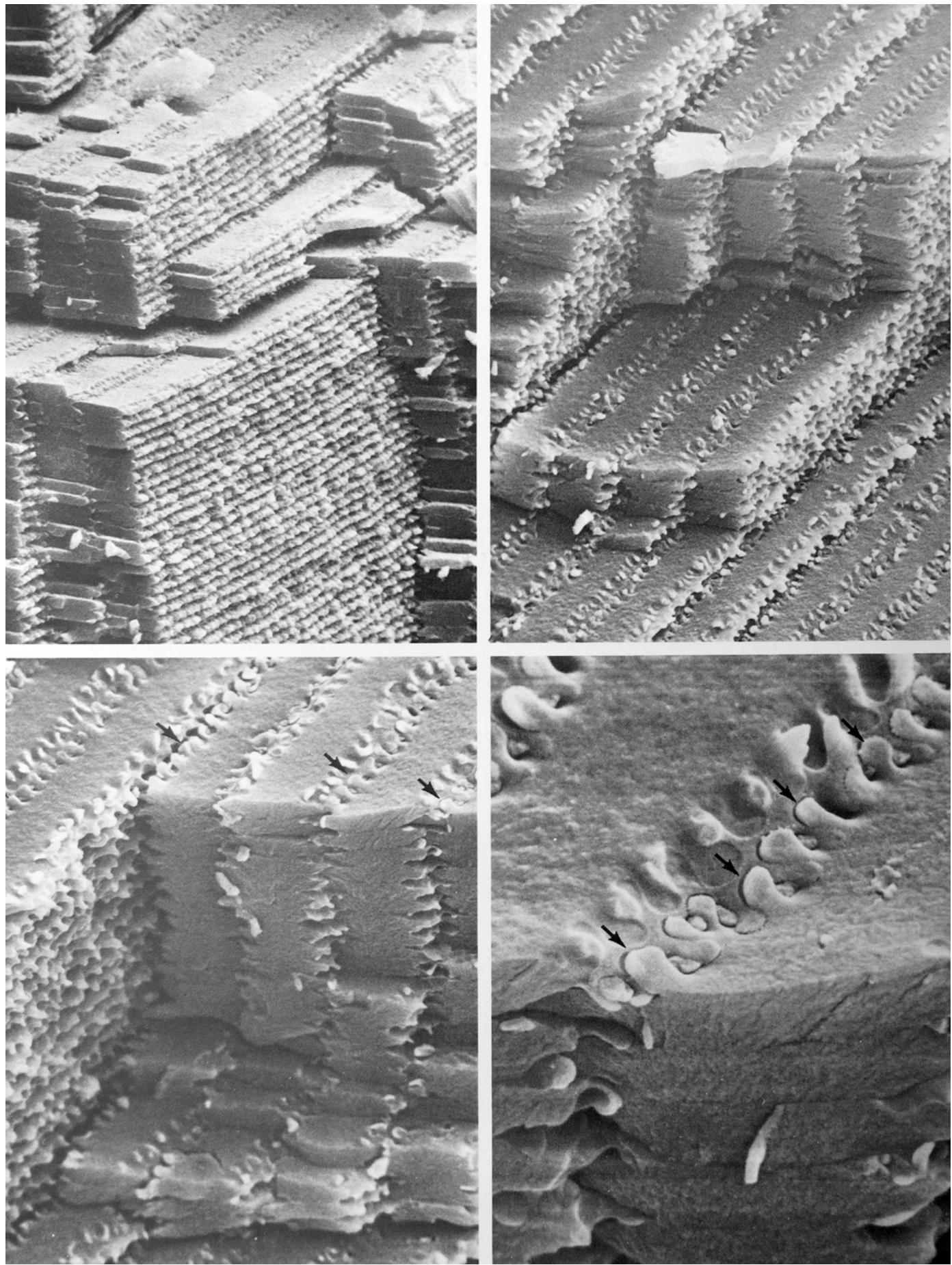


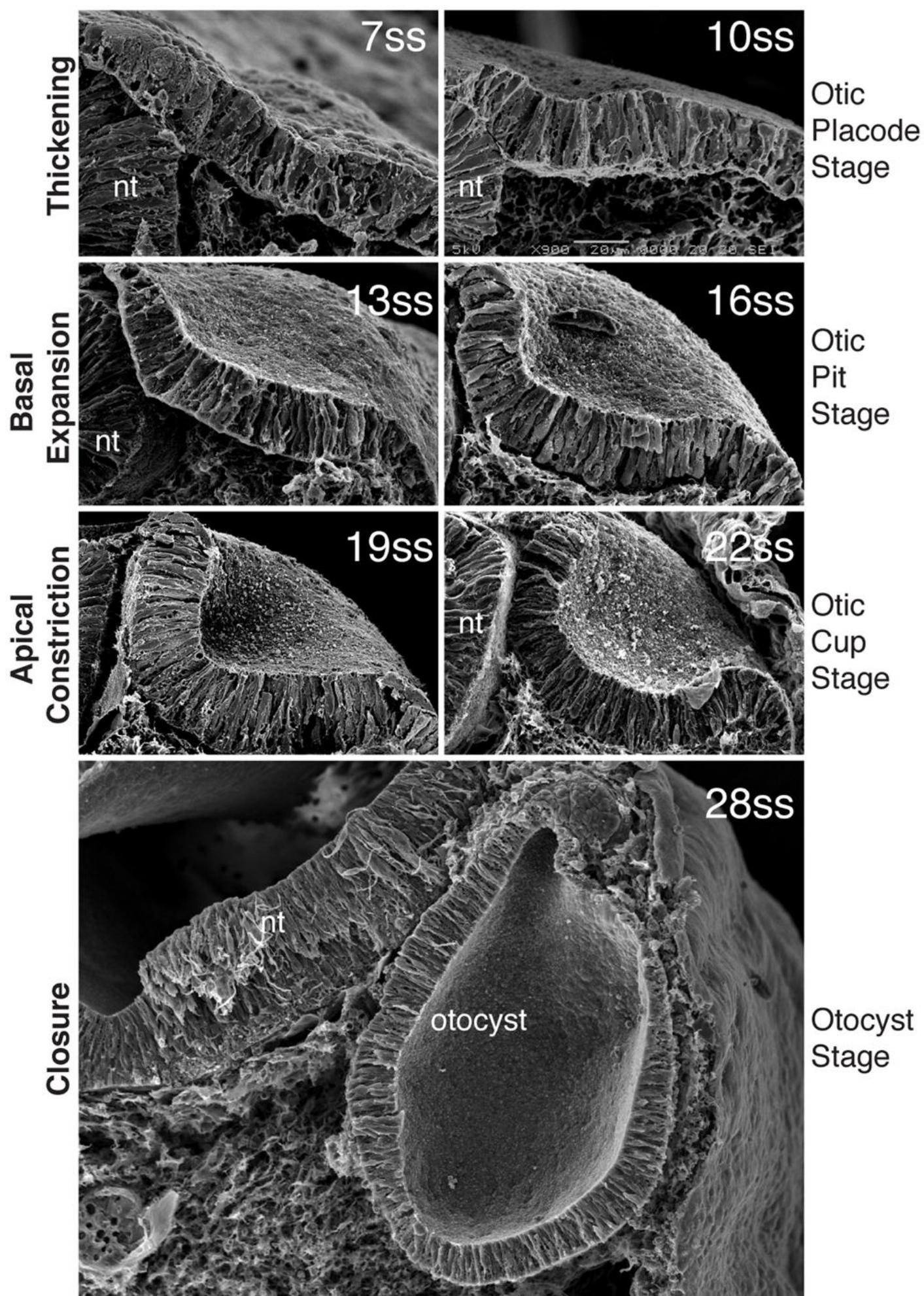
B**C****D**





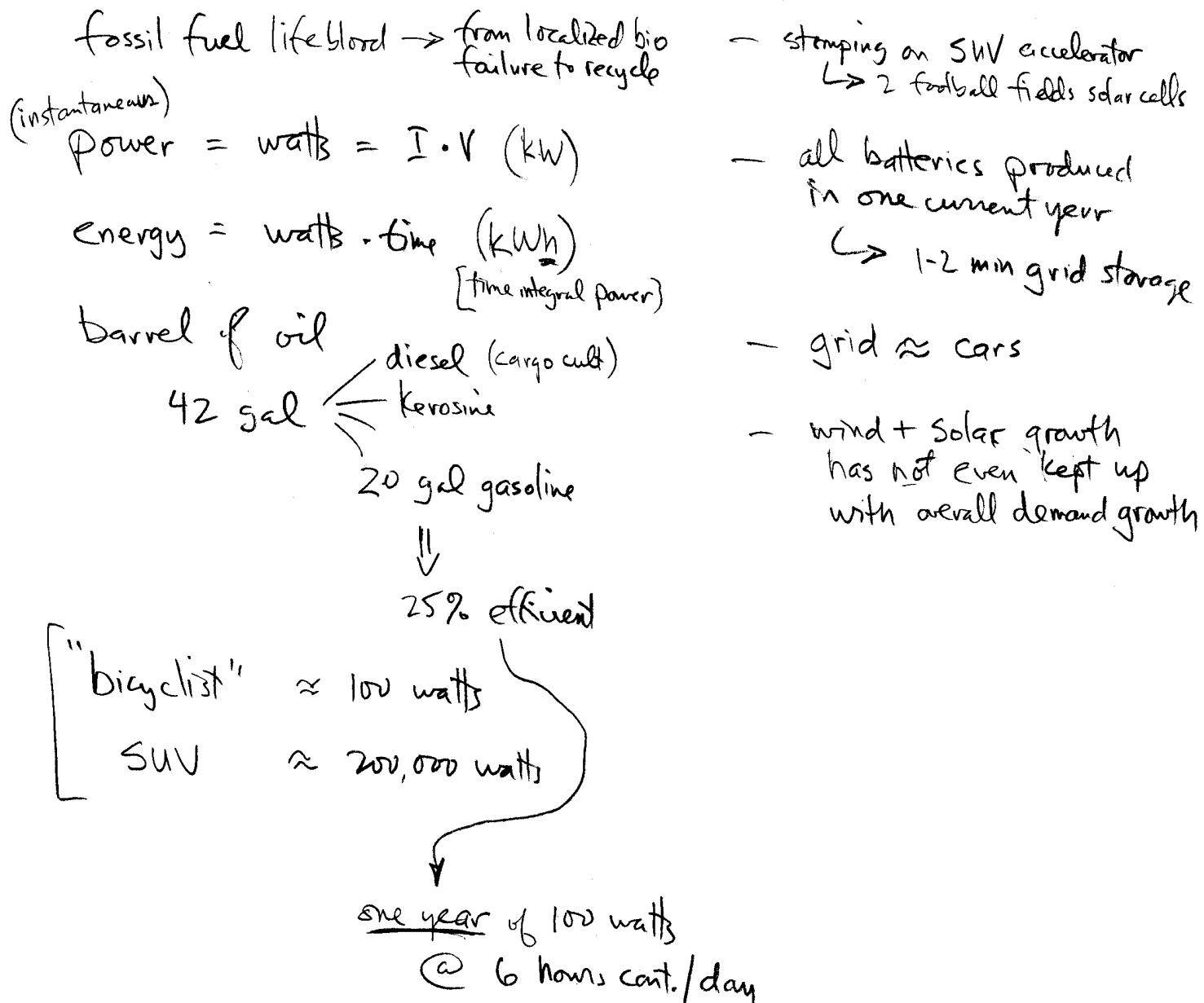
Inner Outer
layer layer
of optic cup

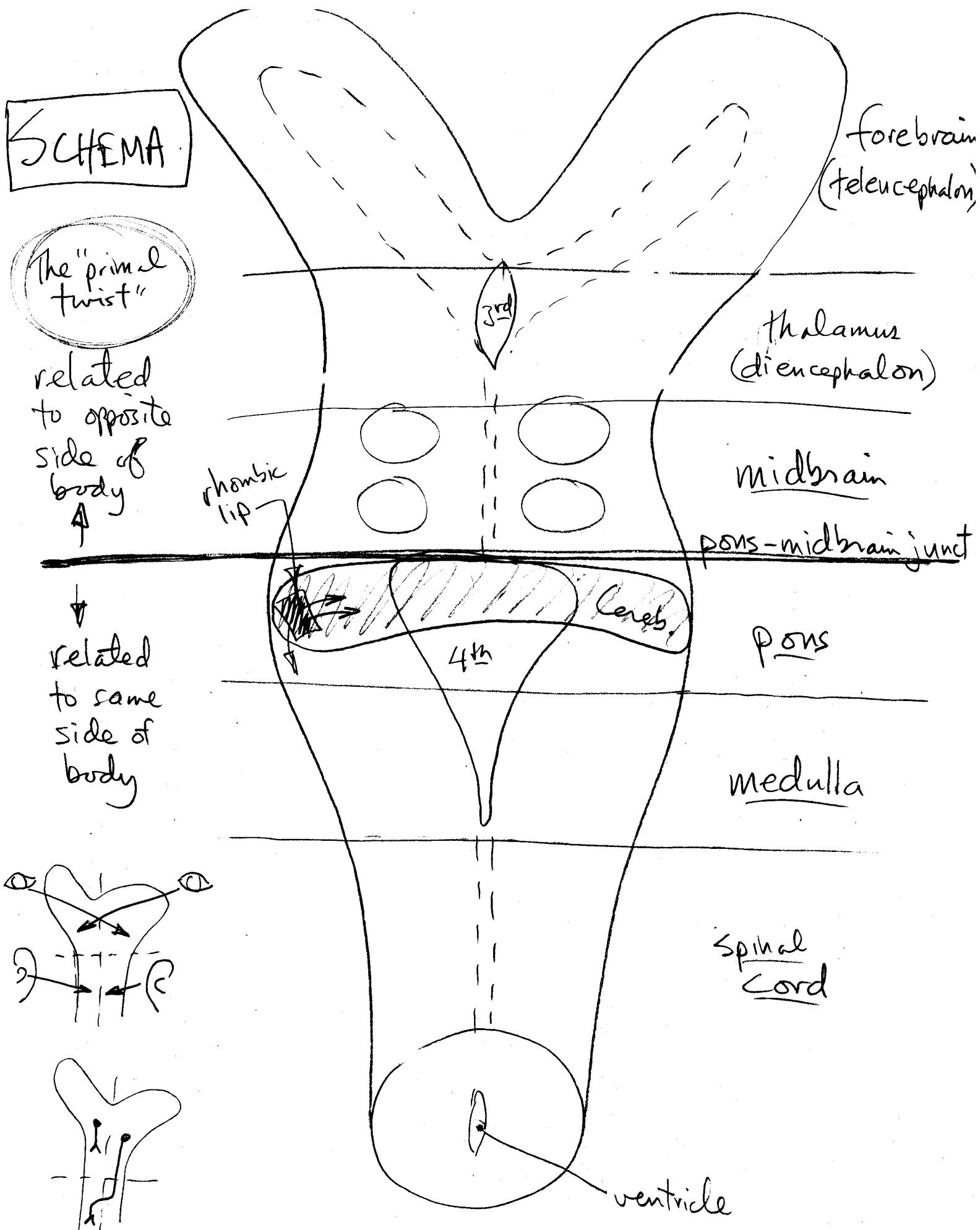


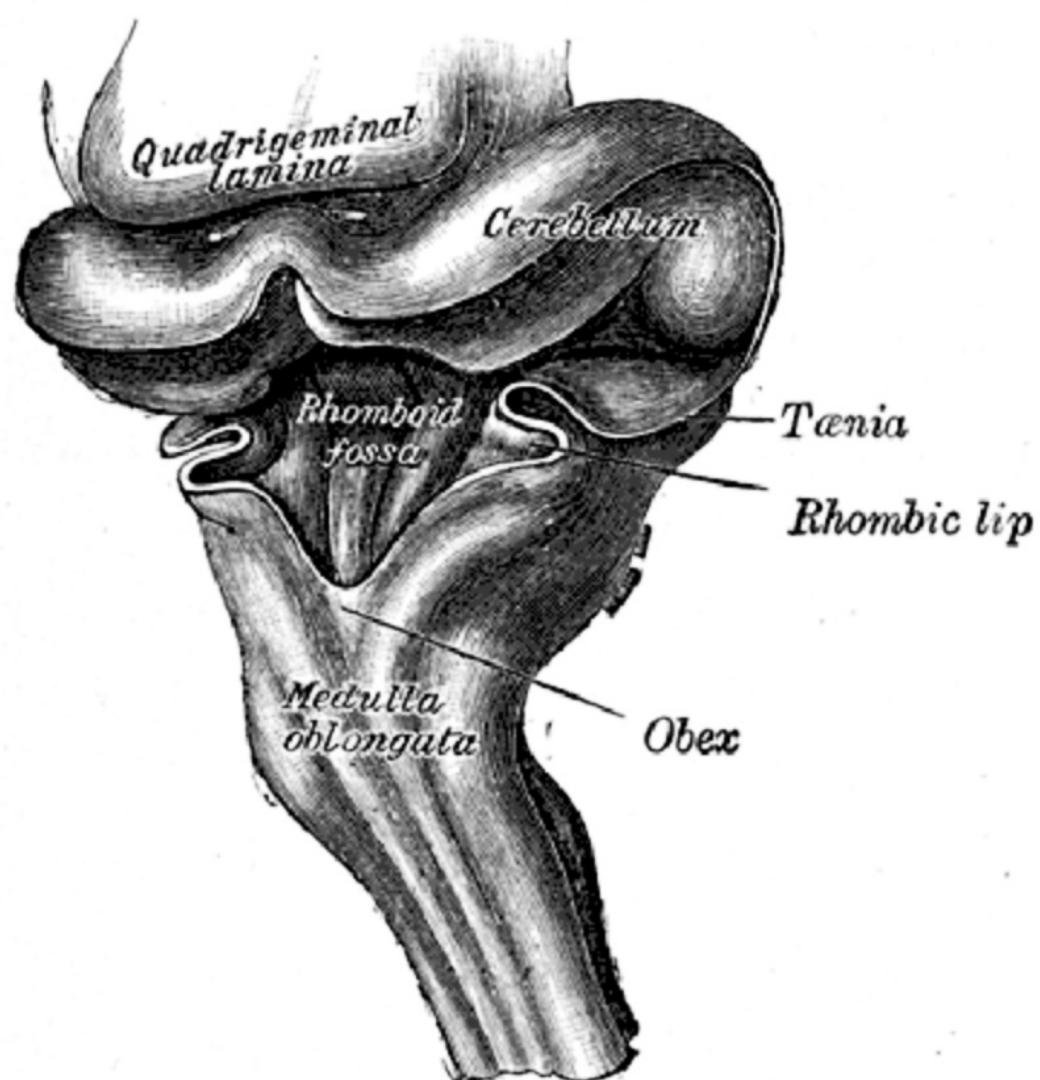
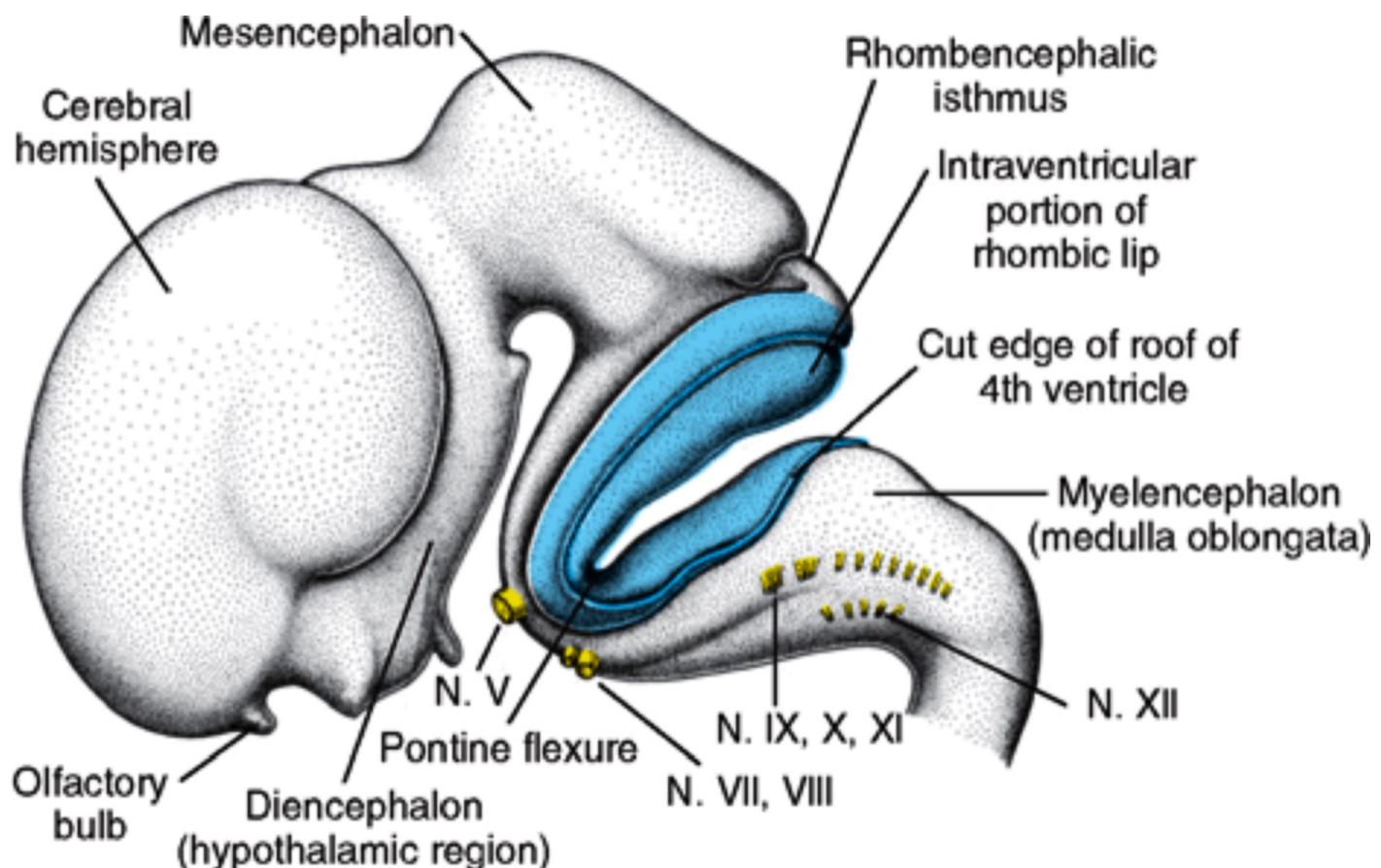


"5 Minutes on Energy" :-}

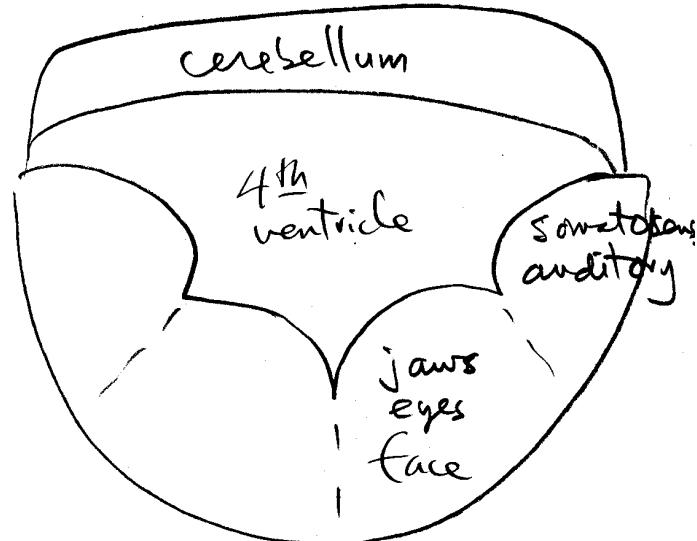
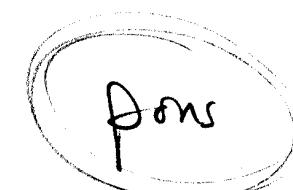
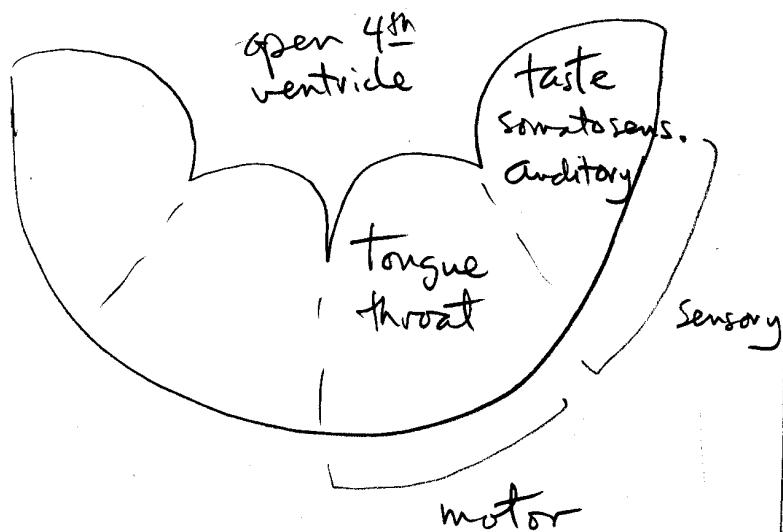
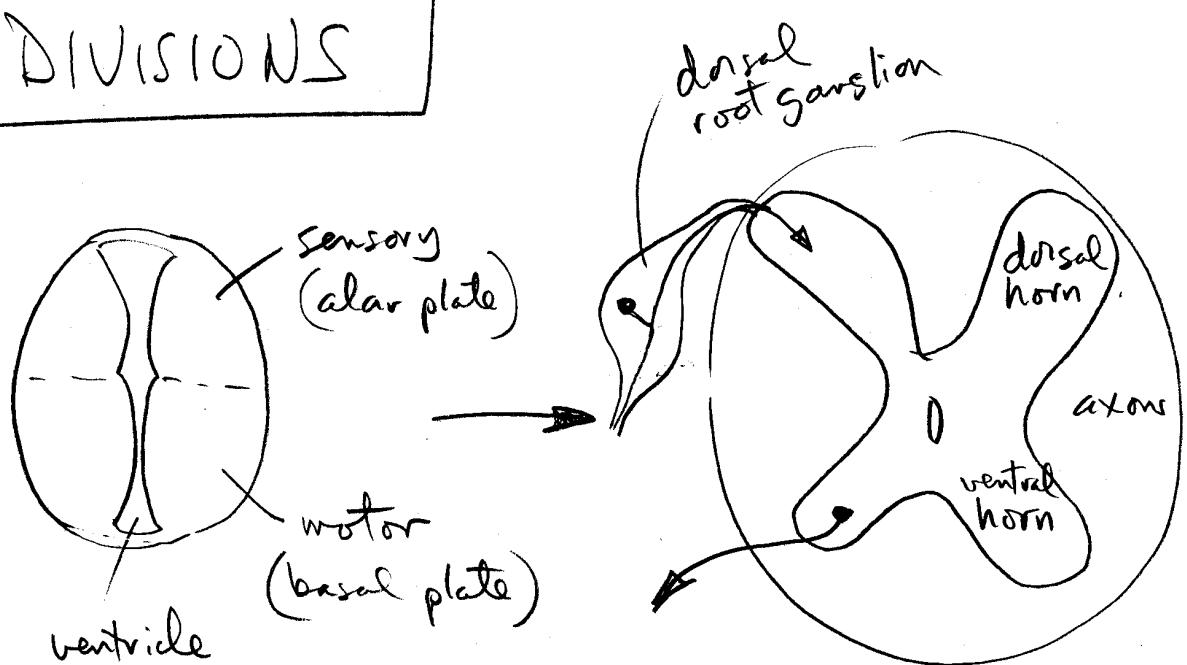
Basic average numbers to remember for solar electric. Power hitting the atmosphere is 1366 watts/m². Practically available average power considering atmospheric losses, oblique, diurnal and weather variation *before* conversion to electricity is about 190 watts/m². Power available after conversion including defects, soiling, and inverter and spacing losses is about 15 watts/m² (=1.4 watts/sq ft). For scale, a standard sized car is a 100,000 watt device (135 horsepower) at maximum output, and can cruise smoothly at highway speeds on about 20,000 watts. Therefore, to directly power a car cruising at 60-70 mph (no storage), you need the average output of 14,000 sq feet of solar cells -- an array 120 feet by 120 feet, which is 1/4 the area of an American football field. At current prices, such an array (at volume discount) would cost over \$300,000. For strong acceleration (like they do here in London trying to pass me on my bicycle on their way to a red light), you need a million-dollar full football field's worth of solar cells. That's why people are going to eventually be driving smaller, lighter cars at lower speeds, and accelerating less -- which is excellent news for cyclists



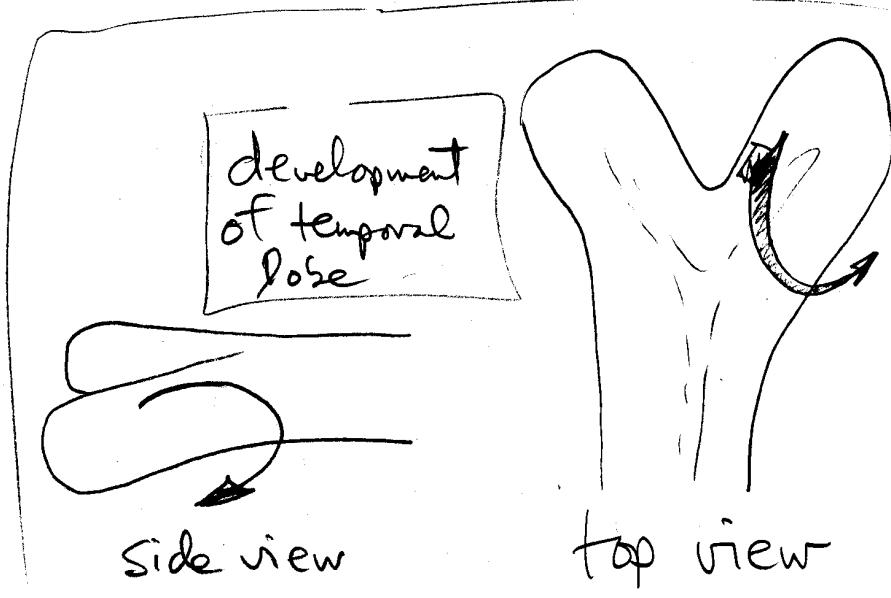
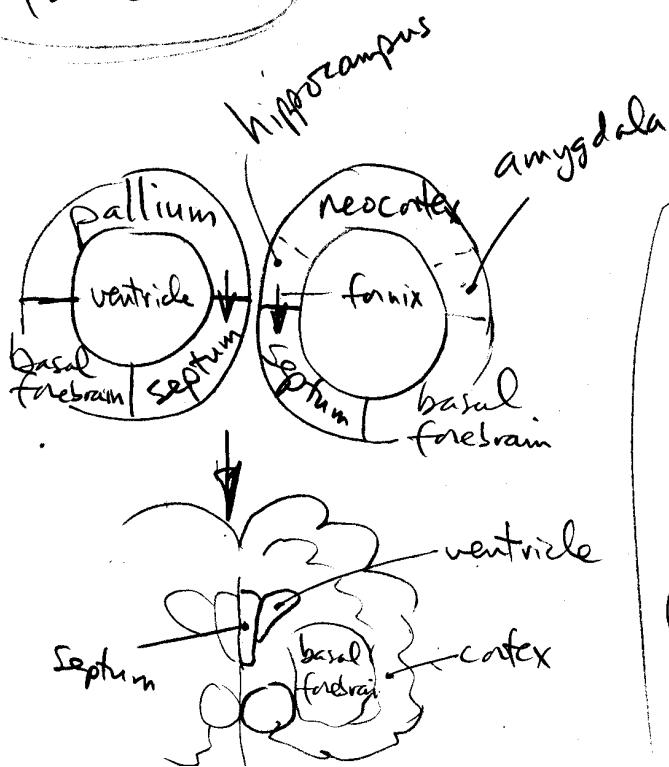
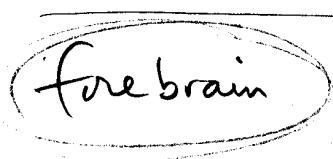
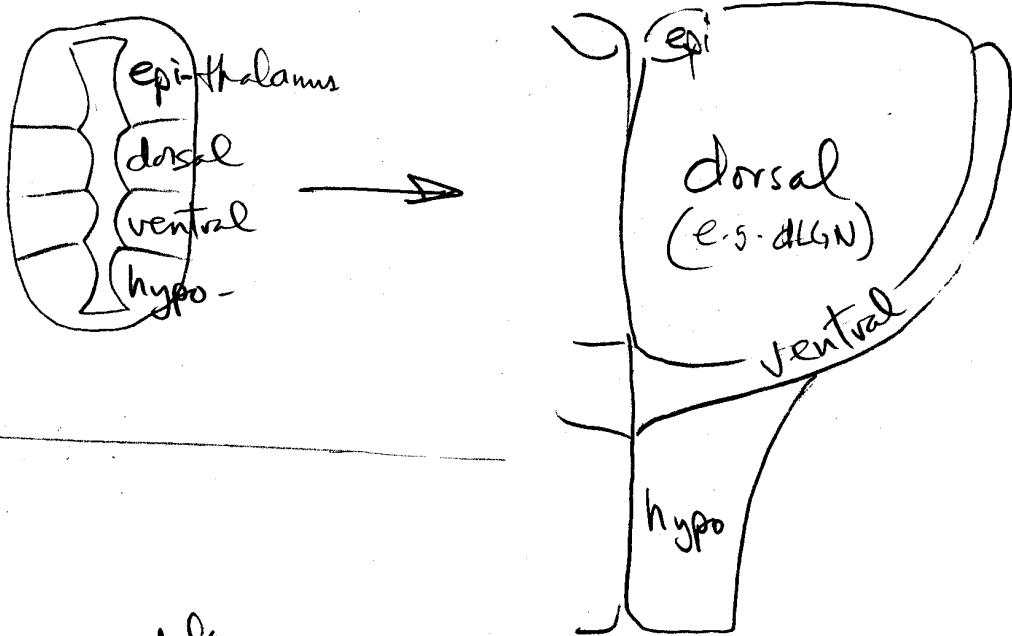
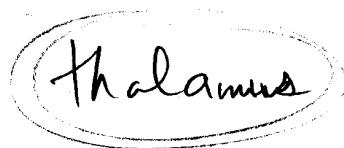
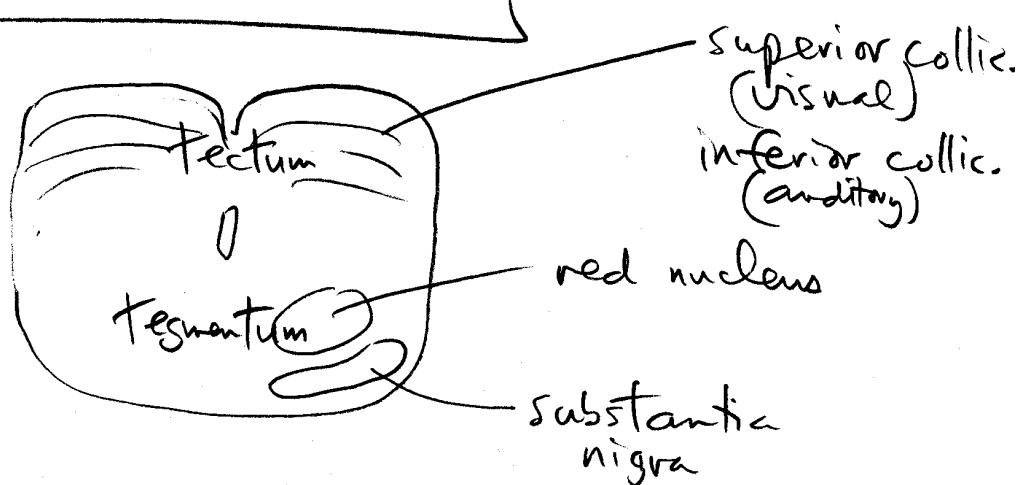
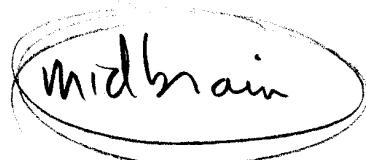


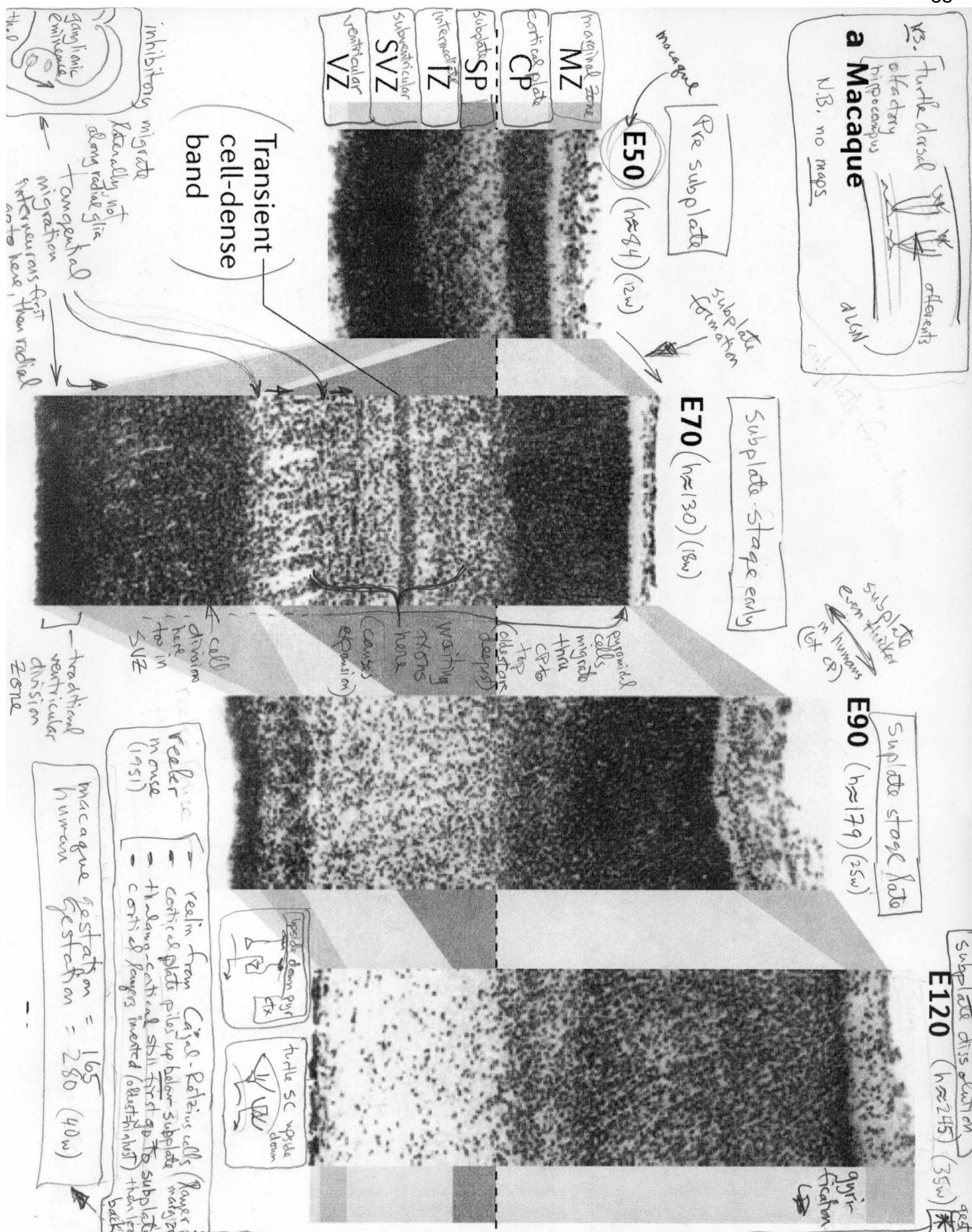


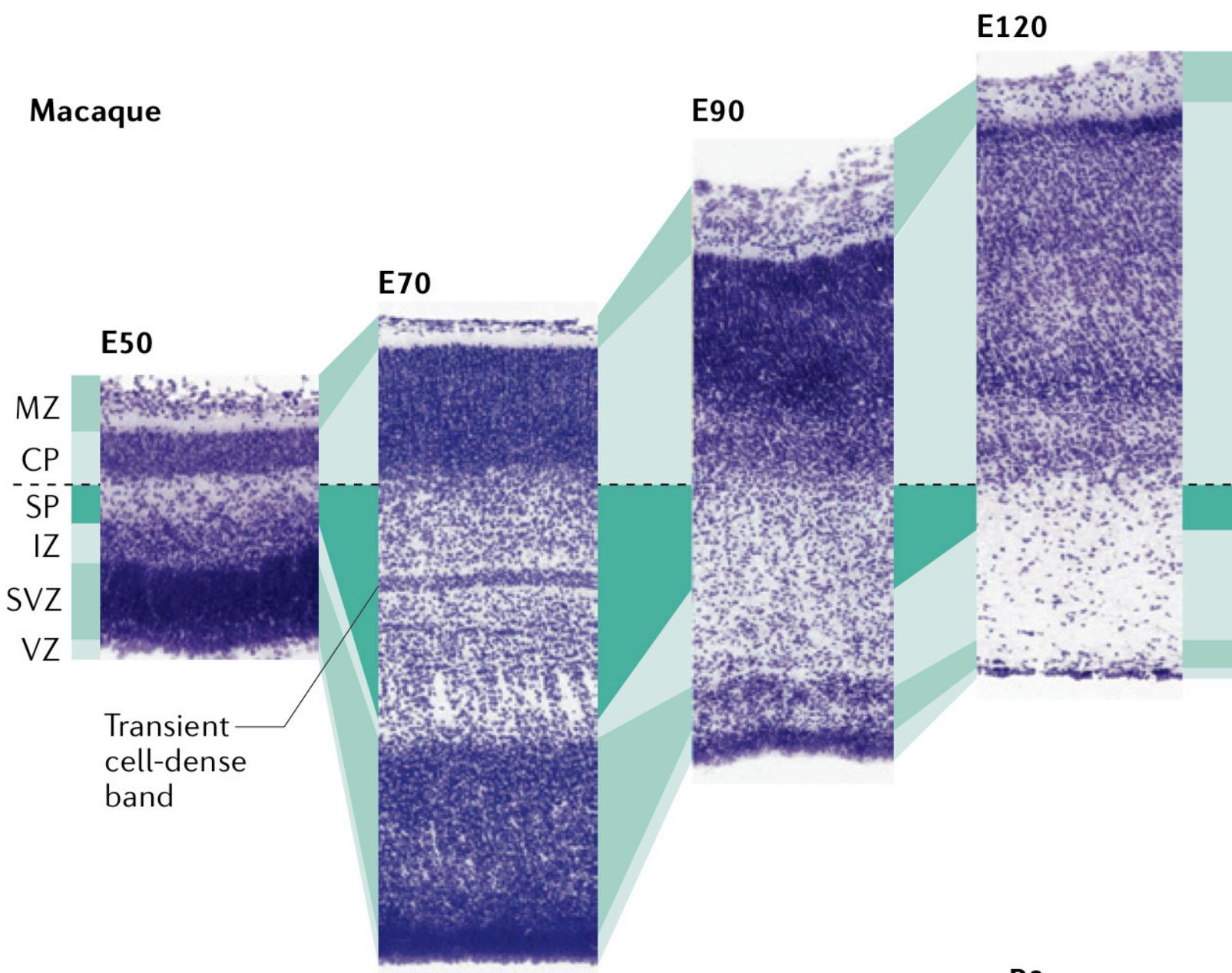
MAIN DIVISIONS

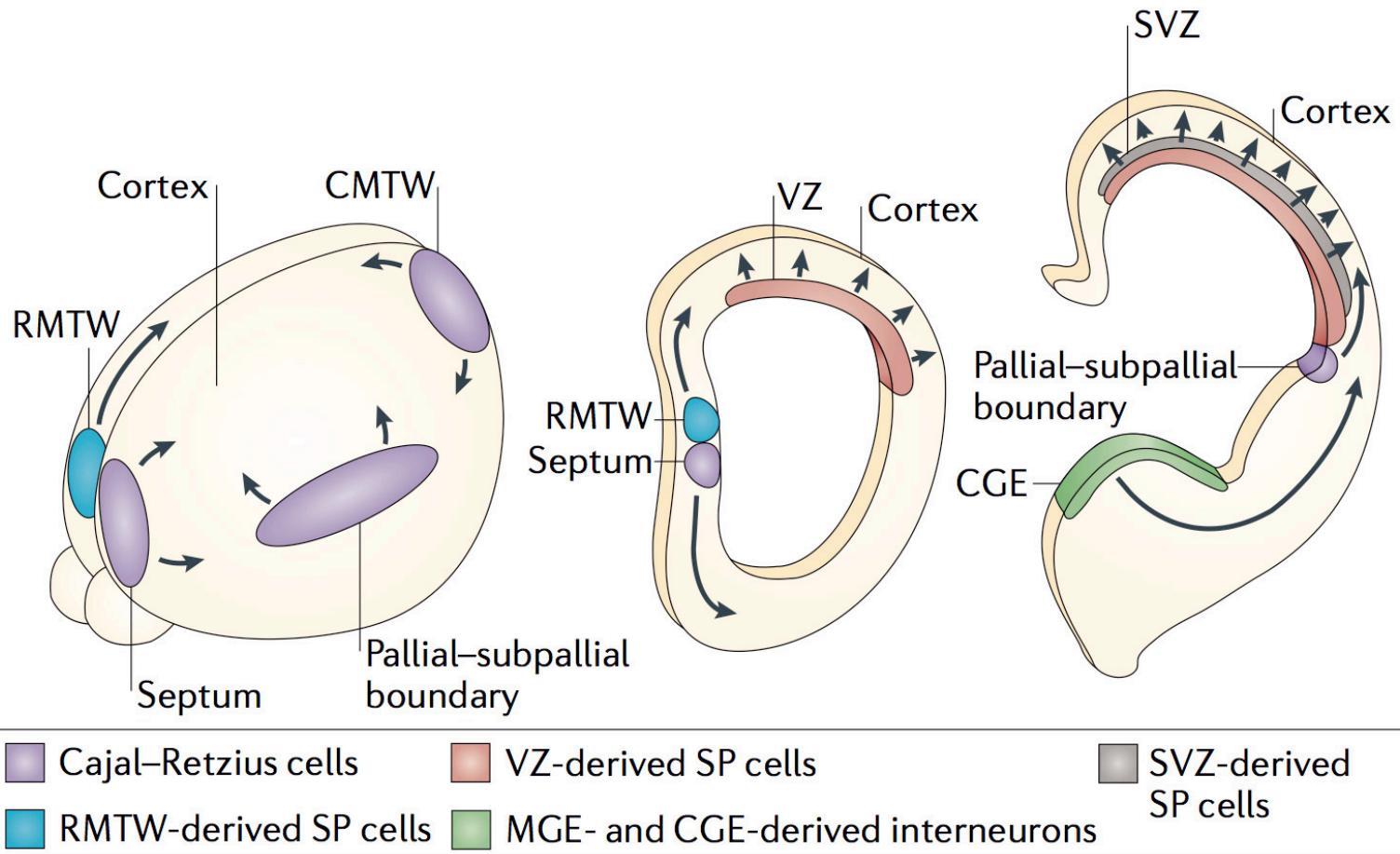


MAIN DIVISIONS (cont)



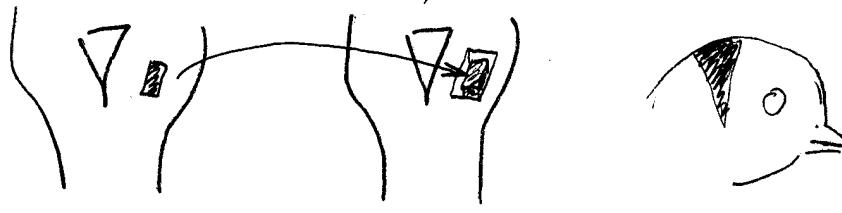






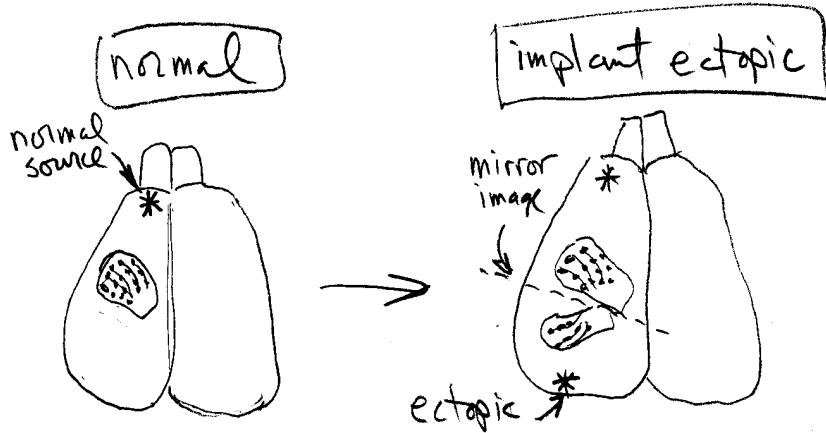
DEVELOPMENT EXPTS

chick/quail embryonic brain grafts



- operate on egg
- transplant vocalization "sound"
- transplant vocalization head movement

ectopic fgf8 source implant (mouse) (fibroblast growth factor #8)



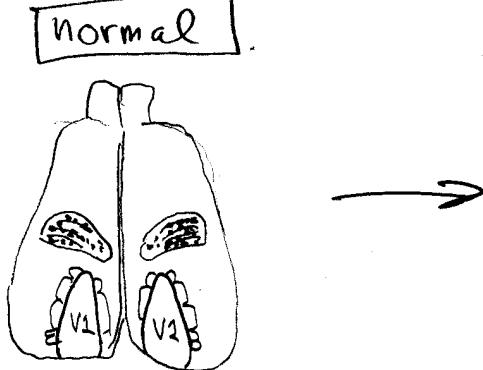
- many cortical areas are mirror-images, e.g.:
 - V1 / V2
 - SII / SII
 - AI / R
- Evol. of areas fusion splitting
- what is cortical area? cf. ear bone evolution function can move!

tailless (mouse mutant)

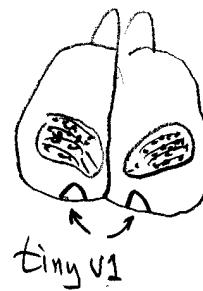
[$f_{ly} = Tll$
 $mouse = Trx$]

flies : structures from segments beyond 8th abdominal

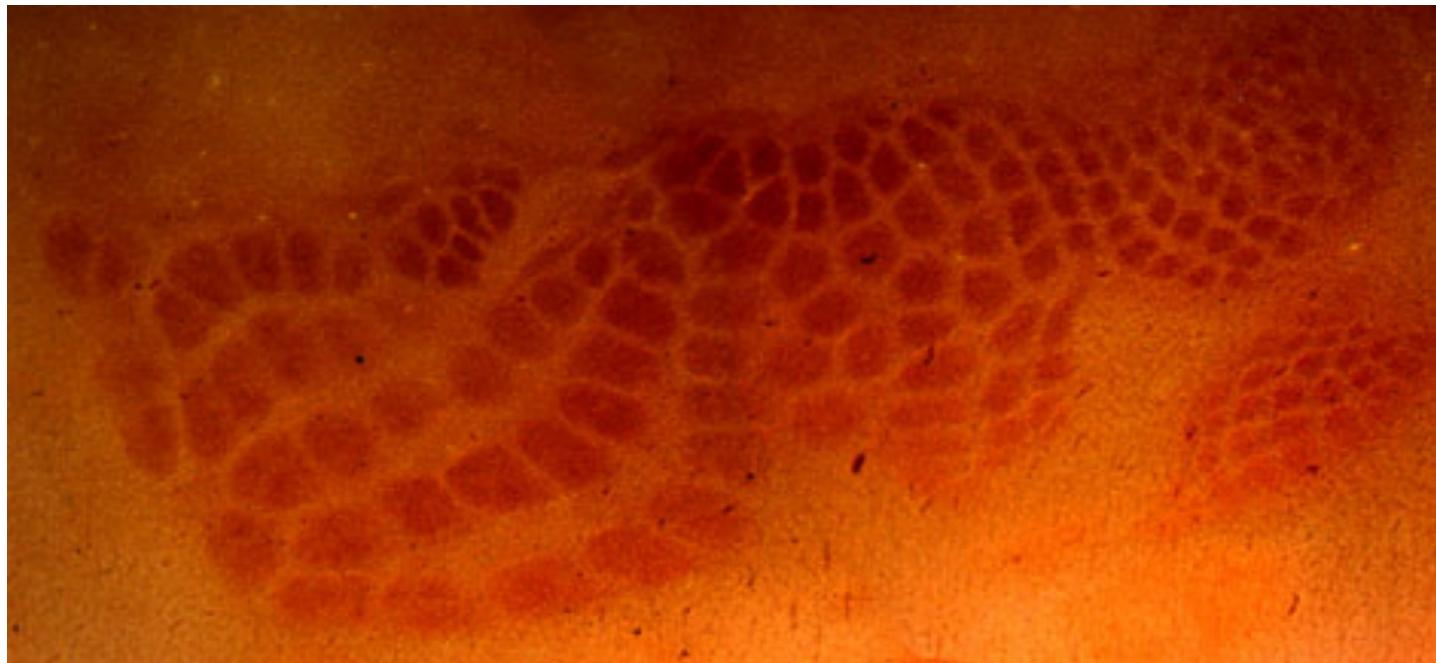
mice : missing tail, small cortex, aggressive behavior, no neurogenesis in subventricular zone



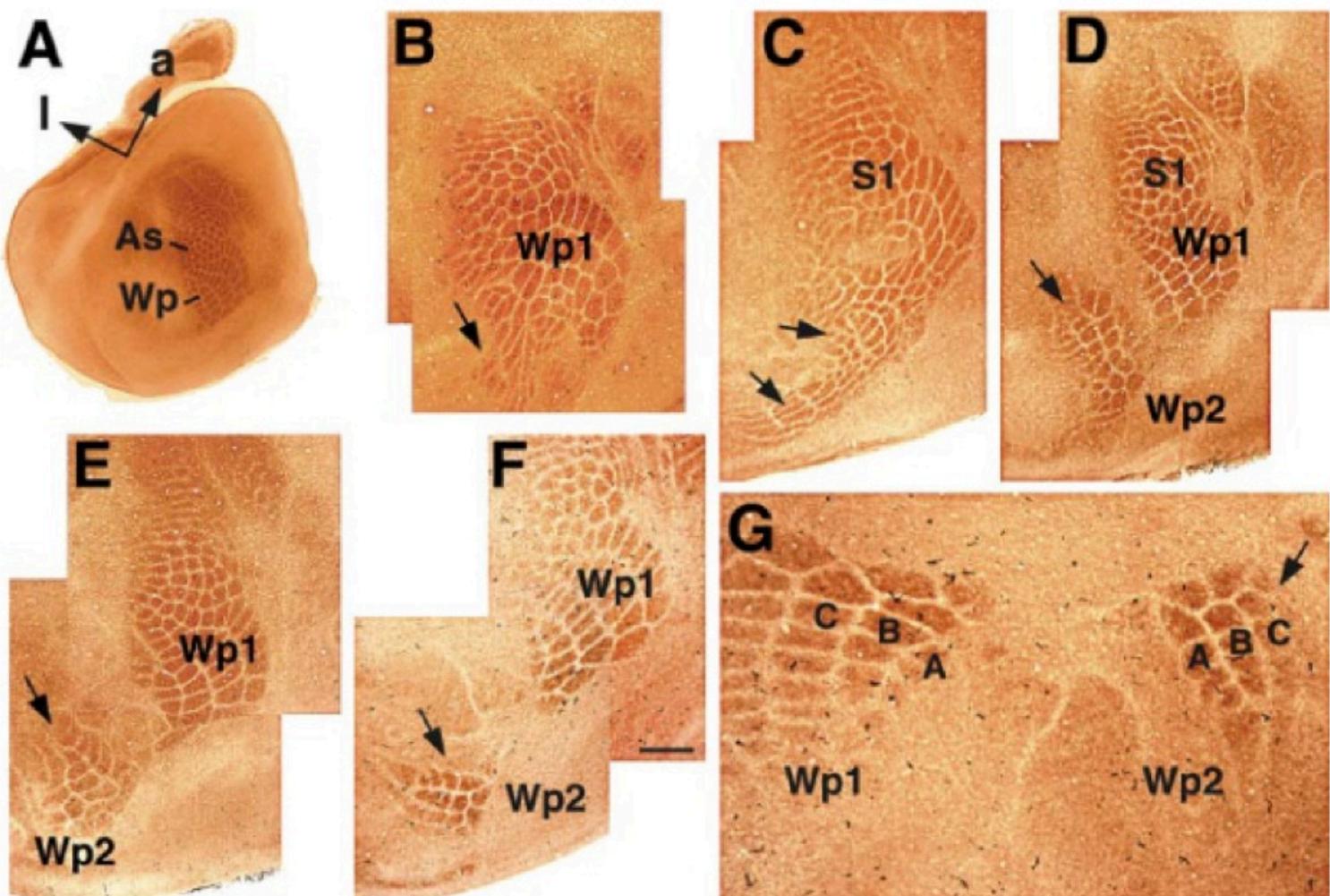
Tlx (tailless)



- normal size S-I
- tiny V1
- cortex does not "regulate" (adjust to give similar proportions)



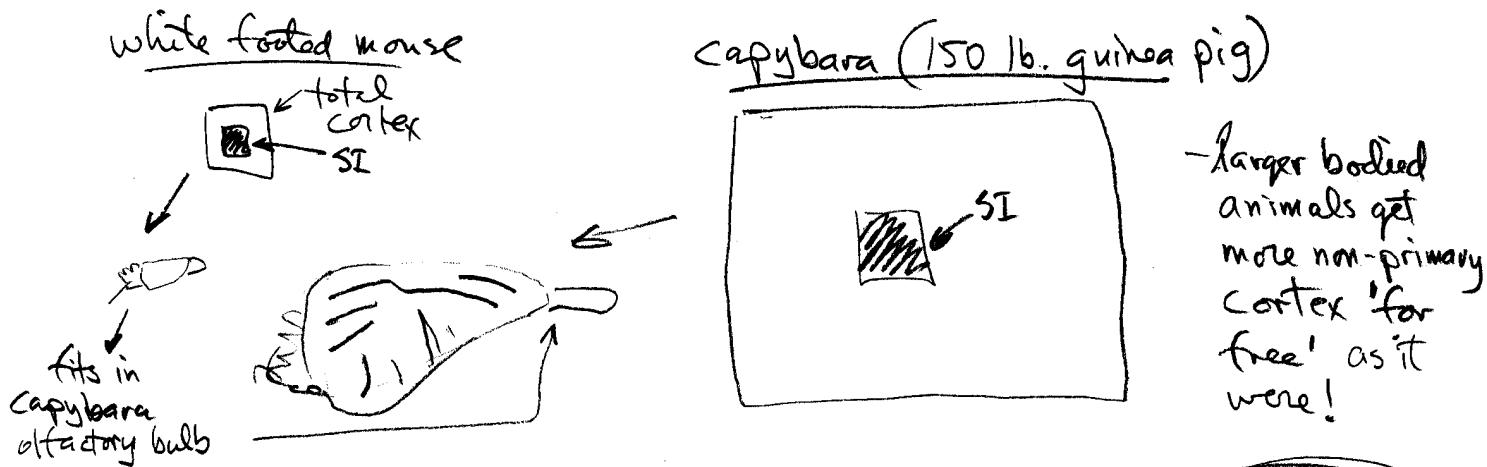
fgf8: Duplication and Divergence



FORMATION OF GYRI/SULCI

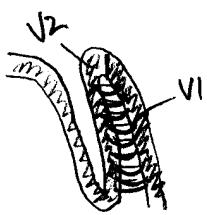
Brain size

- 95% of brain size variance explained by body size
- correlation body size and first PC of brain size measurements: 0.998
- 0.96 corr. of indiv. structures w/ total brain size
- as body/brain size increases, primary sensory areas occupy smaller percentages of total cortex

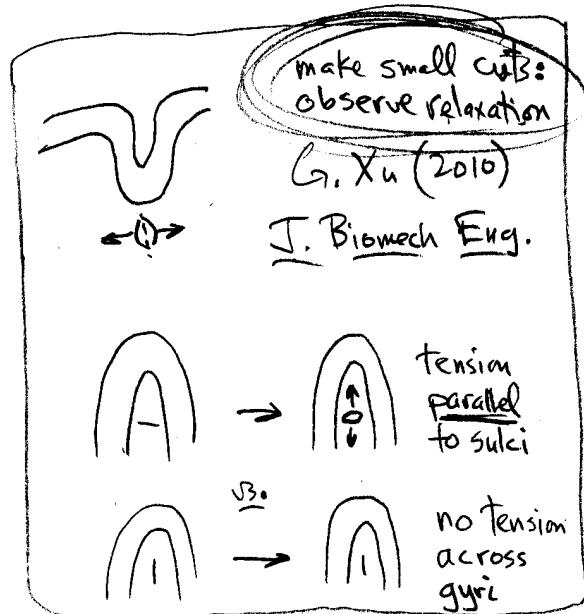
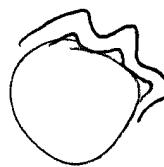
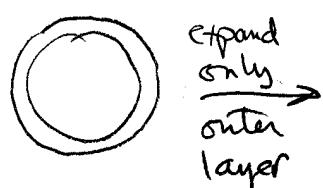


Formation of gyri (Tallinen)

Van Essen tension idea

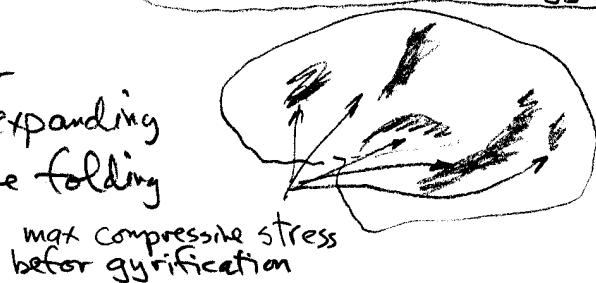


physical expts. suggest opposite



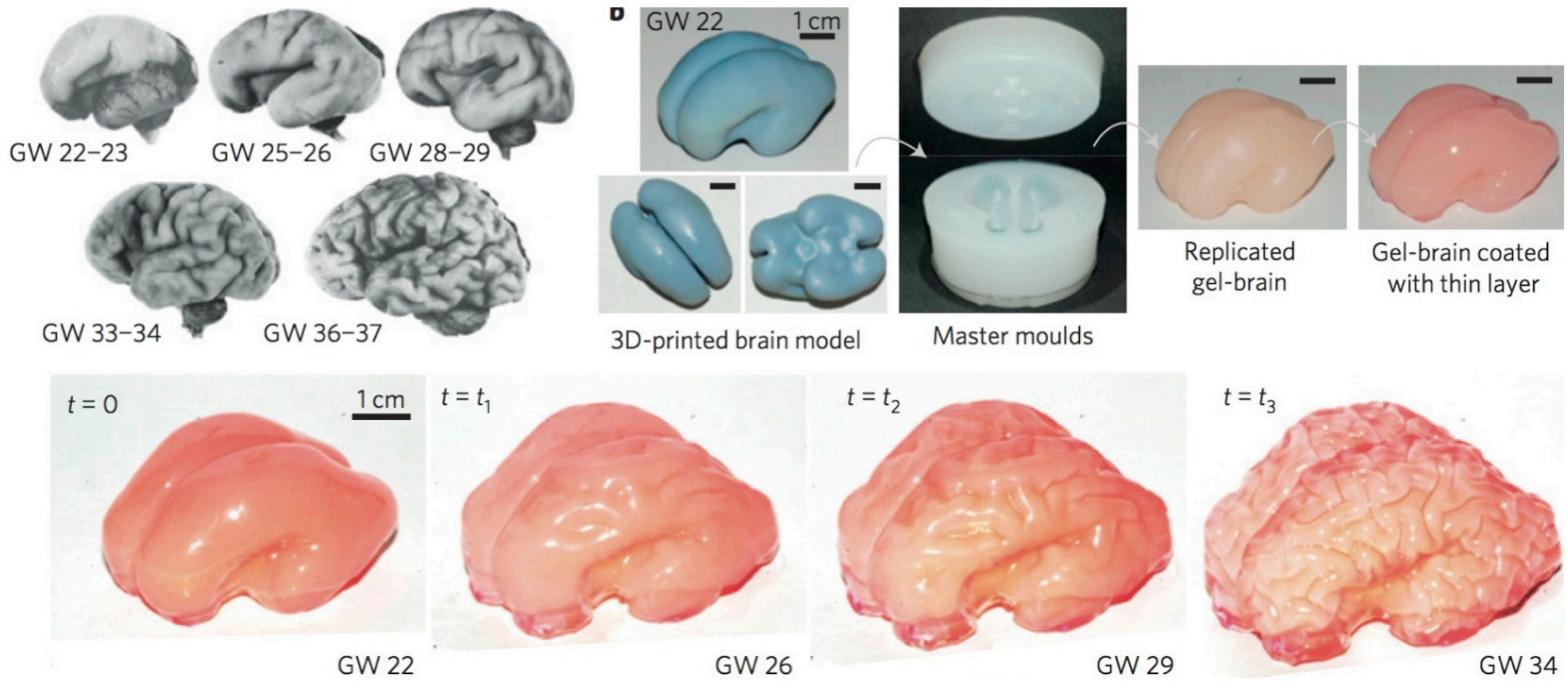
Tallinen (2016)

- coat expanding gel (ctx) over rigid WM-shaped non-expanding armature \rightarrow human-like folding
- also, finite element models



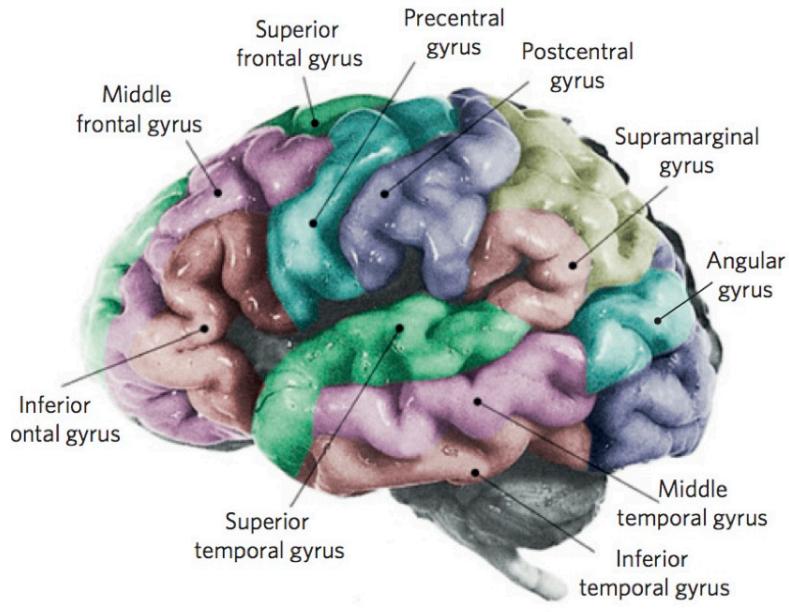
Physical Simulation

(thin outer elastomer gel layer expands over non-expanding gel brain base)

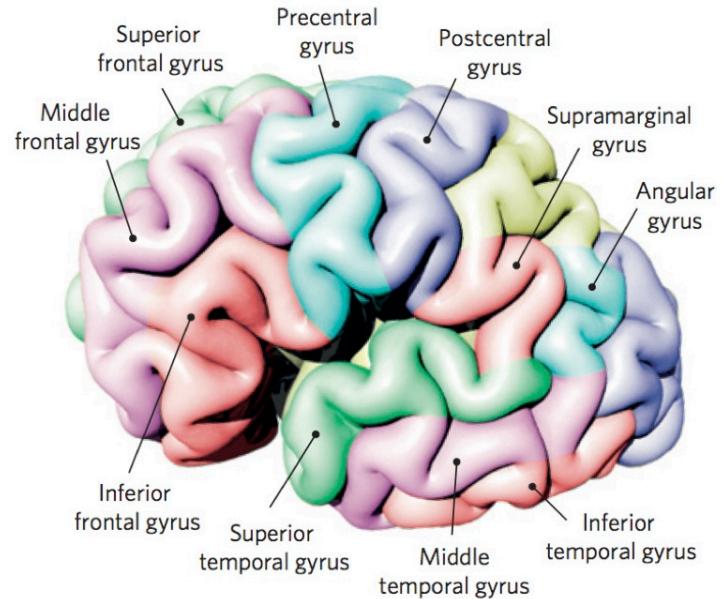


Tallinen et al. (2016) *Nature Physics* (doi:10.1038/NPHYS3632)

Real Brain

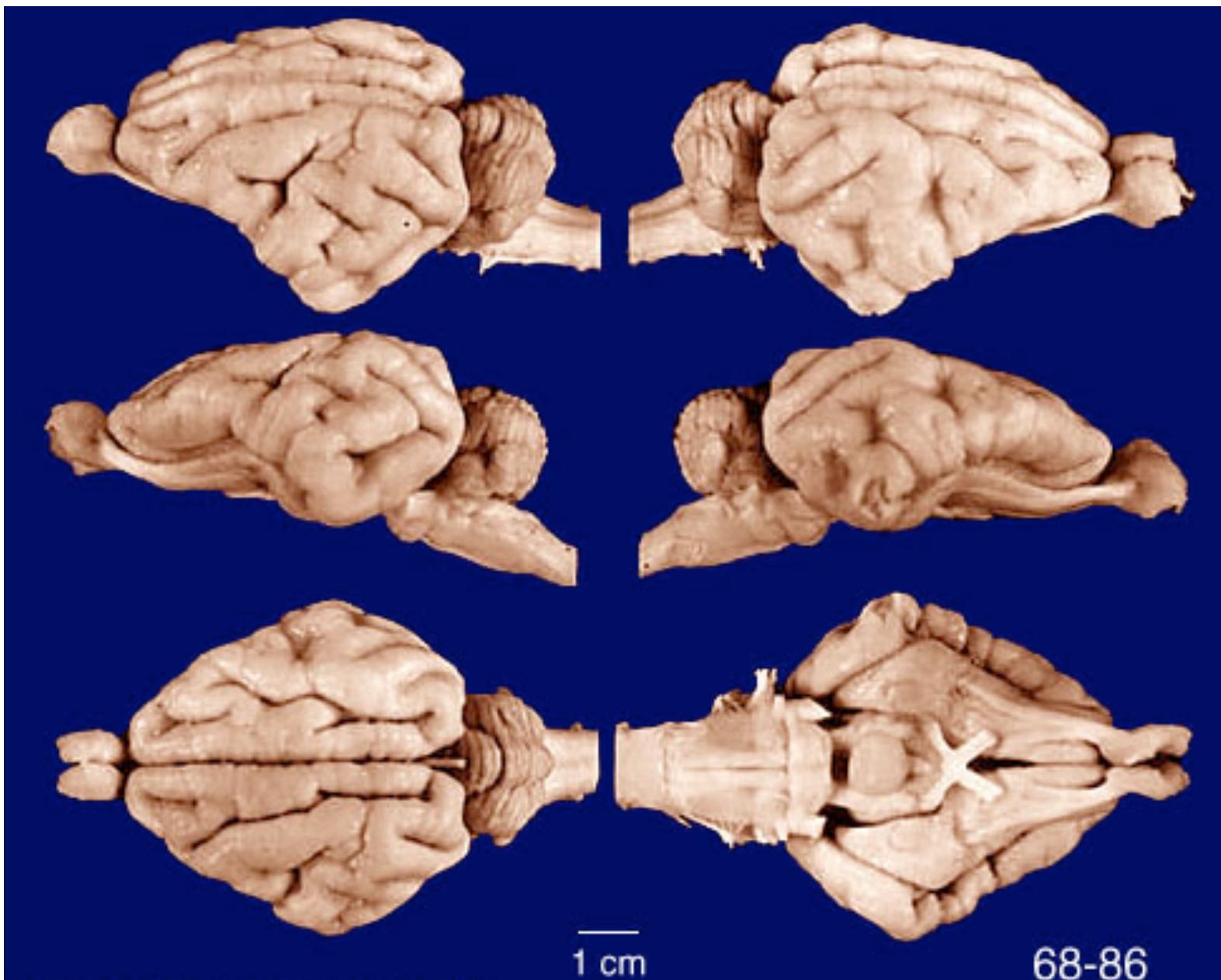


Numerical Simulation



Capybara

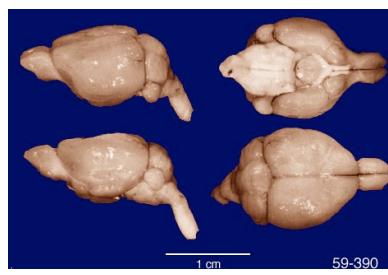
(150 lb. guinea pig)



68-86

Lab Deer Mouse

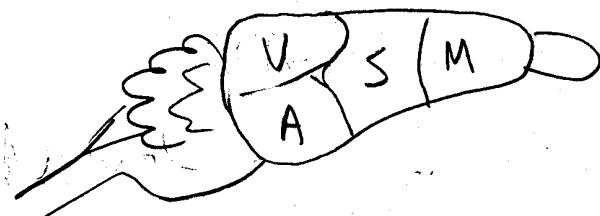
(same scale)



59-390

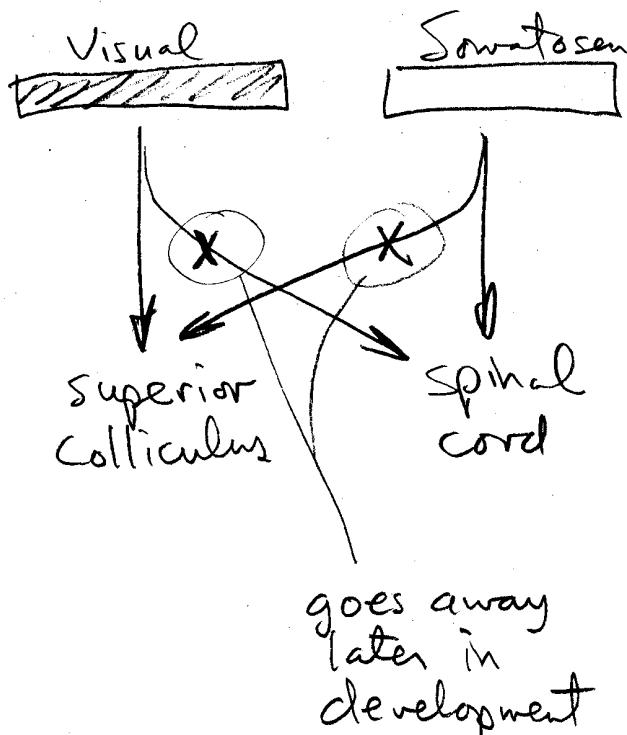
4)

O'LEARY EXPT

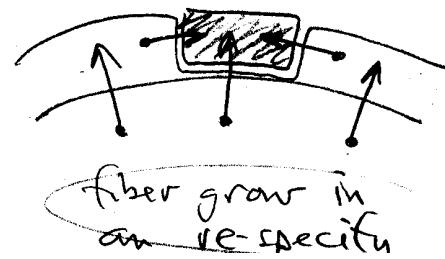
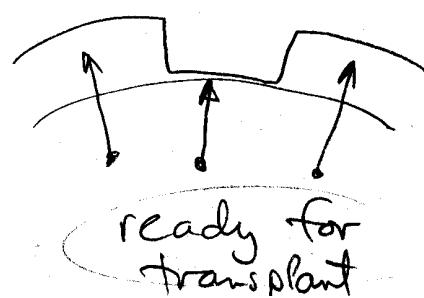
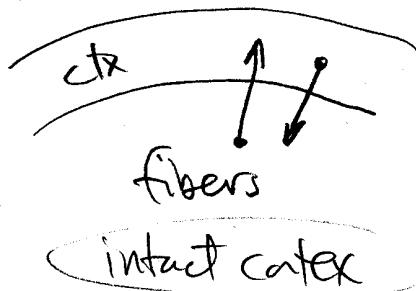
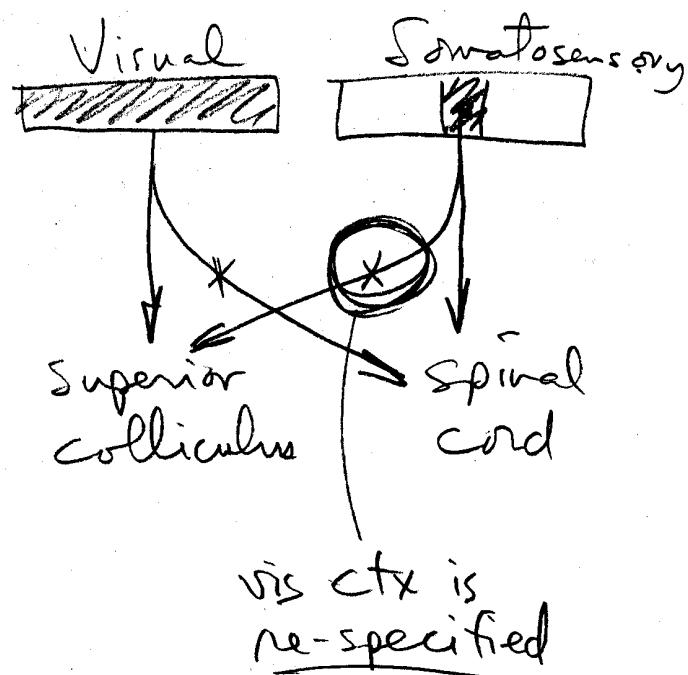
also

- Shatz LCN Ramamurti
- Cohen-Tannoudji gene expt

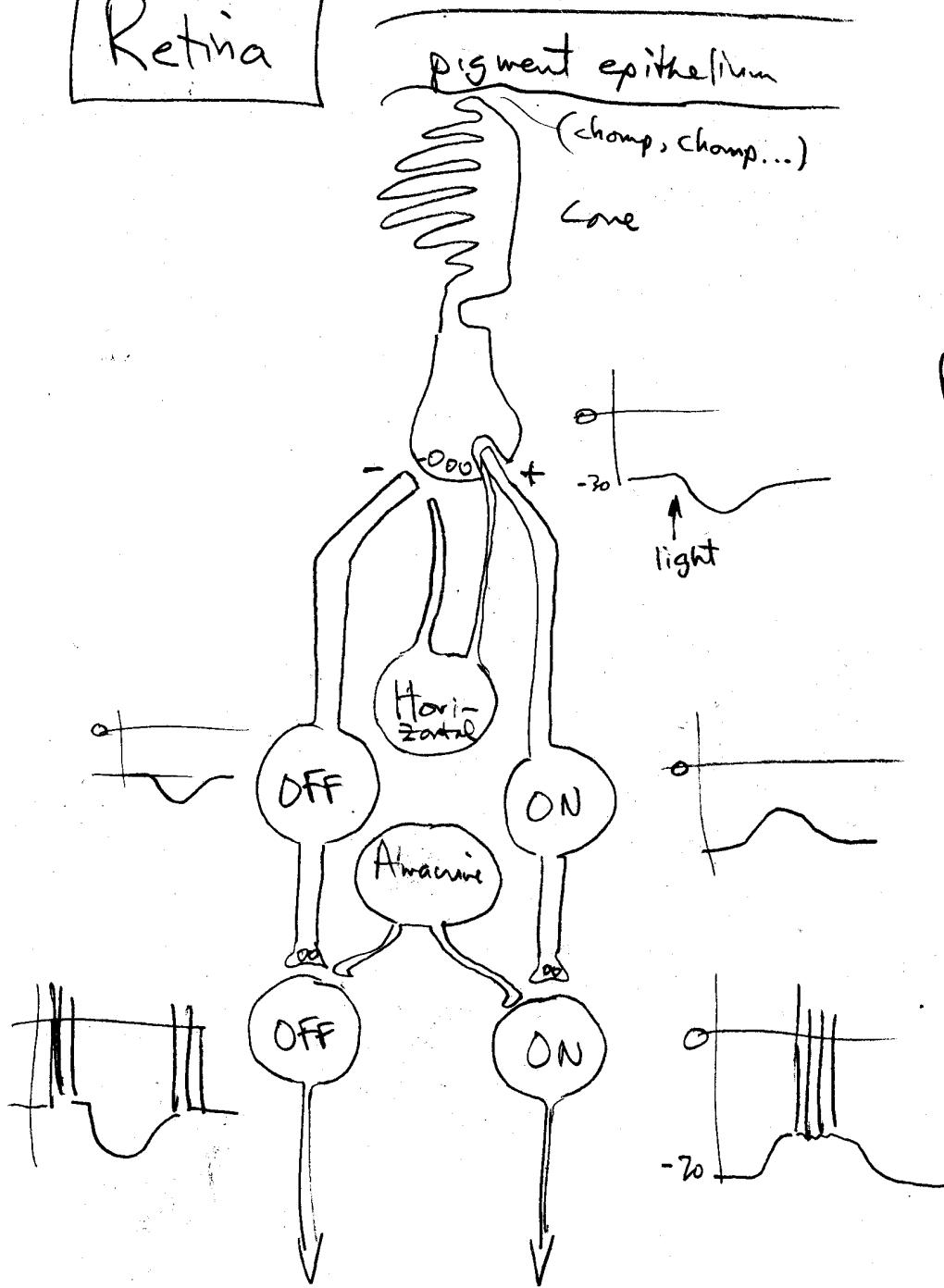
Normal



Transplant



Retina



X - Parvo
Y - Magno
blue - Konio

Photoreceptors (4 types)

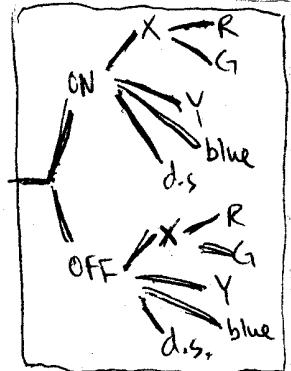
[dark current
light closes channels → repolarization
repolarization → less neurotransmitter release]

Horizontal cells (2 types)

Bipolar cells (3 types) (ON, OFF, rod)

Amarine cells (20 types)
AII, A1-25...
startburst

Ganglion cells (10+ types)



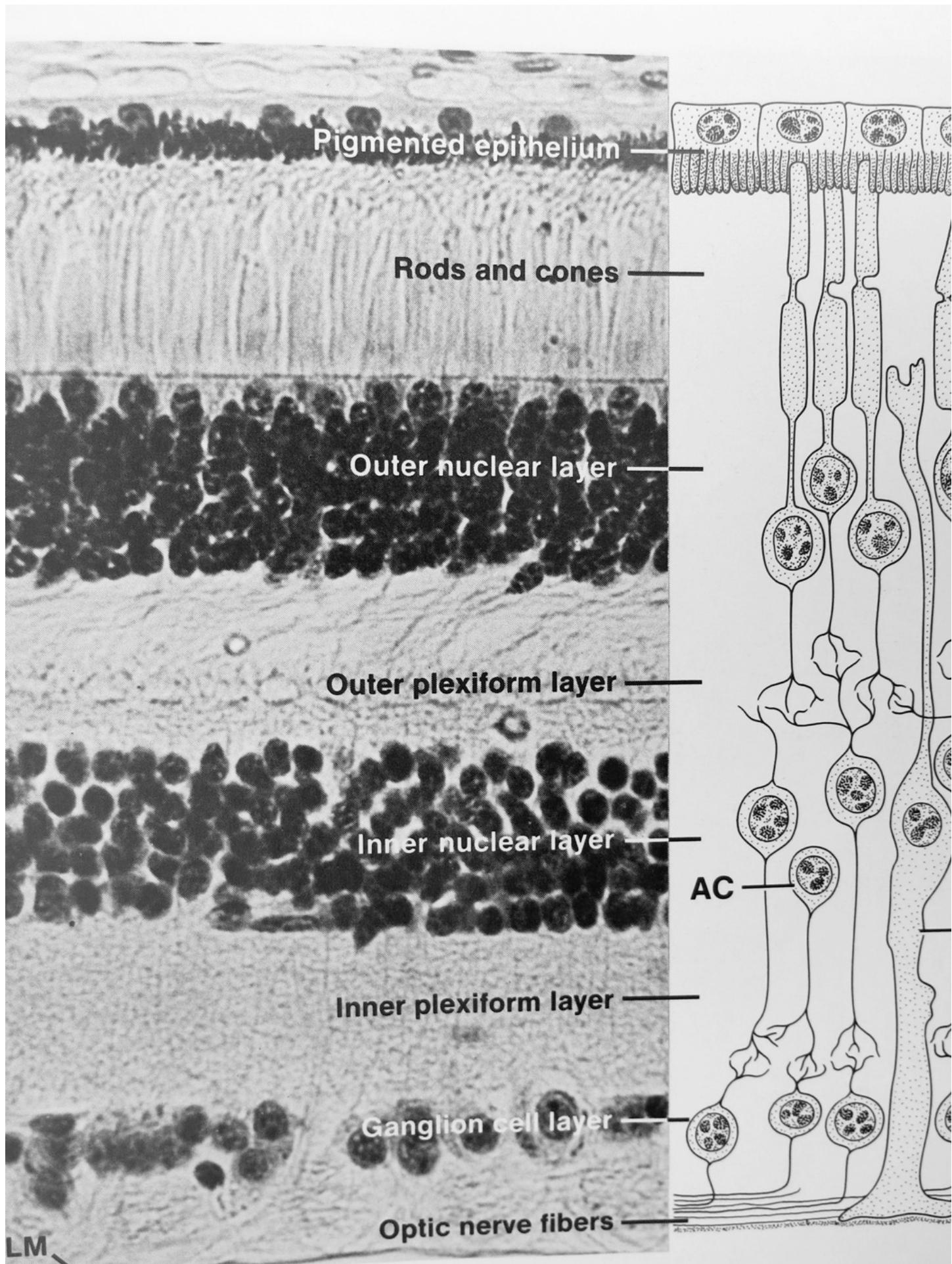
ganglion cell types

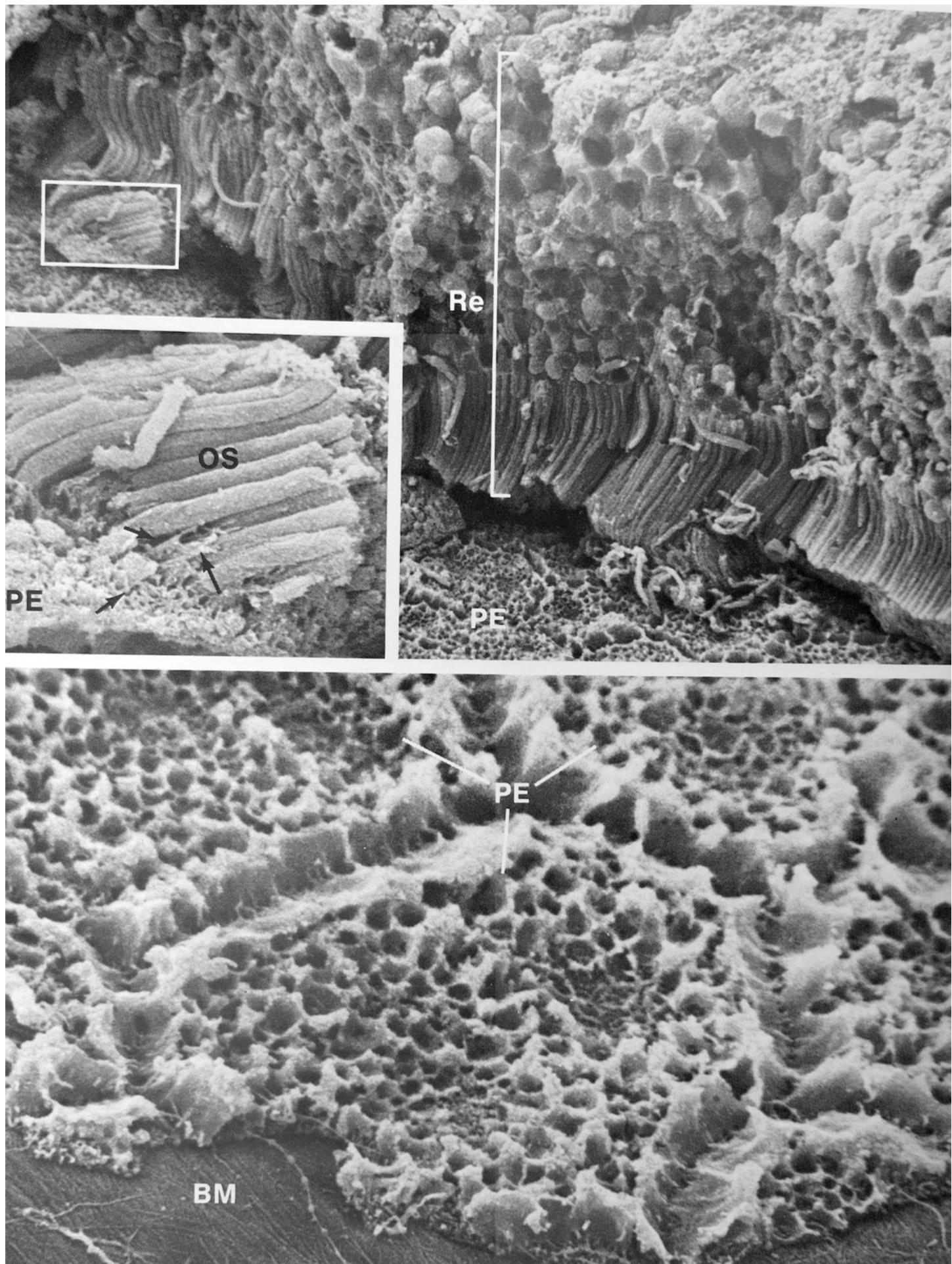
X
sustained
color selective
smaller RF
linear

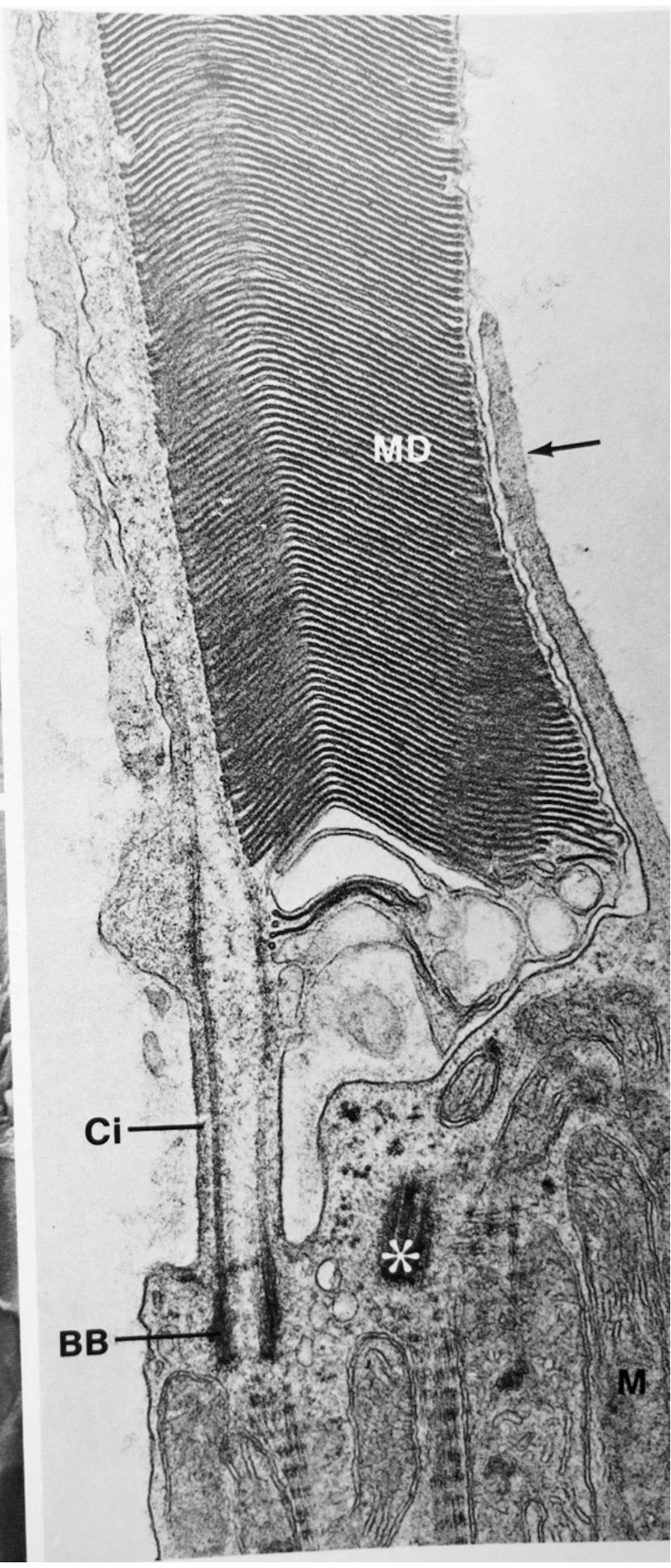
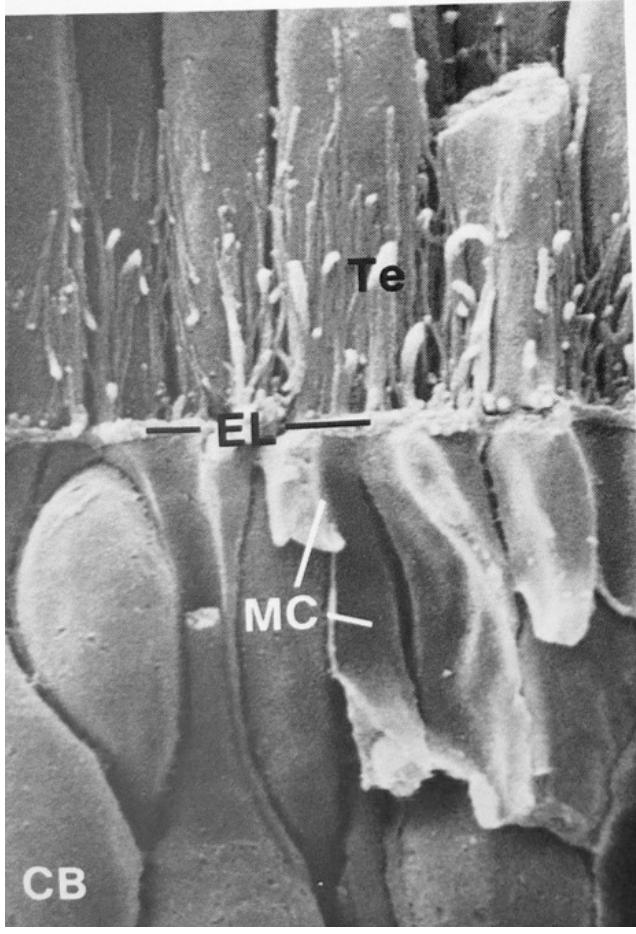
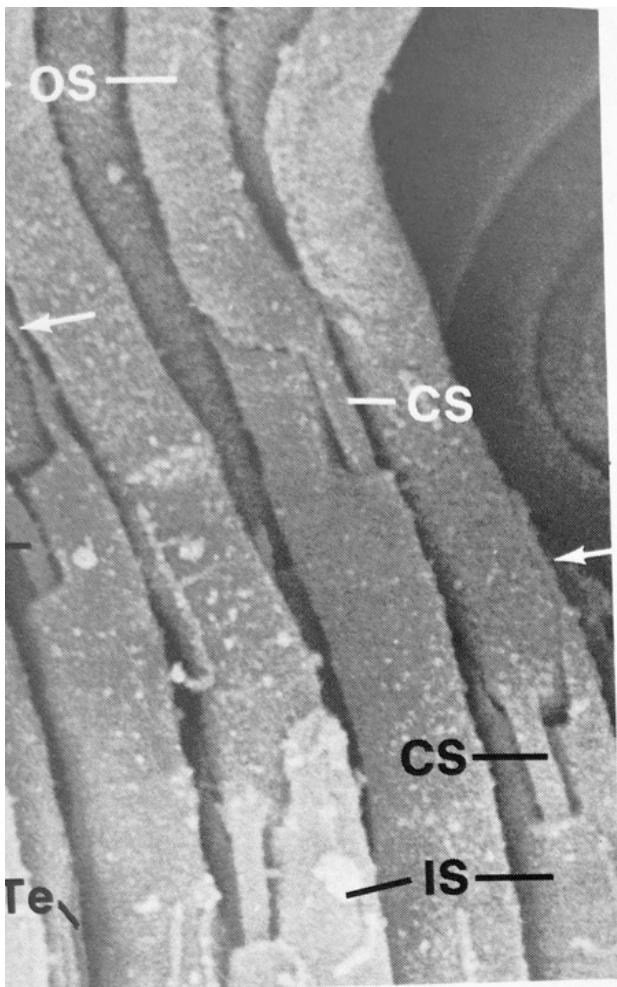
- cf.
Somatosens.

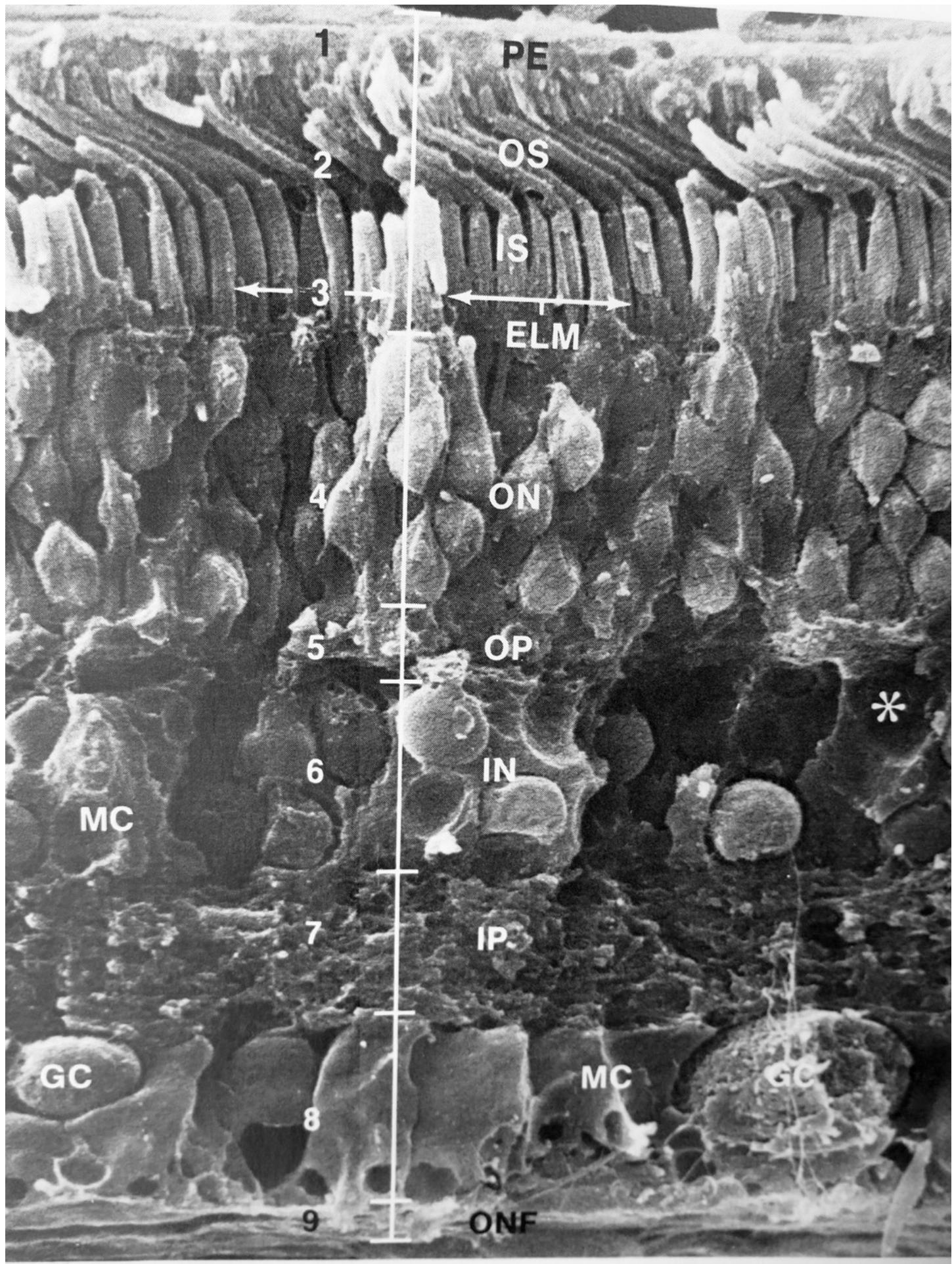
Y
transient
not color-selective
larger RF
non-linear

(midget)
X (ON) R cent
OFF G cent \Rightarrow LGN
Y (ON) R cent
OFF G cent \Rightarrow LGN, colliculus
blue ON OFF \Rightarrow LGN, colliculus
W (ON/OFF) \Rightarrow LGN, colliculus
(cat)
direction selective \Rightarrow pretectum (MTN, LTN)
thorn cells \Rightarrow colliculus

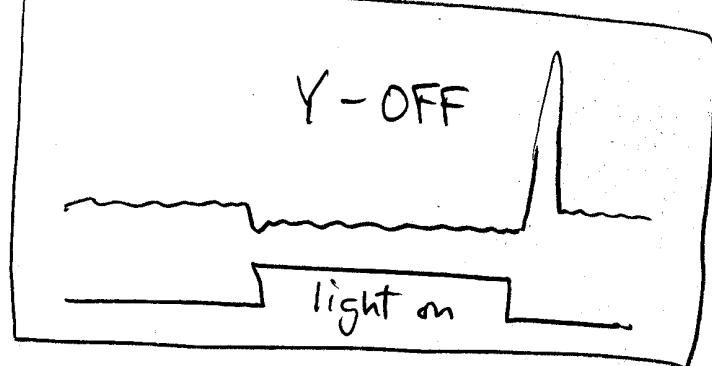
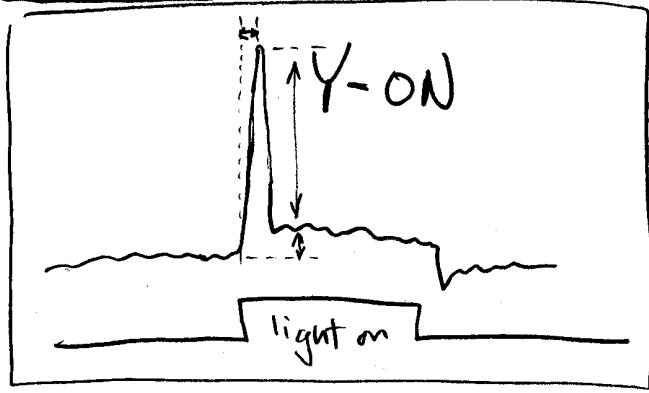
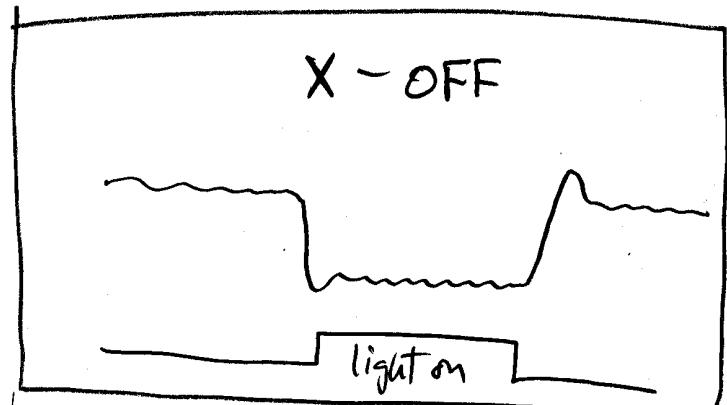
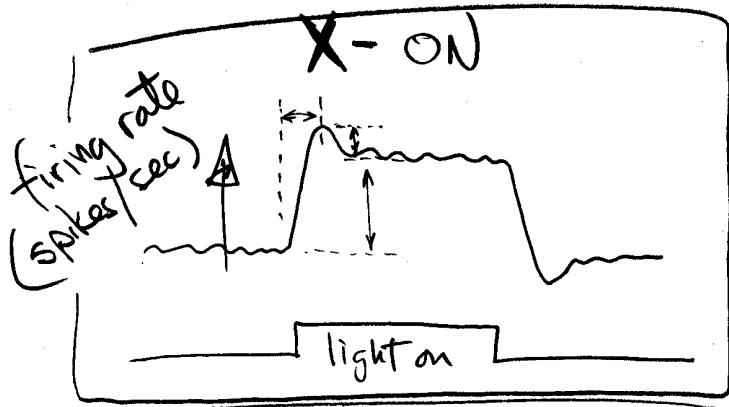








Ganglion Cell Types



- this is like brightness & derivative of brightness

$$f(x) = \begin{cases} 0 & x < 0 \\ 1 & x \geq 0 \end{cases}$$

$$f'(x) = \begin{cases} 0 & x < 0 \\ 1 & x = 0 \\ 0 & x > 0 \end{cases}$$

→ typical cells have both, but emphasize one

Other features of ganglion cells

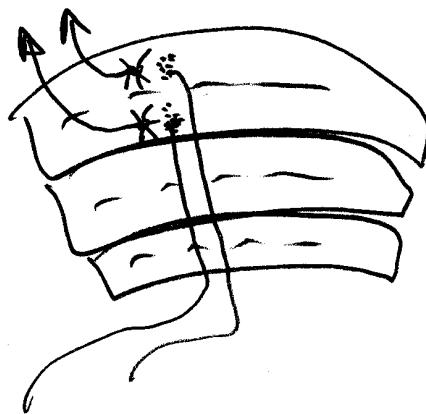
- colliculus cell types

- | |
|-----------------------------|
| red center / green surround |
| green center / red surround |
| blue |

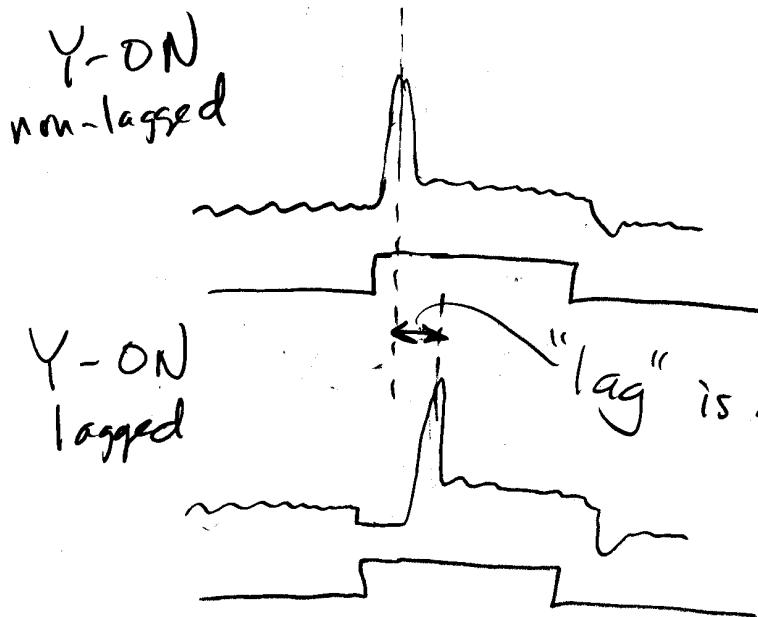
luminance (Y cells - red \neq green equal if brightness)

dLGN additions (not too many)

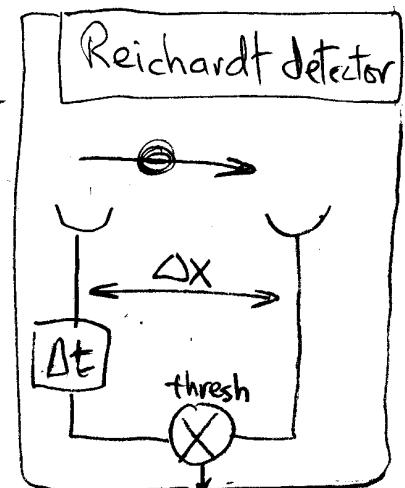
- cells are still monocular in dLGN despite being lined up



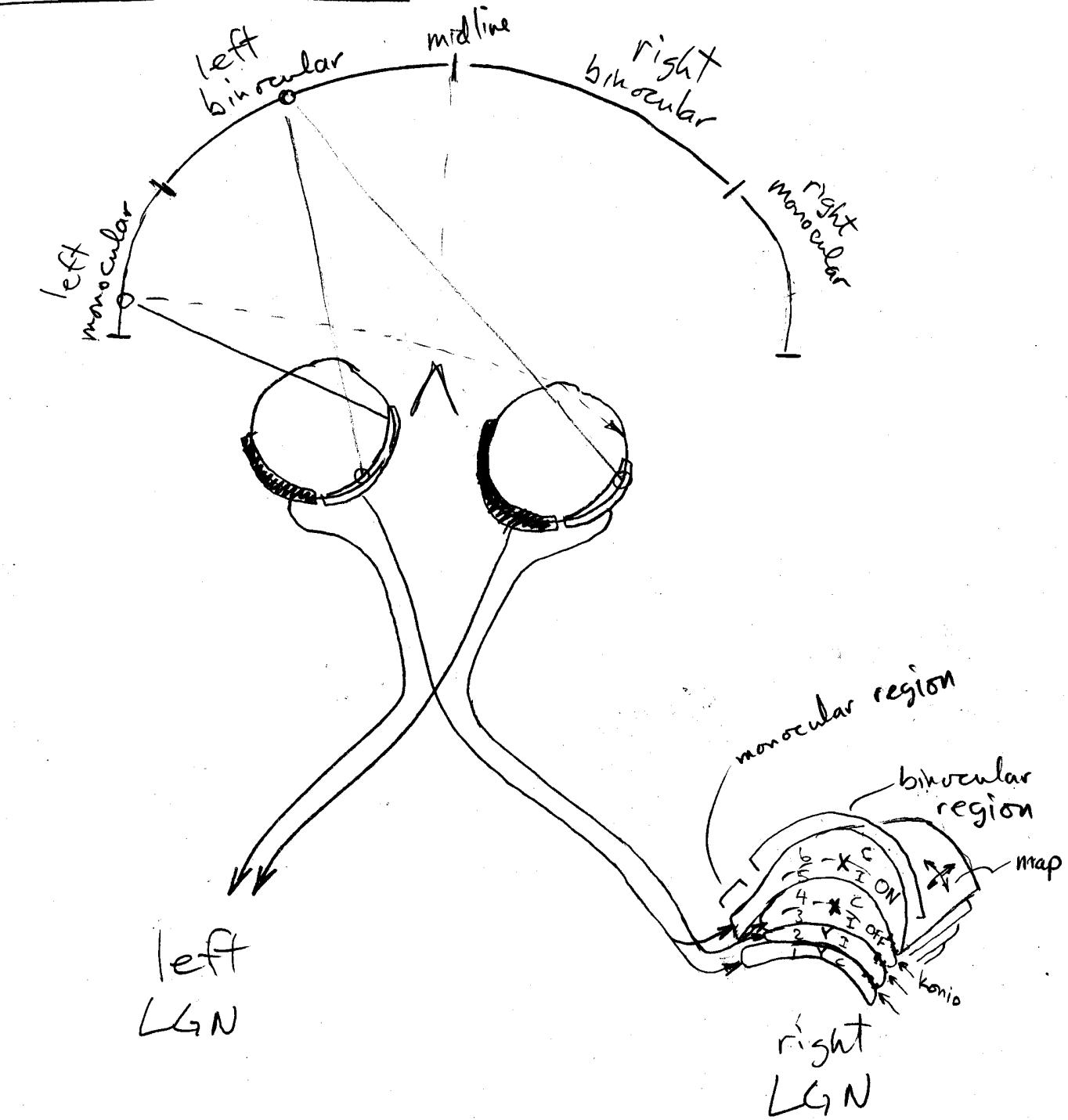
- non-lagged vs. lagged (cat dLGN)



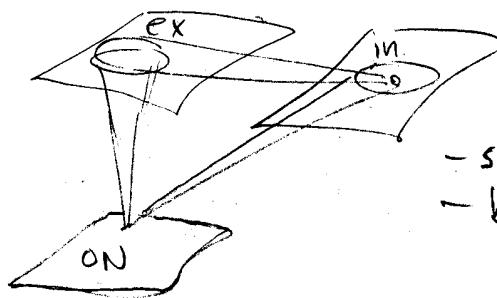
N.B.: in primates,
this operation is
postponed until the
cortex (V1) $4C\alpha \rightarrow 4B$



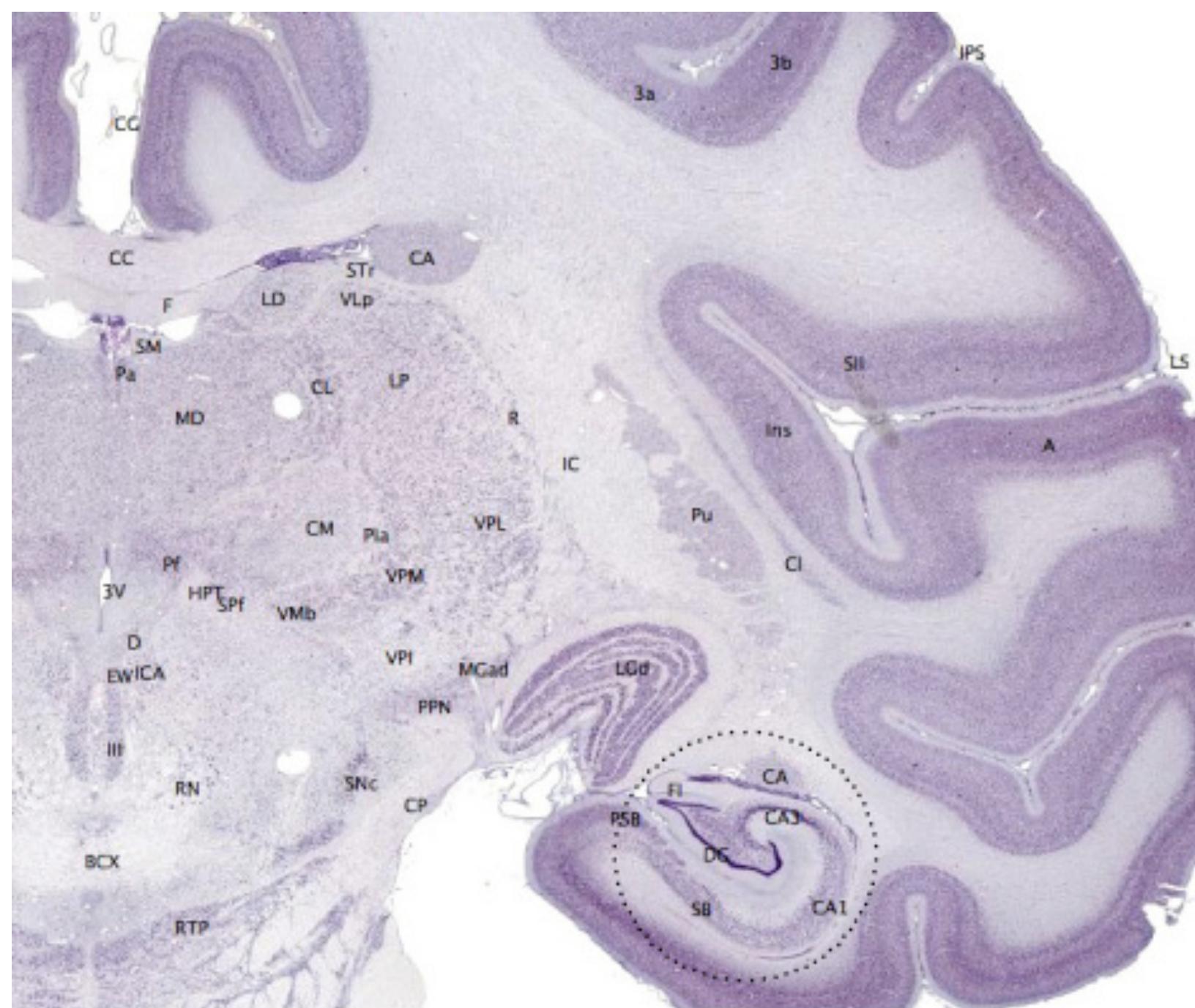
Retino geniculate

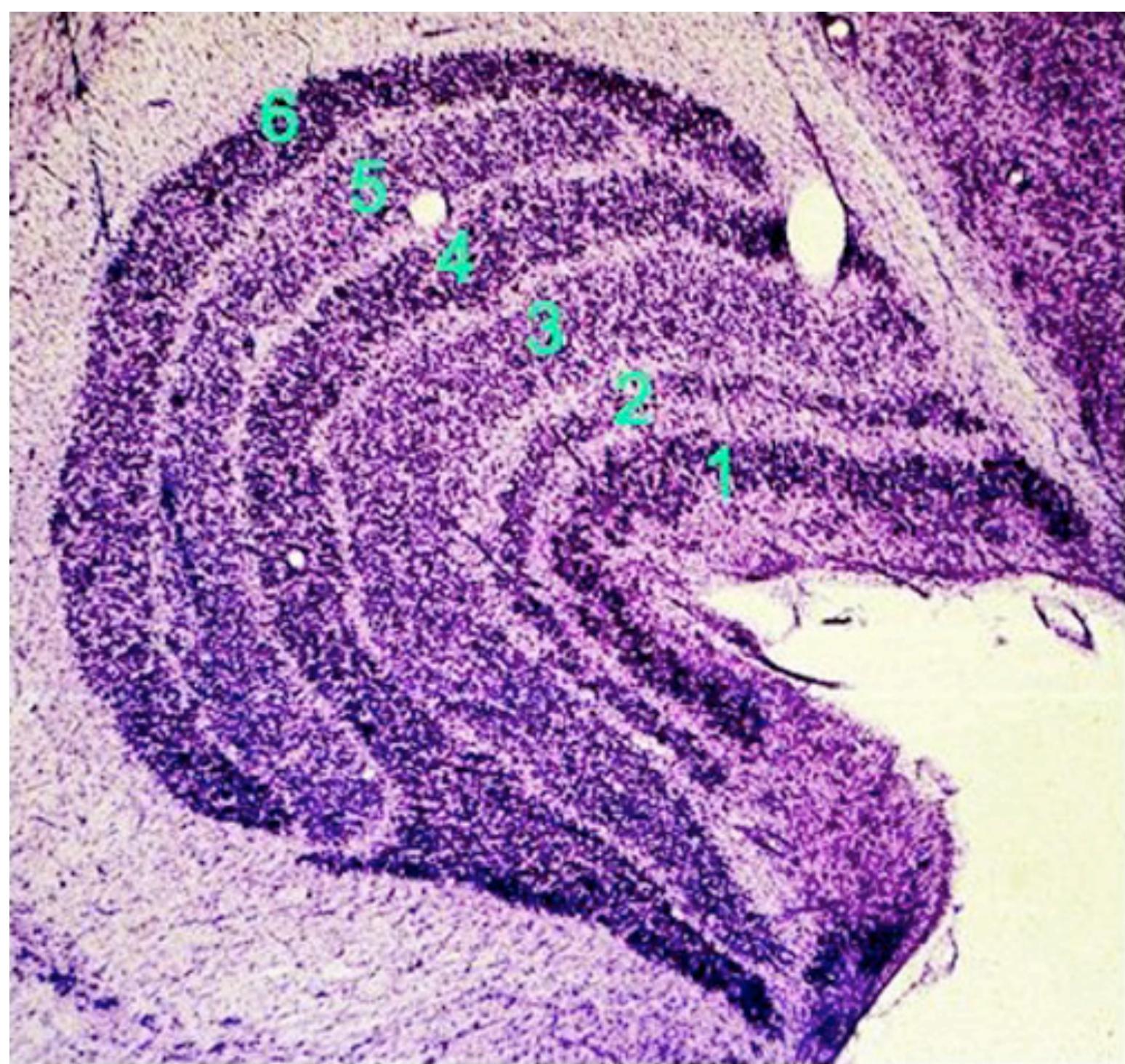


c.g.



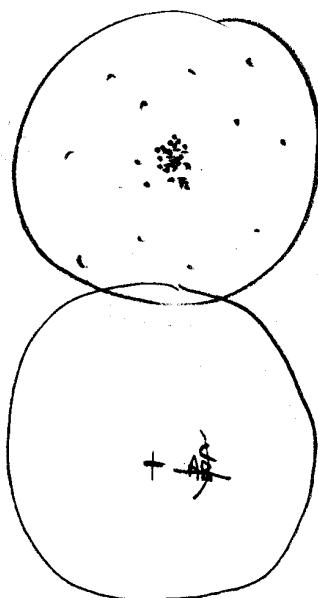
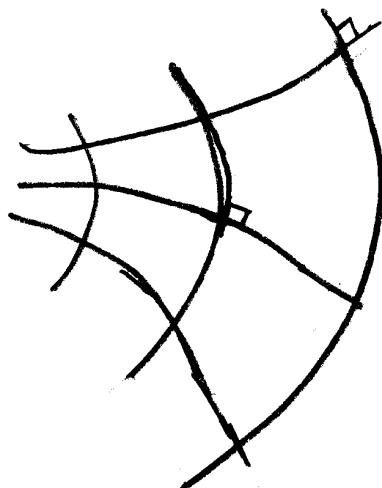
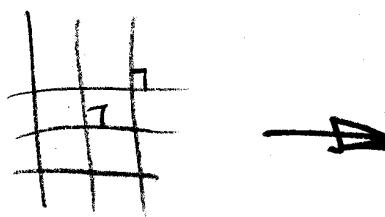
- surround is inhib (not OFF)
- bigger because feed forward inhib. is disynaptic



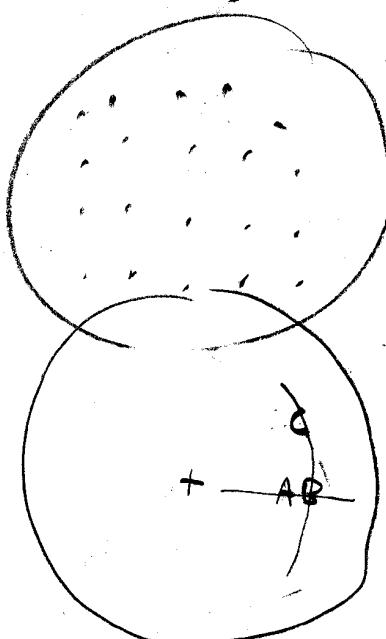


Maps

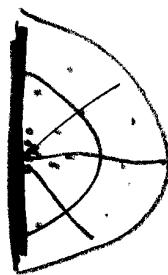
Conformal Maps → angle preserving



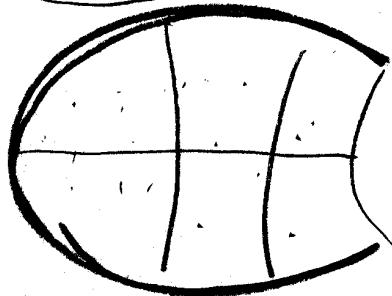
can't be conformal
w/o cuts
w/ force

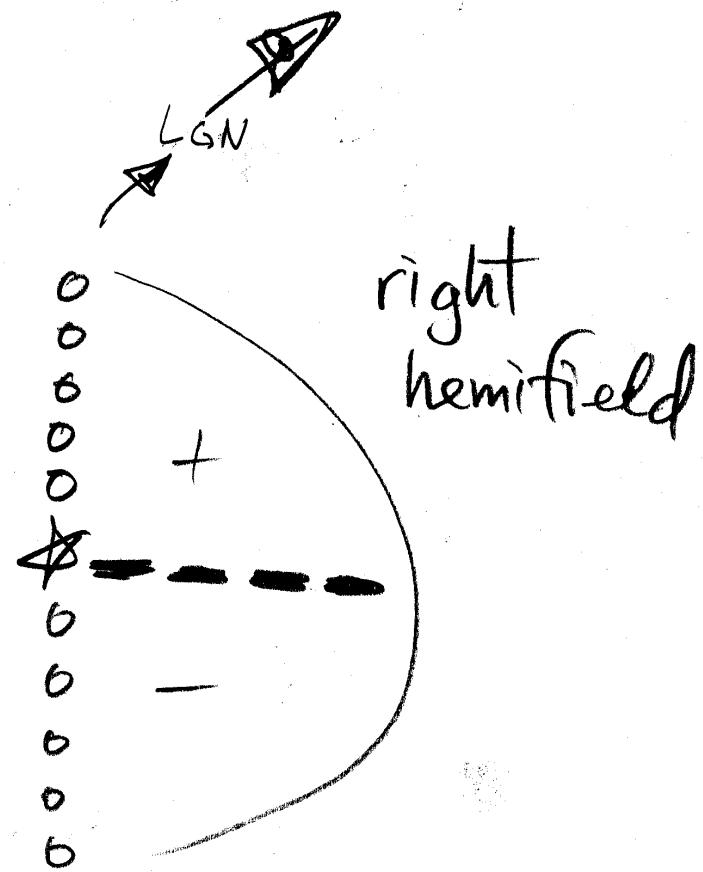
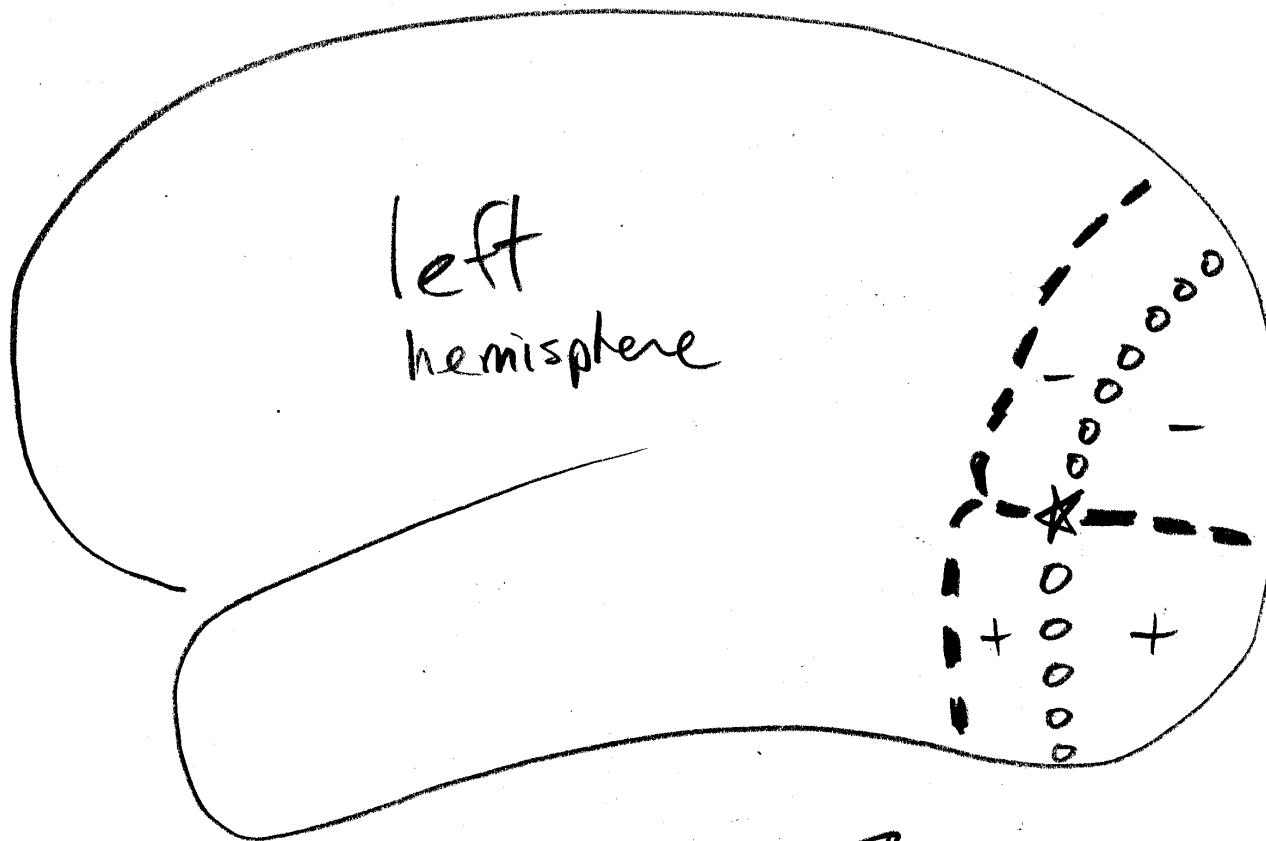


circumferential stretching
greater than radial stretching

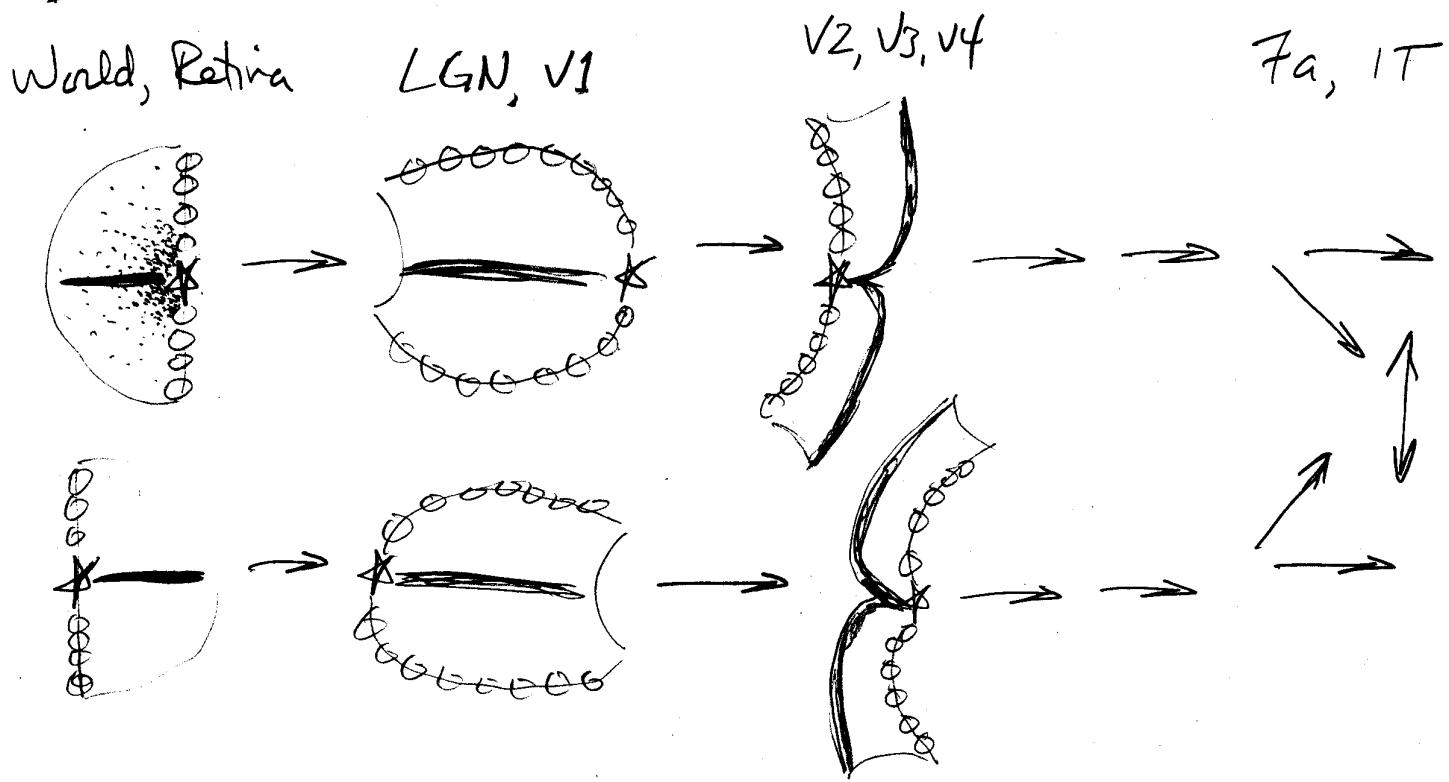


a hemifield
can be nearly conformal





Maps

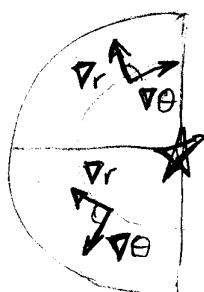


Mirror-image \cong Non-Mirror Image

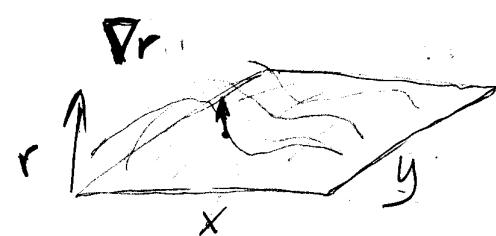
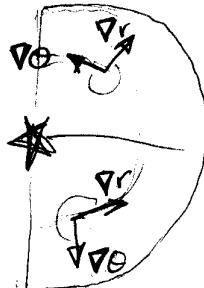
Map: Retinal Position (r, θ) \Rightarrow Cortical Position (x, y)

like $r = f(c)$
but r, c both have
two dimensions

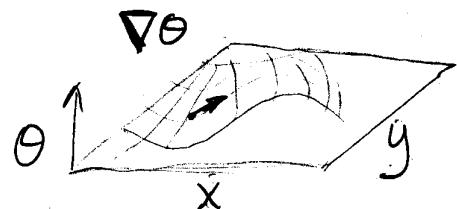
non-mirror image
cortical map of
left hemifield



mirror image
cortical map of
left hemifield



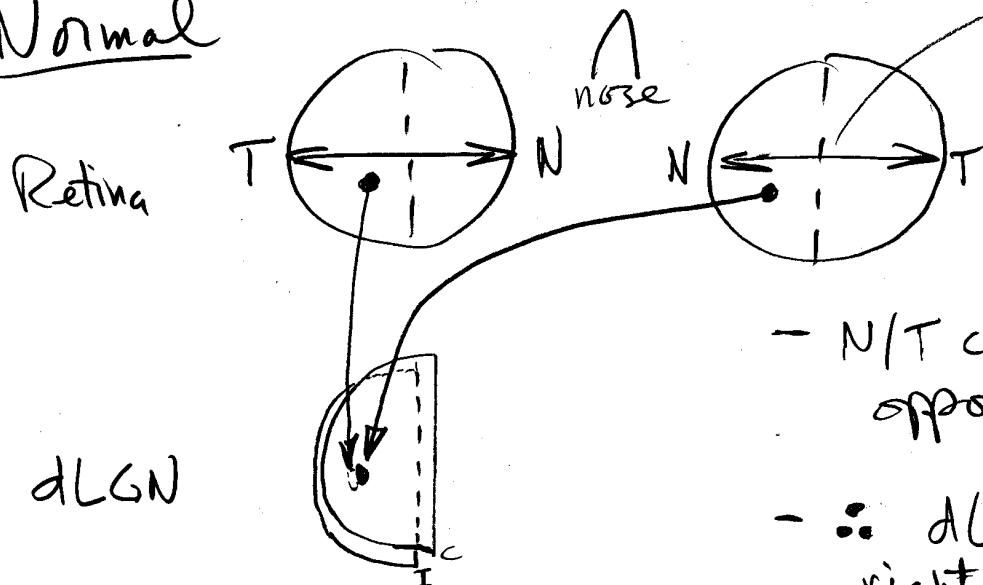
- steepest
uphill
direction
of r at
each x, y



- steepest
uphill
direction
of θ at
each x, y

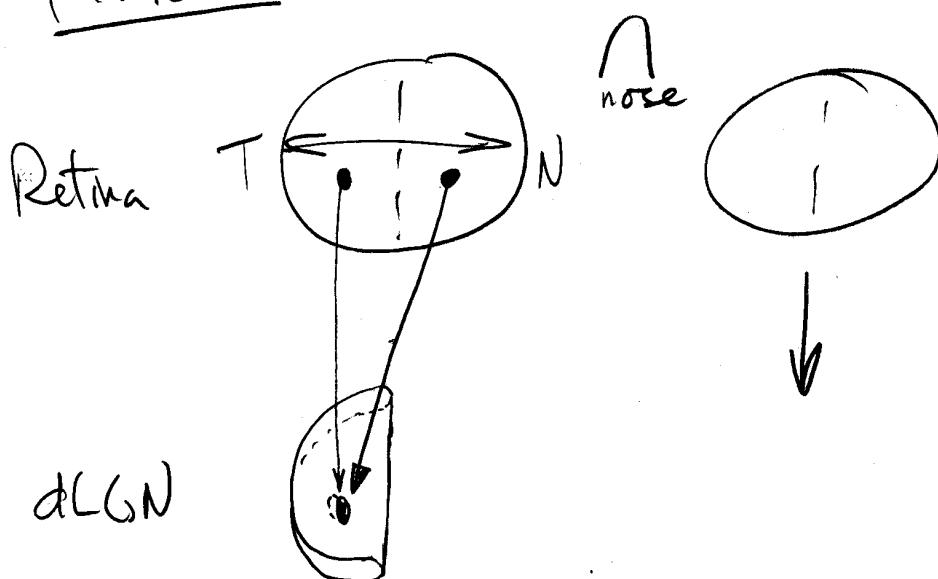
Mutant Belgian Sheepdog

Normal



- N/T coordinate has opposite "sign" in two eyes
- ∵ dLGN aligns left & right retinal points w/different N/T coordinates in C & I layers

Mutant



- retina still goes to same N/T coordinate
- lack of chiasm then leads to a "taco"
- ∵ map generated by labels not activity

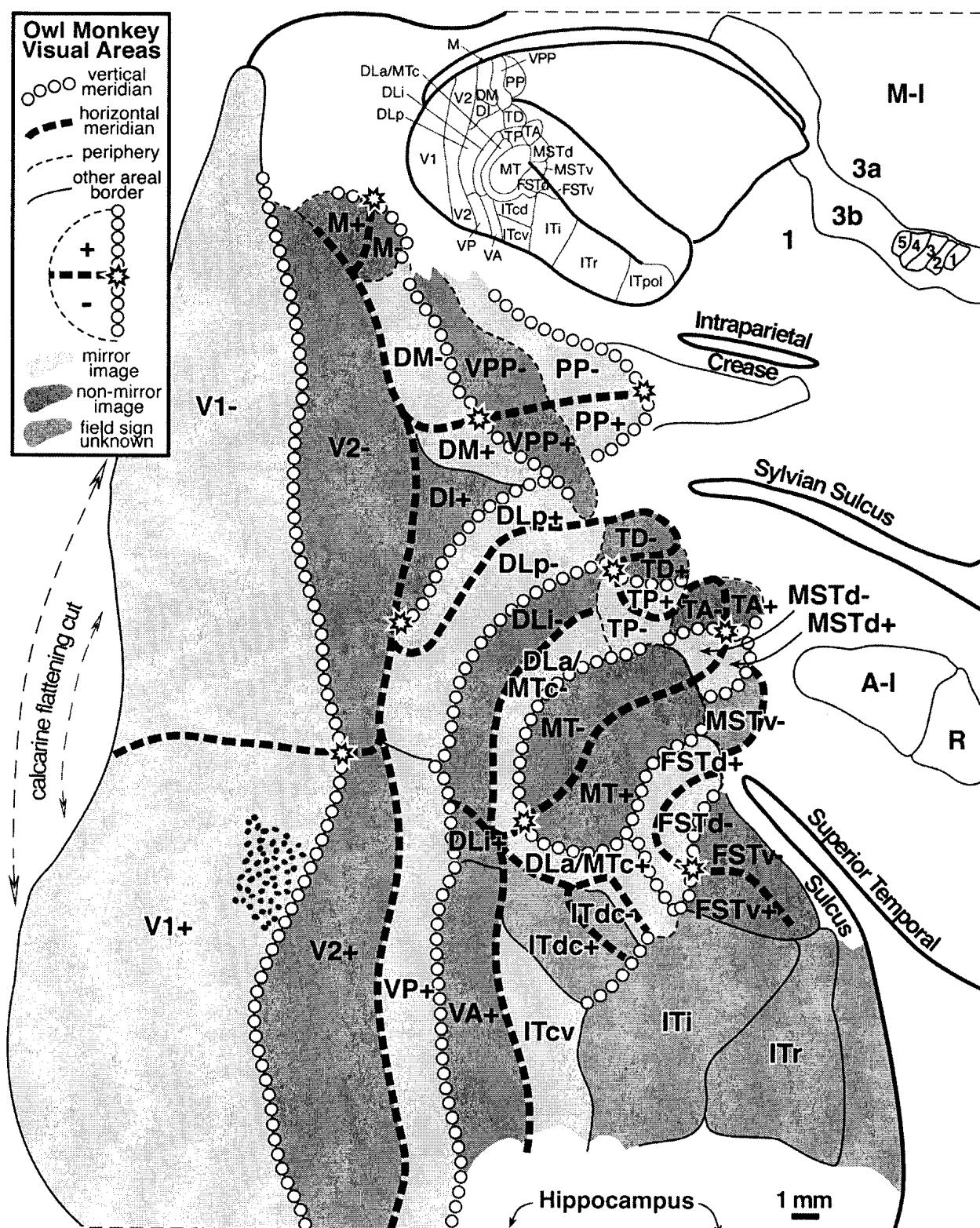
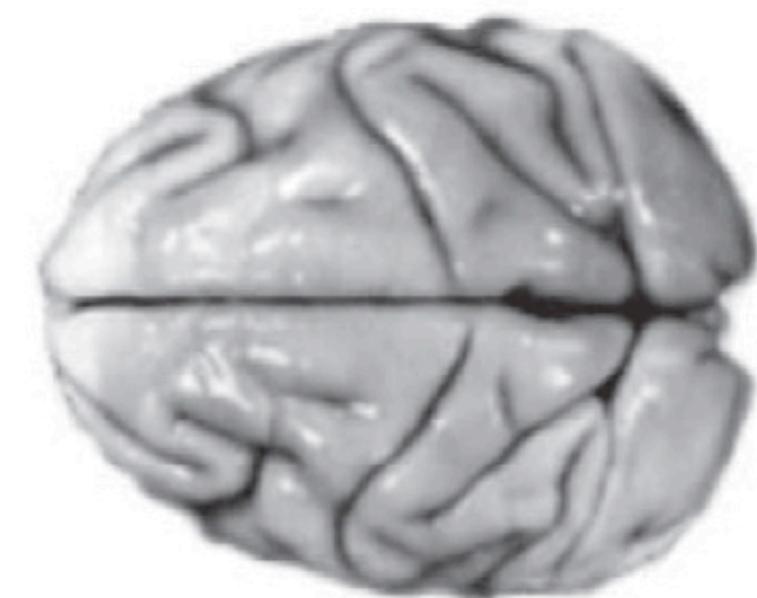
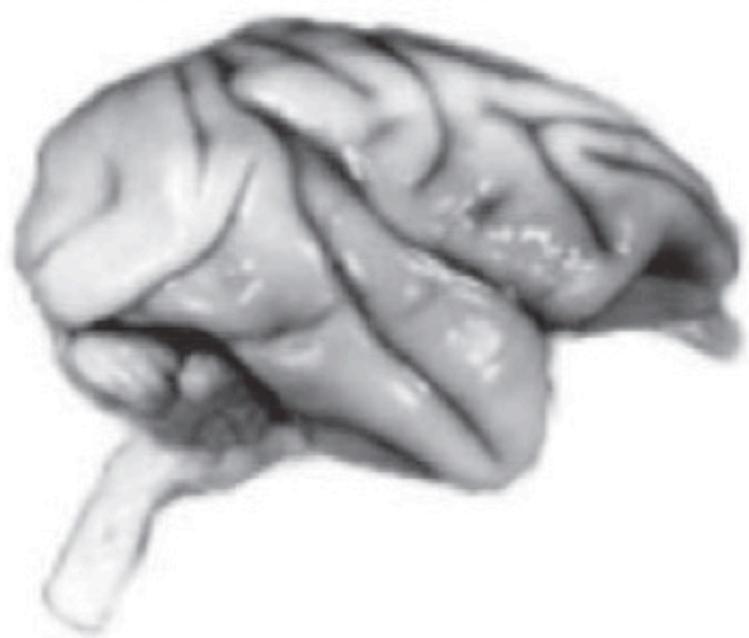
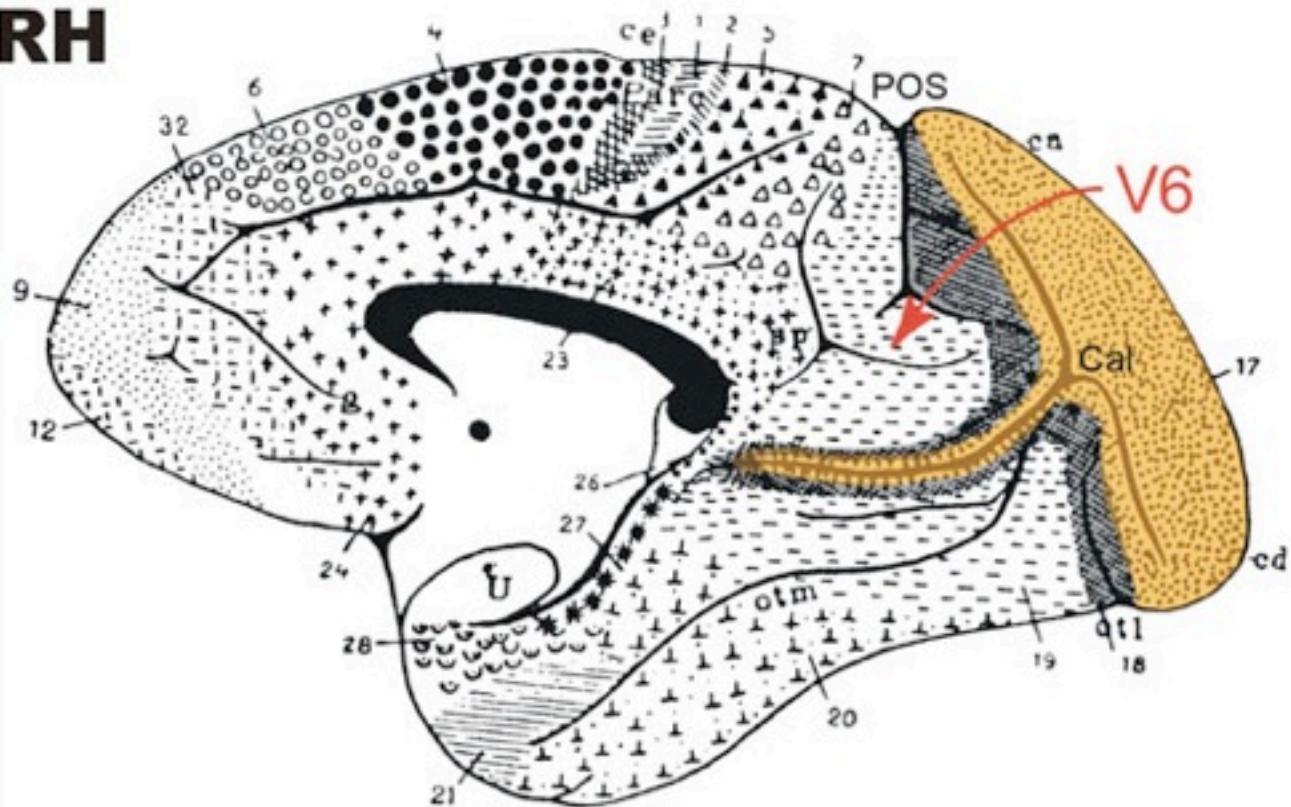


Fig. 2. Schematic proposal for the retinotopic organization of 24 owl monkey visual cortical areas drawn using a myelin-stained flatmount. Anterior to V2 near the dorsal convexity of the brain are 3 areas with alternating field sign—DM, VPP, and PP. DM and VPP share a center of gaze and vertical meridian representation. VPP and PP share a periphery representation. The upper fields of DM and VPP curve anteriorly, away from V2. Just lateral to DM, there is another series of strip like areas with alternating field sign—DI (which unlike DM, has the same field sign as V2), DLp, DLi, DLa/MTC, and finally MT. DLa/MTC bends laterally away from MT along a shared vertical meridian border with FSTd. The complex topography anterior to MT is best visualized as two pairs of areas—TD/TP and TA/MSTd. Each pair of areas shares a center of gaze and a vertical meridian representation. In all four areas, the upper field is anterior to the lower field. Anterior and lateral to MT are FSTd and FSTv, which share a center of gaze and a vertical meridian. ITd contains mainly an upper field representation lateral to DLa/MTC. Retinotopy of M, ventral VP and VA, and ITi, and ITr were taken from Allman and Kaas (1975), Newsome and Allman (1980), and Weller and Kaas (1987).

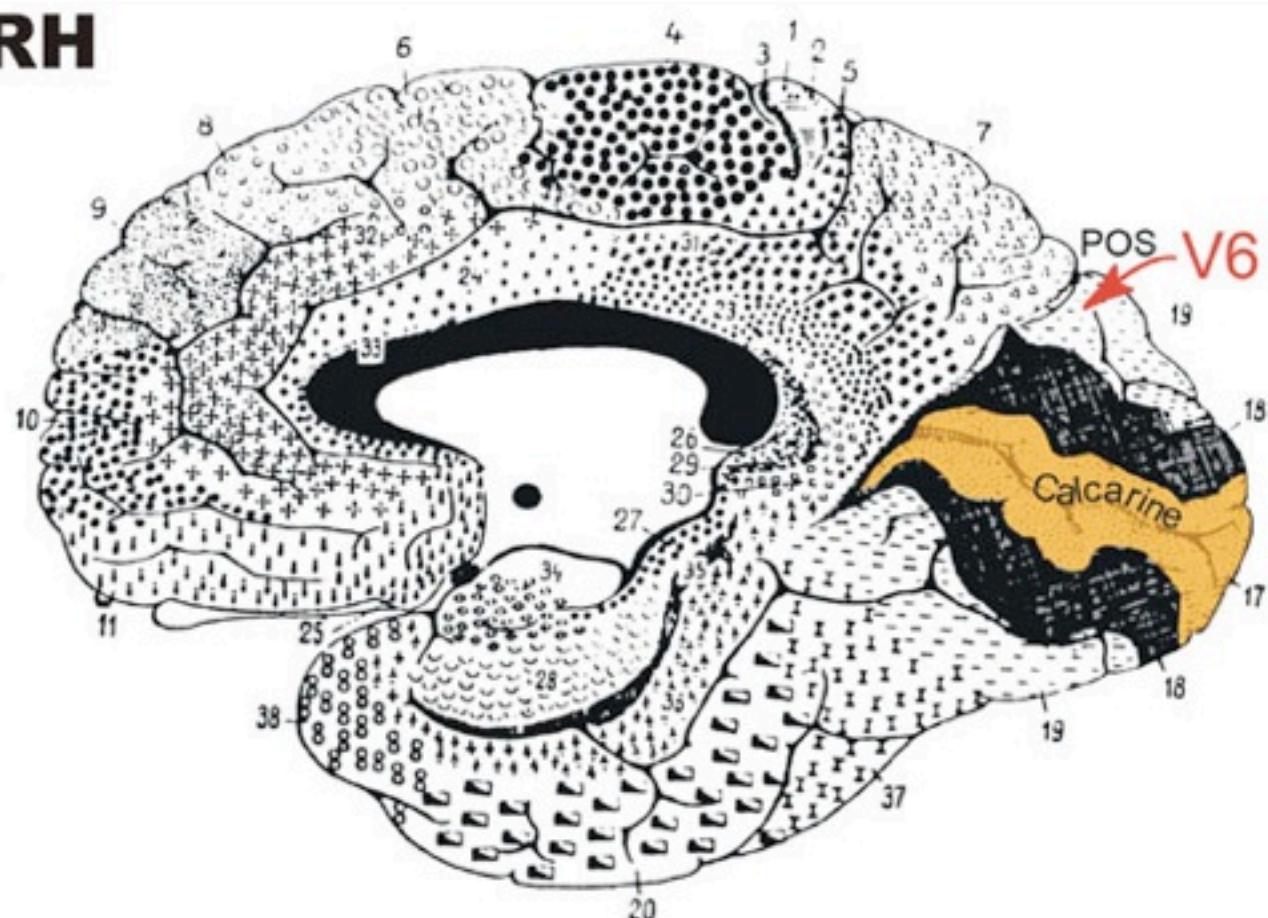


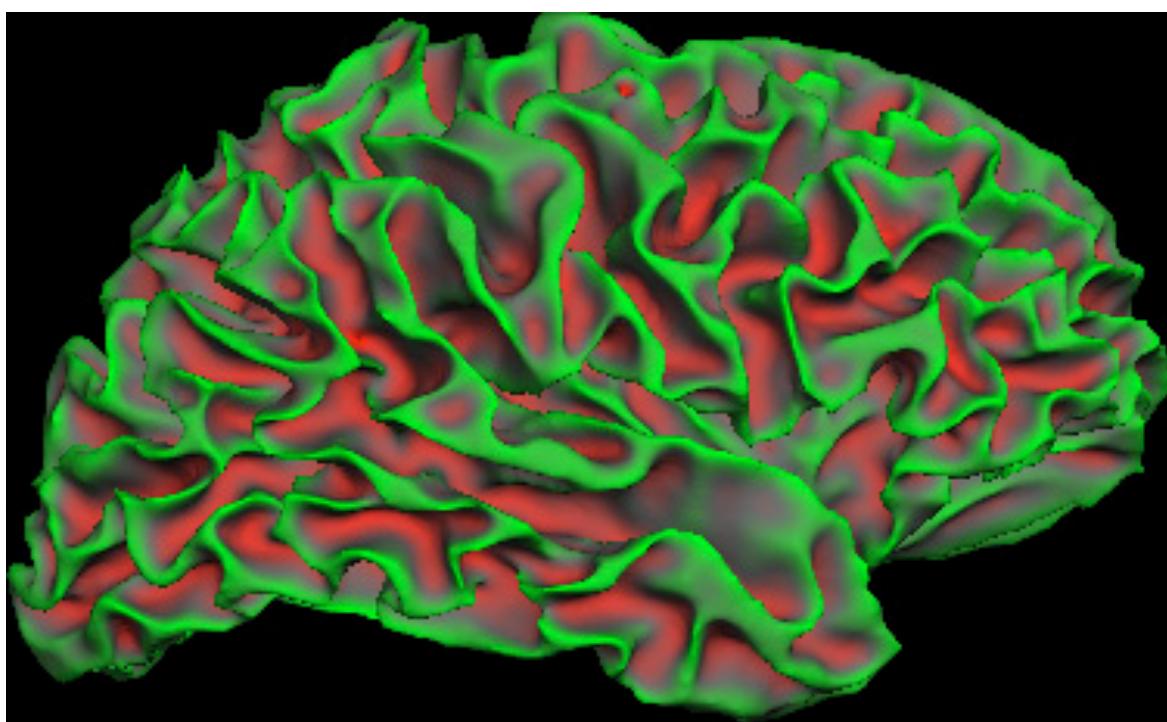
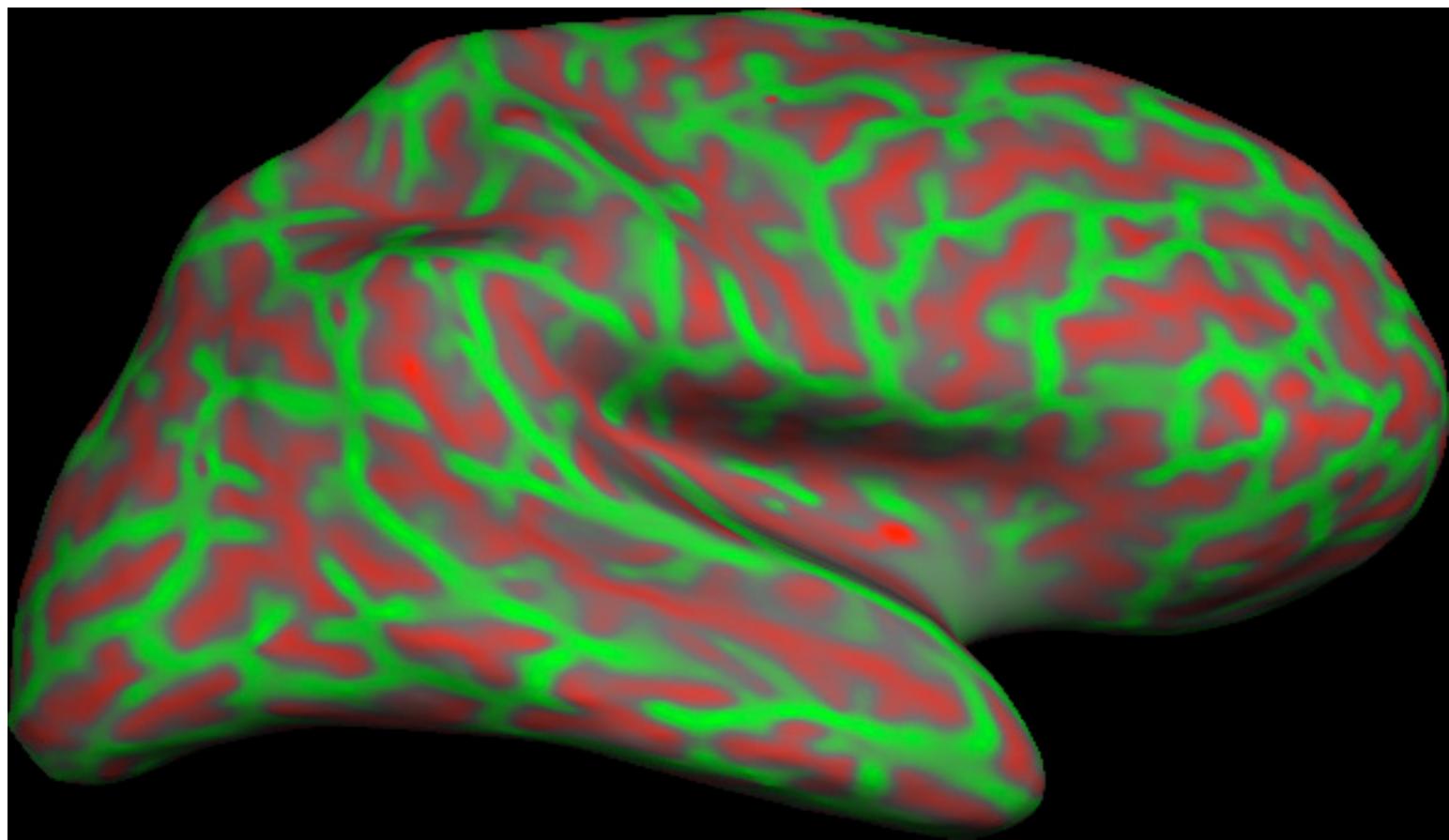


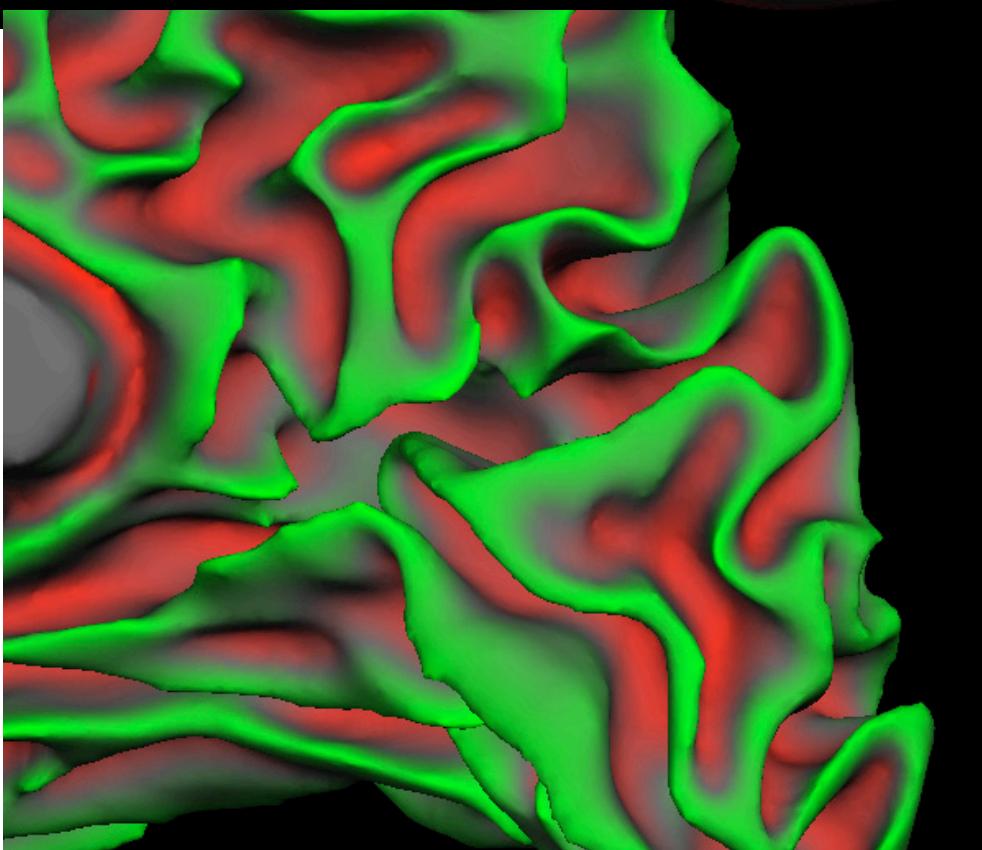
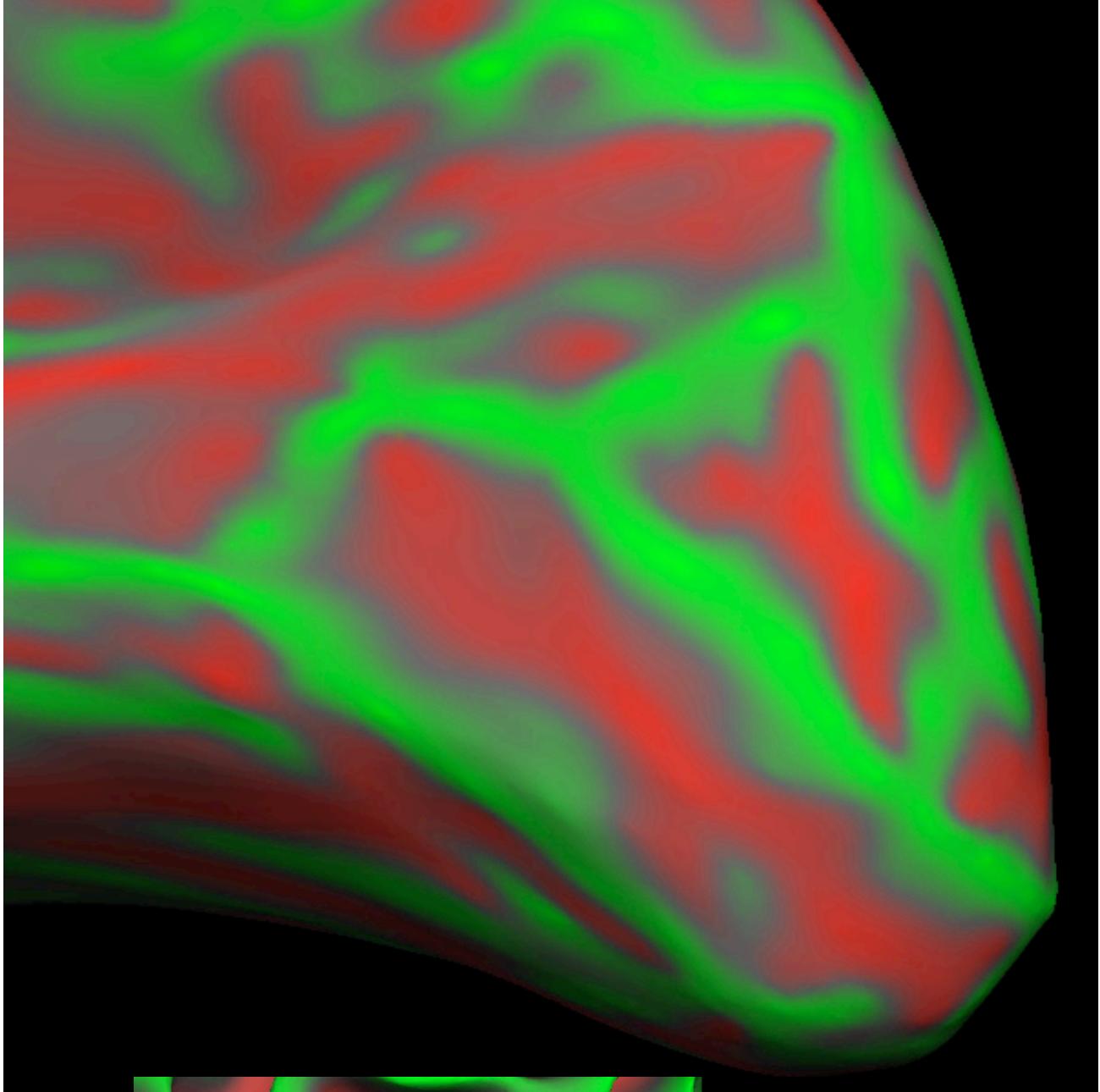
RH



RH







Visual System Overview

Areas

- 30 cortical < 50% of total neocortex
- 10 thalamus
- 10 midbrain

10 cortical < 10 - 100 million neurons each

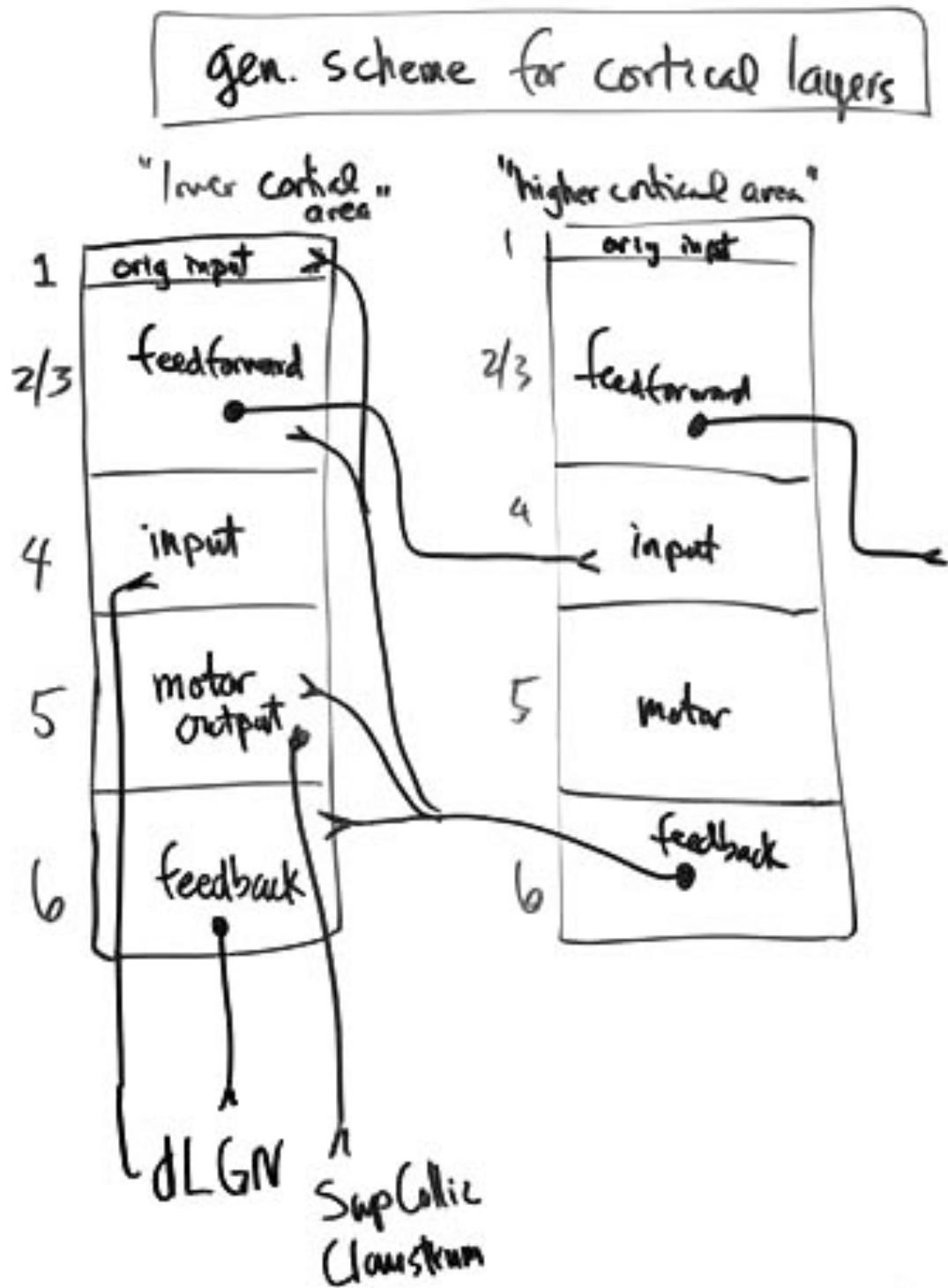
10 thalamus <

- retic
- sup pulv
- inf pulv
- IT pulv

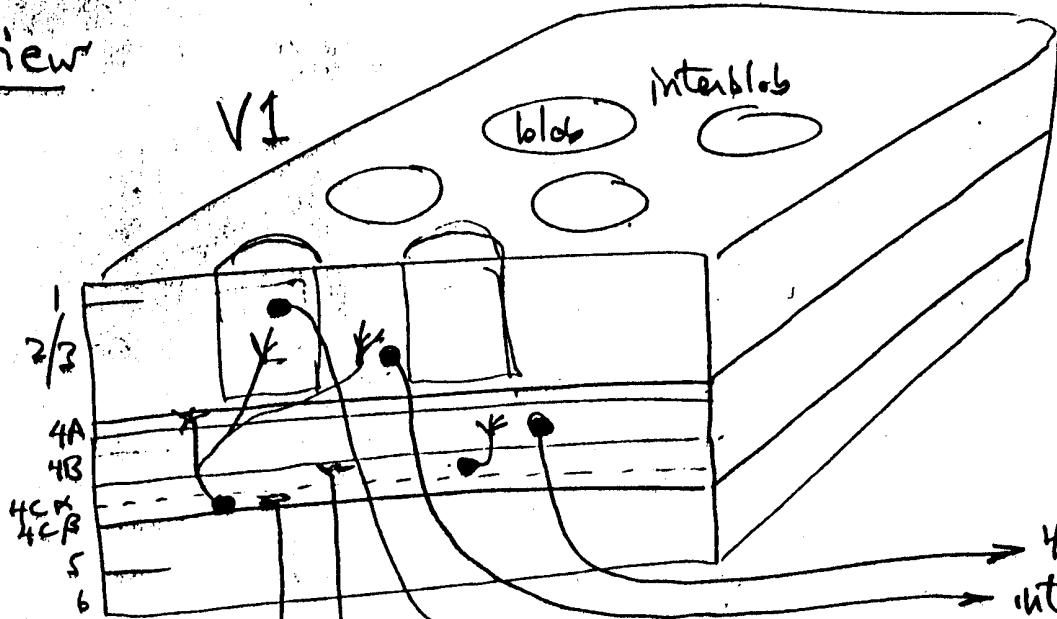
Connections

- each area connects to ~10 cortical
- perhaps 800 interareal tracts ~5 thalamus
usu. reciprocal
- each cortical area has outputs to several of these:

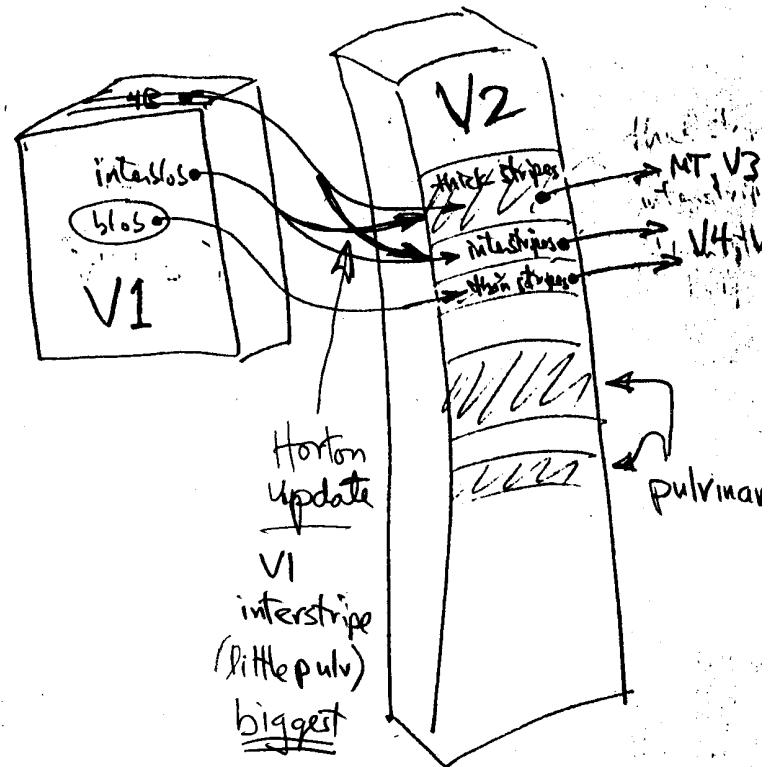
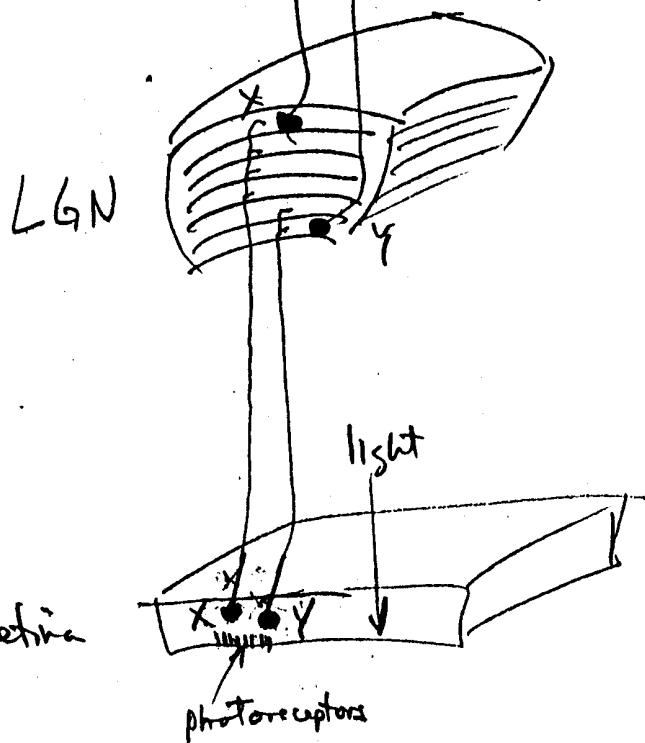
frontal
striatum
Sup. collic
pontine nuc.



Review



$4B - Y \rightarrow$ parietal
 \rightarrow interlob - X
 \rightarrow blob - X \rightarrow inferotemporal



"dorsal" MT, V3 \rightarrow MST \rightarrow 7a

— crude summary of ~300 corticocortical connection bundles before areas.

"ventral" V4, VP \rightarrow PIT \rightarrow AIT

— also, remember the thalamus

5 mm

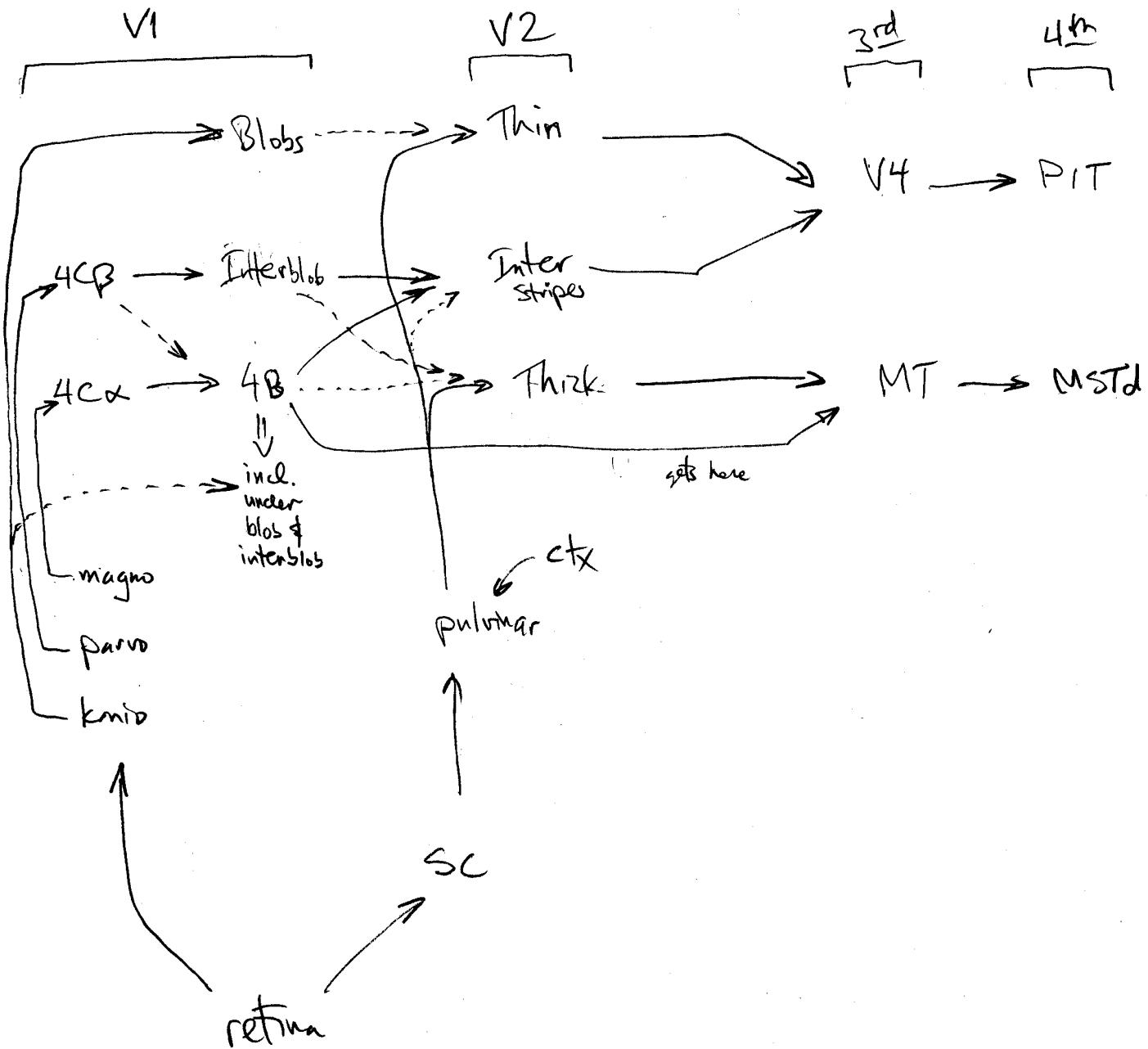
blobs
and
interblobs

thin inter
thick { inter
 inter
 inter

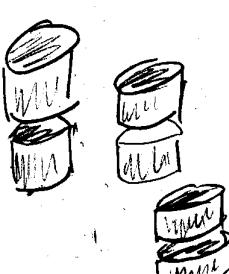
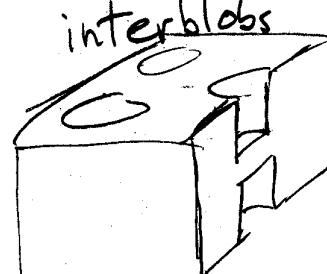
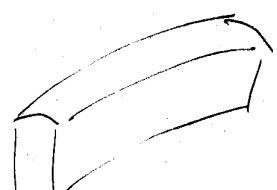
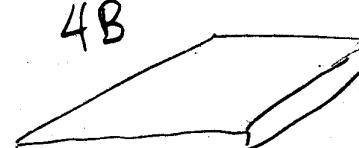
thin inter
thick { inter
 inter
 inter

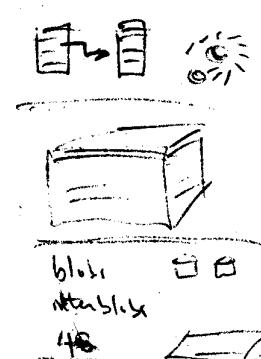
thin inter
thick {

Blobs, 4B, Interblobs/stripes - NEW

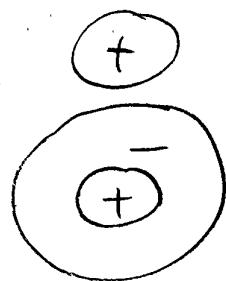


- differentiate, then remix
- motion OK w/complete retinal paraxial lesion (w/ contrast > 10%)

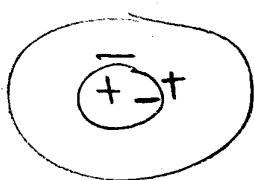
V1	V2	properties
<p>blobs</p> 	<p>thin stripes</p> 	<ul style="list-style-type: none"> - color selective (but cf. Gologo) → <u>brightness</u> - large dynamic range (0-150)
<p>interblobs</p> 	<p>interstripes</p> 	<ul style="list-style-type: none"> - <u>orientation</u> selective - not color selective but can use color to detect orientation - small dynamic range (0-50)
<p>4B</p> 	<p>thick stripes</p> 	<ul style="list-style-type: none"> - <u>direction</u> selective



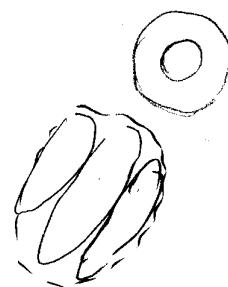
Basic Visual Physiology



non-opponent



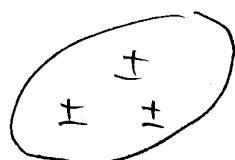
simple-opponent



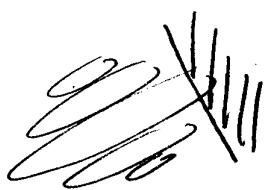
double-opponent



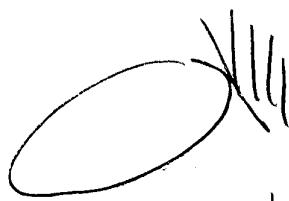
Simple cells



Complex cells



"Simple
hypercomplex" cells



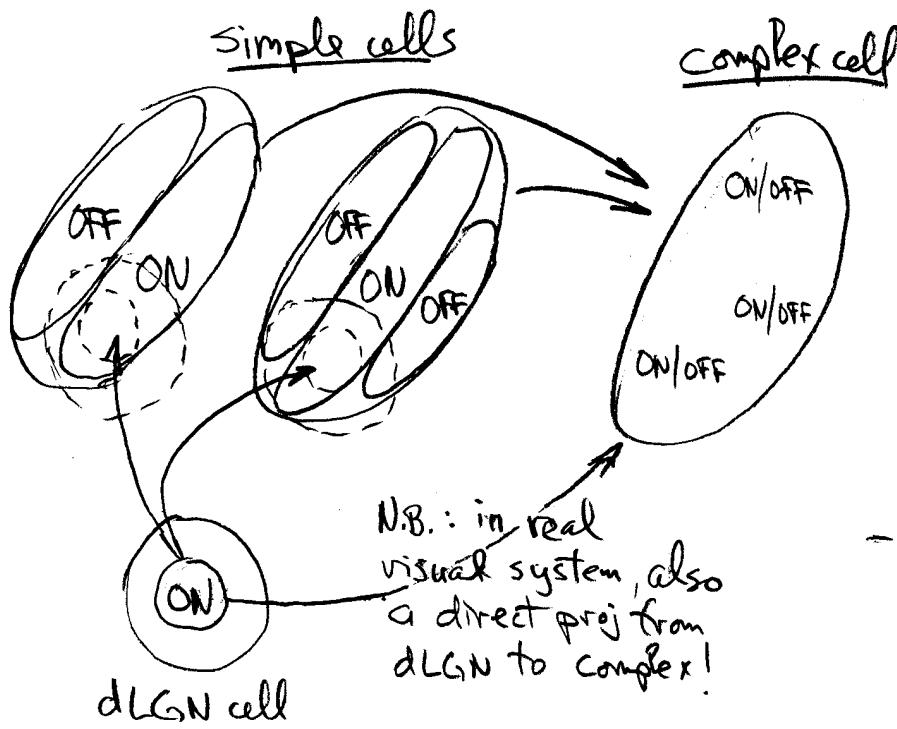
Complex
hypercomplex cells



special complex (=anti-hypercomplex!)

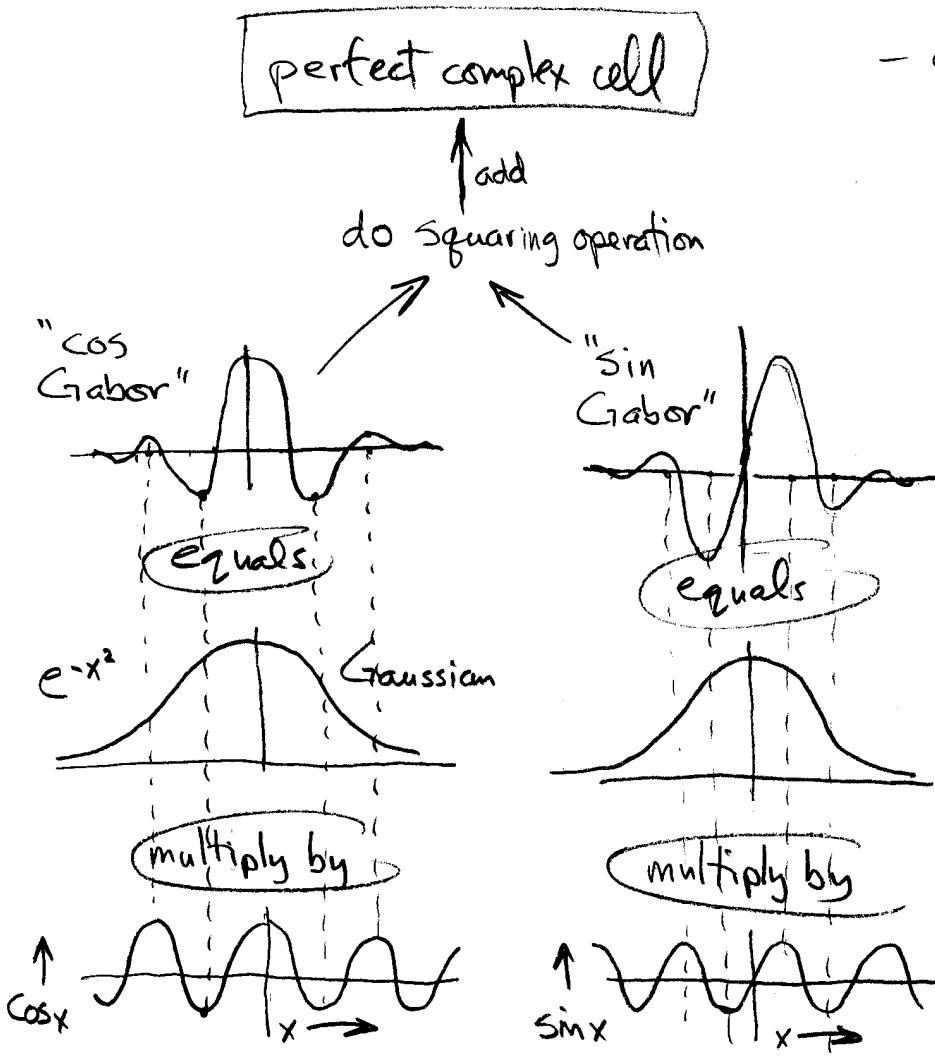
moon $\approx 1/2$ deg vis angle

COMPLEX CELL MODEL - GABOR FILTERS



- N.B.: real simple cells have excitatory responses to ON & OFF (no negative responses in real cells)

- model simple cells by convolution of visual stim with Gabor filters (heavily used in machine vision)



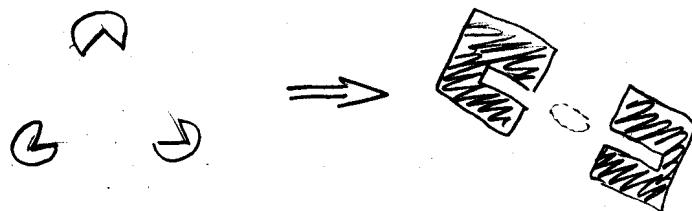
- convolving with individual Gabor finds oriented stripes in stim (cross prod = corr) but response varies with stim positioning in R.F.

- N.B.: trig identity $\cos^2 x + \sin^2 x = 1$

- Combining cos & sin Gabors eliminates response variation

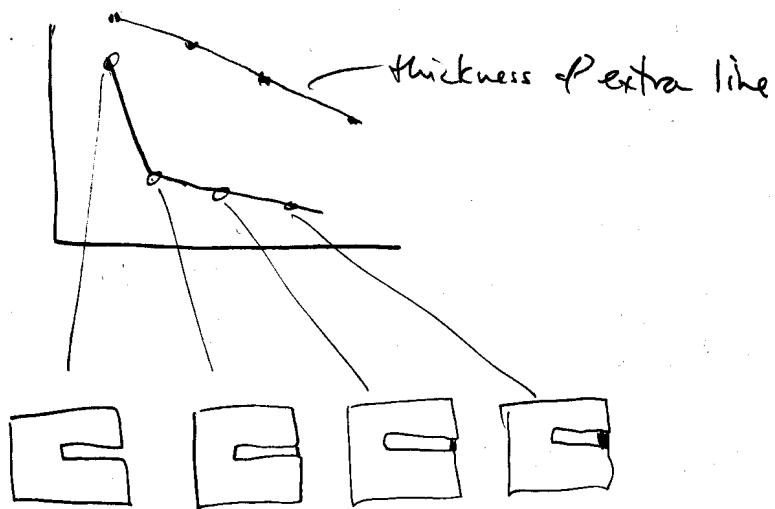
- * output detects ripple-i-ness in stimulus independent of exact position of ripples in R.F.

Van der Heydt et al

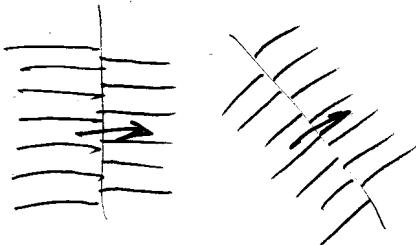


V1 - cells respond only to real contours

V2 - some cells respond to illusory



cf.

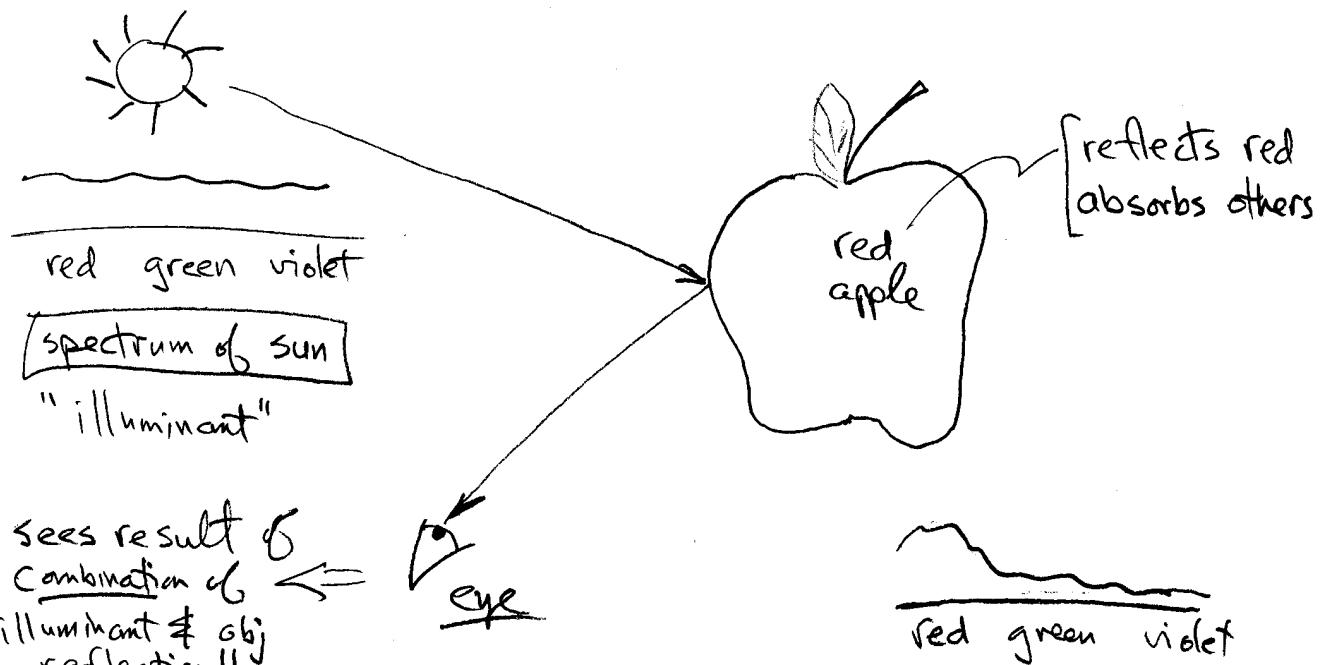


APERTURE PROBLEMS IN GENERAL

- to process local detail, need local views
- like viewing the world through straws
- V1 is "dumber" than you think
- a general problem
 - across modalities
 - across different stimulus features
 - across levels of processing
- other examples
 - { Object color
 - Pattern motion
 - Complex motion
 - Texture
 - ITD
- solved by combining information across space
 - ↳ i.e., across receptor map space
- higher areas are "smarter" than they first look
 - (e.g. often have "sloppy"-looking tuning curves to simple stimuli)

DETECTING COLOR AS APERTURE PROB

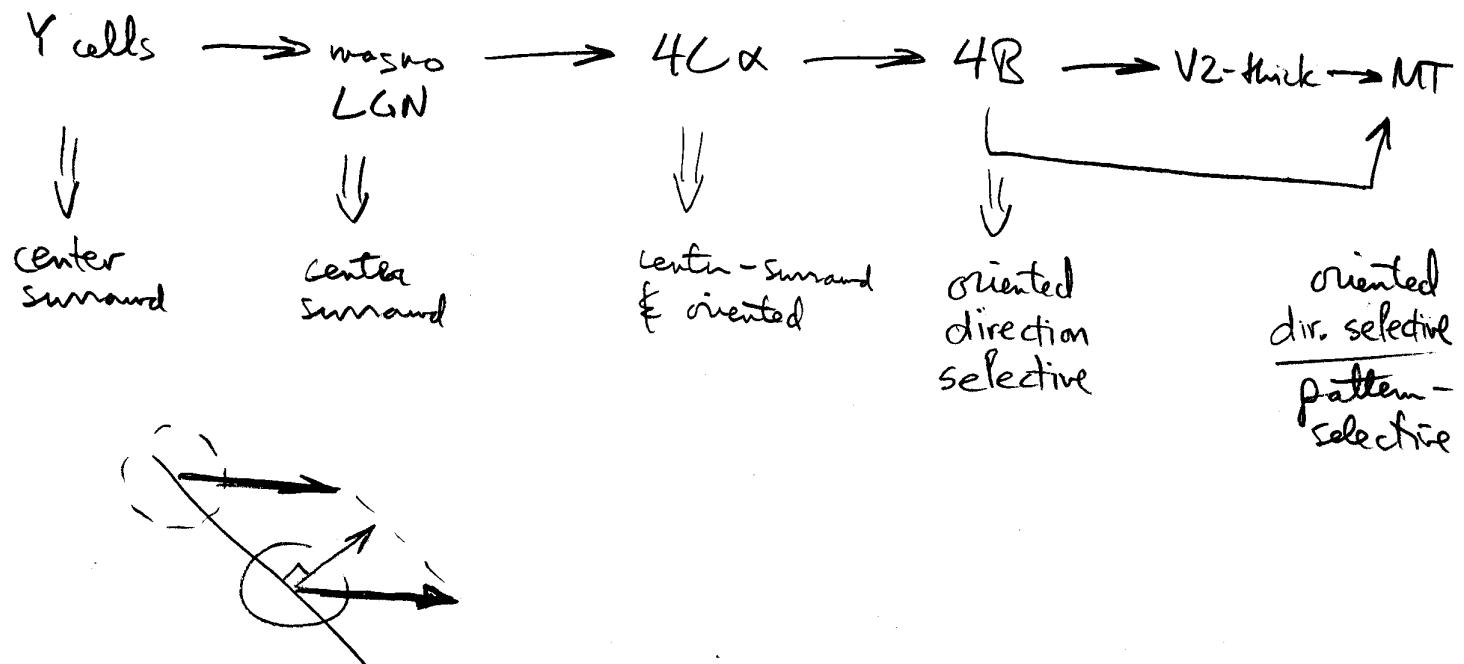
- "color" is spectrum of what is reflected (not absorbed)
- depends on properties of object (e.g. a ripe fruit)
- but spectrum of illumination affects final result



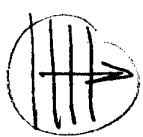
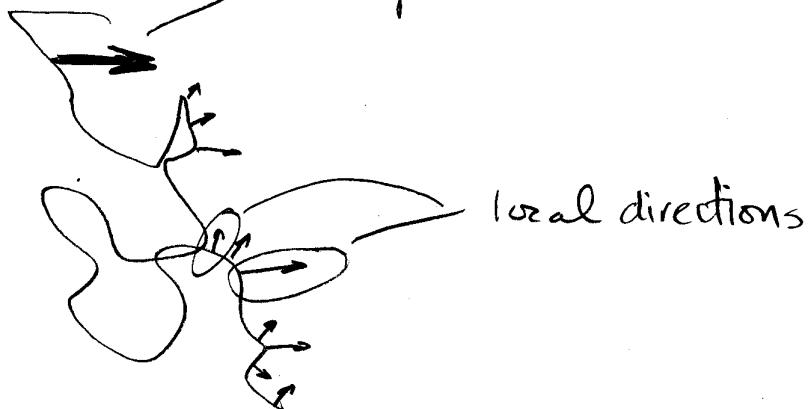
- if spectrum of illuminant is biased (e.g. mostly orange light inside pub) even green objects will be mostly orange (tho less orange than orange obj)
- green objects still look green! → color constancy
- this color constancy will fail in monochromatic light (e.g. Sodium vapor lamp) so req's some range of colors

Aperture Problem For Pattern Motion

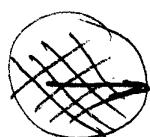
Overview



→ true pattern dir

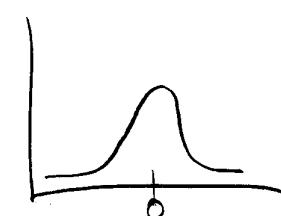


etc

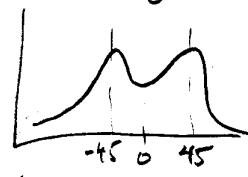


etc

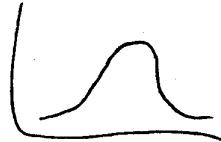
Saito



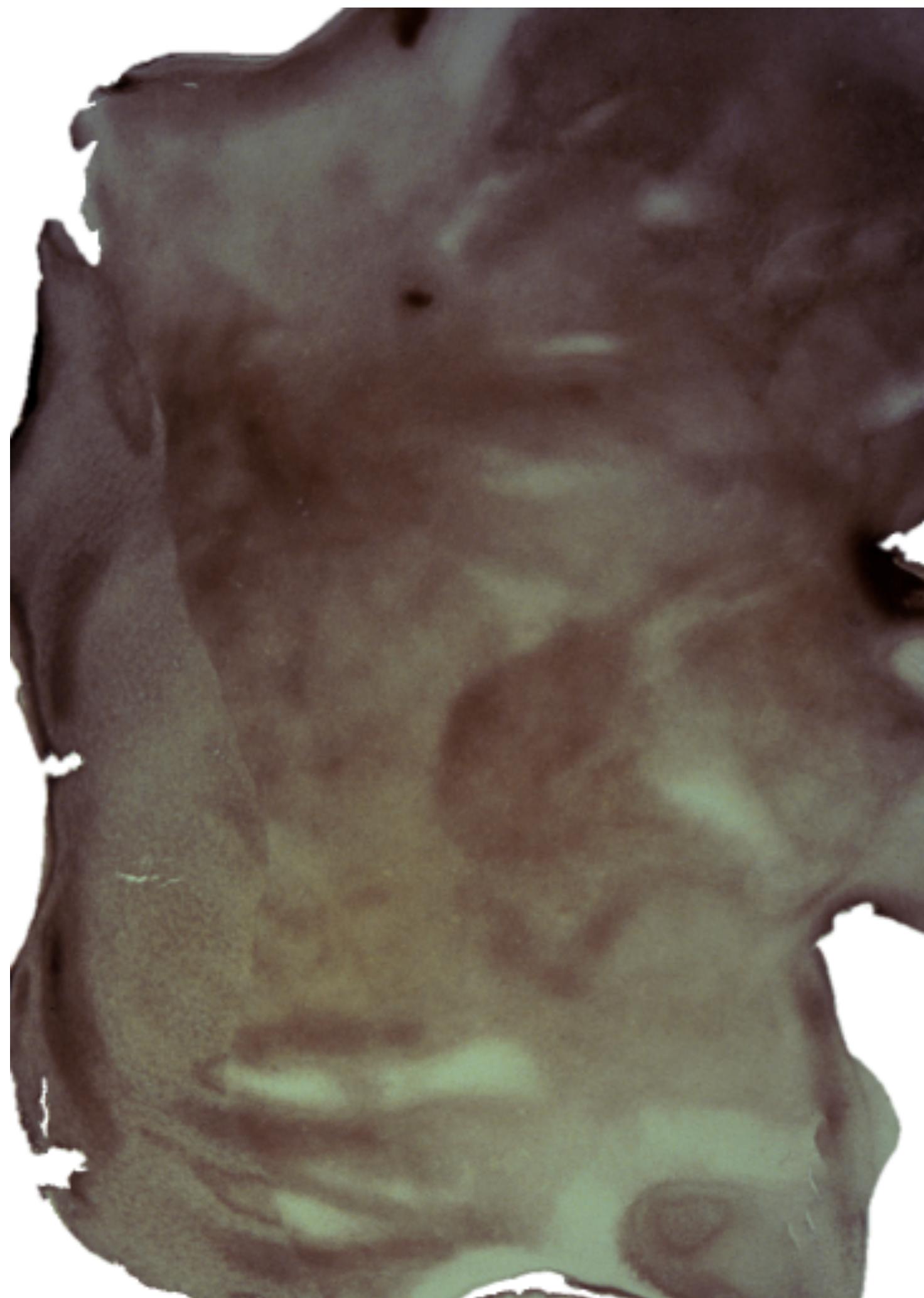
component selective



component selective

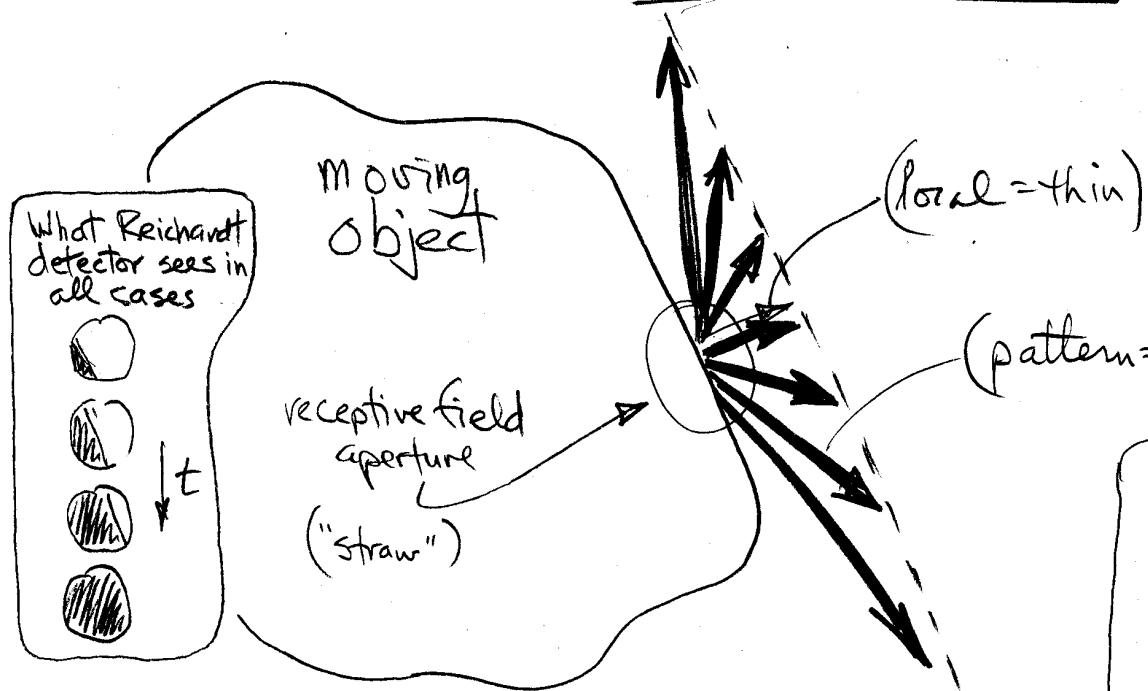


pattern-selective

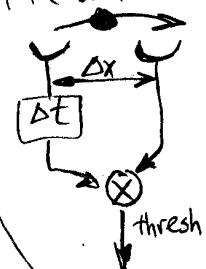


View from V1

Family of pattern directions that could have given rise to this one local direction



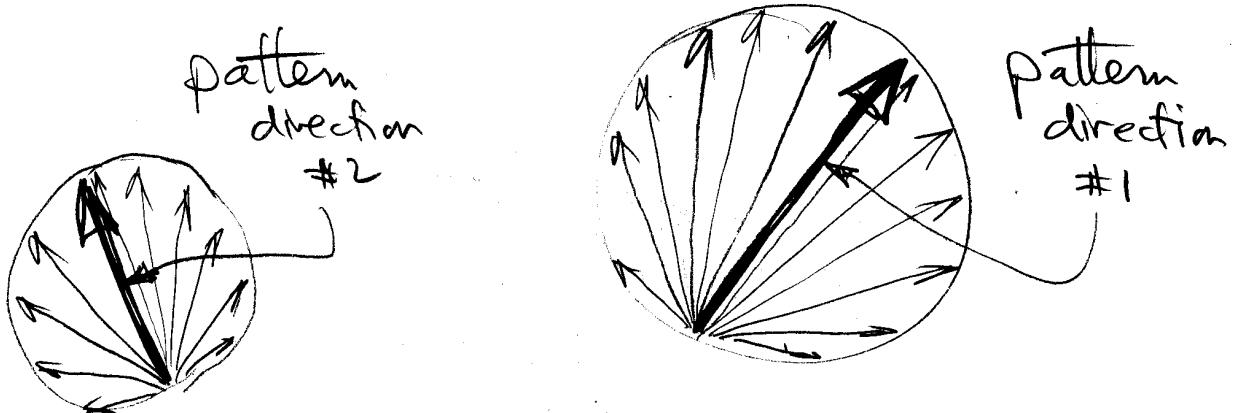
this is a property of any Reichardt-like detector:



View from MT

Family of local directions consistent with one pattern direction

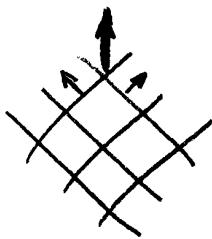
↳ i.e. V1, 4B detects component of pattern motion \perp to contour



Why the average local direction won't work in general

average works

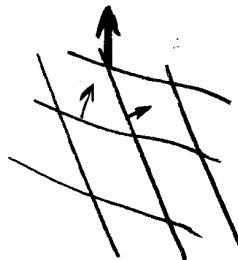
(in same direction
tho speed wrong)



↑ actual
pattern dir

↑ average of
local dirs

average doesn't work

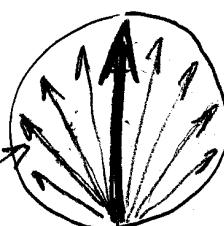


↑ actual
pattern dir

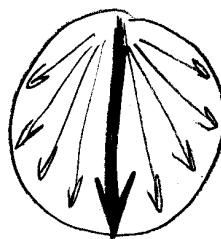
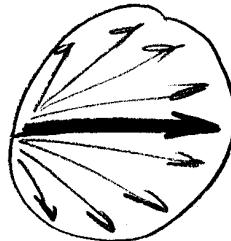
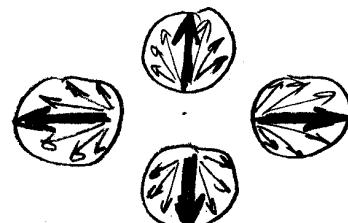
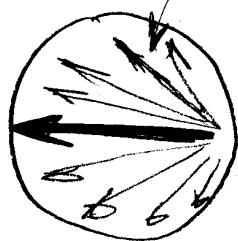
↑ average of
local dirs

Model of map point in MT that can detect
8 different pattern directions

N.B.
single
local motion
same



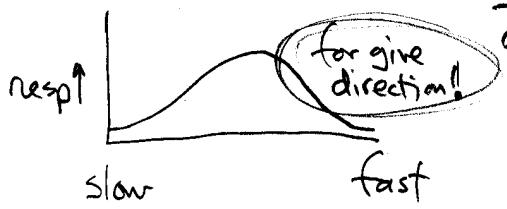
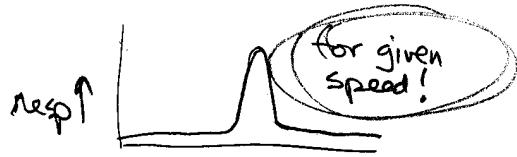
[8 thick — pattern dirs
many thin — local dirs]



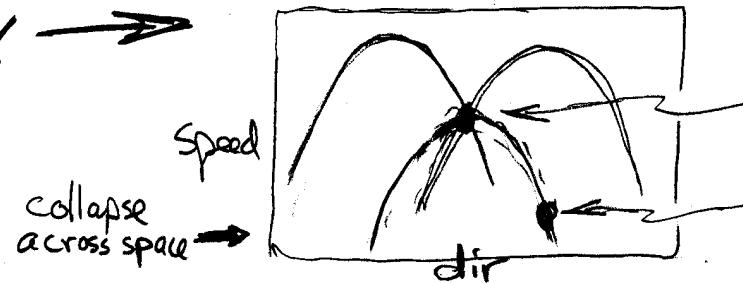
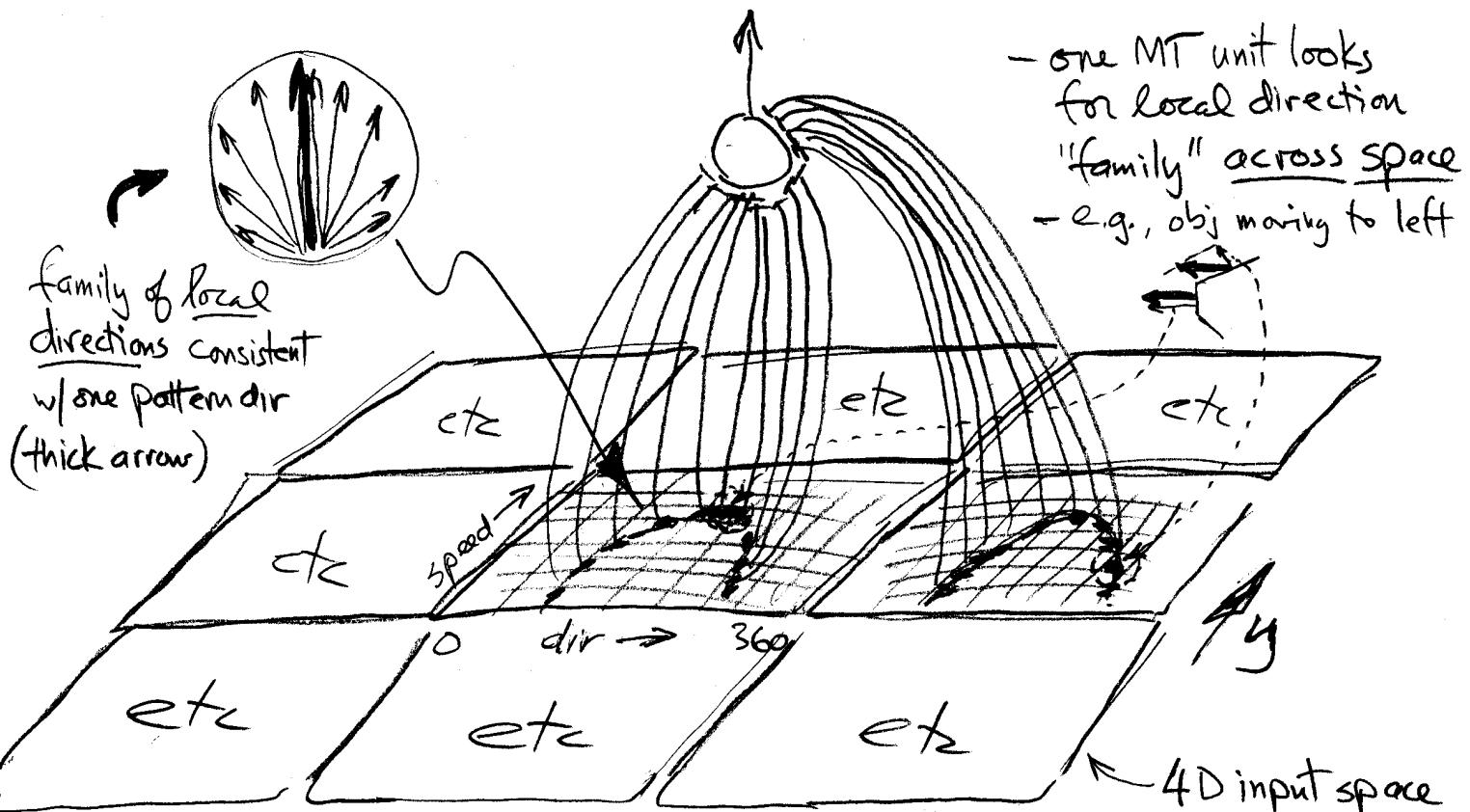
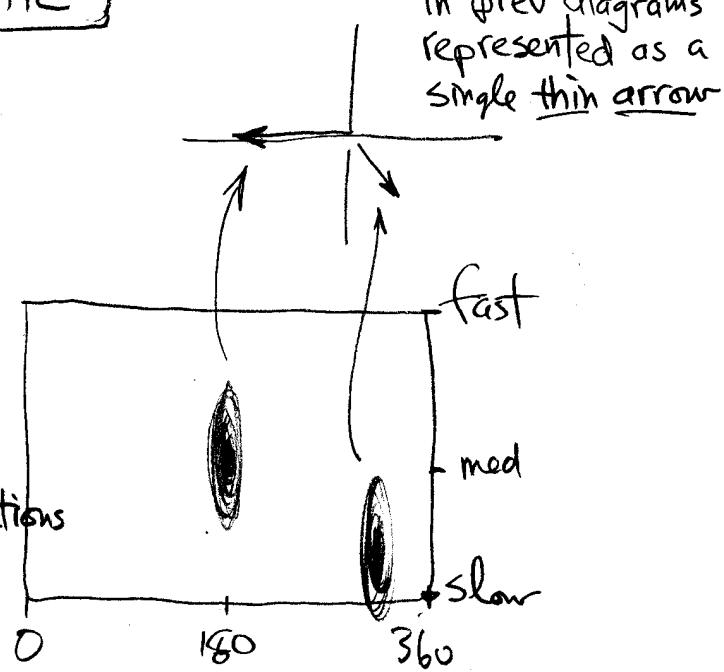
- each MT neuron gets input from multiple V1 locations
- overlapping RF's for all 8 units

MT MODEL IN DETAIL

V1, layer 4B



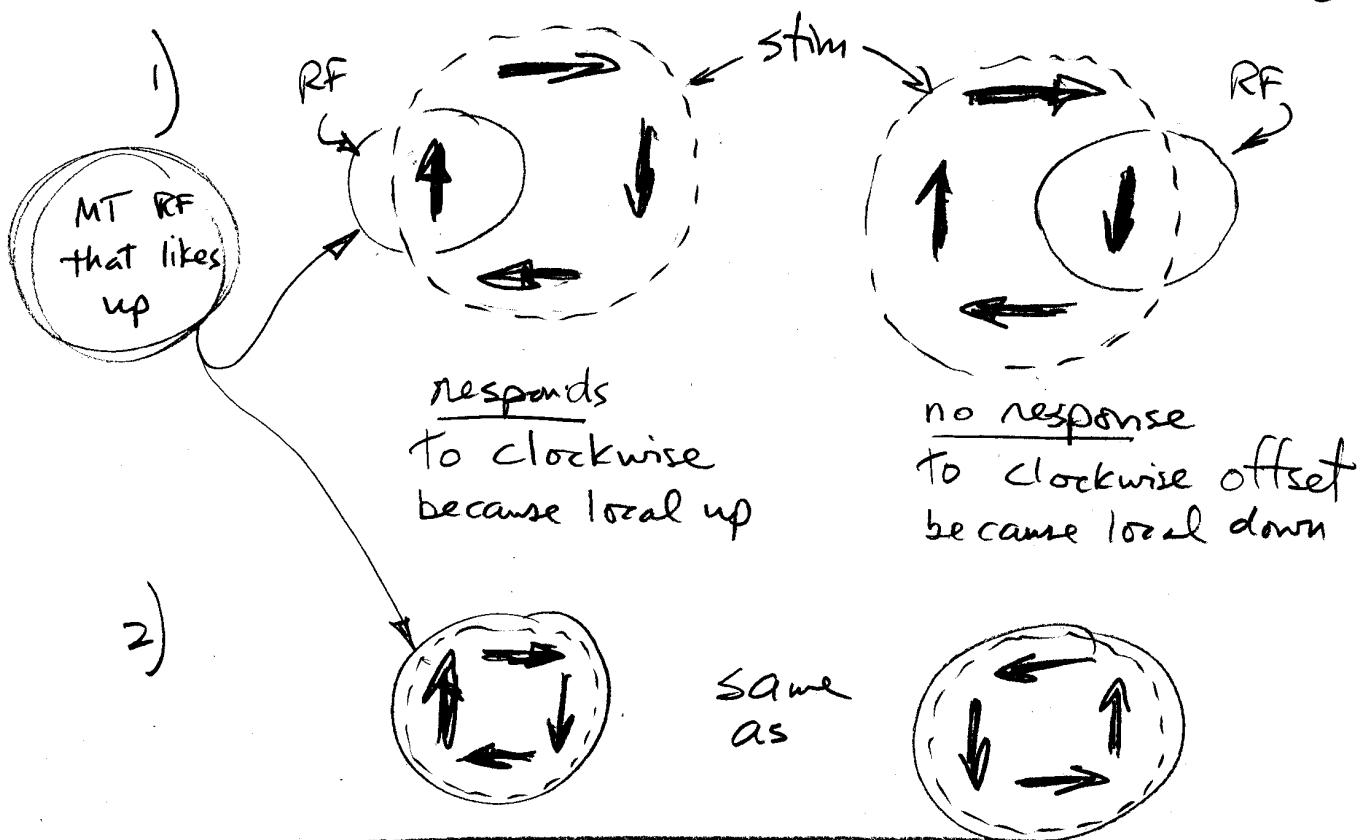
need to test
all combinations



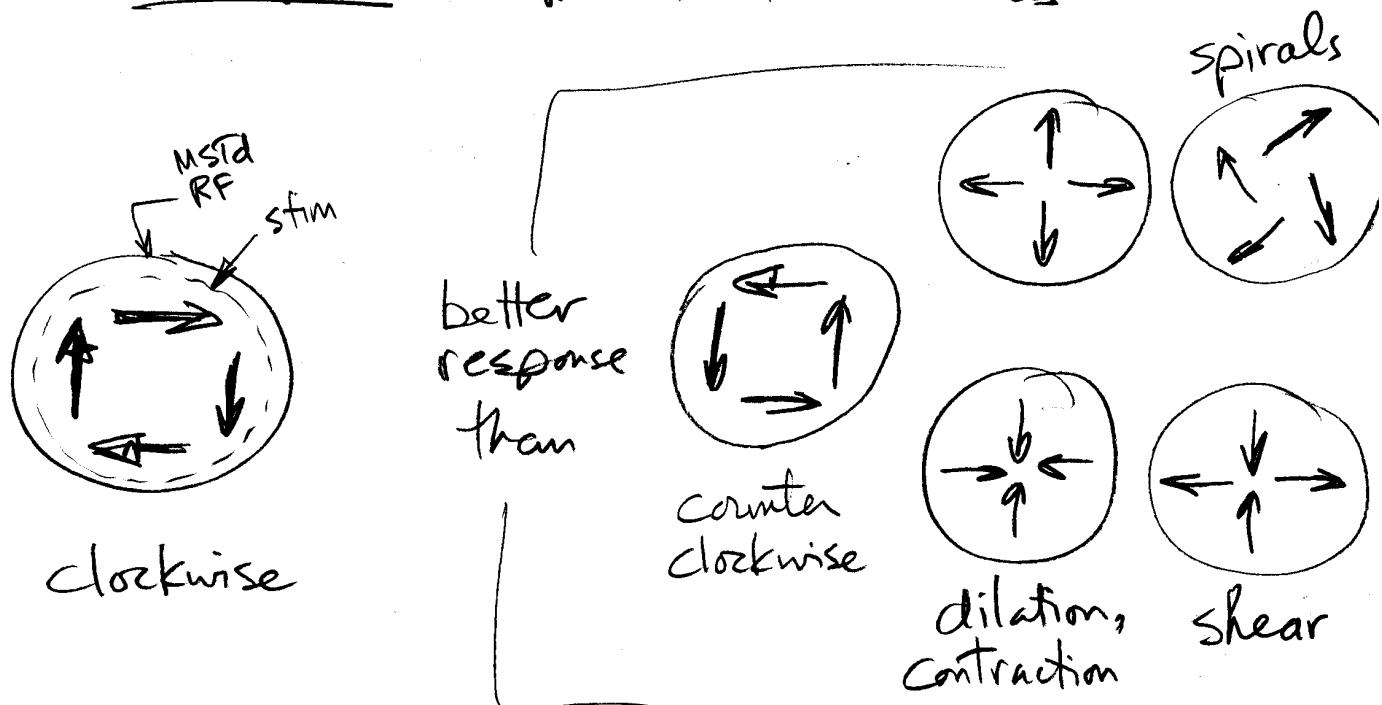
single stim consistent with several families
2nd local dir disambiguates

MSTd

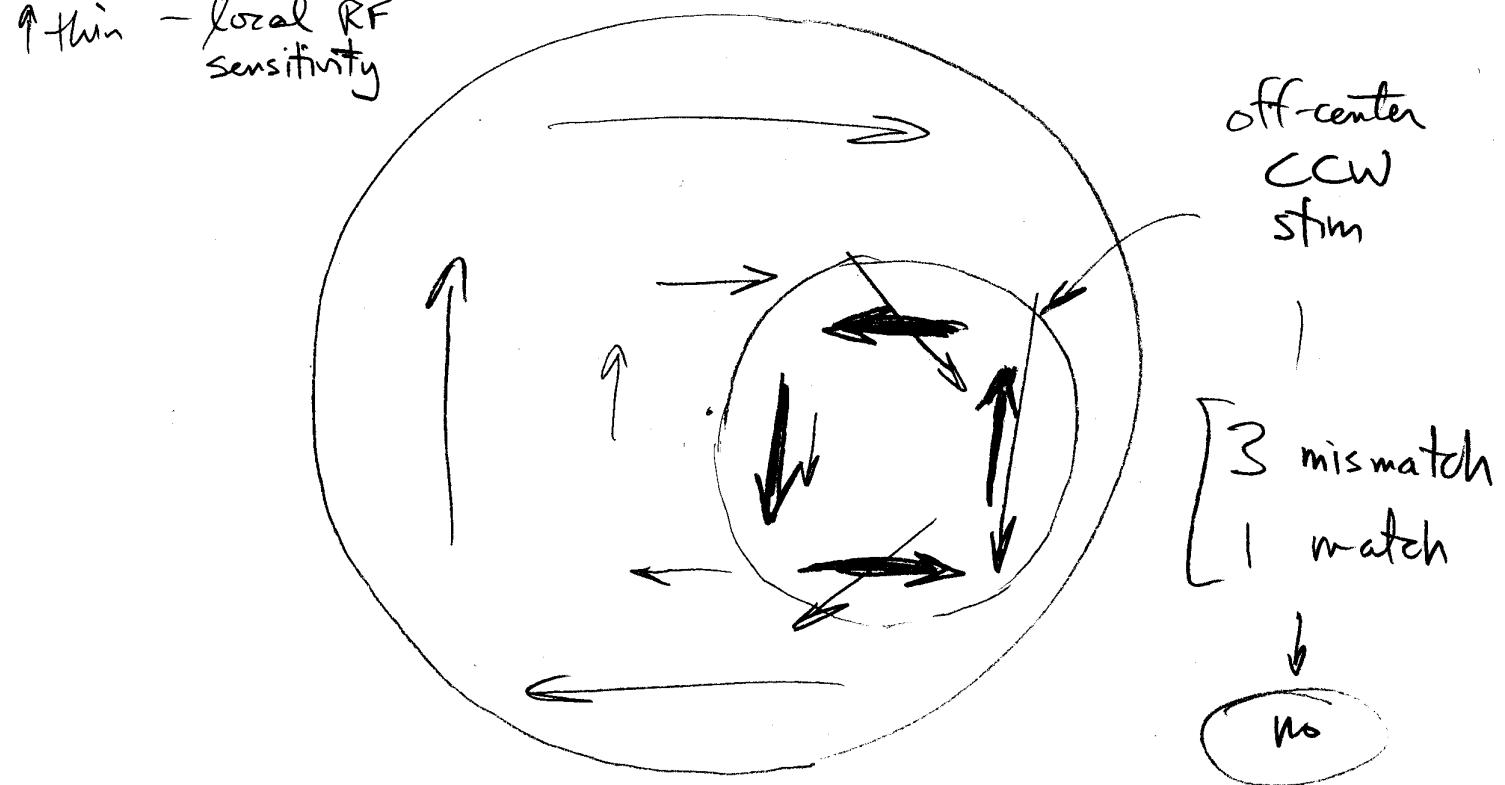
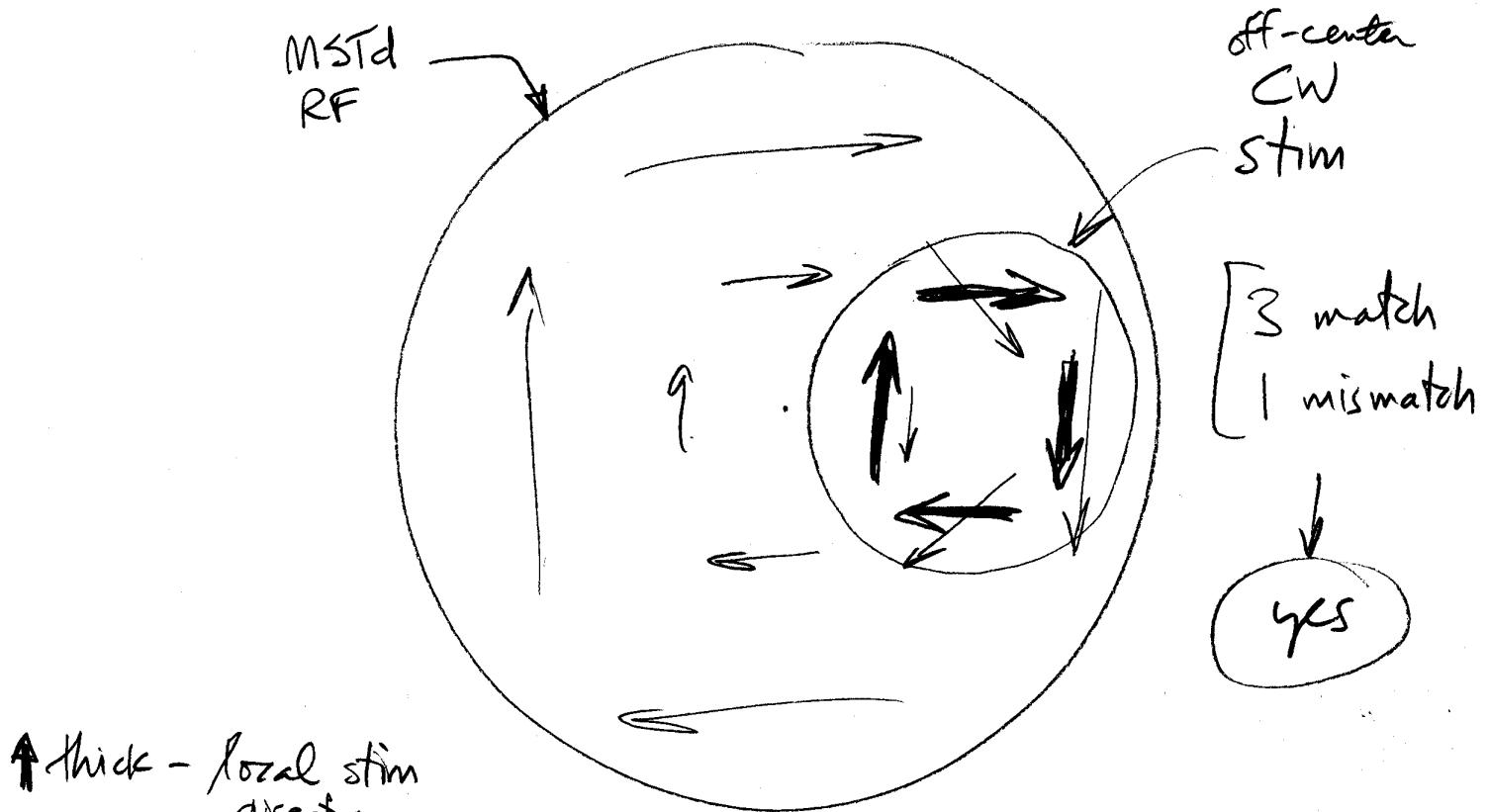
MT responds to rotation/dilation but not selectively



MSTd distinguishes different flow fields



How a fixed template actually works



DETERMINING OPTICAL FLOW

Horn & Schunck, 1981
"gradient model"

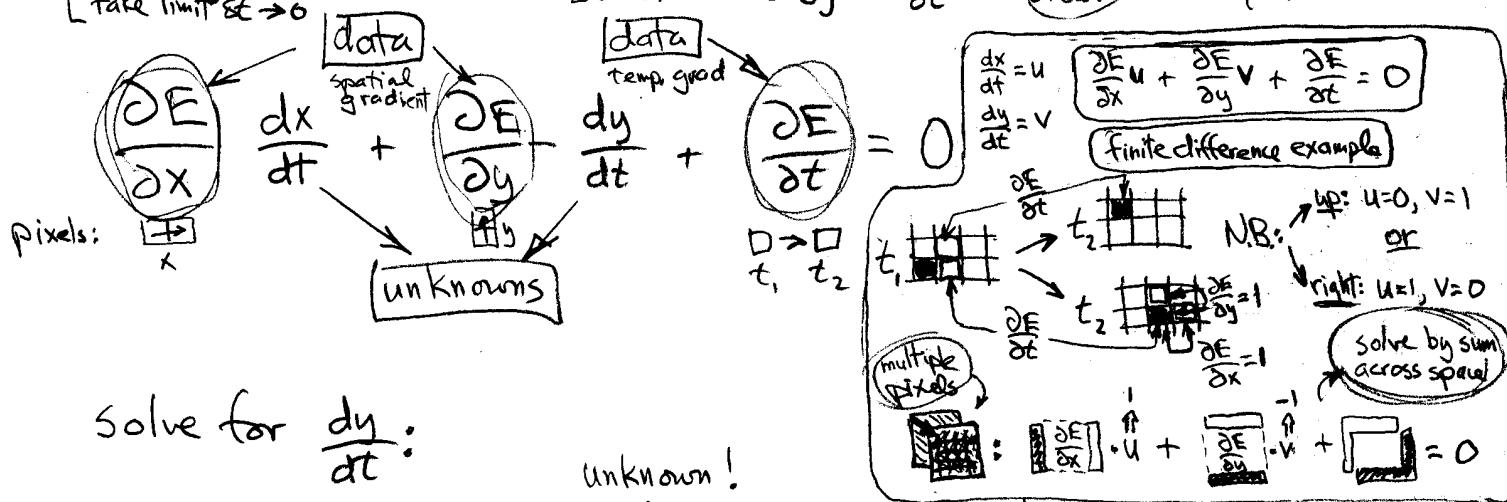
$$\text{brightness of image point} = E(x, y, t)$$

assume brightness doesn't change w/move: $\frac{dE}{dt} = 0$ [no shadows, no lighting change]

[movement] $E(x, y, t) = E(x + \delta x, y + \delta y, t + \delta t)$

[Taylor expand around this point] $E(x, y, t) = E(x, y, t) + \delta x \frac{\partial E}{\partial x} + \delta y \frac{\partial E}{\partial y} + \delta t \frac{\partial E}{\partial t} + \dots$ higher order Taylor terms

[Subtract $E(x, y, t)$, divide by δt , take limit $\delta t \rightarrow 0$] $\frac{\partial E}{\partial t} = \frac{\delta x}{\delta t} \frac{\partial E}{\partial x} + \frac{\delta y}{\delta t} \frac{\partial E}{\partial y} + \frac{\partial E}{\partial t} + \dots$ ignore in limit of $\delta t \rightarrow 0$



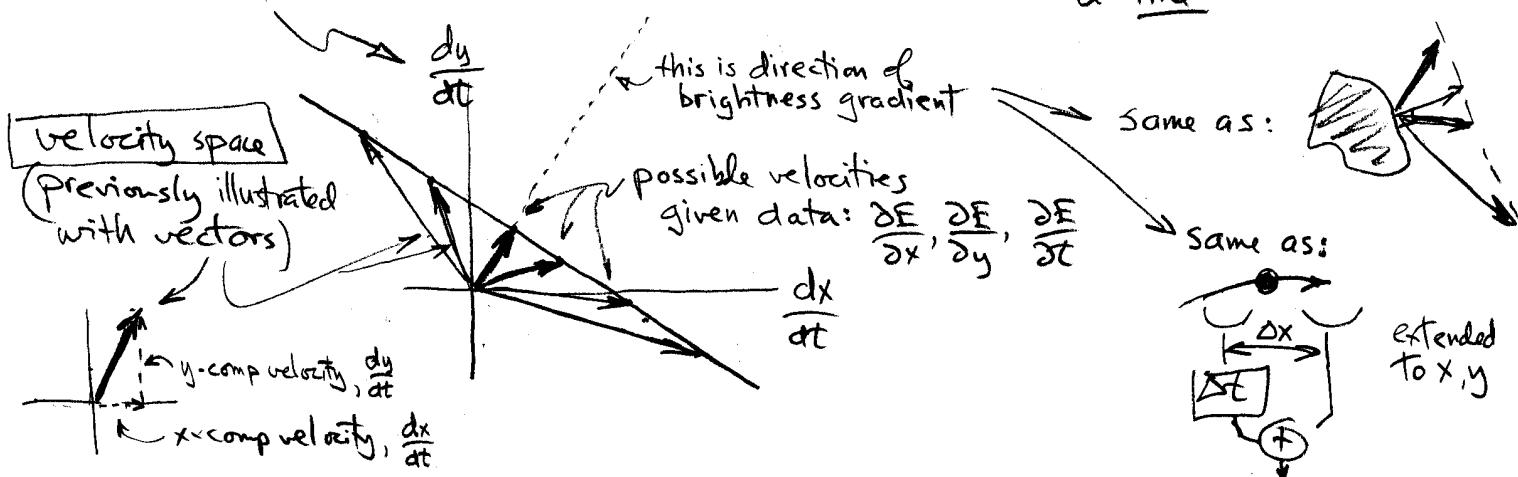
Solve for $\frac{dy}{dt}$:

Unknown!

$$\frac{dy}{dt} = - \frac{\frac{\partial E}{\partial x}}{\frac{\partial E}{\partial y}} \frac{\frac{dx}{dt}}{\frac{\partial E}{\partial x}} - \frac{\frac{\partial E}{\partial t}}{\frac{\partial E}{\partial y}}$$

$\frac{dy}{dt}$ is therefore dependent on unknown x-velocity (i.e., the aperture problem!) ↳ in Cartesian coord form

$$y = mx + b \Rightarrow \text{this is equation for a line}$$



COMPUTATIONAL MOTION

Horn & Schunk (2)

Smoothness

since two variables at each point $(\frac{dx}{dt}, \frac{dy}{dt})$ need additional constraints to solve even for rigid field
add assumption of smoothness (gives completely smooth answer for rigid object, of course)

minimize: error + non-smoothness

should = 0

$C_{err} =$

$$\iint \left[\frac{\partial E}{\partial x} \frac{dx}{dt} + \frac{\partial E}{\partial y} \frac{dy}{dt} + \frac{\partial E}{\partial t} \right]^2 dx dy$$

non-smoothness

$$\iint \left[\left(\left(\frac{\partial u}{\partial x} \right)^2 + \left(\frac{\partial u}{\partial y} \right)^2 \right) + \left(\frac{\partial v}{\partial x} \right)^2 + \left(\frac{\partial v}{\partial y} \right)^2 \right] dx dy$$

$$u = \frac{dx}{dt}$$

$$v = \frac{dy}{dt}$$

$$\min: e_d + \alpha e_s$$

2D velocity

$\frac{dy}{dt} = v$

$\frac{dx}{dt} = u$

Examples -

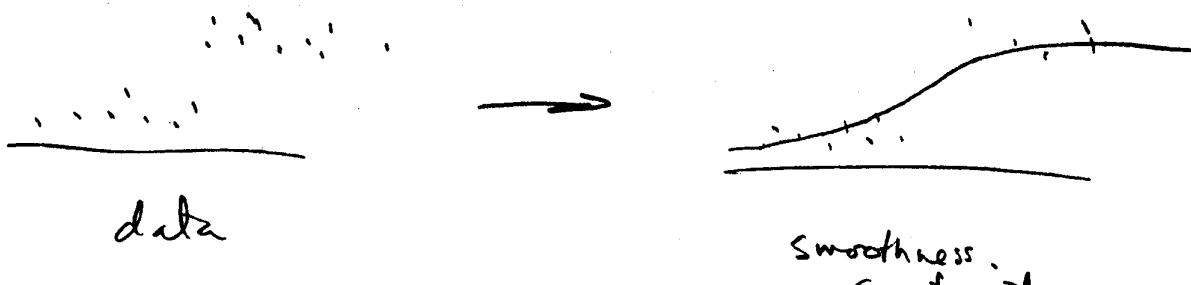
Iterative solution to minimization problem

stable in $\sim 10-20$ steps

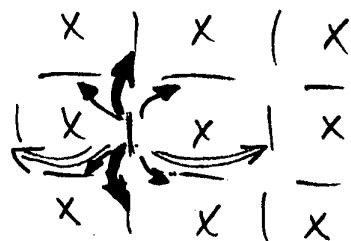
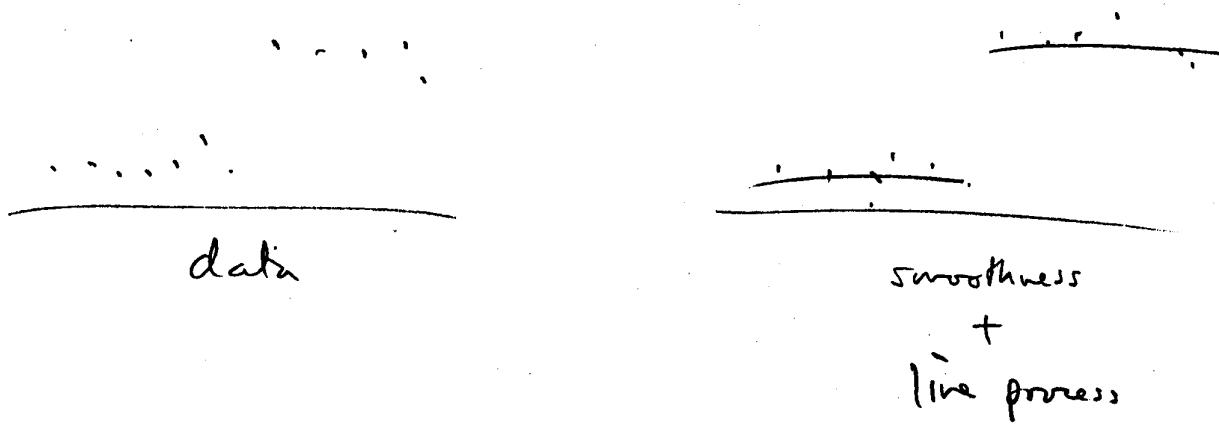
Problems

- people can't detect movement of smooth brightness gradient
- smoothness smears edges

Line Processes



minimize \rightarrow error + non-smoothness + penalty for discontinuity + penalty for non-colinear



EFFECTS OF ATTENTION IN V4

Moran & Desimone, 1985

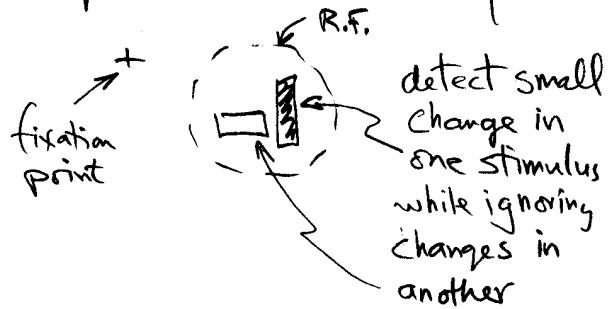
- standard model attention :

↳ this is wrong
attention is everywhere!

- macaque monkeys are pre-adapted for peripheral attention expts

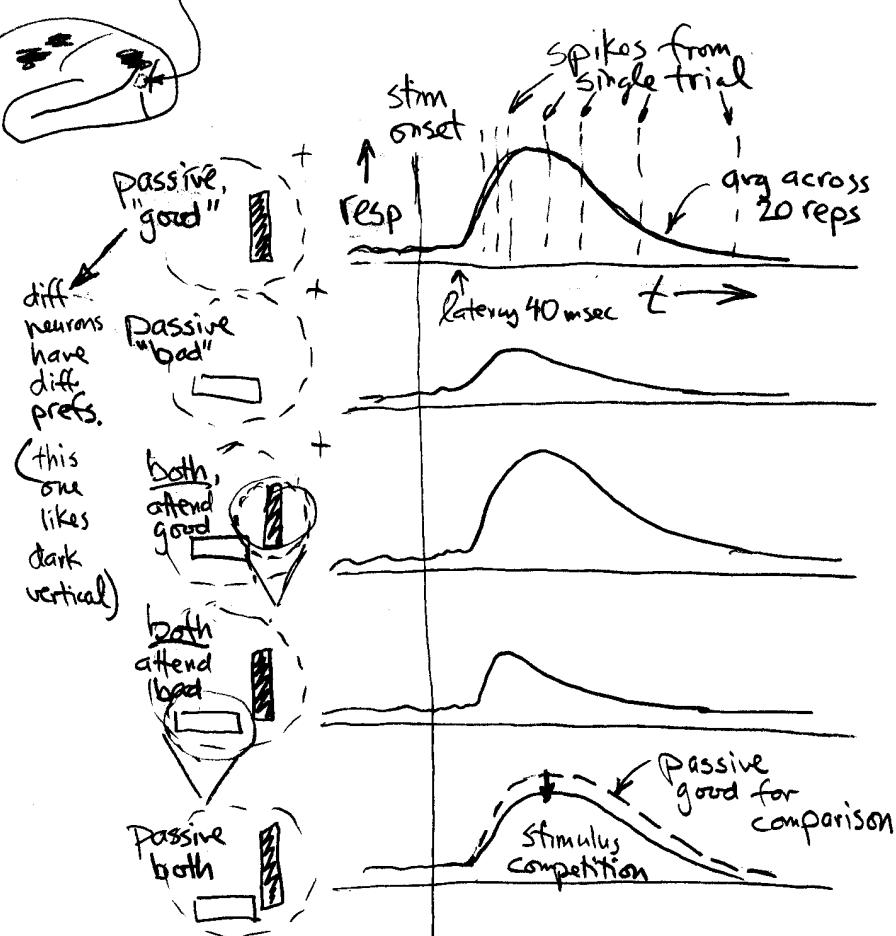
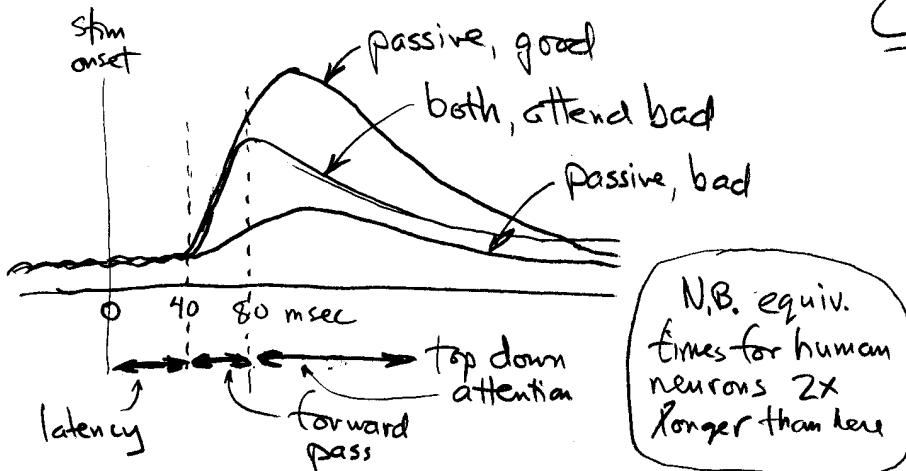
- 1) spend all day not looking directly at attended targets (e.g. dominant male)
- 2) eat many very small food items grass root

- operational definition of attn:



- much harder to train monkeys than undergraduates 😊

- 3) Some evidence that there is a "feedforward pass" before top down attentional filter kicks in



- 1) attention 'rescues' response to attended stimulus from the effect of stimulus competition

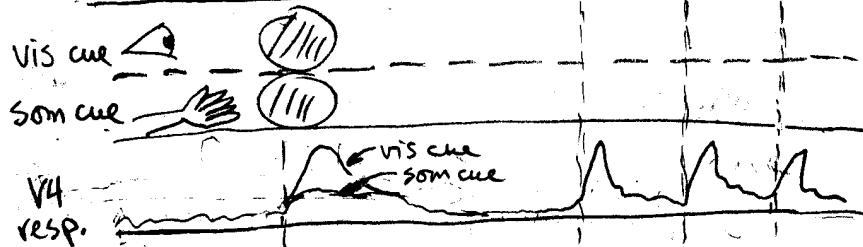
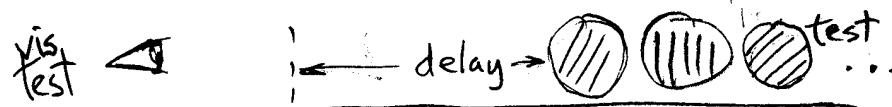
- 2) interpretation is ambiguous :
[attention to spatial location?
attention to features?]

Conclusion

- attention to location and features is present to some degree in every visual area (and somato. & and areas, too)

Haenny & Mansell

somatosensory-visual & visual-visual



test responses

(2) test stim

///	///	=	///
cue	///		
=	✓	✓	✓
==			

(1) test stim

///		✓		
///			✓	
=			✓	
==			✓	

"cue-tuned"

orientation-tuned

(3) test stim

				✓
cue				

(4) test stim

				✓	
cue					

both "cue"-tuned
and orientation-
tuned!

"match cell"

Unencapsulated

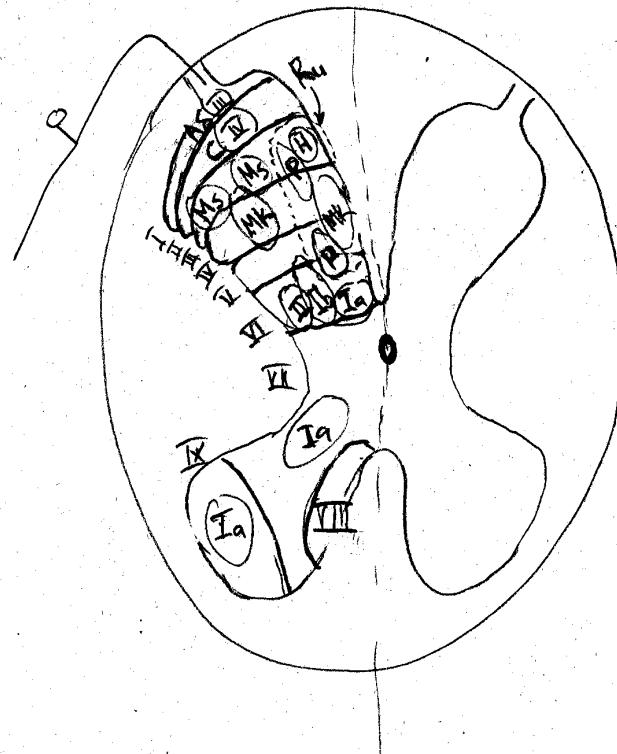
- heat
- cold
- pain < "x", C, sustained, small/unmyelinated
- pain < "y", A_S, transient, large/myelinated
- slow stroking

Receptors

Touch

Superficial

- Merkel disks — sustained (= slowly adapting)
- Meissner's corpuscles — transient (= rapidly adapting)



Deep

- Ruffini endings — sustained
- Pacinian corpuscles — transient
- hair follicle receptors — transient

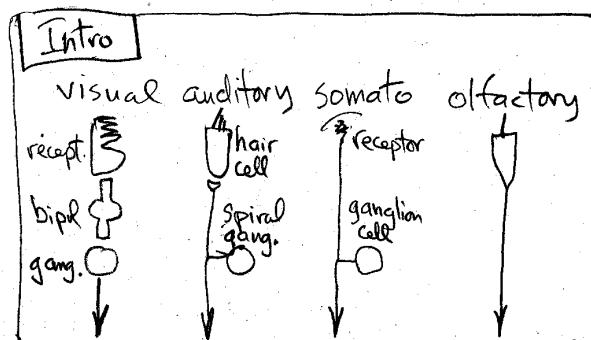
Muscle receptors

"Stretch" → length
↪ "ON" - only

- II spindles — sustained
- Ia spindles — transient
- Ib — Golgi tendon organs → force

Pain topics

- no receptors a prob!
- cognitive pain ("good jam!")
- gating (e.g. machine shop)
- opiate-induced hyperalgesia
- opiate "don't care"

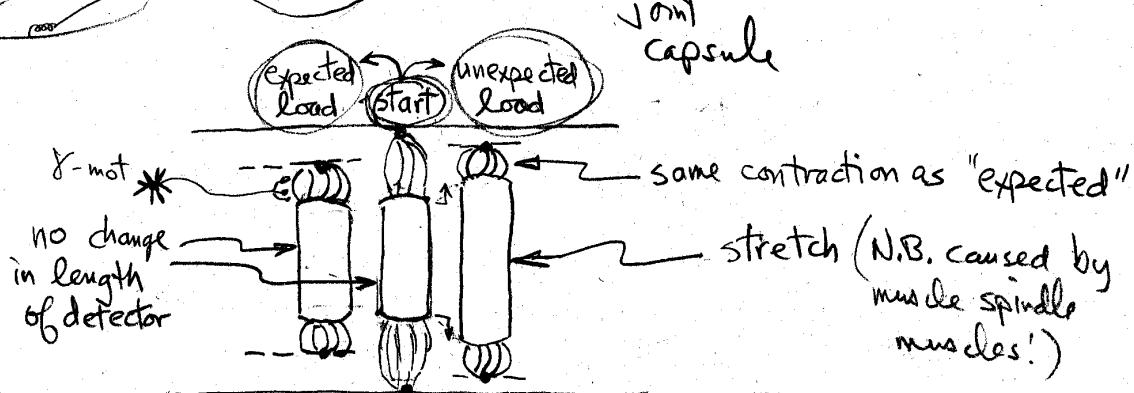
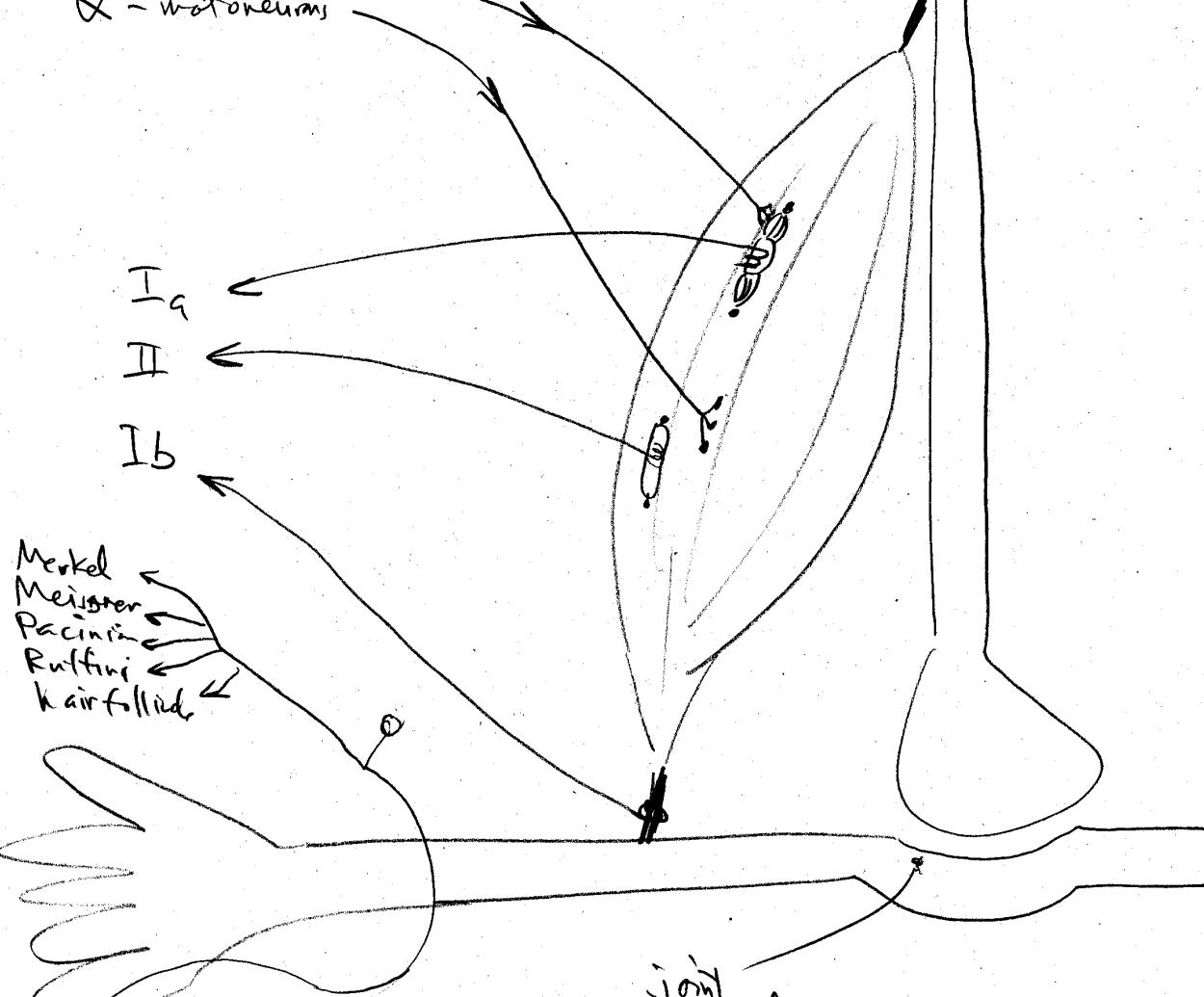


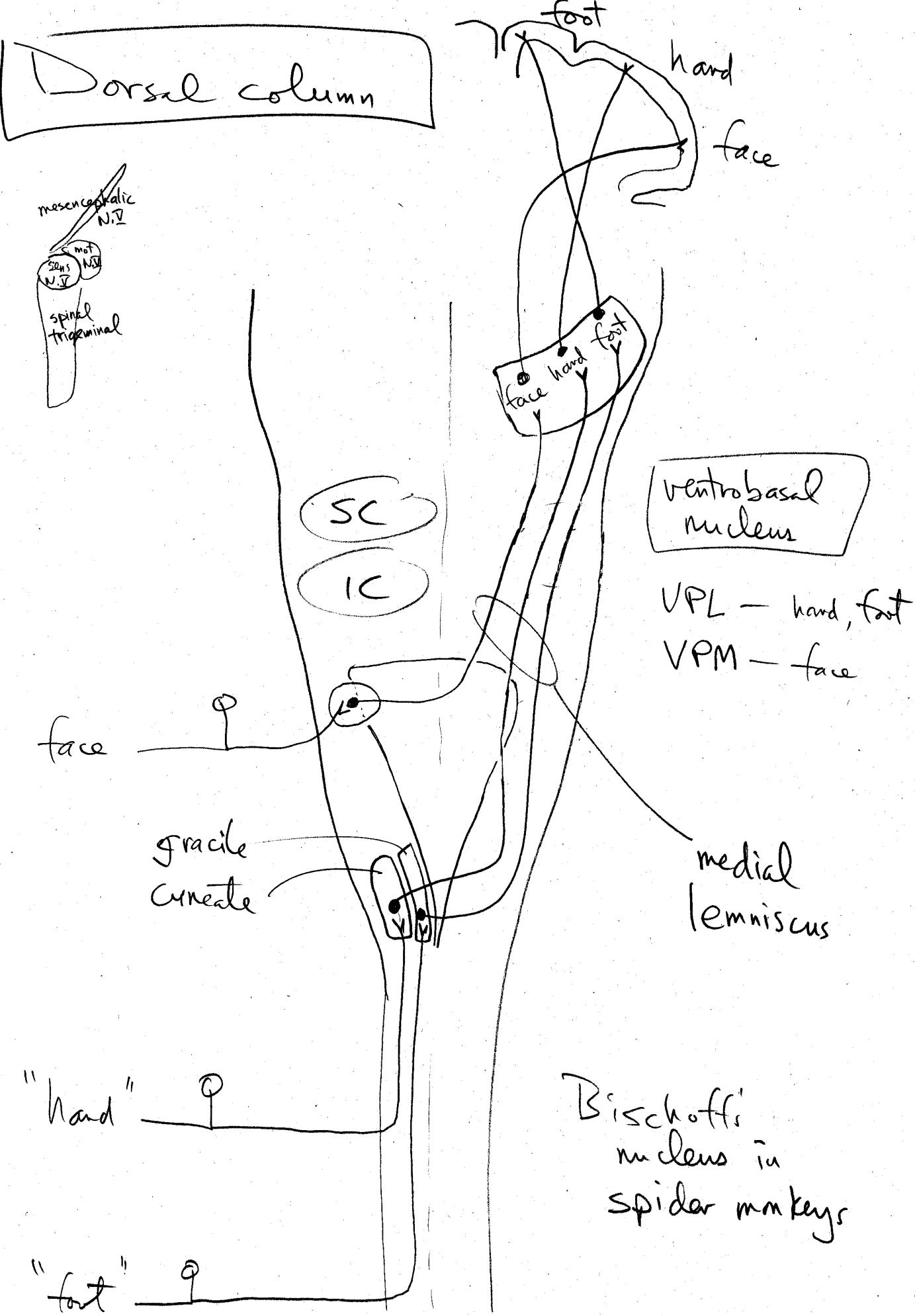
Co-contraction of α , γ motoneurons to detect deviation

- EMG
- length tells state of body!

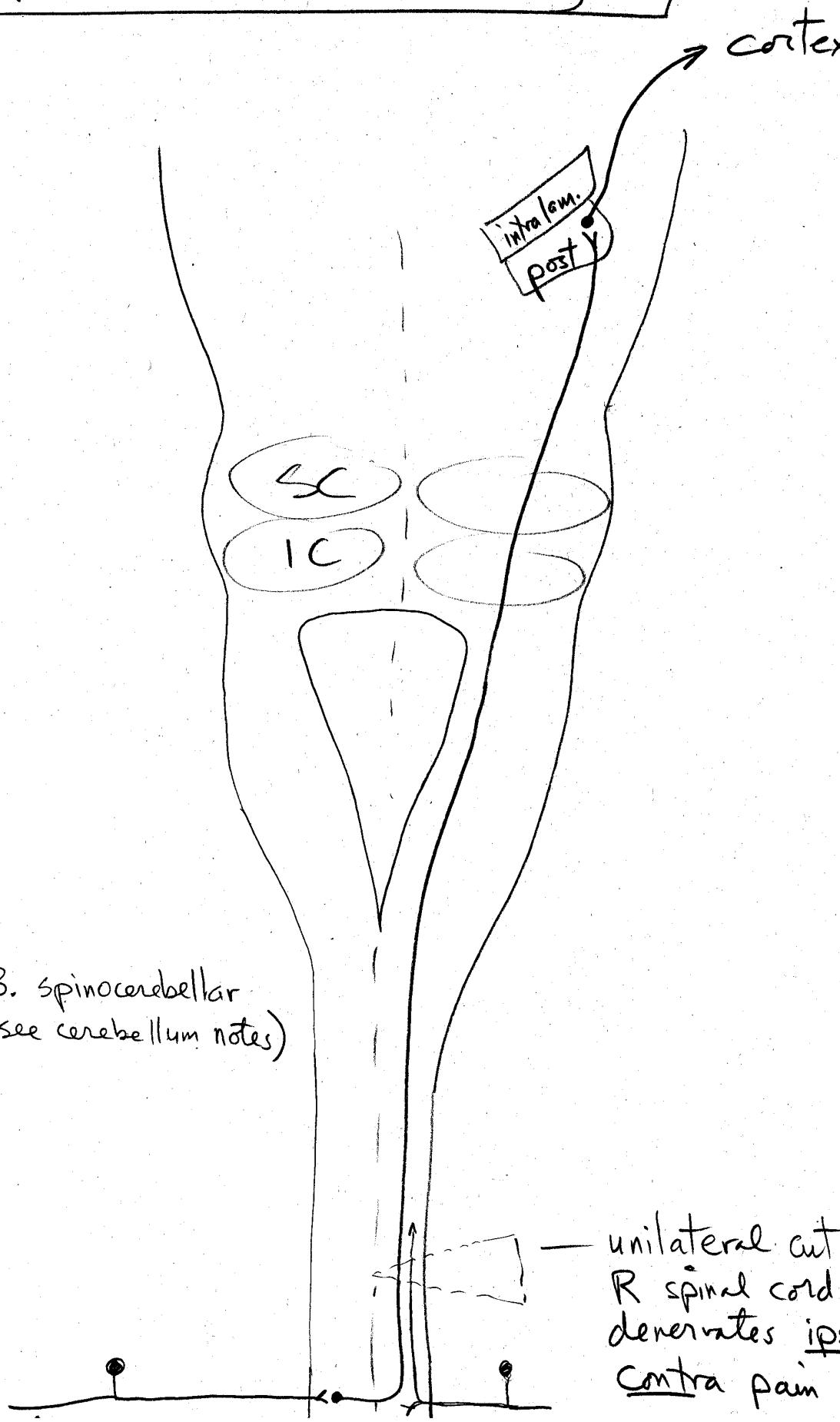
γ - motoneurons

α - motoneurons

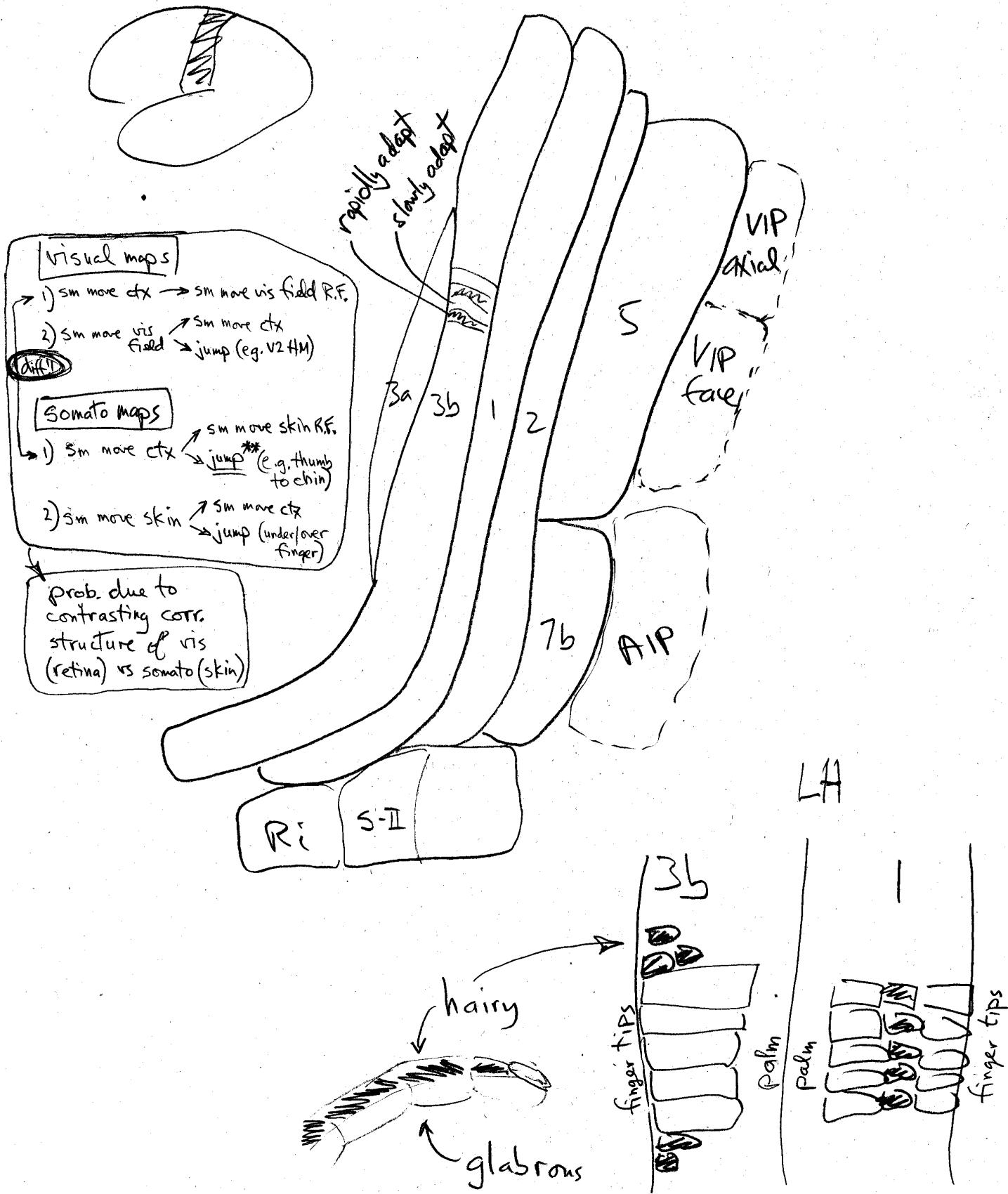


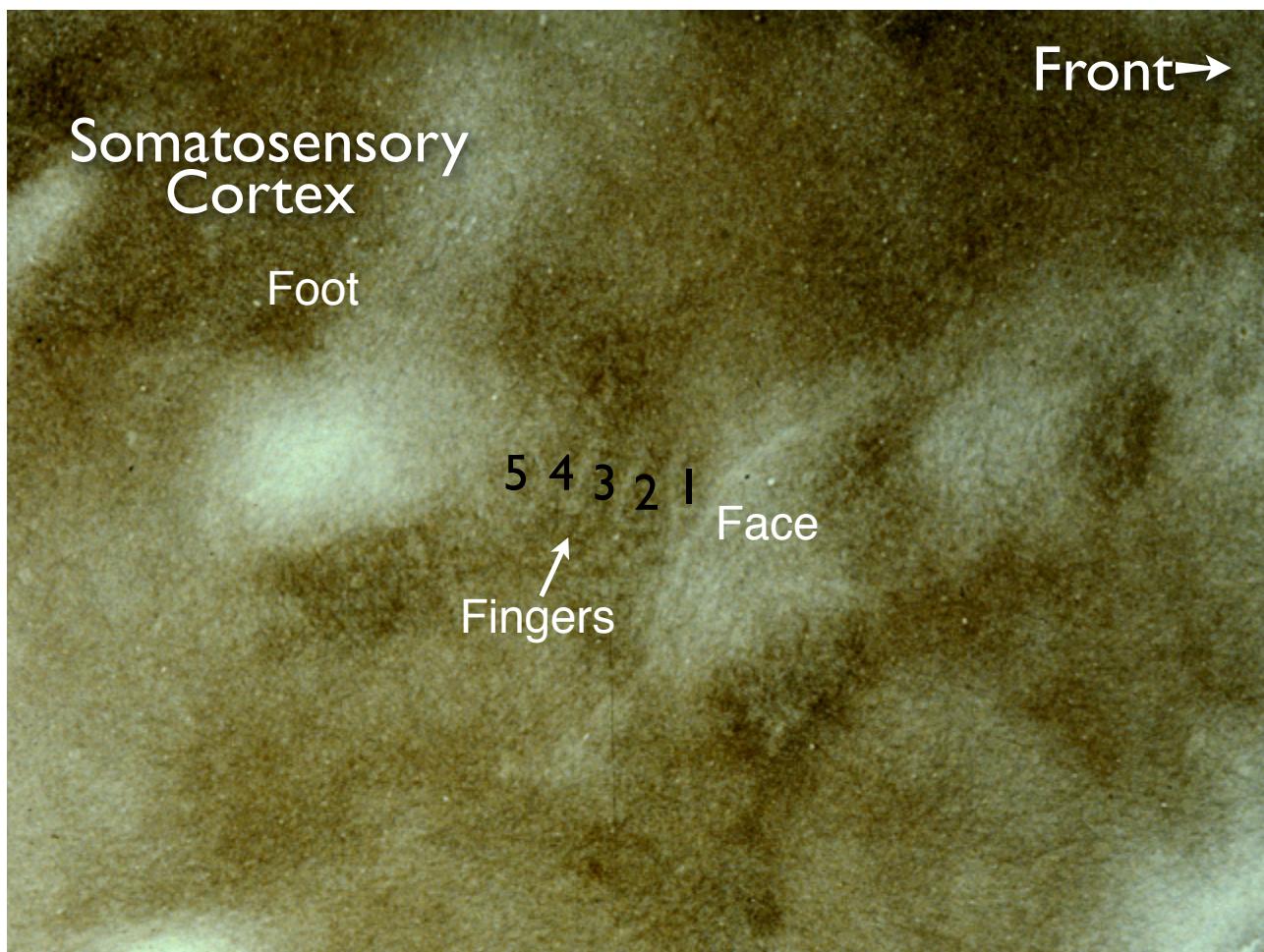
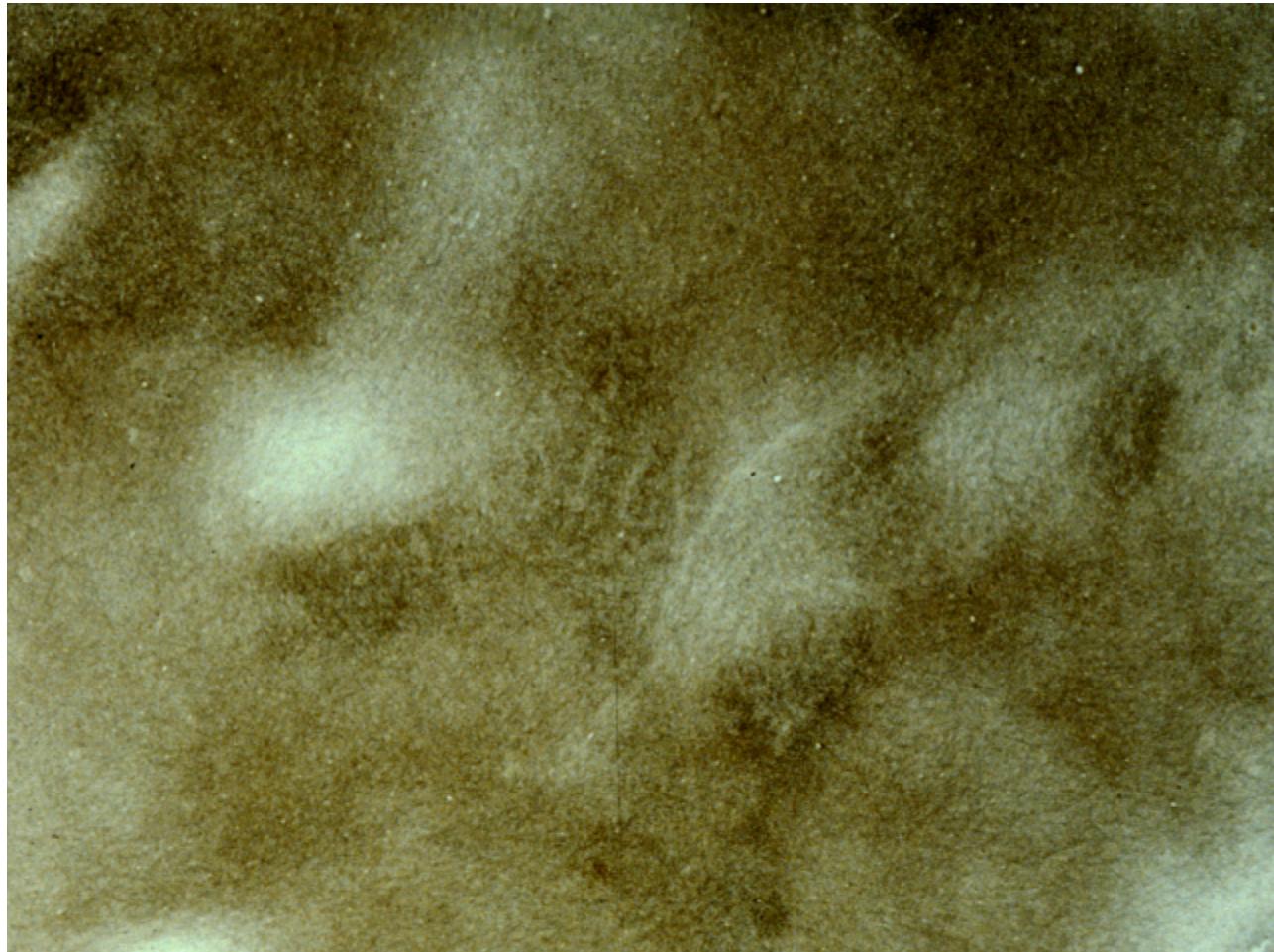


Spinothalamic Pathway

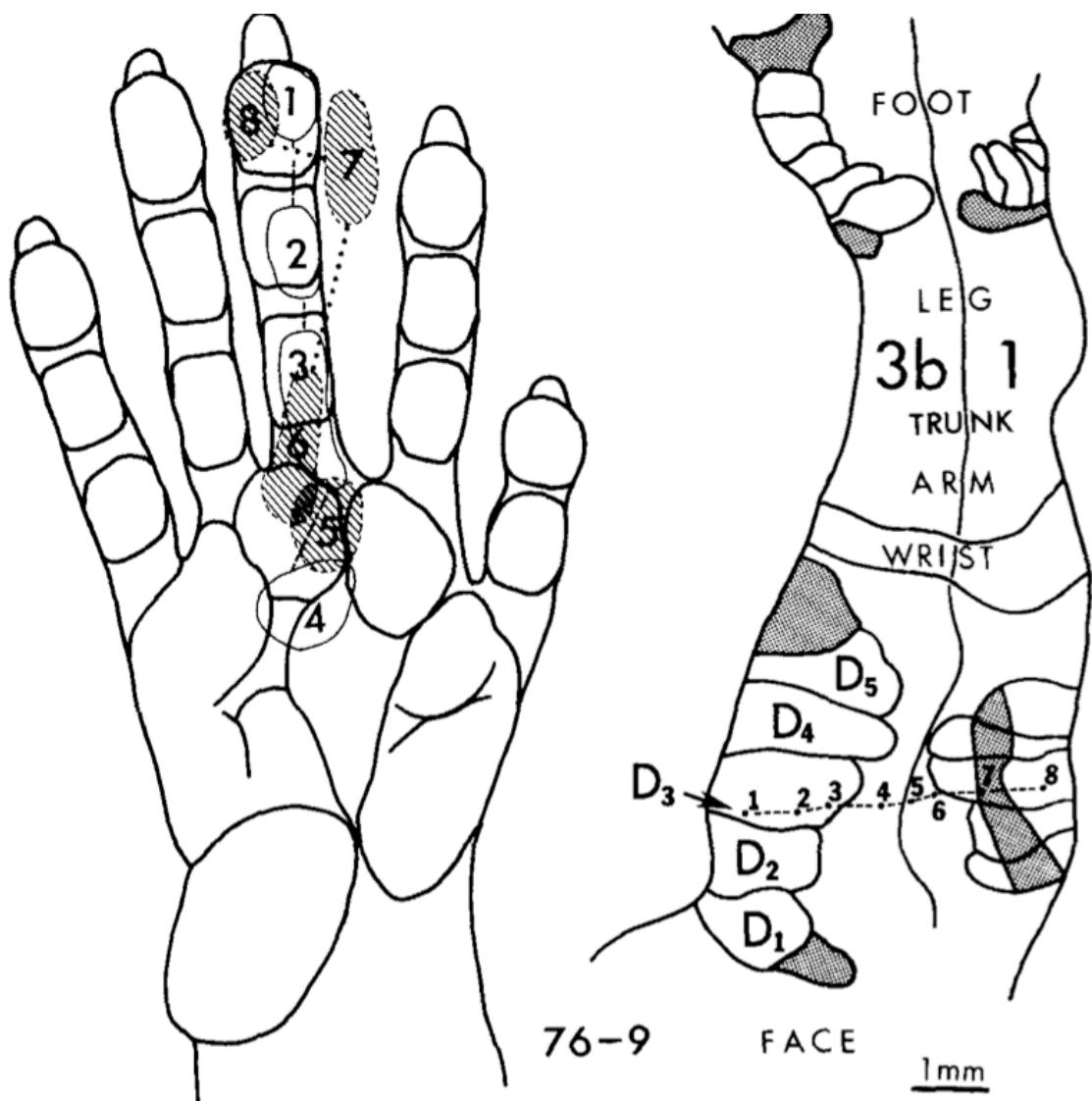
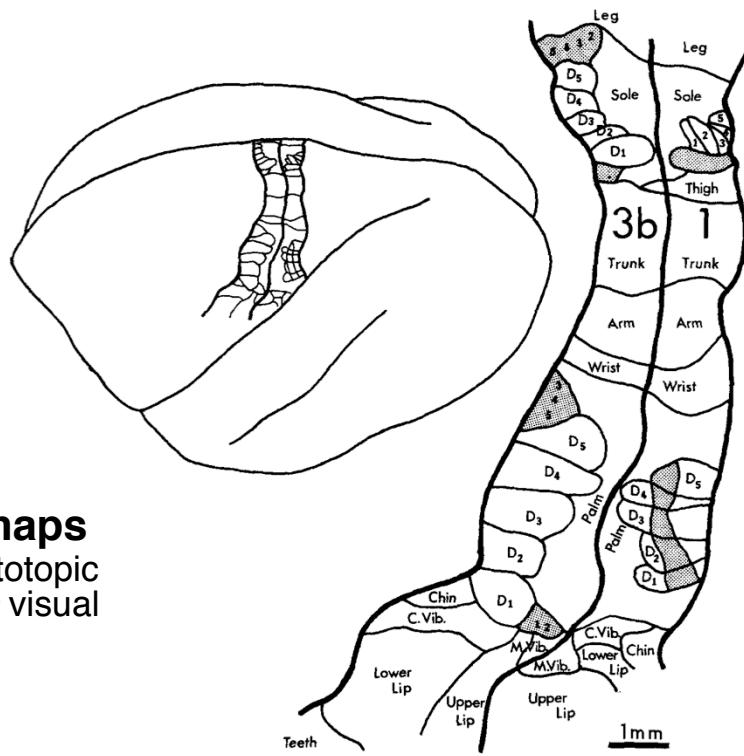


Somato Sensory Cortex



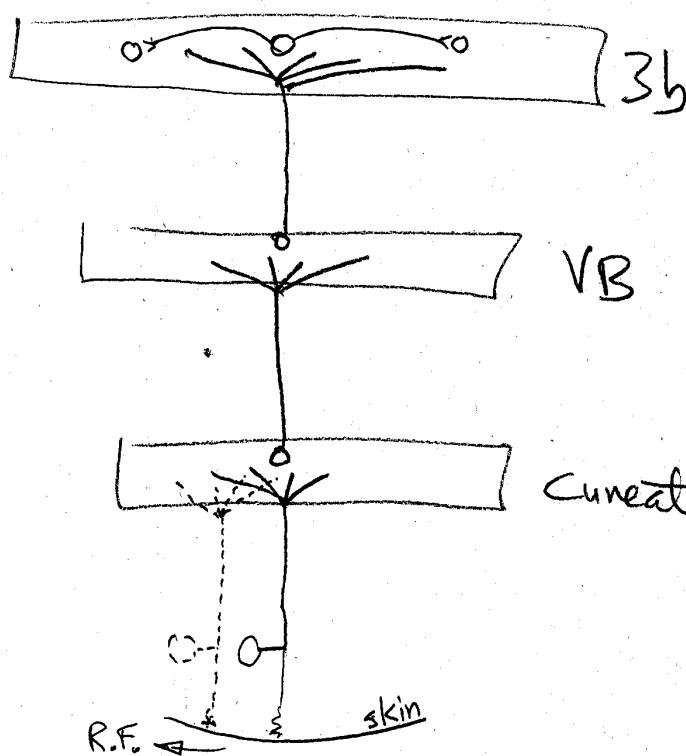


2D maps
somatotopic
versus visual



Plasticity Expts

- denervate radial
- transplant patch
- syndactyly
- just stim — cf. and. Allard
- Silver Spring monkeys — face invades hand
1+ cm



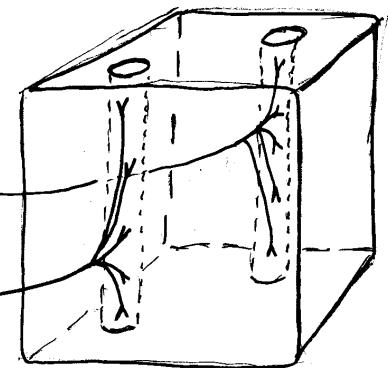
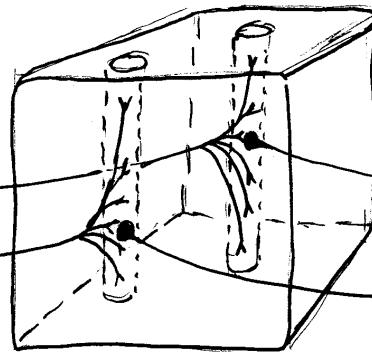
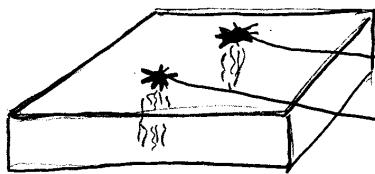
- [- turn up terminals on R
- [- moves RF to L
- [- no plasticity corticocortical!!
- [- no plasticity VB → 3b !!

General Principles — Auditory

- A) sensory surface is 1D
- B) connection difference — convergent-divergent within isofrequency domains
- C) fine-grained temporal info
- D) no "direct" rep of stim location
- E) cortex is further away (6th order vs. 4th order)

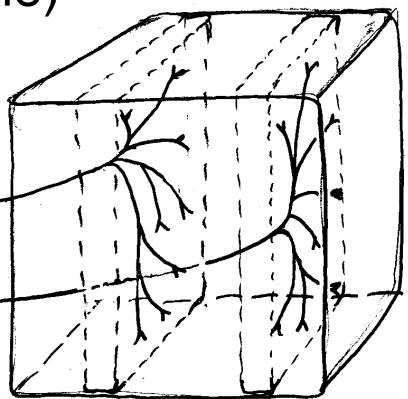
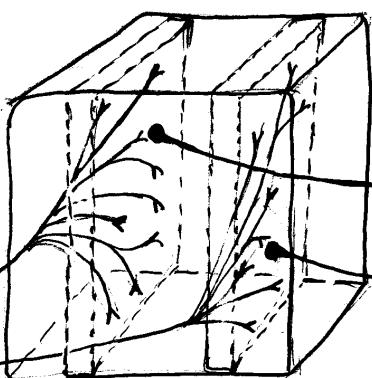
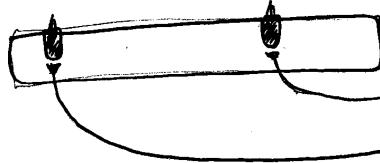
Point-to-point (point-to-line)

2D receptor sheet
(visual, somatosens.)

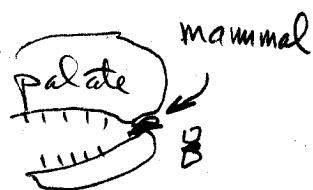
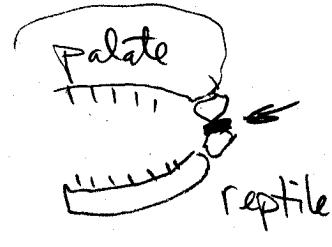
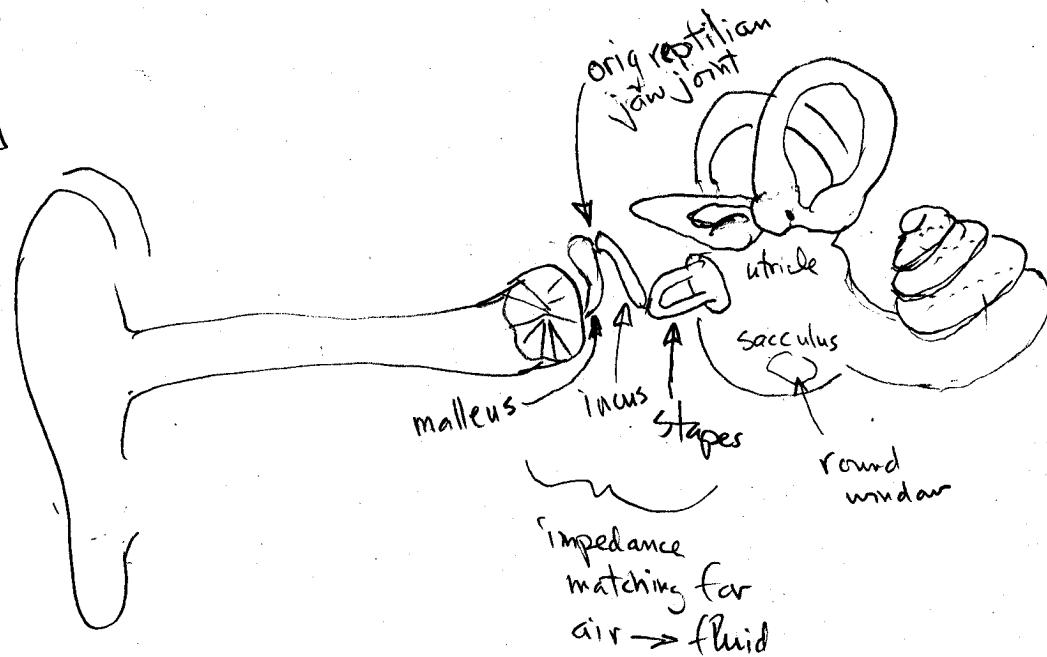


Point-to-line (point-to-plane)

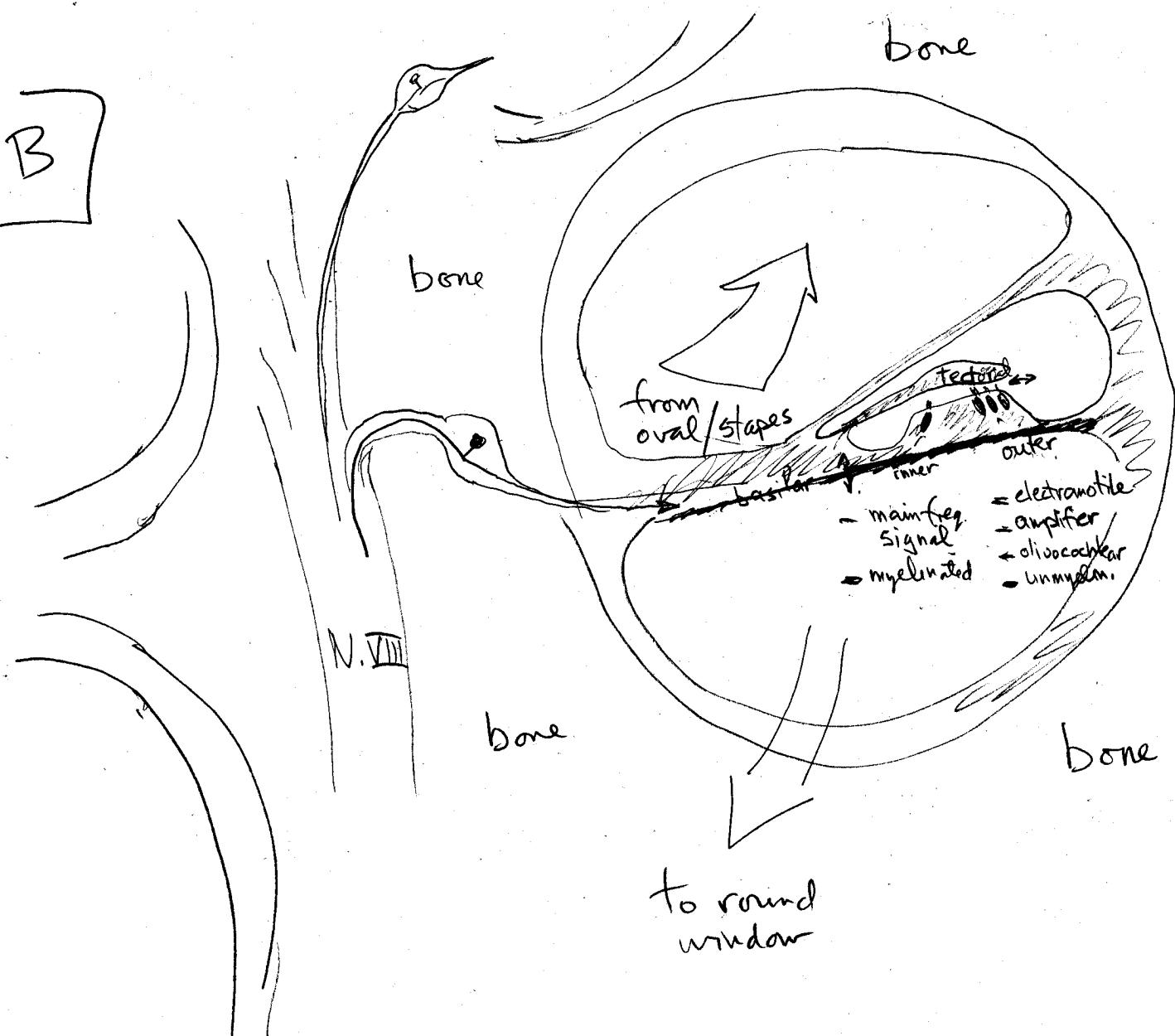
1D receptor sheet
(auditory)



A



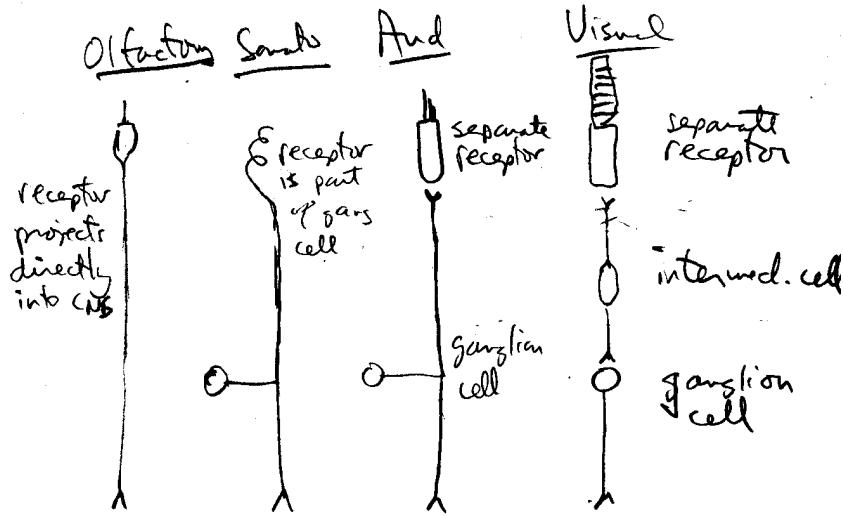
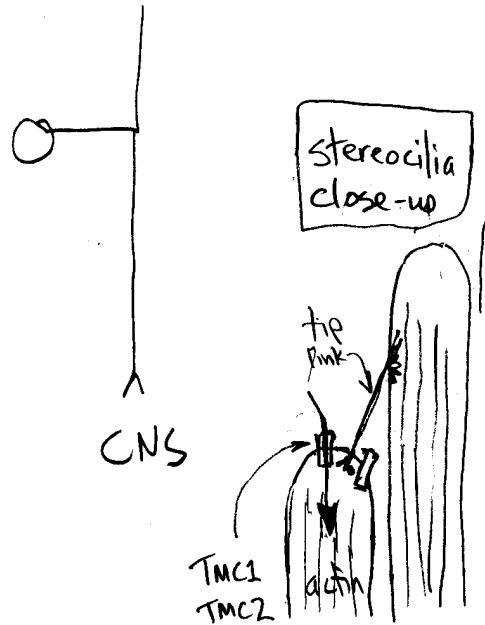
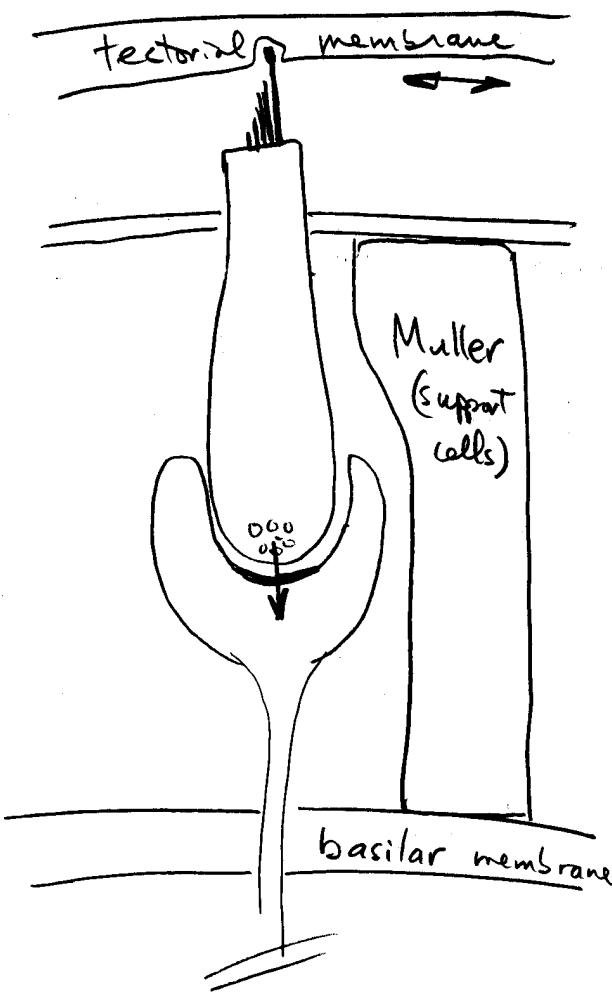
B



C

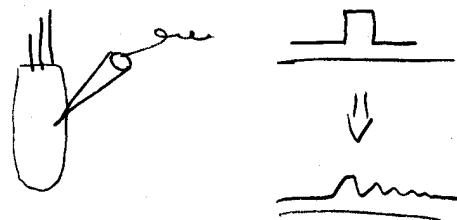
Receptors - Auditory

Hair cells

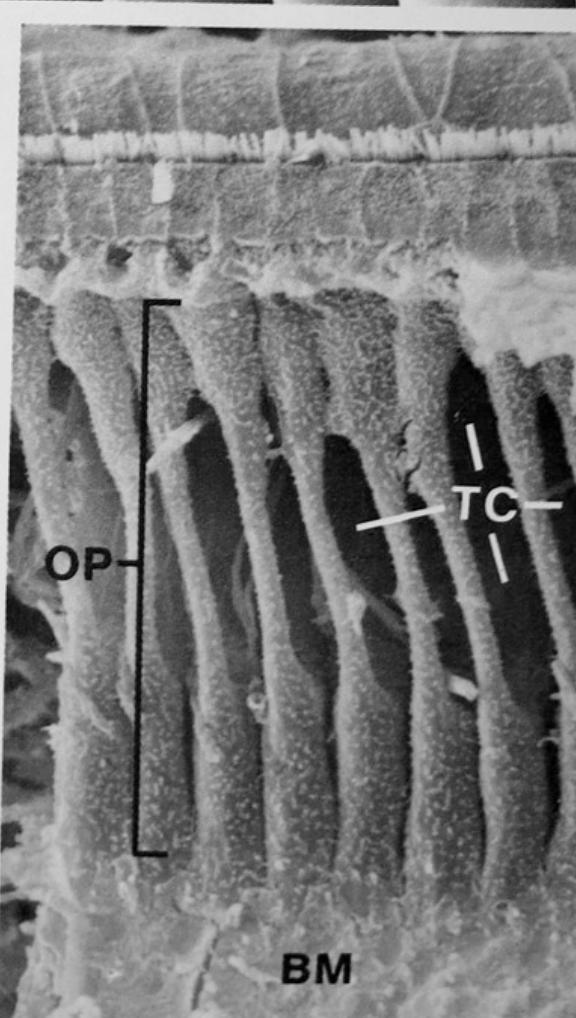
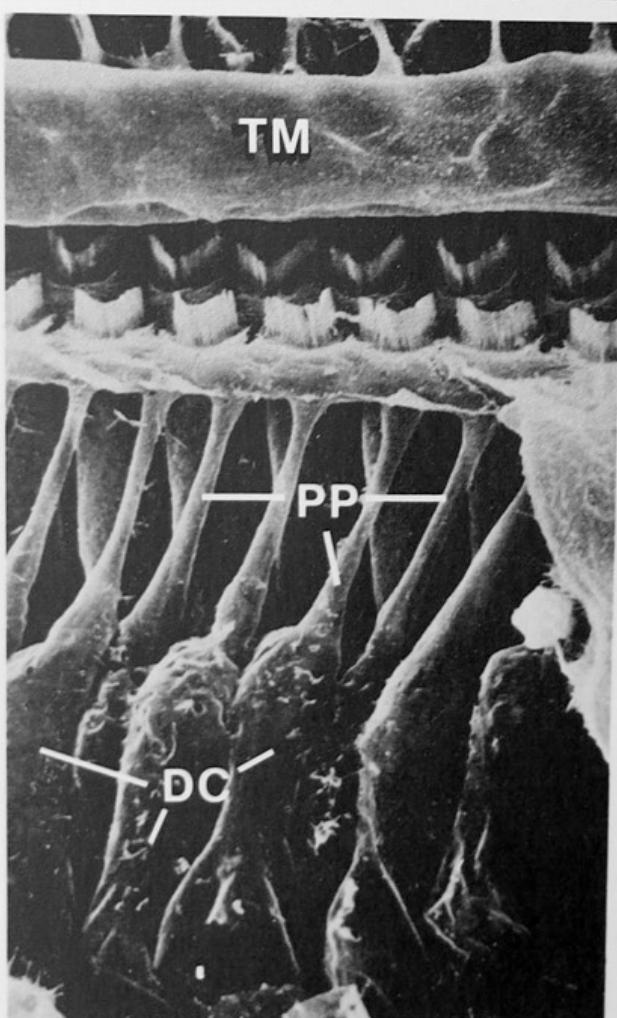
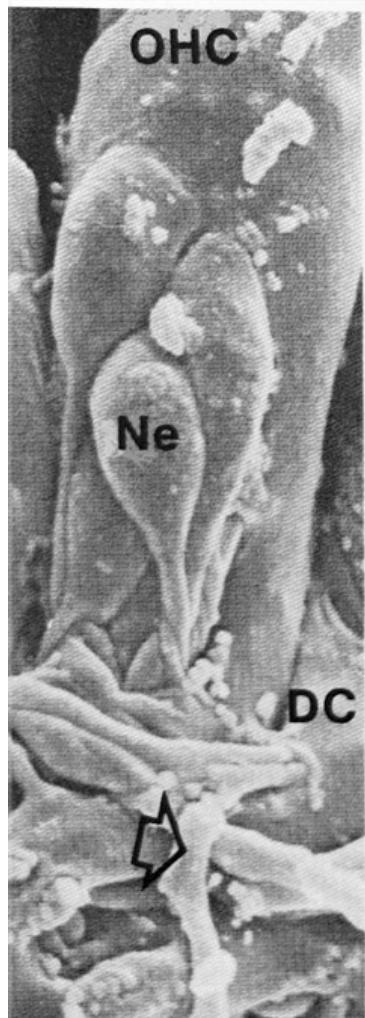
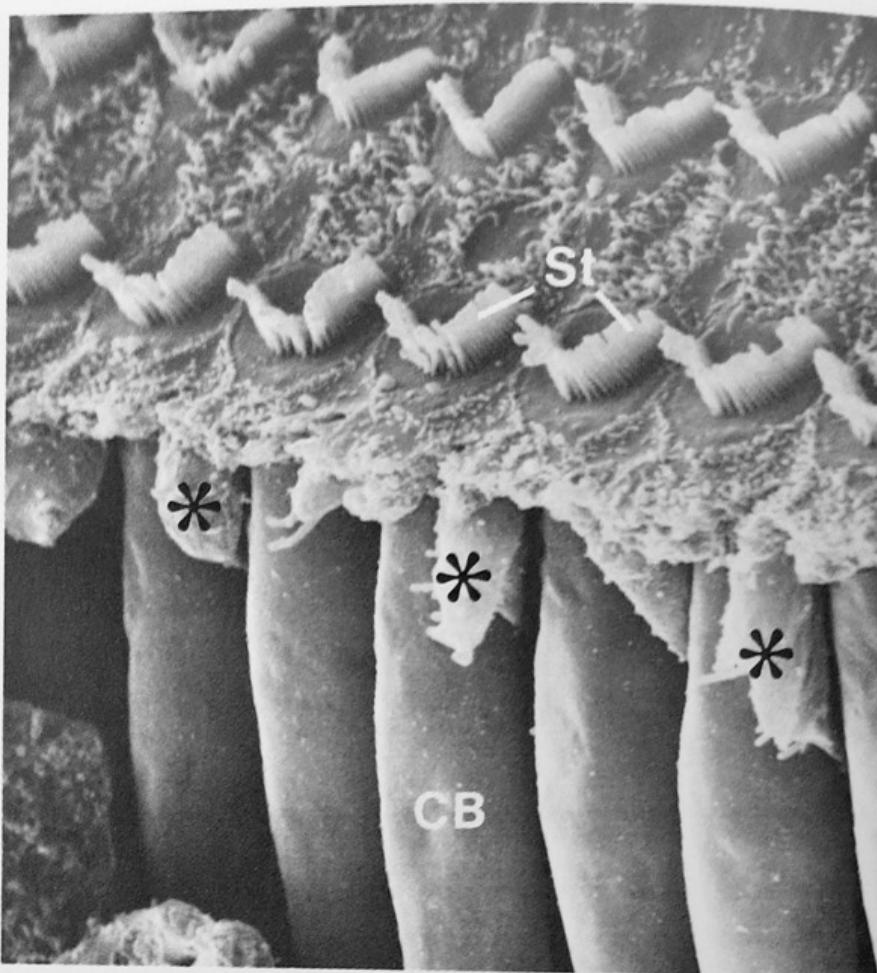
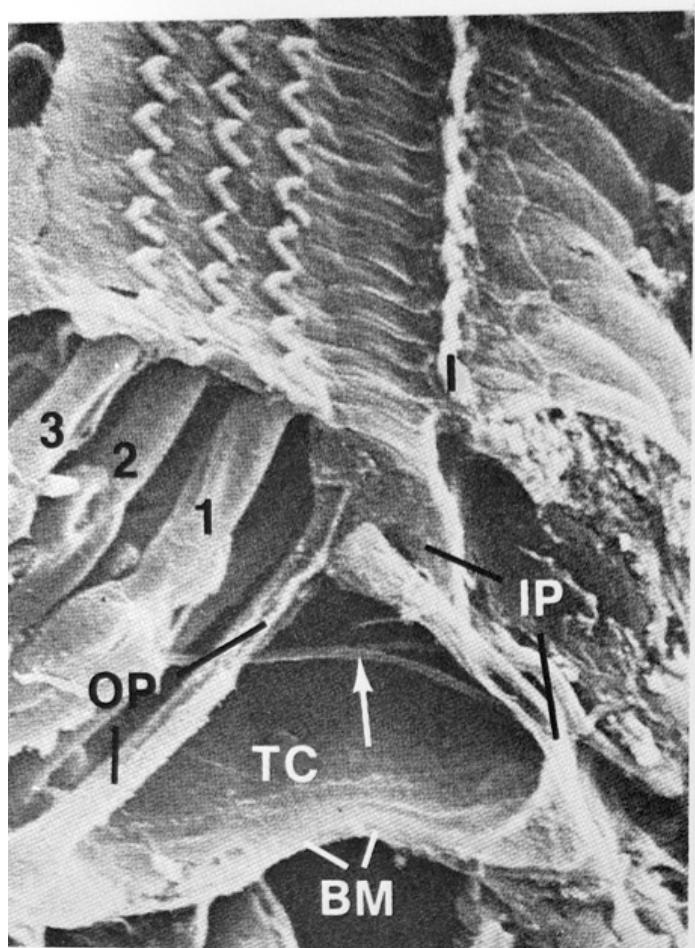


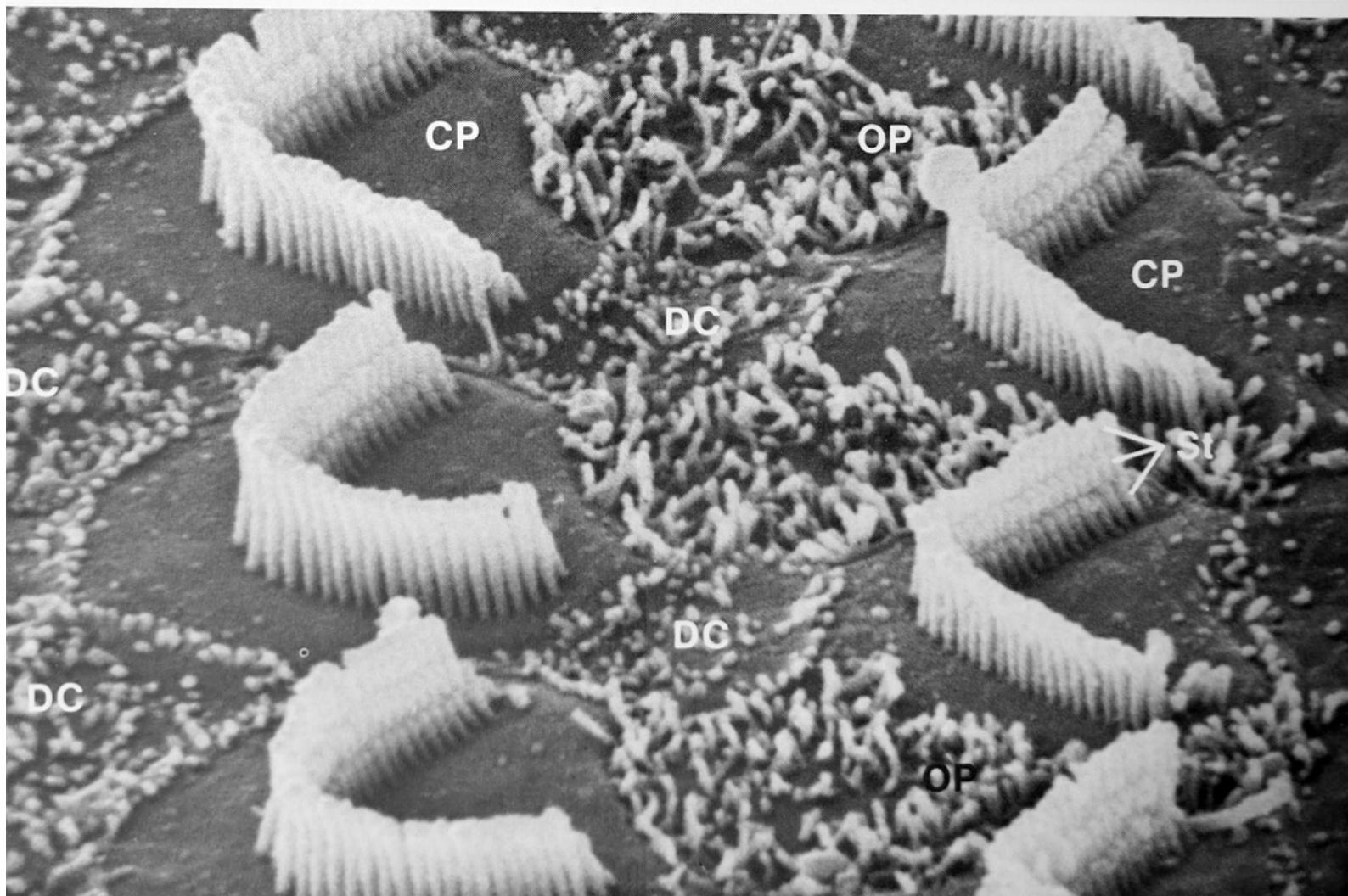
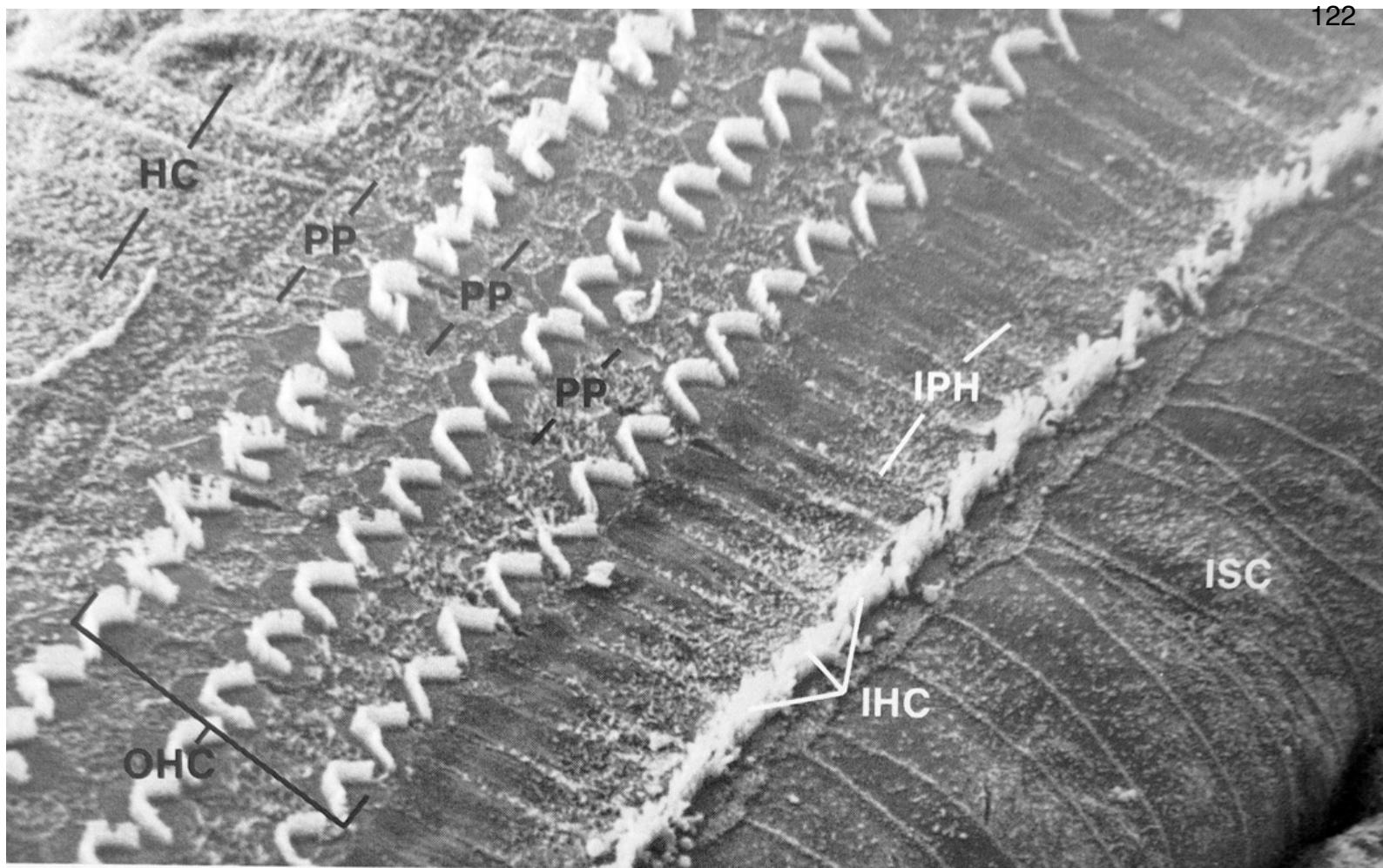
How hair cells become frequency selective

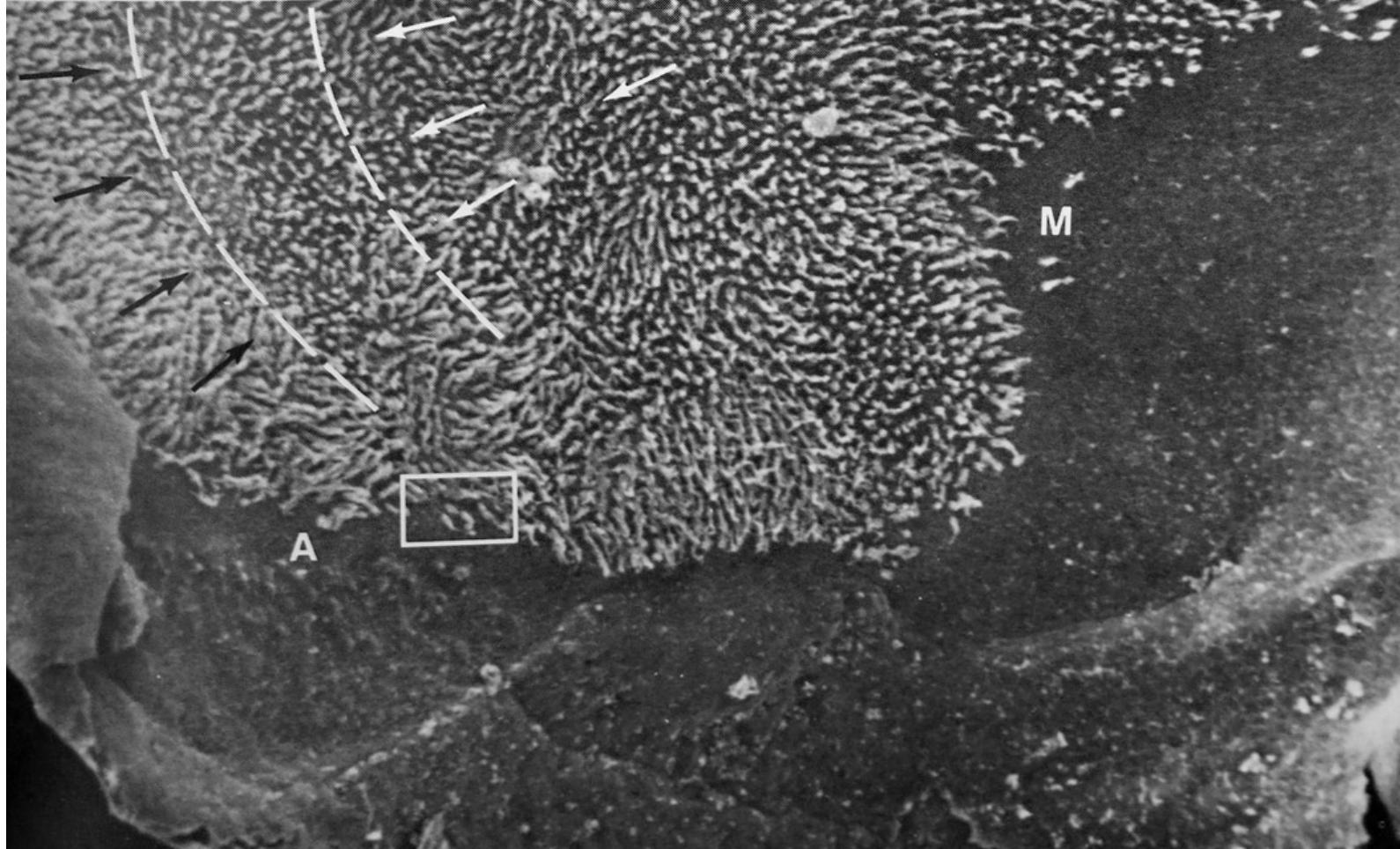
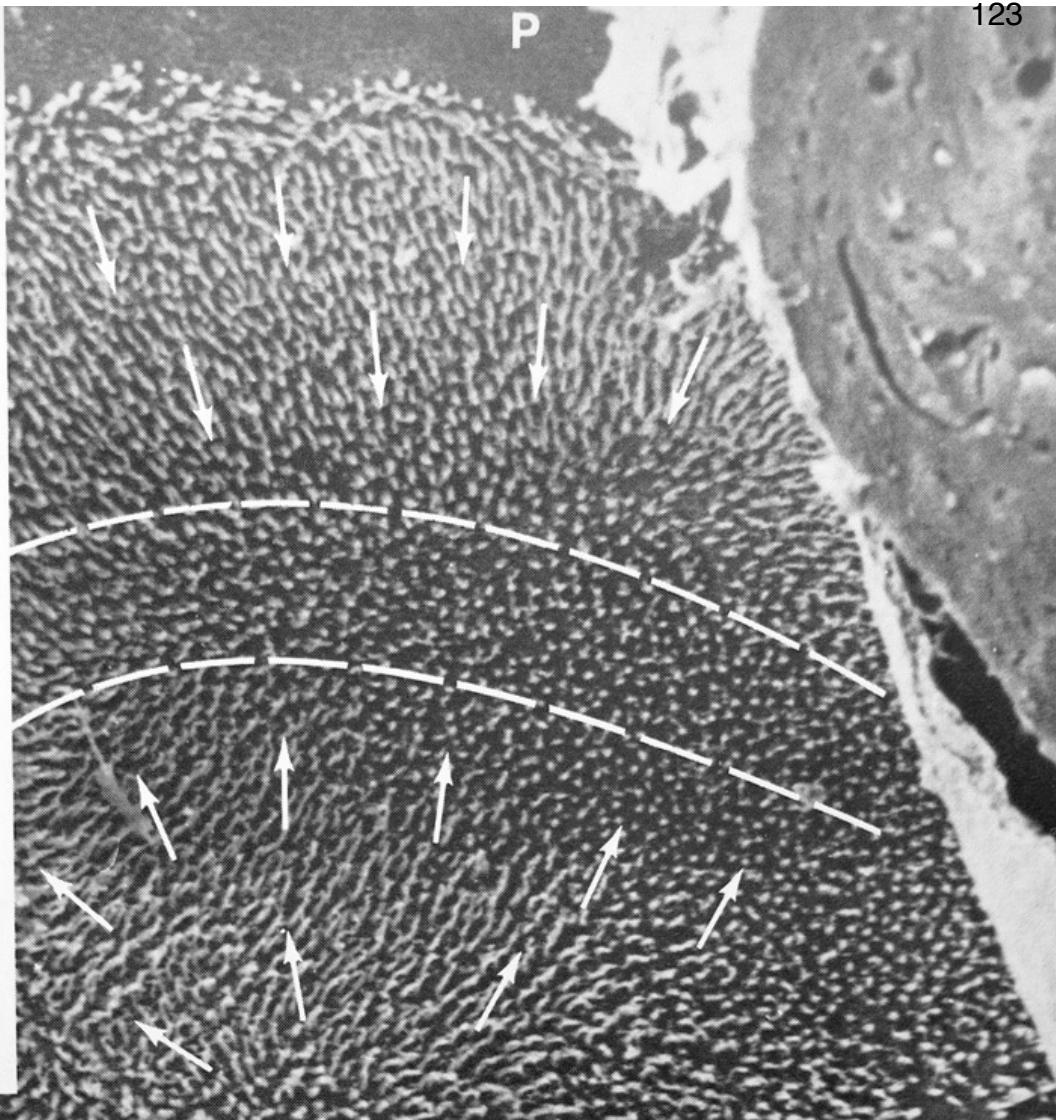
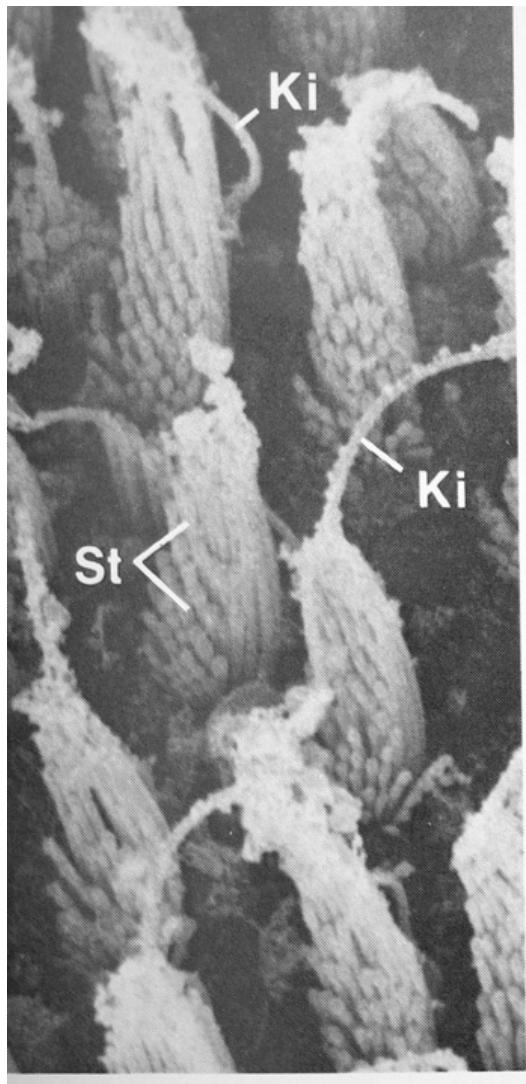
- 1) gradient in stiffness of basilar membrane
→ max in traveling wave
- 2) electrical tuning
- 3) mechanical tuning of cilia base
- 4) active physical response to sound



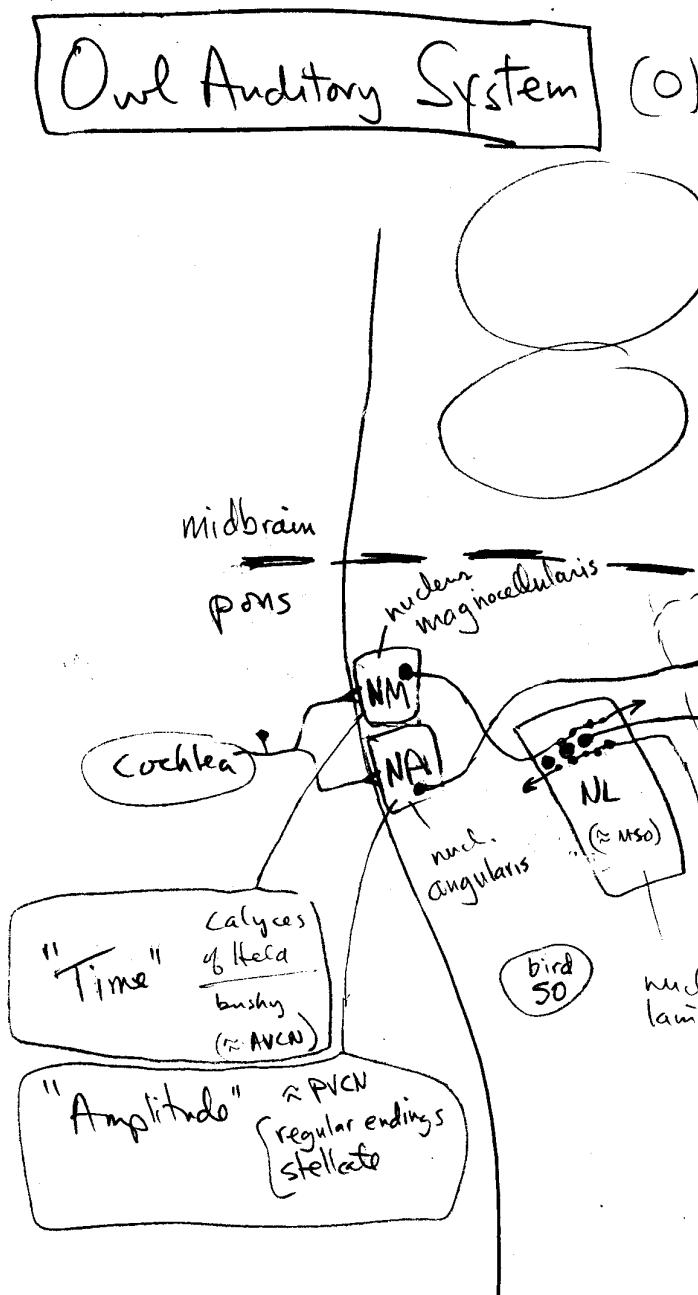
- 1) gradient in stiffness of basilar membrane
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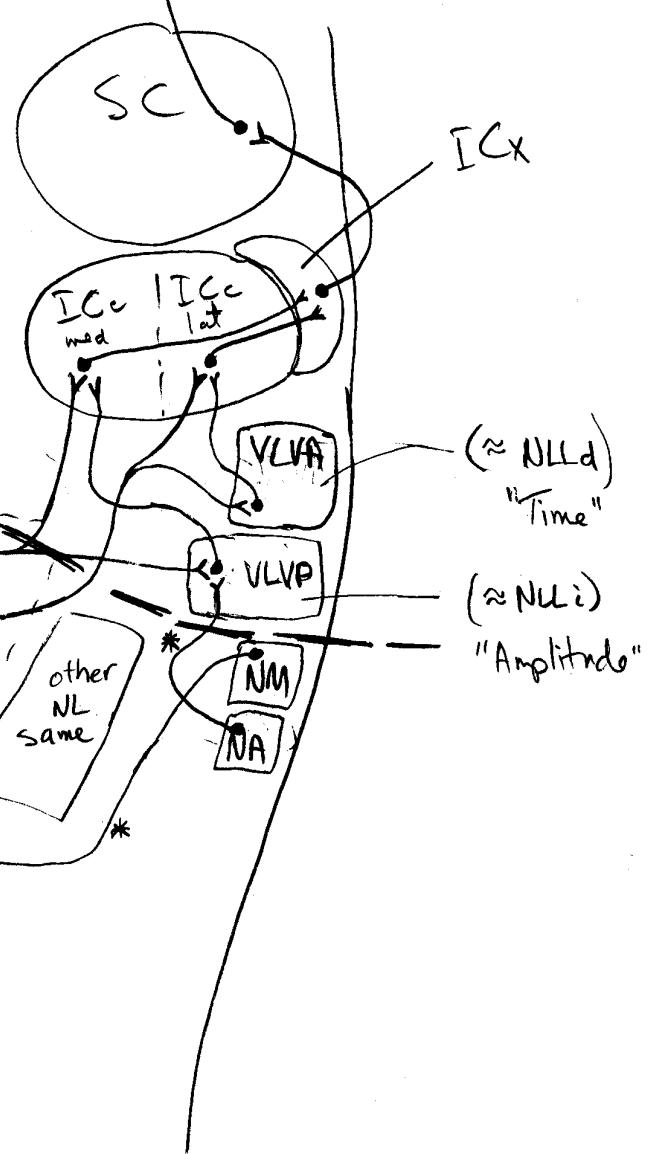


#11



(O)

(orig Wulst space map)

**ICx** \approx NLLd
"Time" \approx NLLi
"Amplitude"

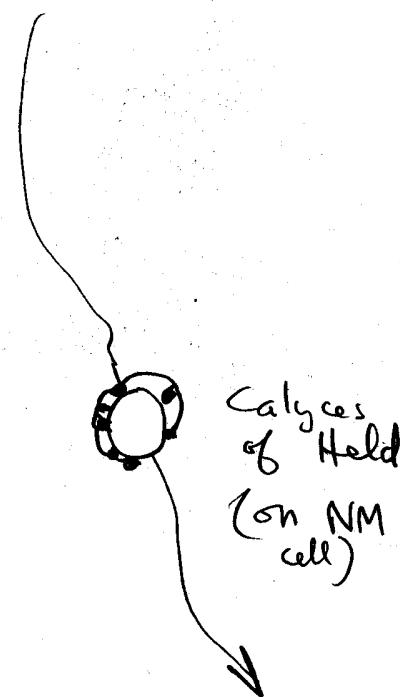
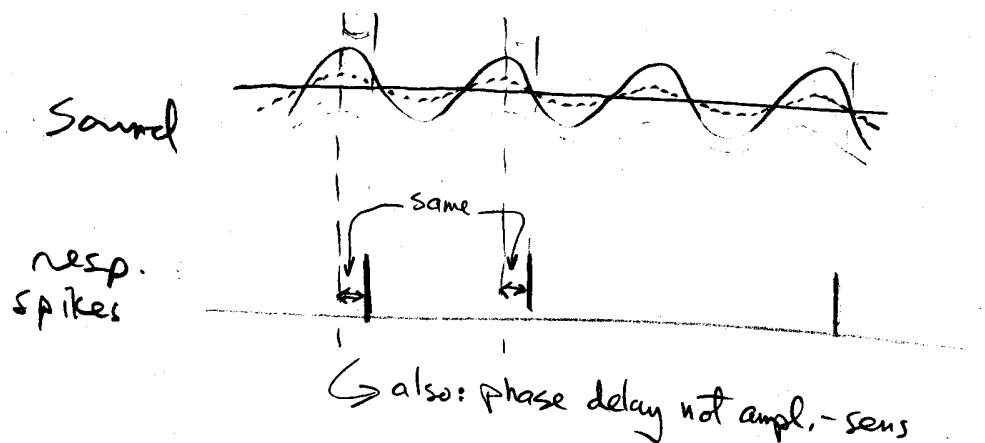
*breaks rule of Sereno

Owl Physiology (1)

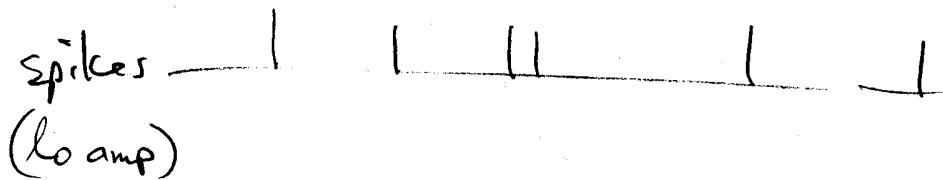
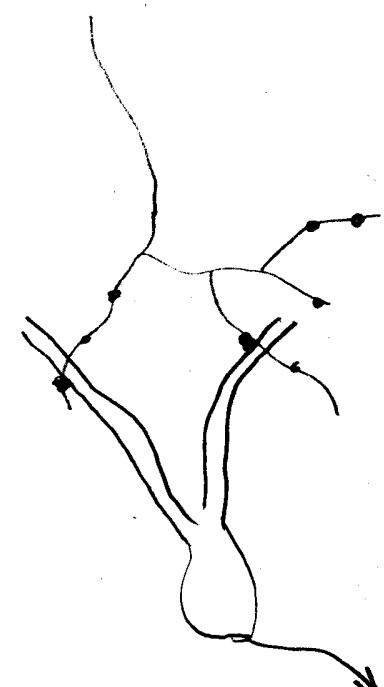
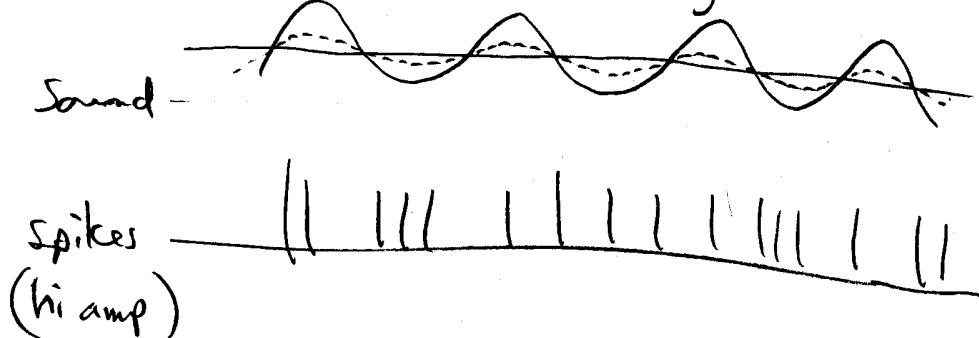
into to sound

frequency
amplitude
phase

NM — phase-locked



NA — amplitude coding

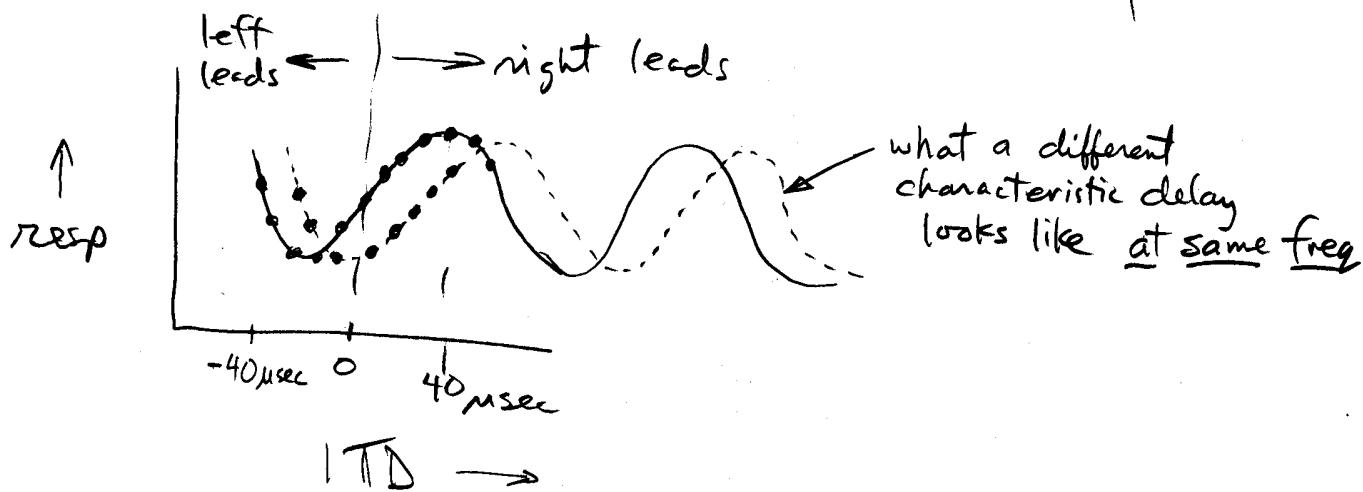
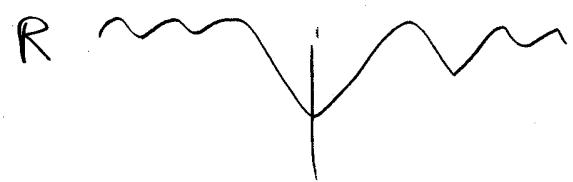
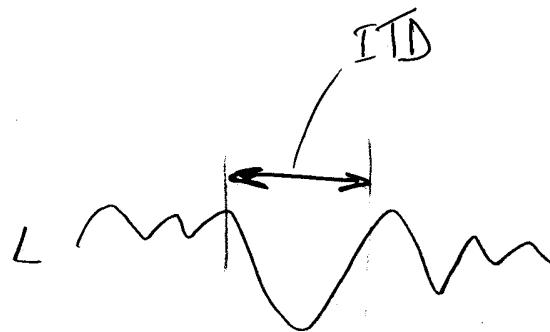


Owl Physiology

(2)

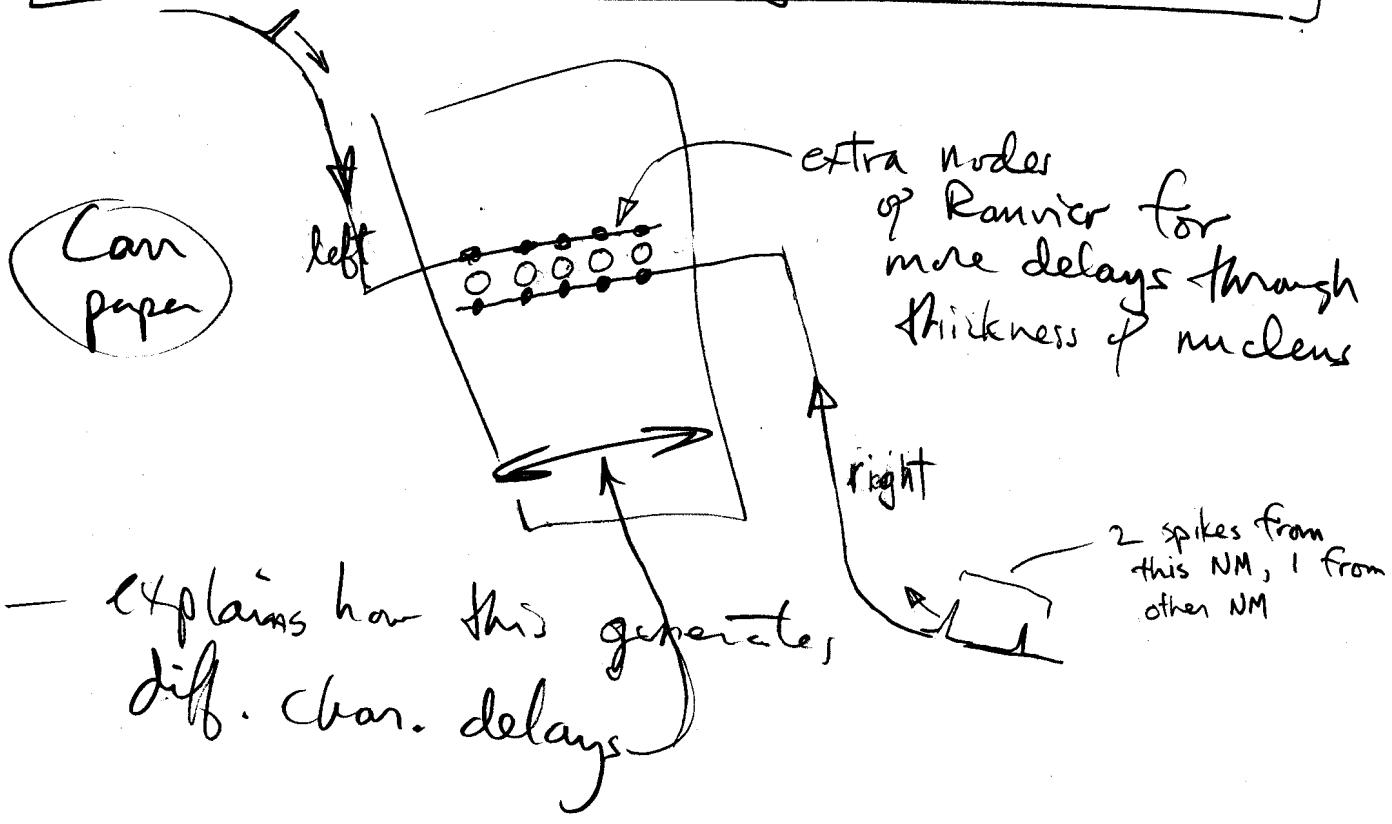
NL (\approx MSO)

- first binaural
- ITD's



VLVP (\approx NLLi)

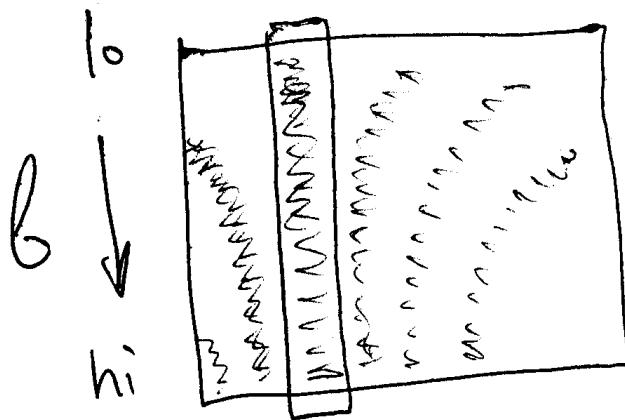
How diff. char delays start in NL



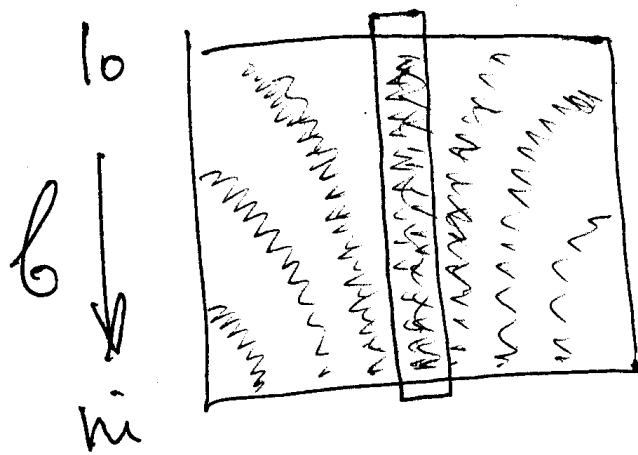
Owl Physiology | (3%)

Patterns in ICc lat w/white noise & ITD's

different



- white noise at
one ITD



- white noise at
a different ITD

char. delay
(same as ITD if summed across f)

- this is view of one ITD
across space

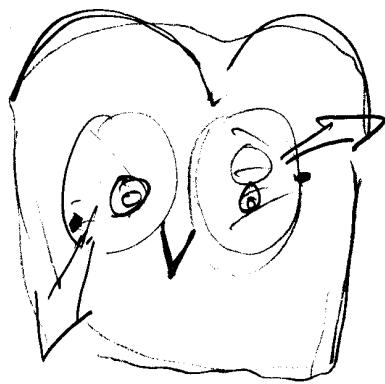
versus previous:

graph of many ITD's for
one NL/ICc lat neurons

Owl Physiology (4)

ICx

- Space map
- amplitude vs . time delay

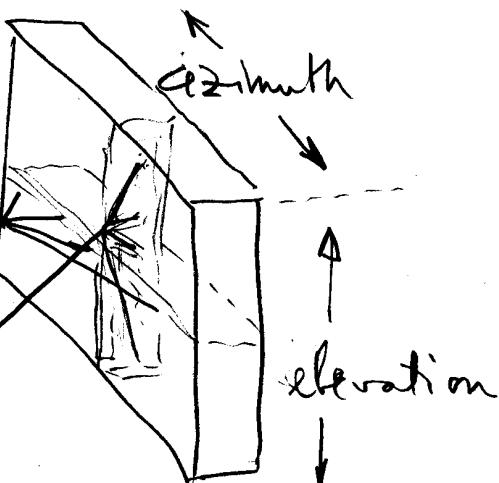


one characteristic
amplitude diff

b₁

b₂

b₃



voilà!

a space map

one characteristic
delay

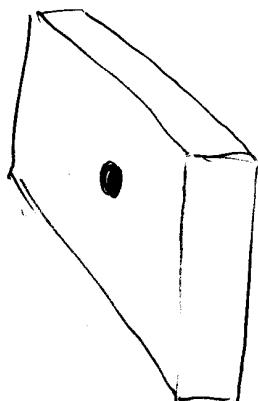
b₁

b₂

b₃

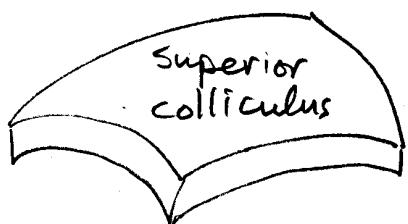


- activation in response to white noise coming from a small speaker



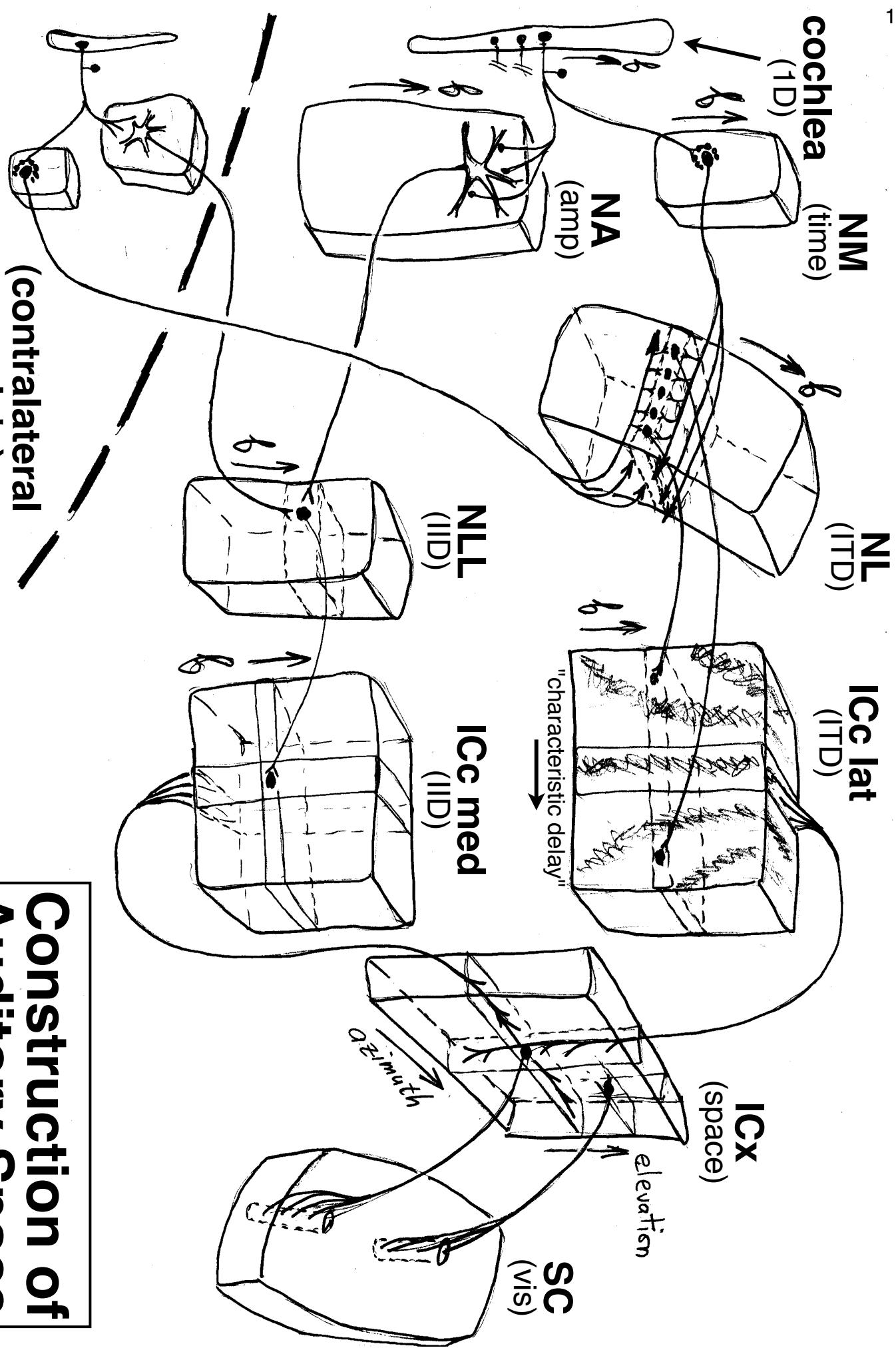
point-to-point projection

eye and head movement control



Construction of Auditory Space

(contralateral
equiv's)



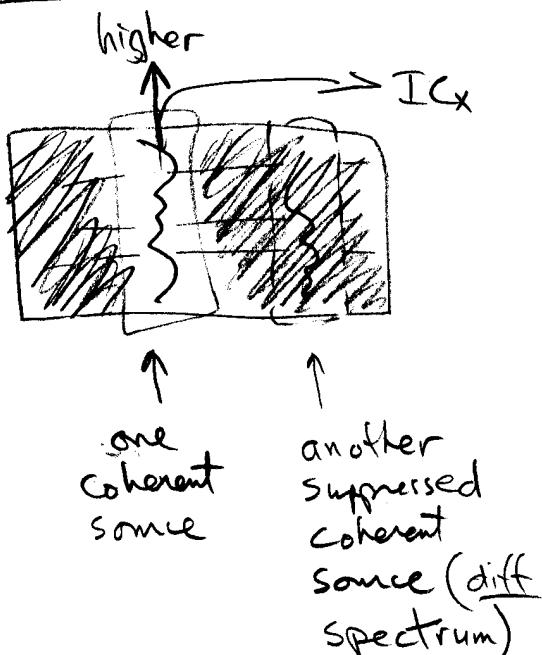
Implications

Why study?

- not all owls do elevation by asymmetric ears 
- why study, then?

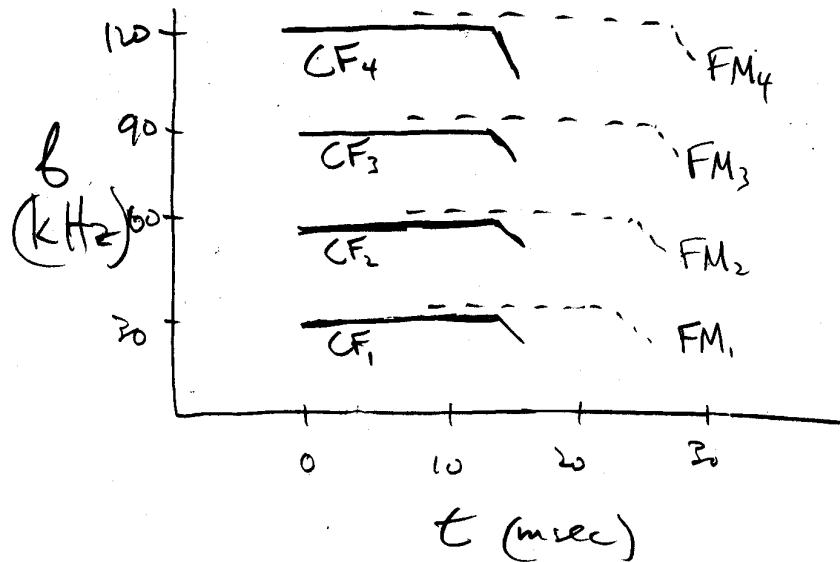
→ well-worked out example of how brain computes w/major
 → likely many other examples not yet worked out

Spatial Auditory Attention



- selectively amplify signals coming from particular spatial position
- depends on spectral differences between sources

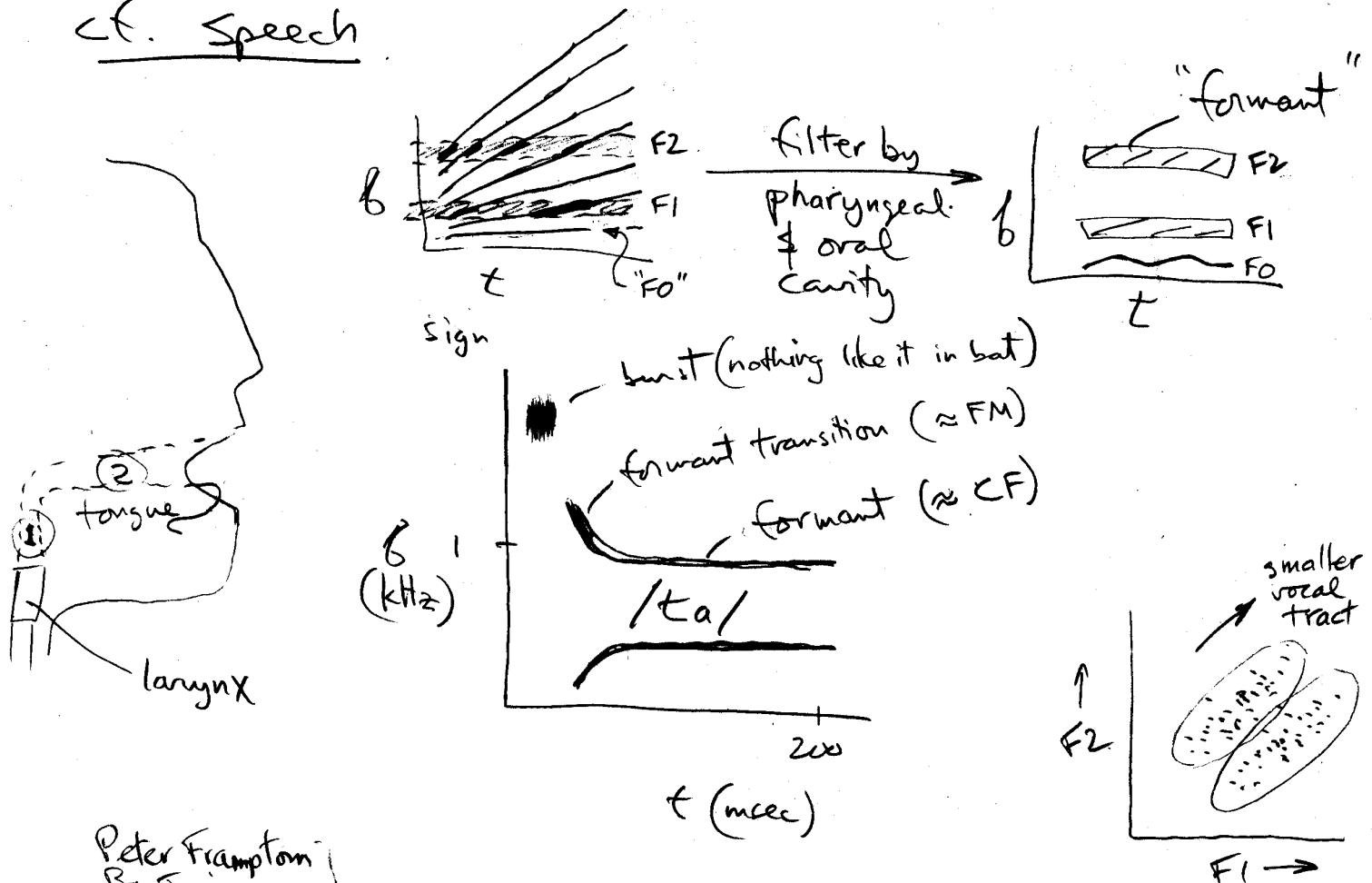
Bats (1) the signal



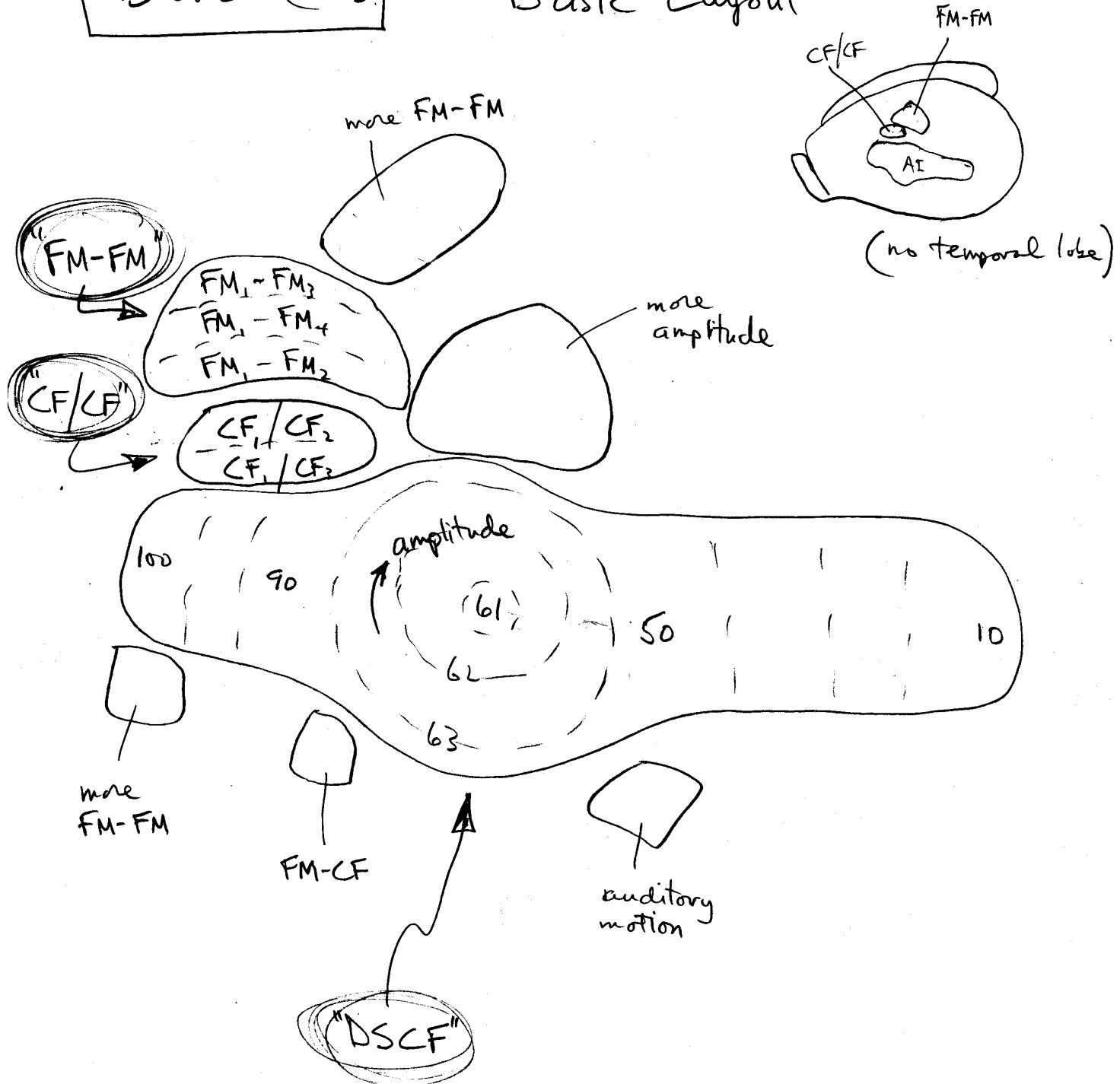
- This is approximately what the auditory nerve sees

- bat emits scream while "gritting its ears" then releases muscle on stapes to hear echo

cf. Speech



Bats (2) - Basic Layout



$\approx 15 \text{ mm}^2$
of auditory cortex
 $(\approx 1,500,000 \text{ neurons})$
 $(\approx 15,000,000,000 \text{ connections})$

Bats (3)

DSCF & Dopp. Shift Comp.

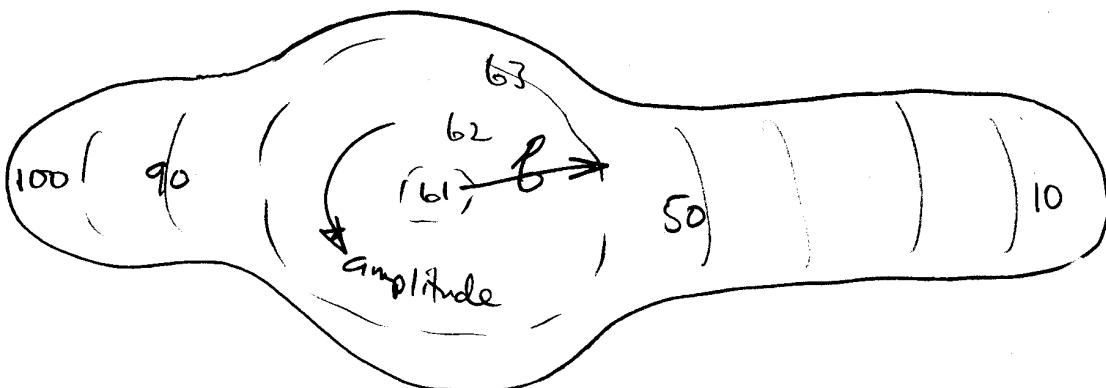
Bat behavior

- what bats do
 - catch insects
 - navigate
 - communicate
- Doppler-shift compensation behavior
 - compensation
 - increase rate at attack
- the "acoustic force"

[- starts in cochlea
 [- by cortex, amplitude is separated
 from frequency]

DSCF area

- frequency & amplitude axes at CF_2 frequency



Bats (4)

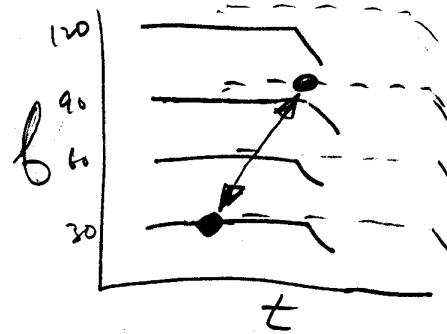
CF/CF

Doppler shift components

- bat movement
- flutter

→ first just consider
bat-target relations

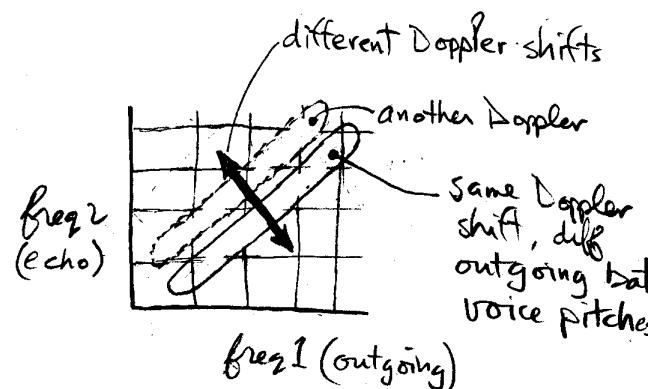
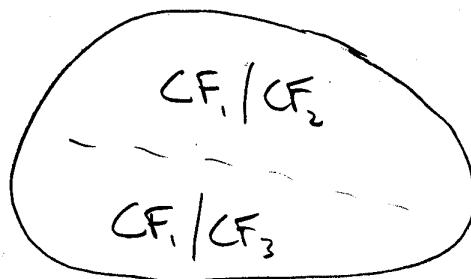
$$CF_{\text{emitted}} - CF_{\text{echo}} \sim \text{relative velocity} \quad (\text{not distance})$$



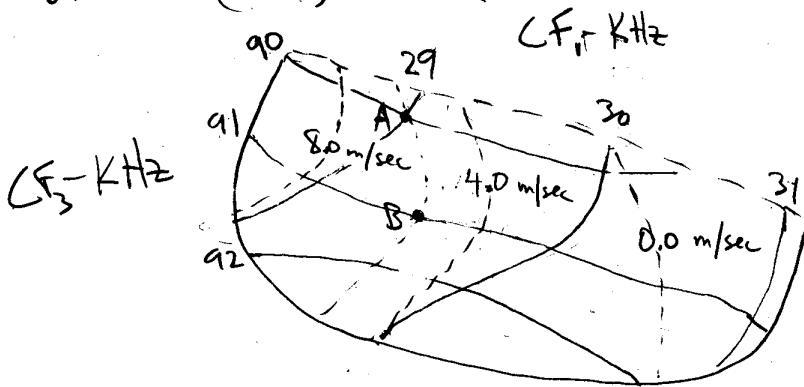
e.g. CF_1/CF_3

$$\begin{cases} y = x + 2 \\ y = x \\ y = x - 2 \end{cases}$$

CF/CF area



A: $90 - (29 \times 3) = 3 \text{ kHz}$
B: $91 - (29.6 \times 3) = 3 \text{ kHz}$



- units are relatively level-insensitive
- respond only to pairs of CF's (AND gate)
- pairs of CF's correspond to particular target velocities (shift = $CF_3 - 3 \cdot (CF_1)$)
- cannot detect distance (delay \cong sensitive)

Bats (5) FM - FM

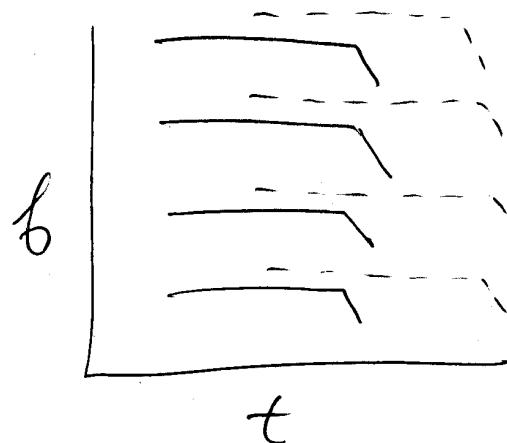
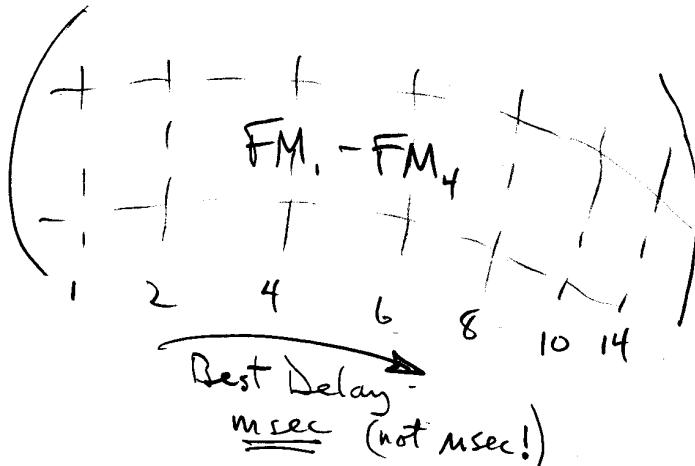
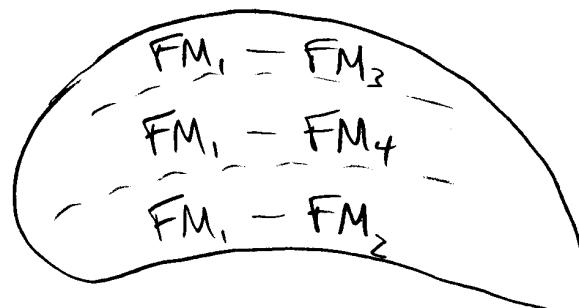
FM info

- good because delays are represented

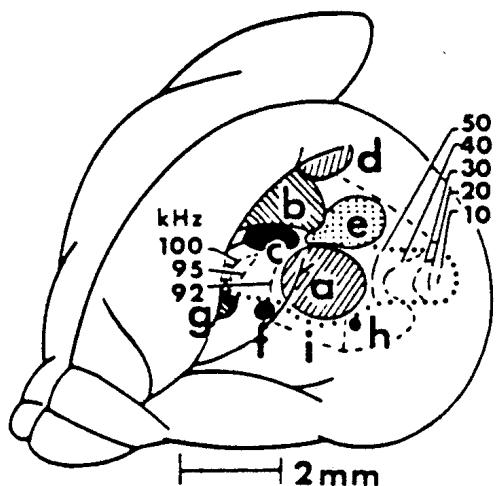
(contrast CF)

- by comparing different harmonics, f overlap reduced
- monaural delay detectors (contrast and)

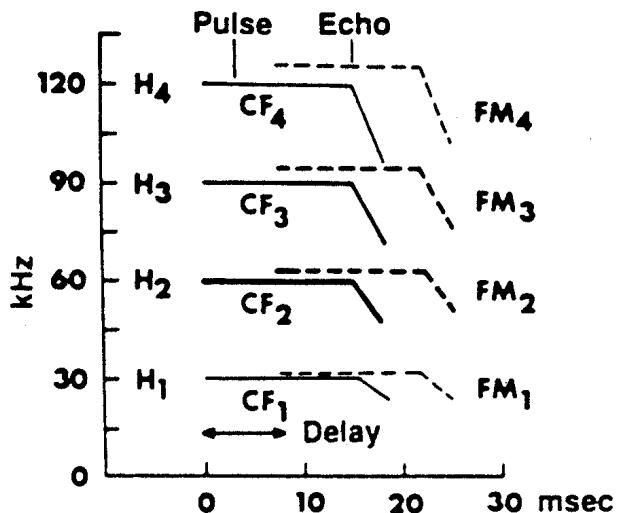
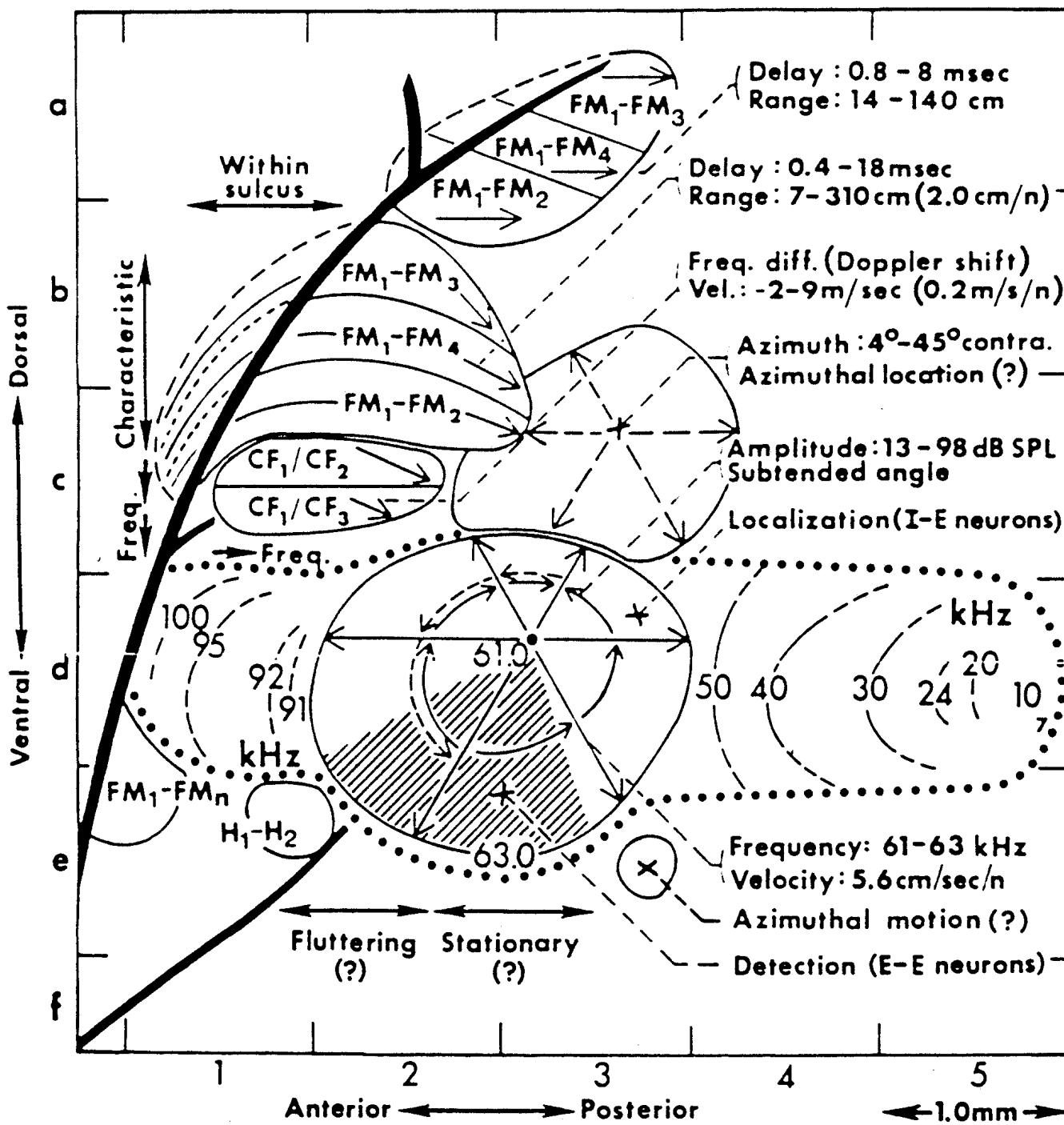
FM - FM area



	no response
	no
	yes
	no
	yes
	no
	yes
	no
	no
	no

A

- a: DSCF
- b: FM-FM
- c: CF/CF
- d: DF
- e: DM
- f: VA
- g: VF
- h: VP
- i: VL

**B**

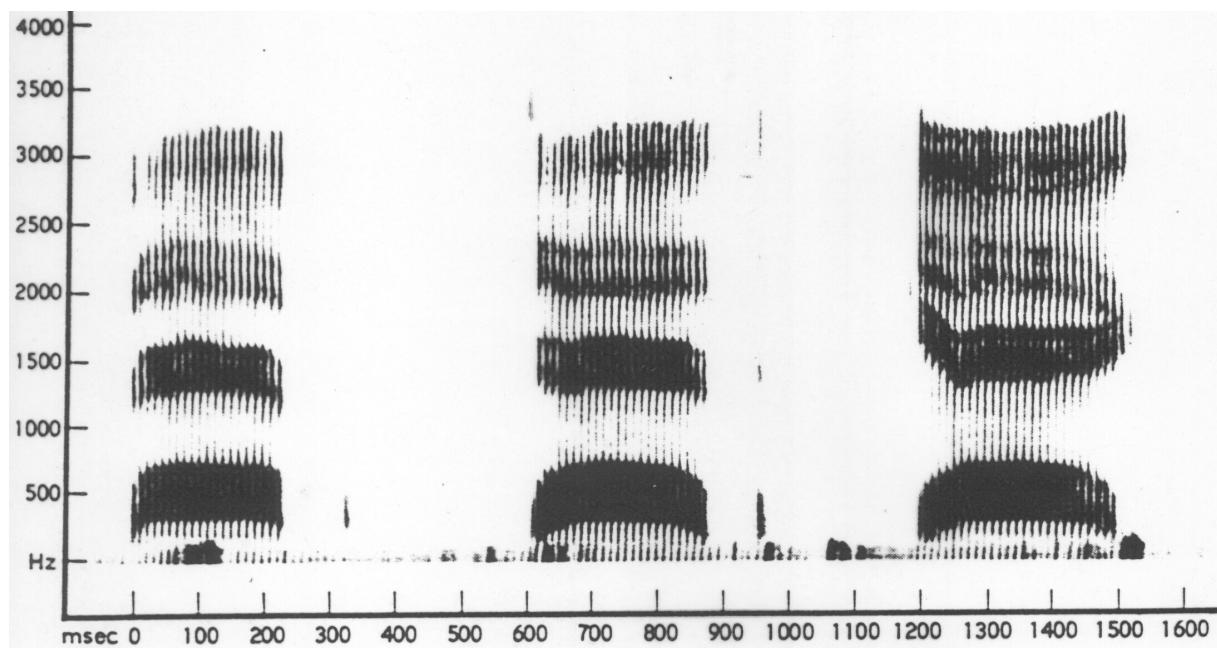


Figure 8.9 A spectrogram of the words "bab, dad, gag" (British accent).

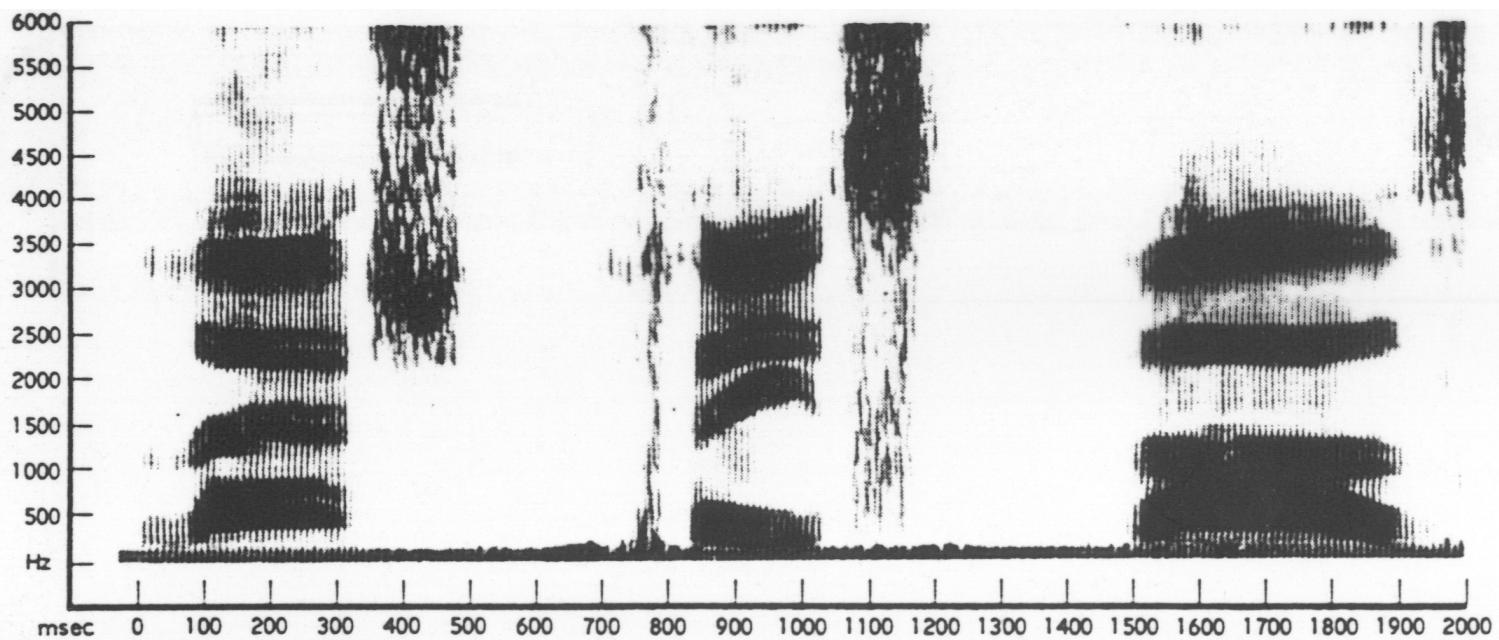
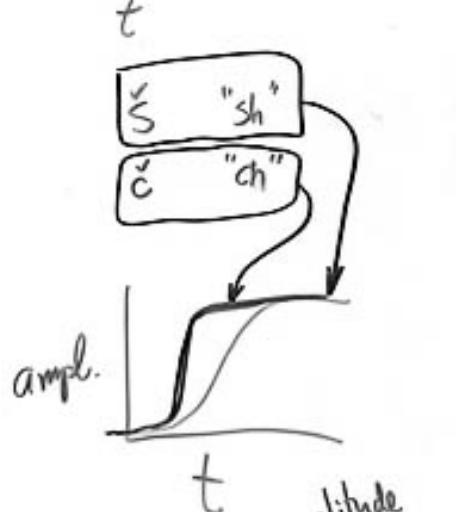
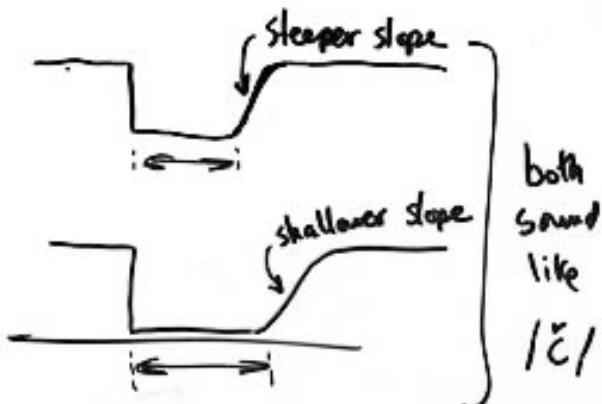


Figure 8.11 A spectrogram of "lash, face, vase" (British accent).

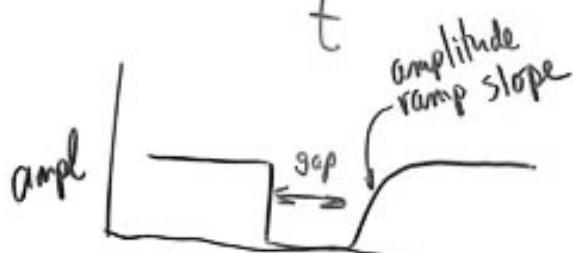
f "trading relations"

- Aud ctr
- complex Snd

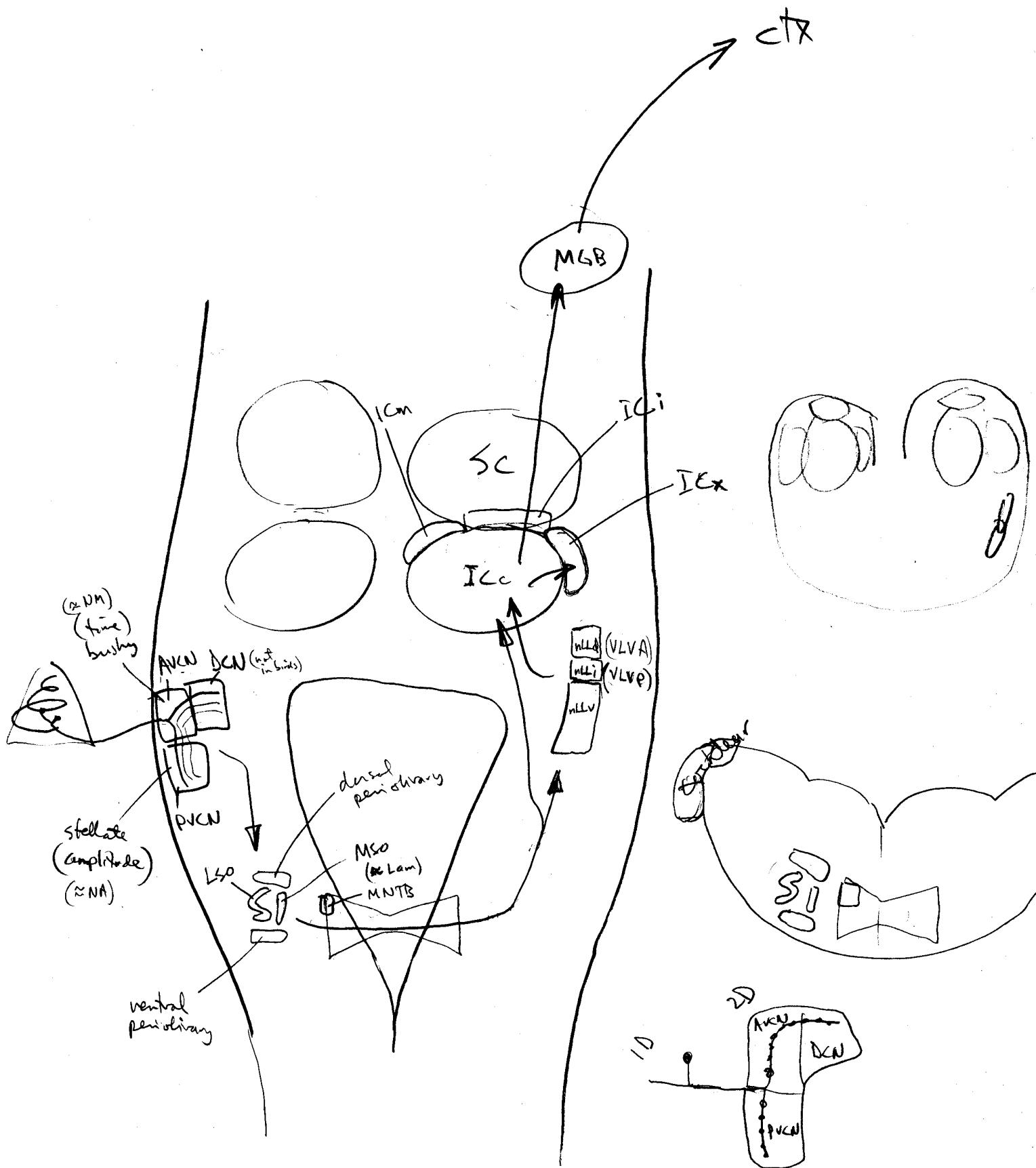


N. VIII

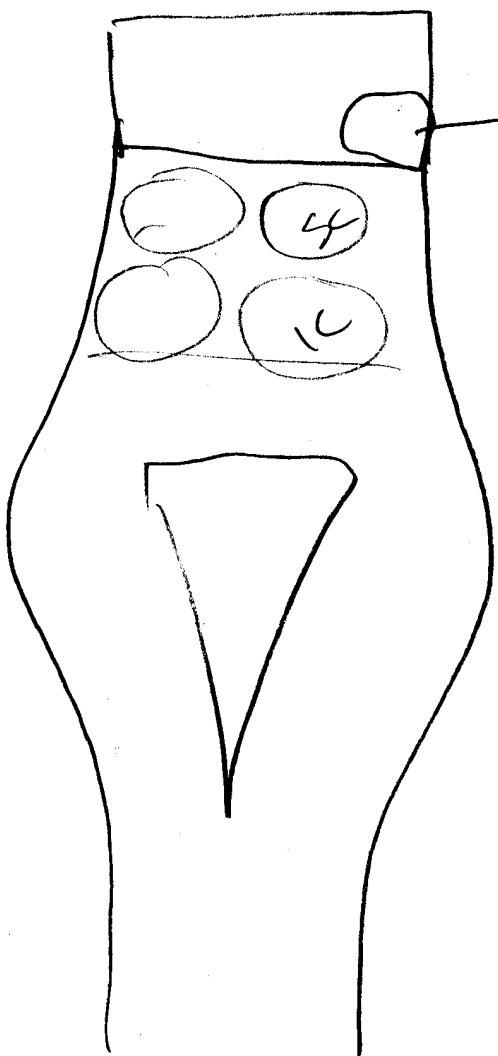
anesthetized cat
larger gap
↳ dishabituation



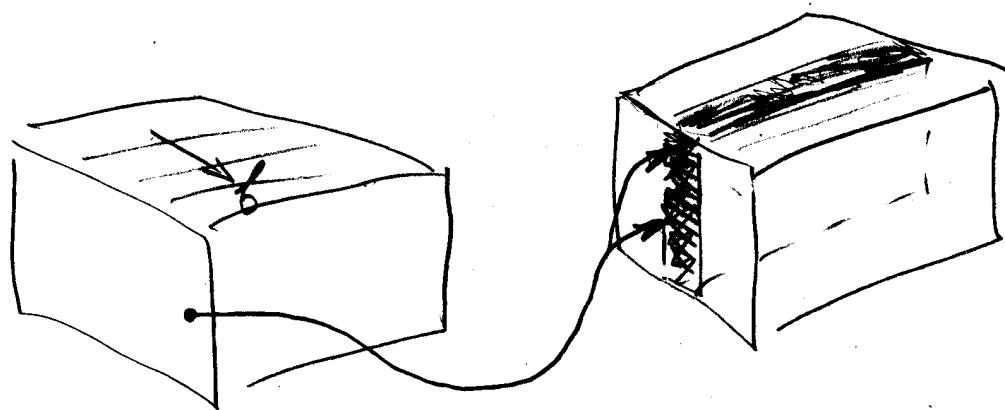
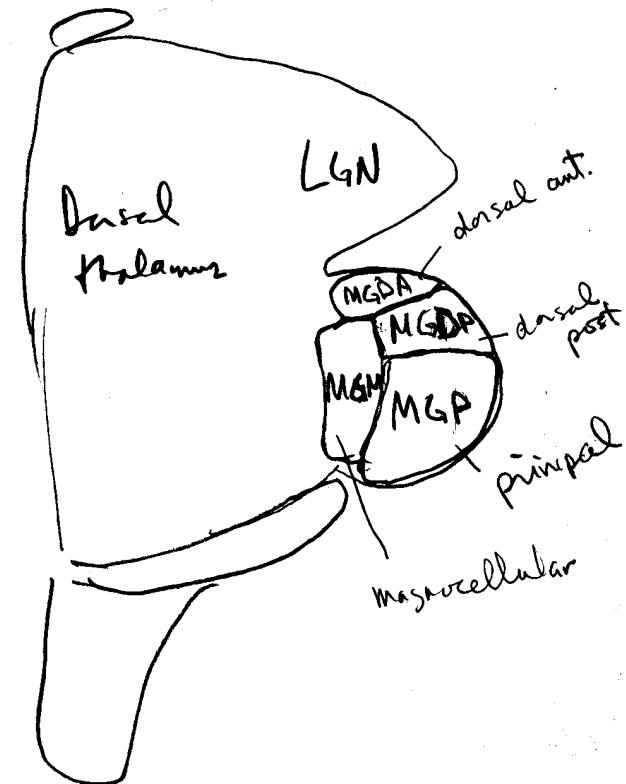
Auditory Brainstem & CtX in Mammals



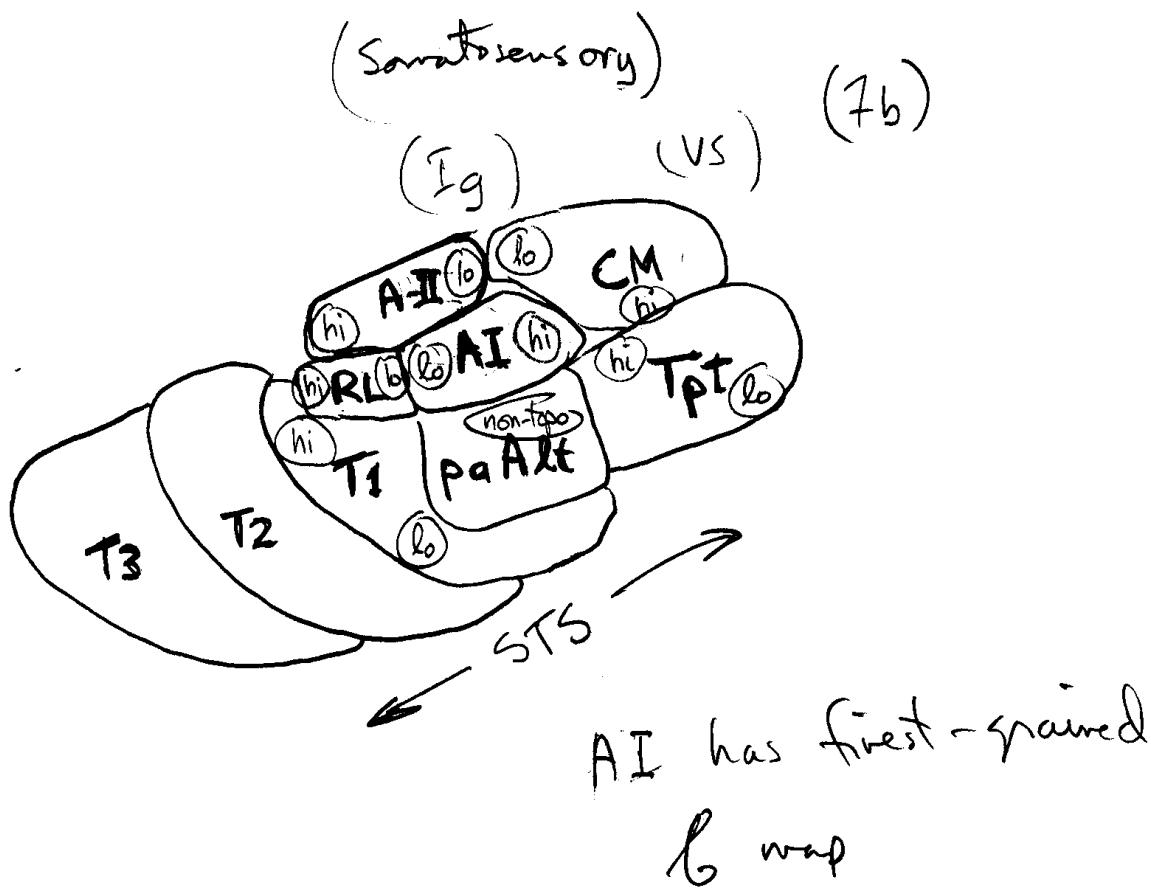
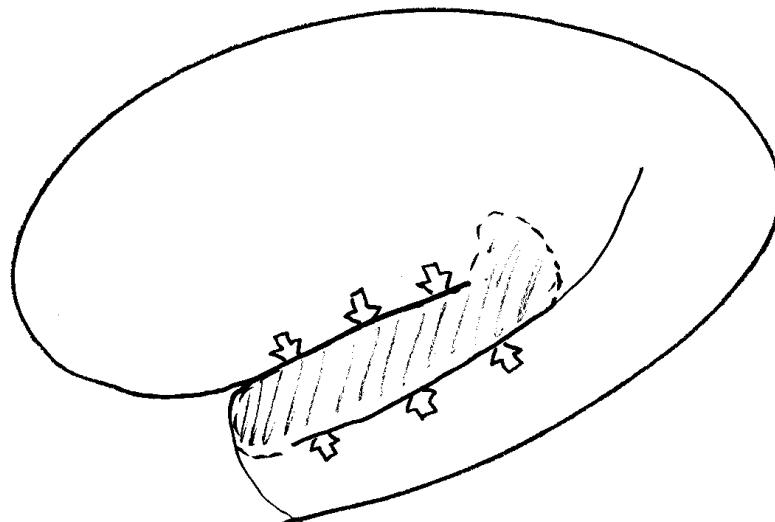
Auditory Thalamus



medial
geniculate \Rightarrow posterior &
lateral in
thalamus



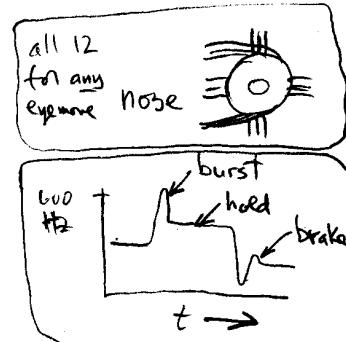
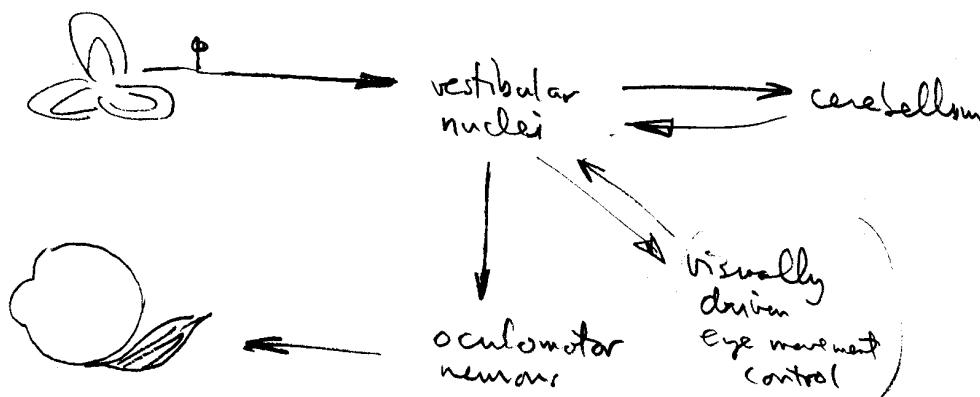
Cortical Areas — Auditory



Eye Movements Anatomy

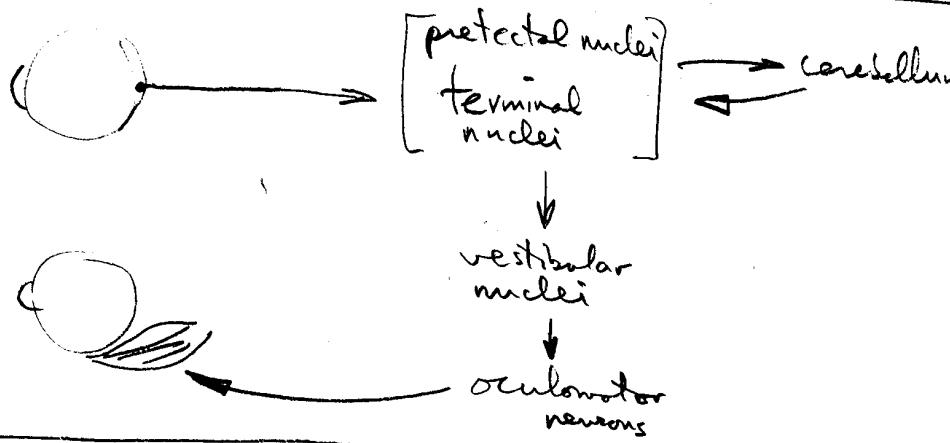
1) Vestibulo ocular reflex

compensating for head movements



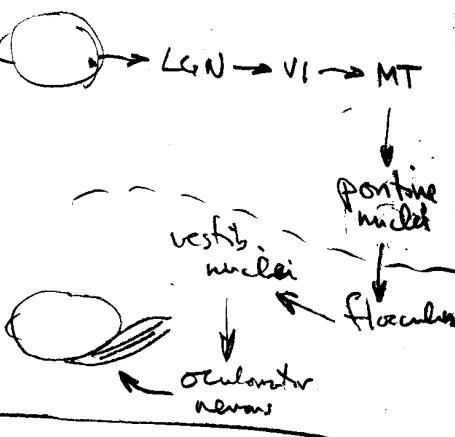
2) Optokinetic Nystagmus

Compensating for visual field movements



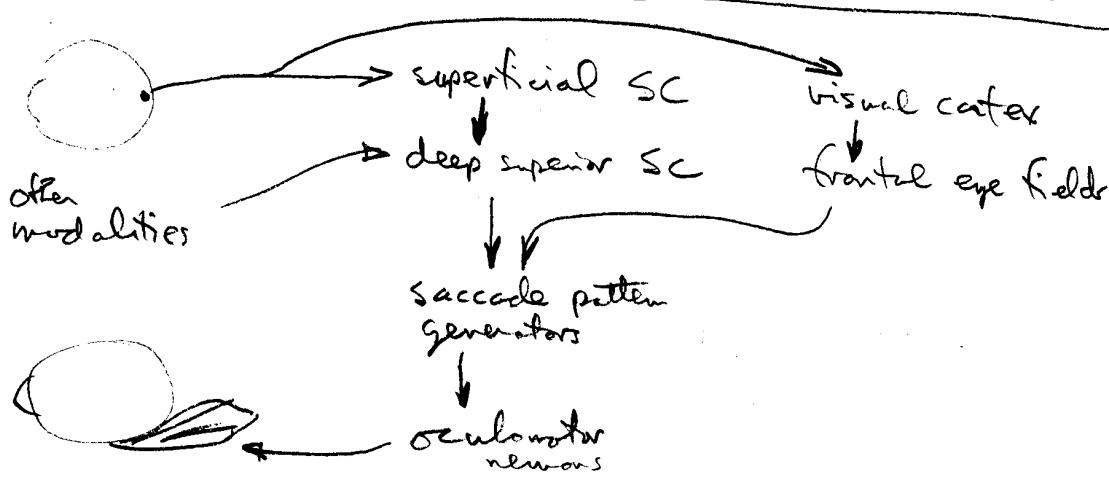
2a) Pursuit

Following a small object

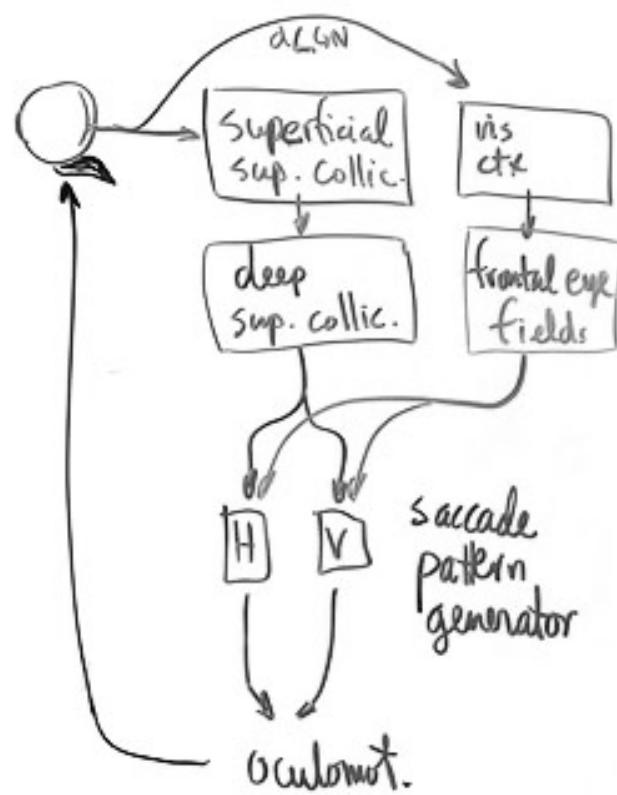


3) Orienting Eye Movements

Looking at new things

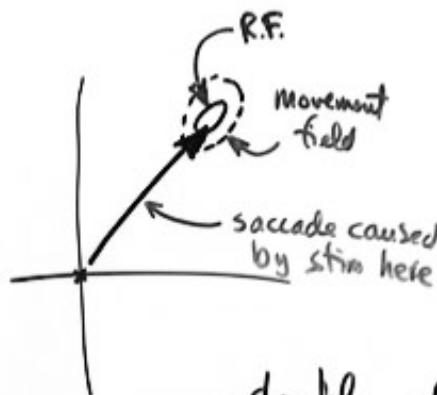
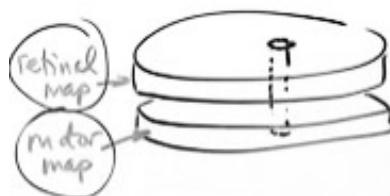


Orienting Eye Movements



Sup. collic.

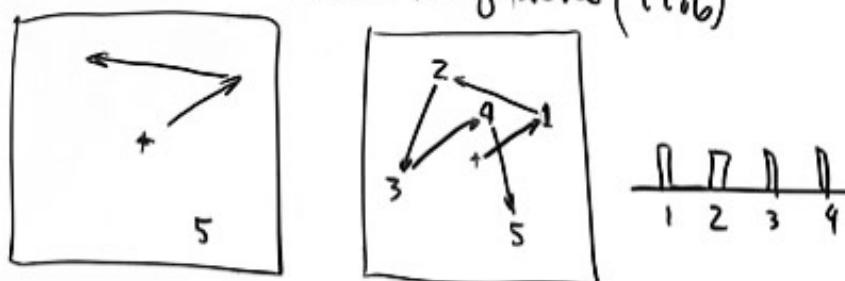
50 cell types
many layers



- double-step saccade

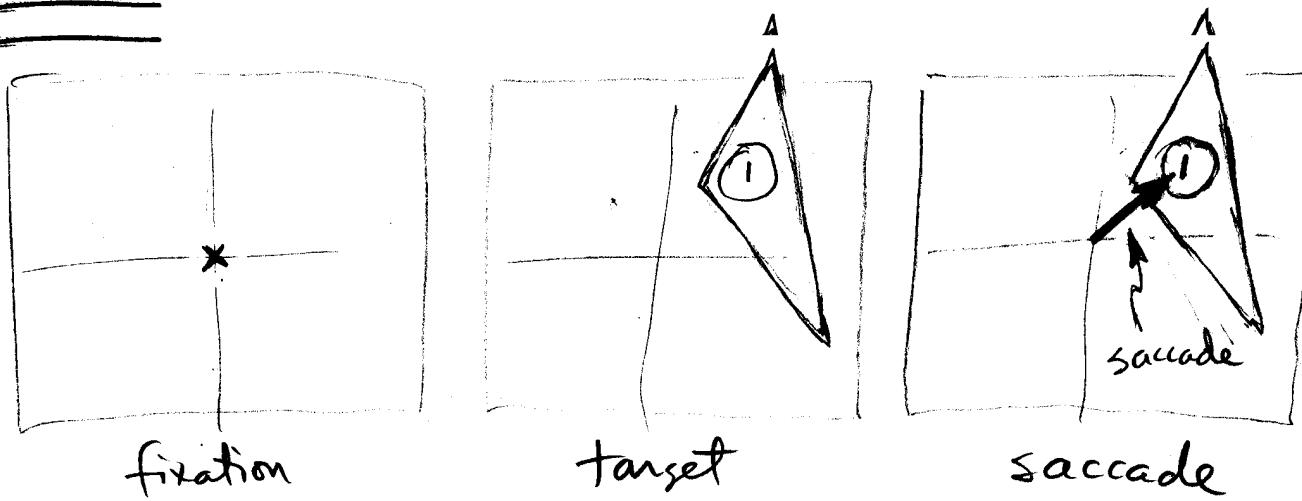
- 1) record receptive field
- 2) record during eye move
- 3) stim

Hallett & Lighstone (1976)

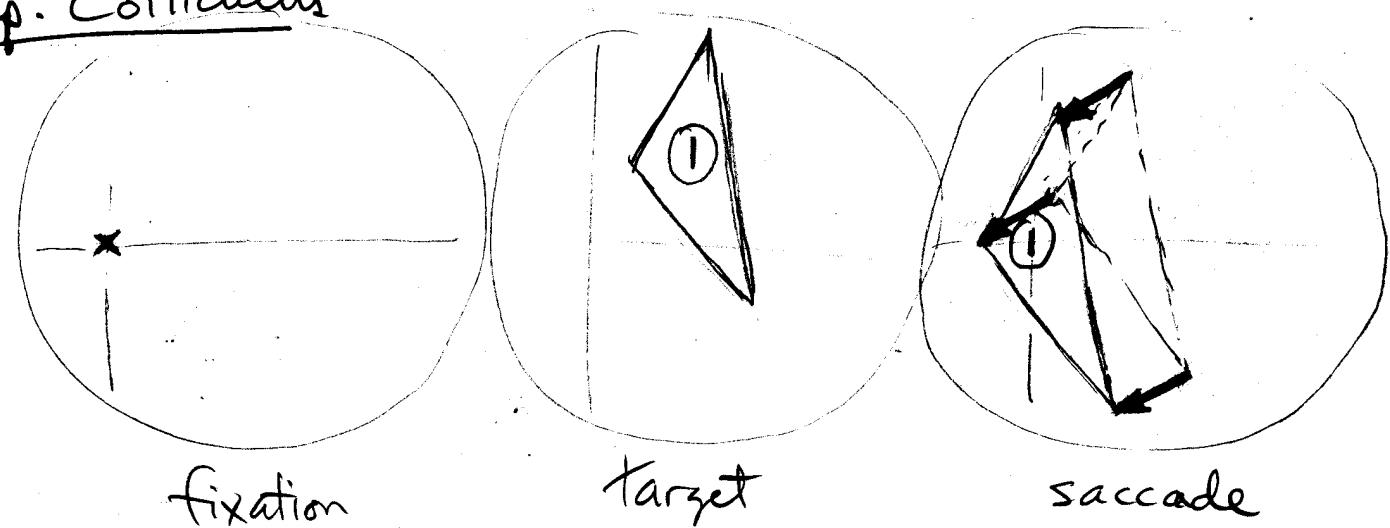


How saccades update stationary targets

World

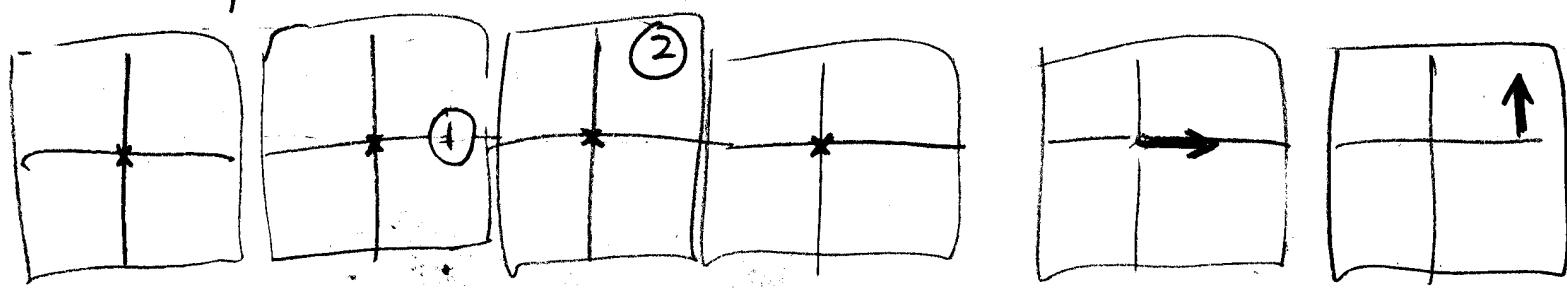


Sup. Colliculus

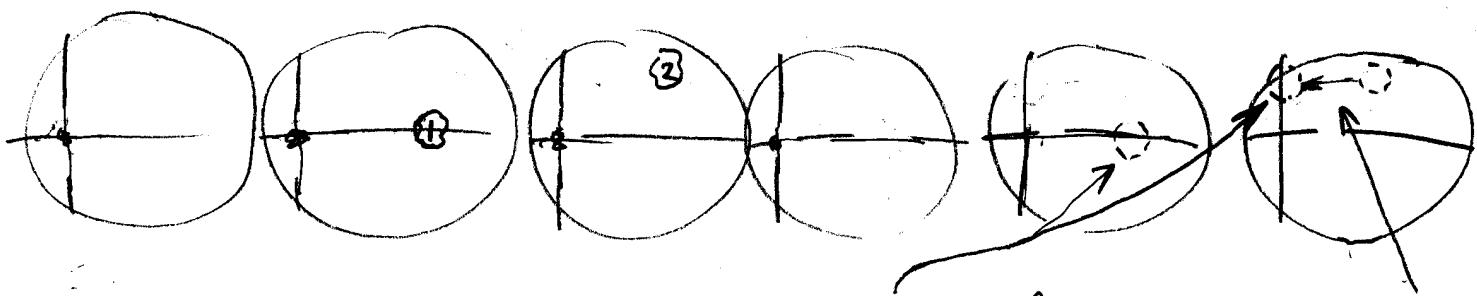


Double-Step Saccade Remapping

World / Behavior

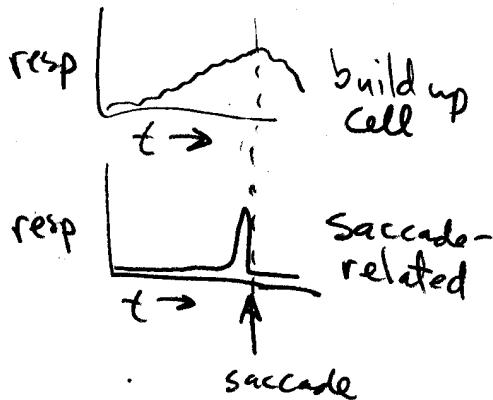


Sup Collic for clarity; reversed (actually right hemifield is represented in left SC)



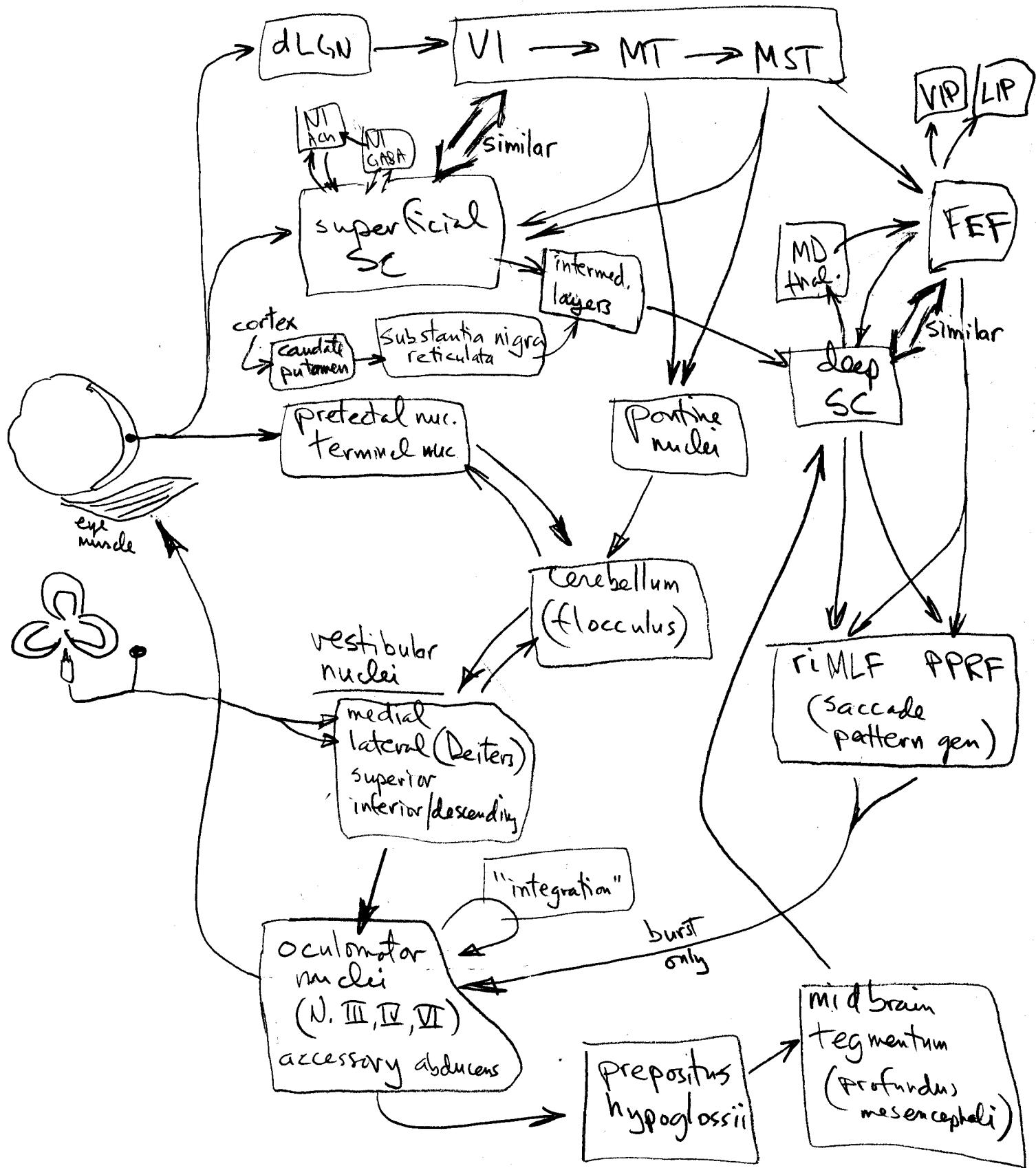
quasi-visual
intermediate, buildup

then deep immed.
preceding saccade



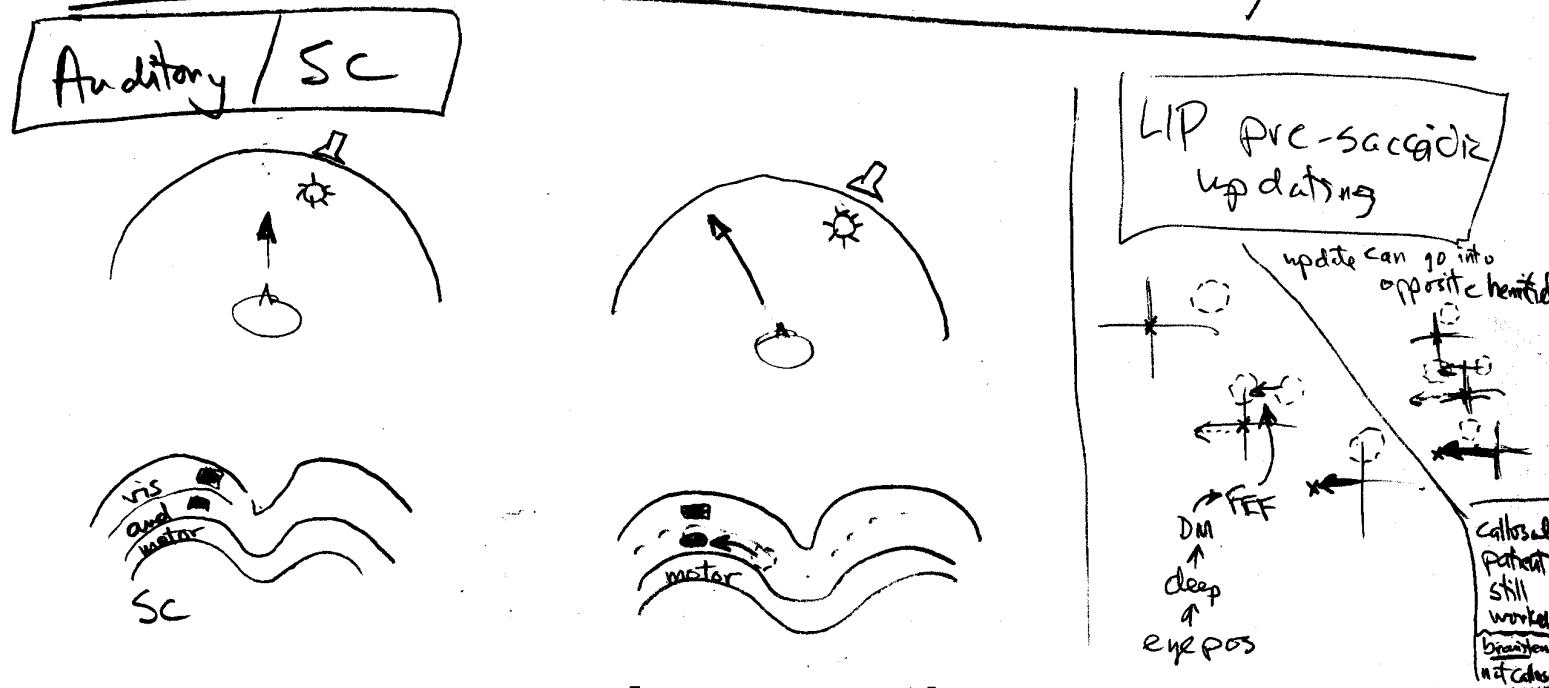
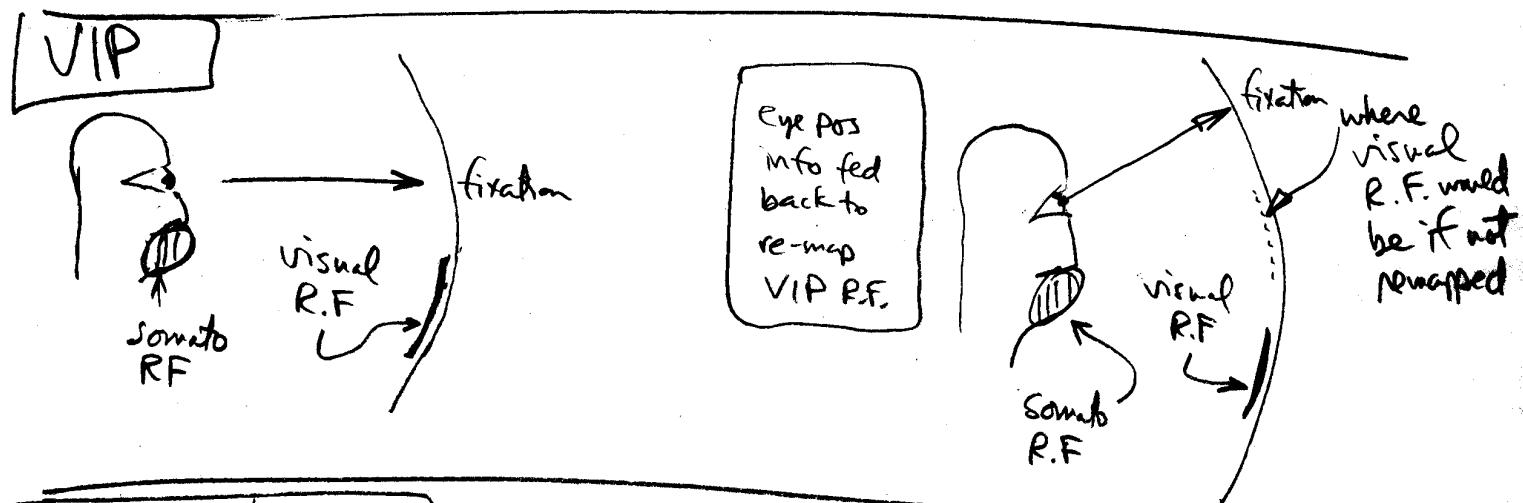
this
translation
performed
by eye pos
feedback
effeference
copy

All-In-One (do after in div.)



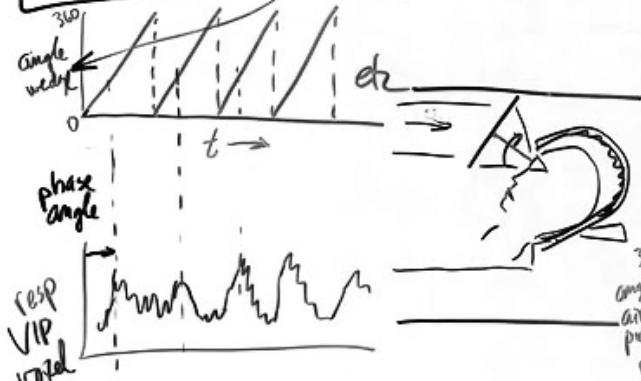
Coordinate Systems for Multi-Sensory Fusion

- SC codes target position by position, not amount of firing
- saccade fall. gen. code position by amount of firing horiz ← vertical
- SC has abstract notion of target — potential target whether or not it's actually there
↳ Saliency map

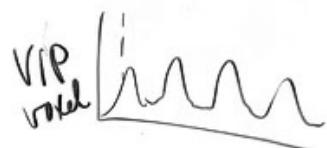
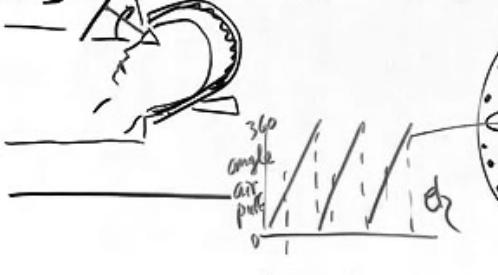


- human VIP

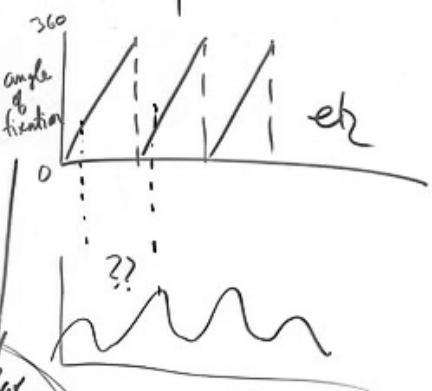
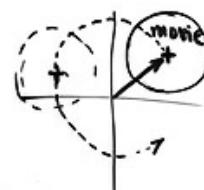
Sereno & Huang (2006)



2) somatotopic mapping

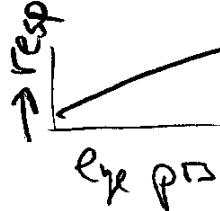
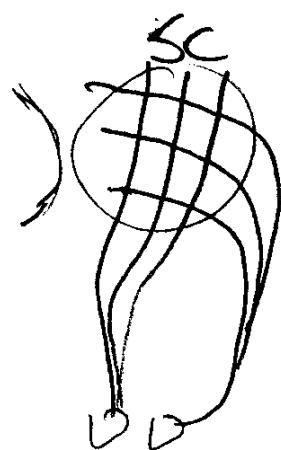


3) face-centered visual mapping

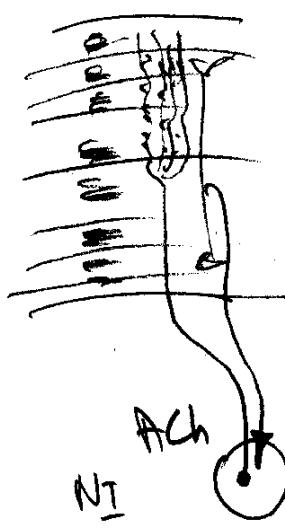


Polar angle vis obj w.r.t. face

"temporo-spatial transformation"

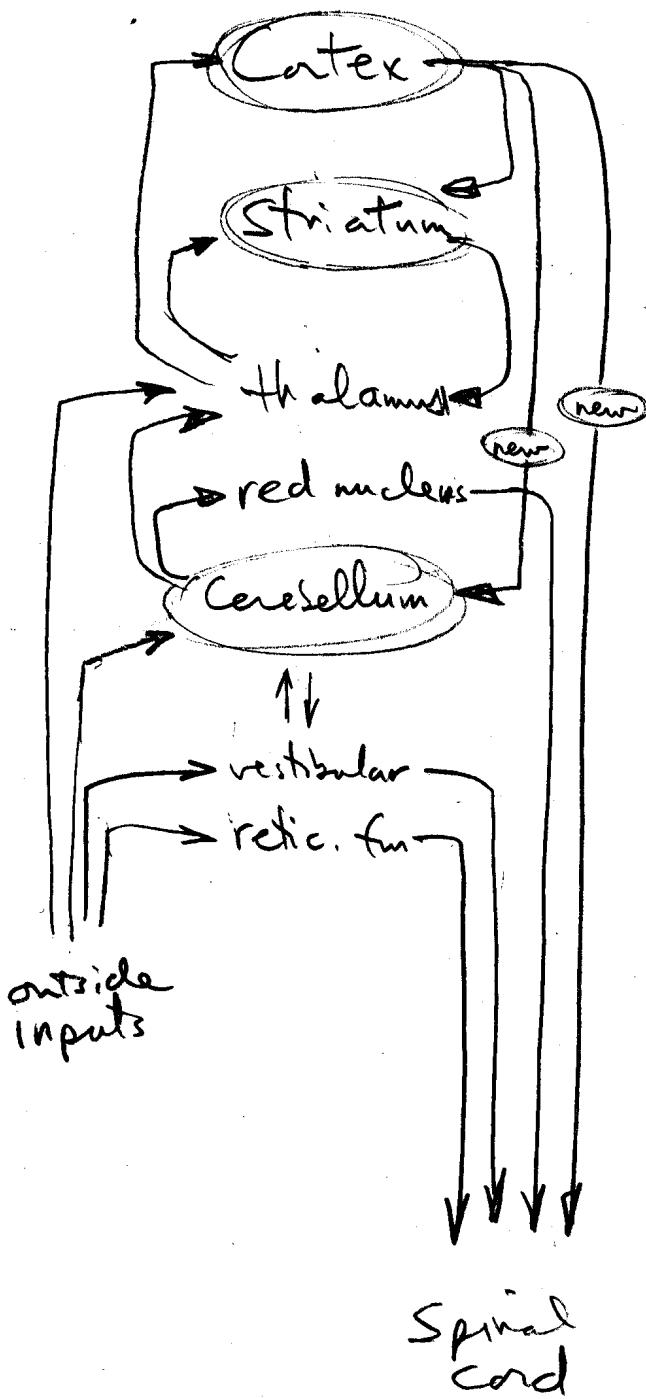


isthmus
&
target choice



ACh
NT
Paradigm

Motor System Basic Plan



excludes SC

↳ visually
directed
orienting
movements

eyes
head
ears
whiskers
shoulders

Motor System "Principles"

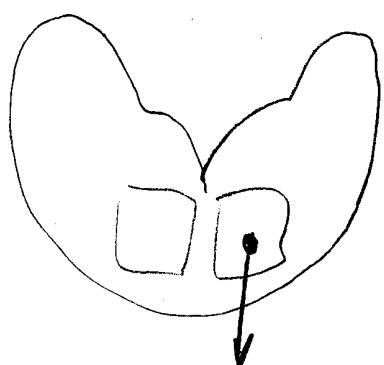
- 1) Short number of synapses from input to output (retic. sp., vestib. sp., rubro sp.)
- 2) Pattern generators — autonomous lower level sensorimotor transformers (spinal cord)
- 3) Additional systems added as input to pattern generators (cereb. striatum)
- 4) Higher level pattern generators may have independent access to motor neurons (corticospinal, corticofugal, speech)
- 5) many "lateral" connections to mediate conflicts between different info sources (suppress VOR during pursuit)

Descending Paths

Reticulospinal

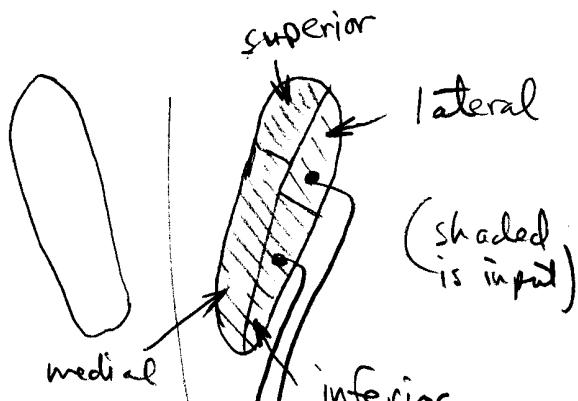
anti-gravity
"extensors"

Pontine & Medullary Retic. fun. — uncrossed
midbrain Retic. fun. — crossed



Vestibulospinal

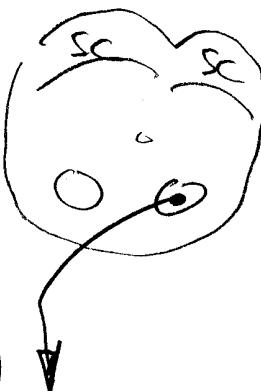
- vestibular nuclei (in pairs)
- all receive semicircular or otolith
- main descending outputs from lateral & inferior



Rubrospinal

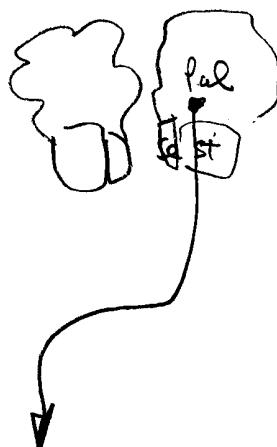
active movement
"flexors"

- red nucleus in mi
- crossed

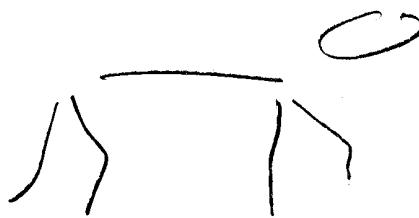


Spinal cord

Corticospinal (only mammals)



Pattern Generators



- spinal cord generates rhythmic, coordinated muscle movement w/o cortex (requires sens. stim)
- cortical control has been superimposed on existing I/O systems
- in primates, raccoons, cortical control bypasses spinal pattern generators
- same thing in birds for song and humans (but not chimps!) for speech

Cats

monkeys, raccoons

Cortex

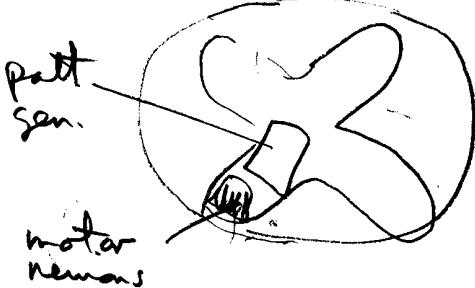
pattern gen

motor neurons

Cortex (new pattern gen.)

pattern gen

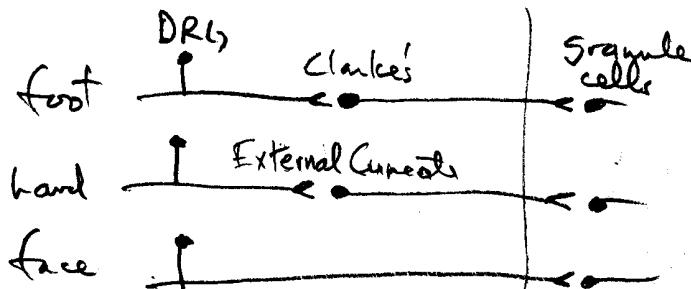
motor neurons



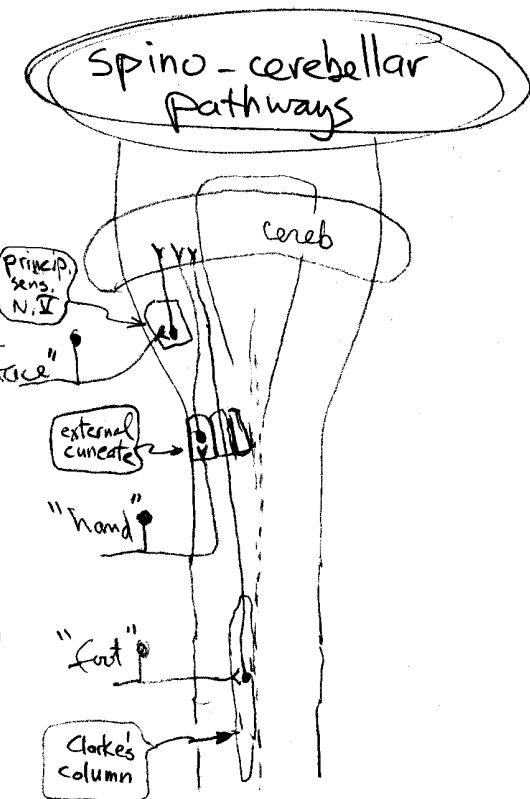
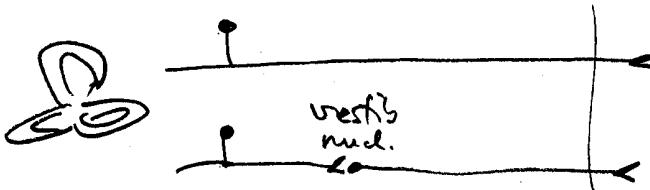
Cerebellum inputs

mossy fibers

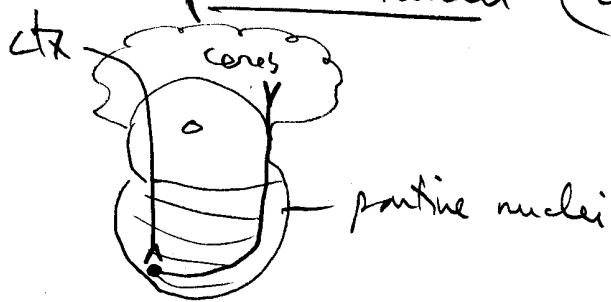
Spinal cord (somatosensory input)



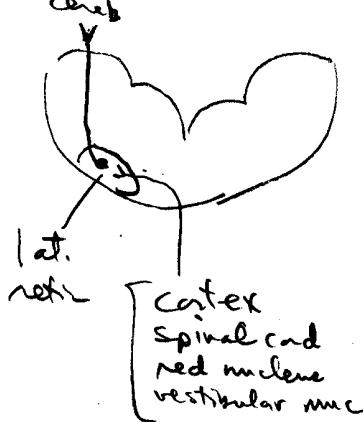
Vestibular system (vestib. input)



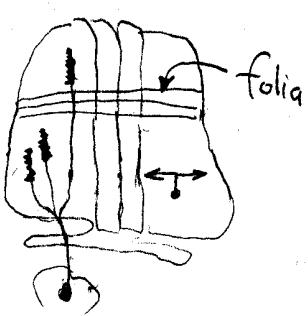
Pontine nuclei (cortical input)



Lateral reticular (mixed)



climbing fibers



inferior olive (spinal, vis. and aud. input)

- main
- medial accessory
- dorsal accessory

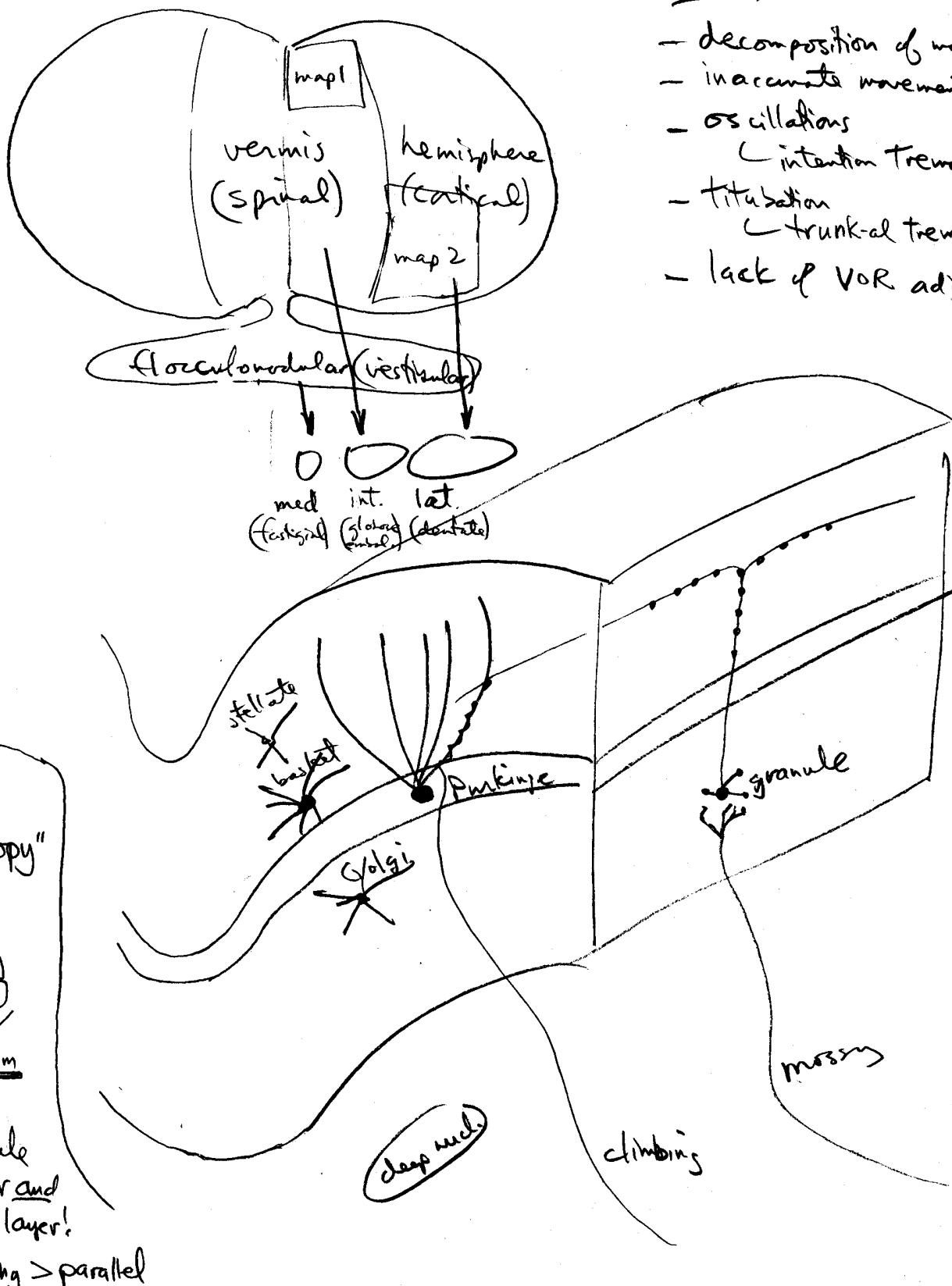
Sensory inputs → IO → climbing fibers

Cerebellum

- Basic Layout

deficits

- decomposition of movement
- inaccurate movements
- oscillations
 - ↳ intention tremor
- titubation
 - ↳ trunk-al tremor!
- lack of VOR adjustment

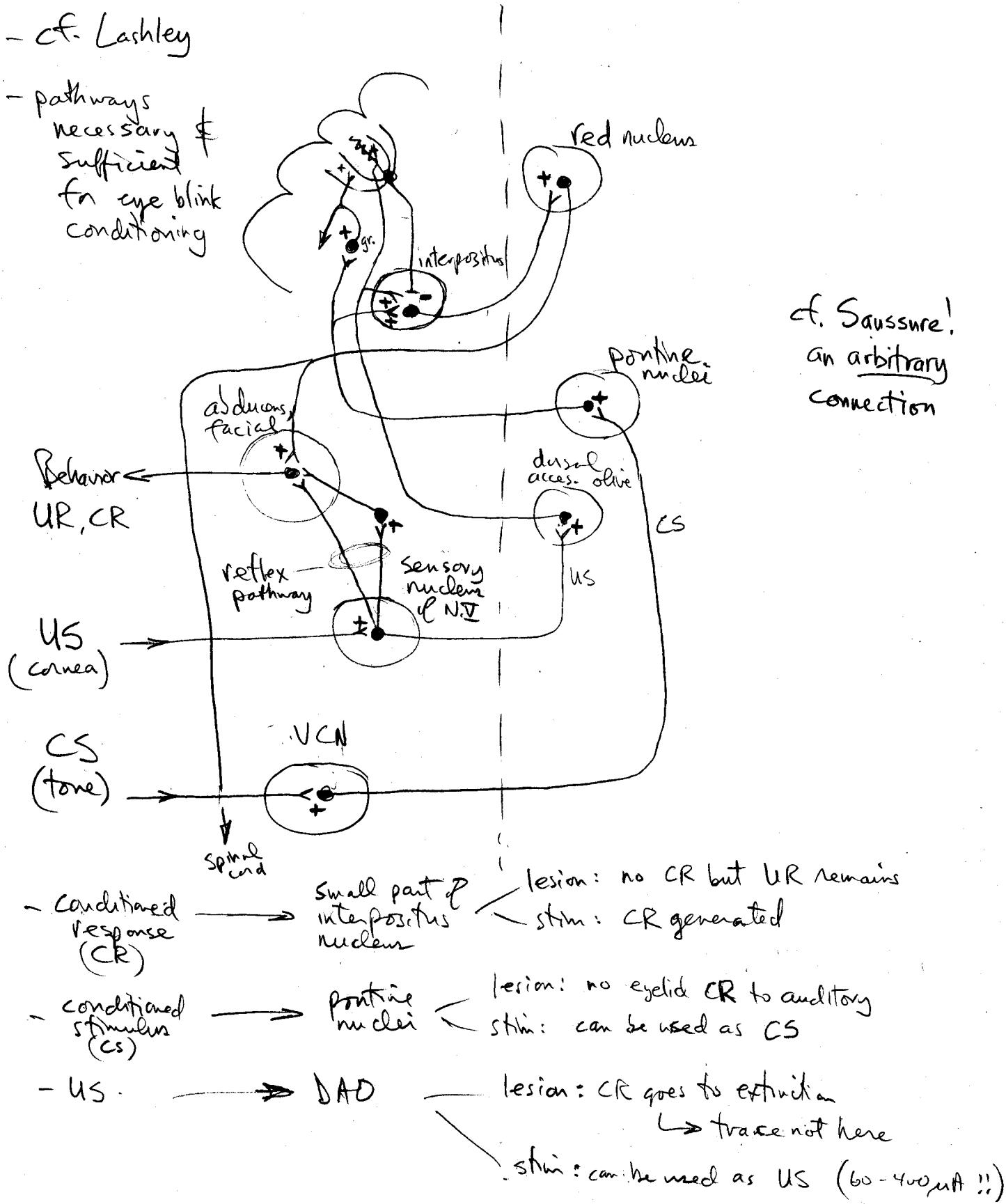


CEREBELLUM & CONDITIONING

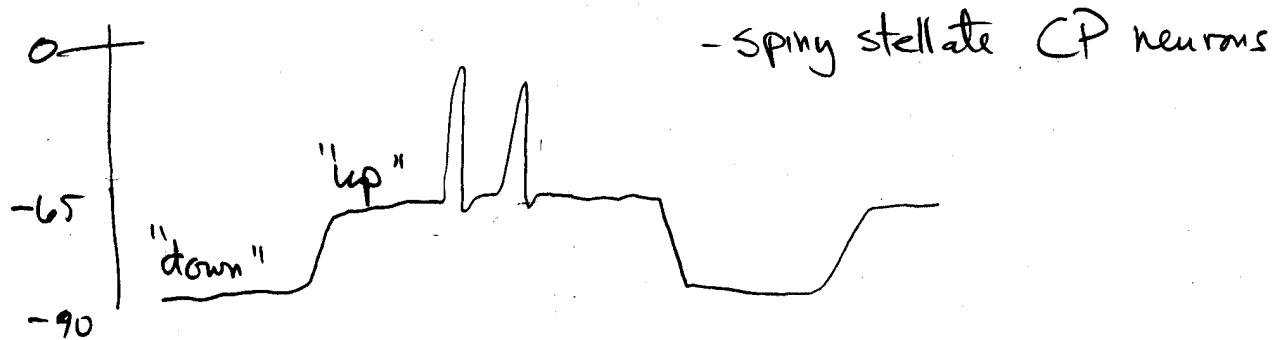
Thompson

- cf. Lashley

- pathways
necessary &
sufficient
for eye blink
conditioning



Up & Down State



- "anomalous" rectifier

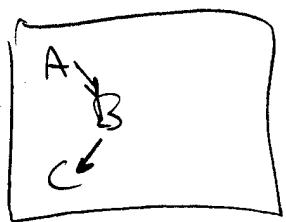
- K^+ current activated by hyperpolarization

- cf. fast Na^+ spike current

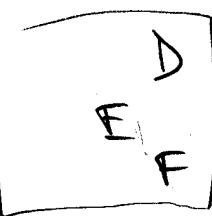
- perhaps a mechanism for switching between motor programs

Basal Ganglia deficits

easy for Parkinson's



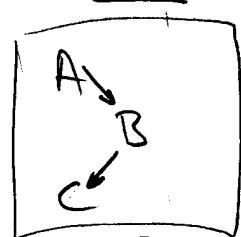
Prog #1



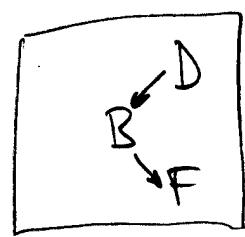
Prog #2

v/s.

hard for Parkinson's



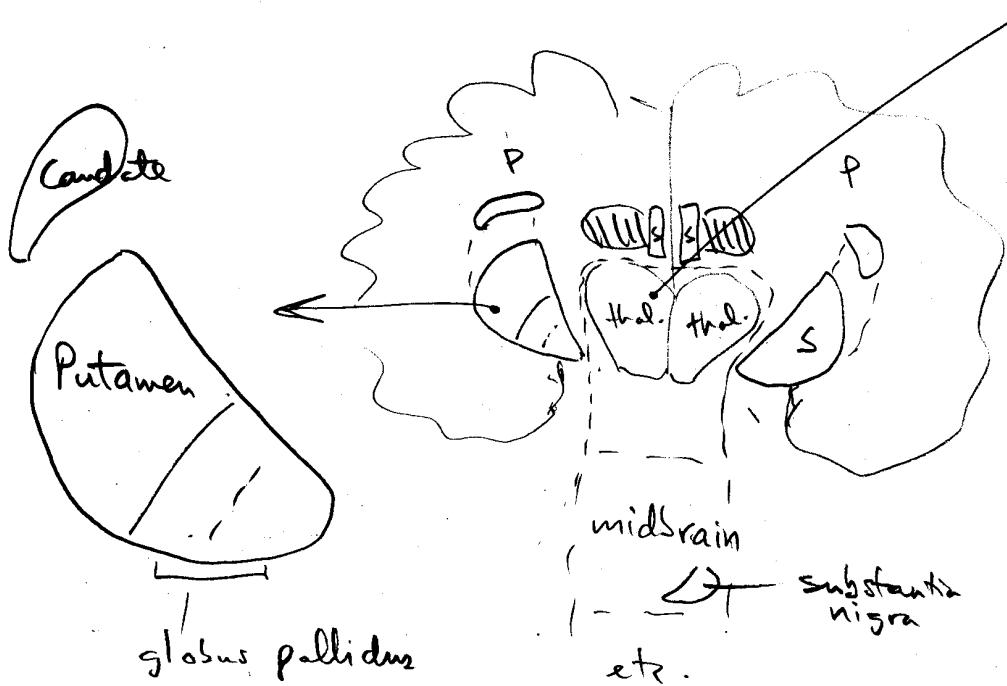
Prog #1



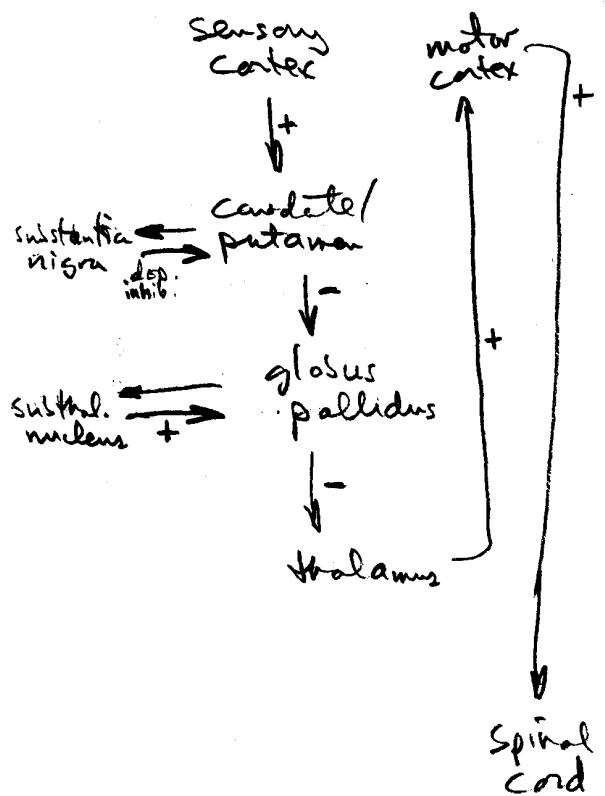
Prog #2

Striatum (= Basal Ganglia)

↳ also used confusingly

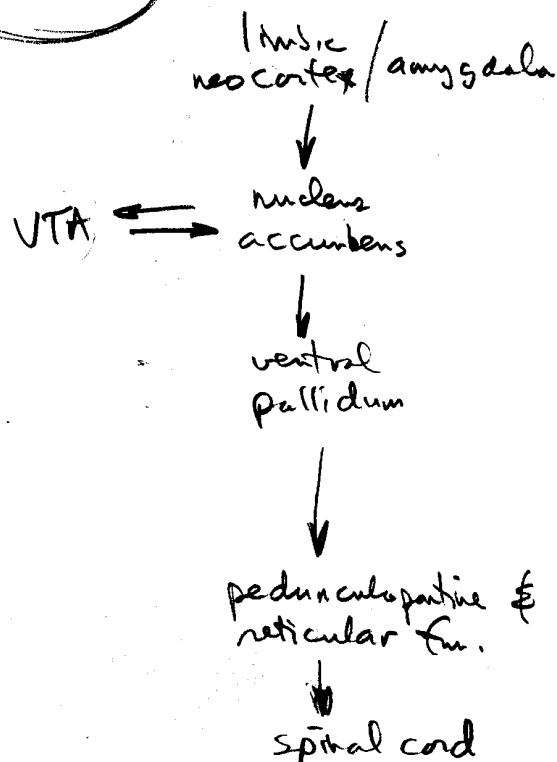


Sensory Striatum



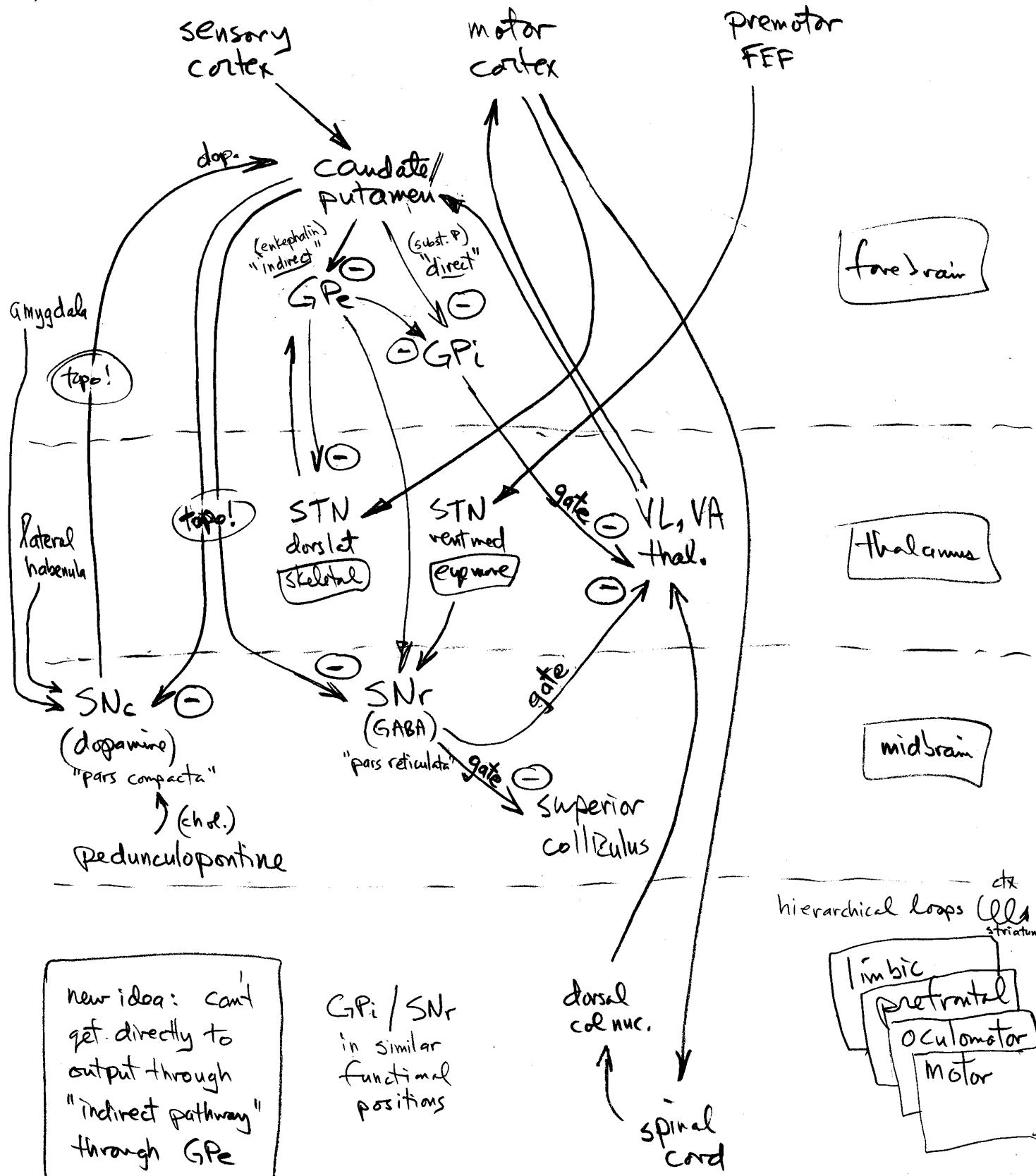
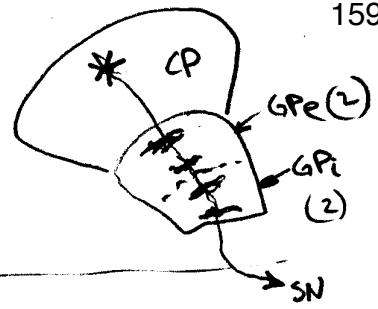
very coarse overview

Limbic Striatum

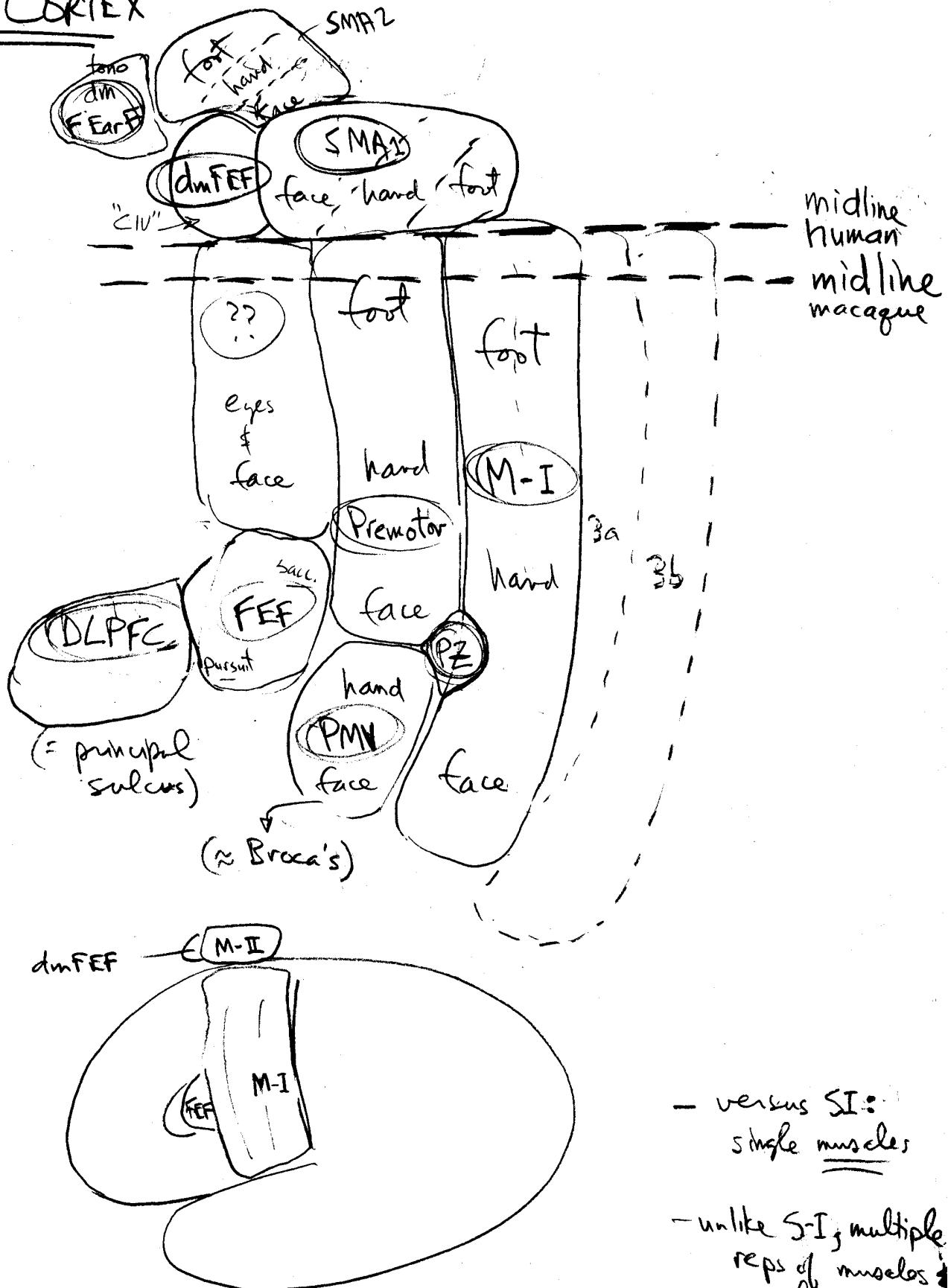


New Striatum

(New A. Parent)



MOTOR CORTEX

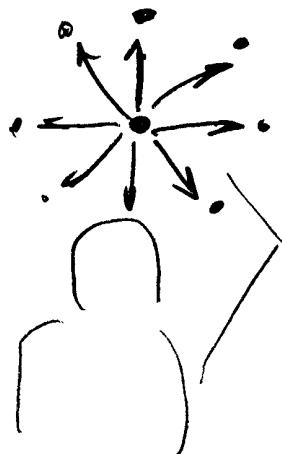


- versus SI:
single muscles

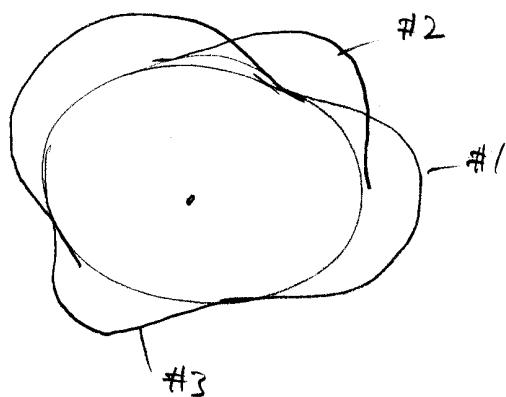
- unlike S-I, multiple
reps of muscles/
muscle groups
(not cut & peel)

- M-I & M-II have
own FEFs

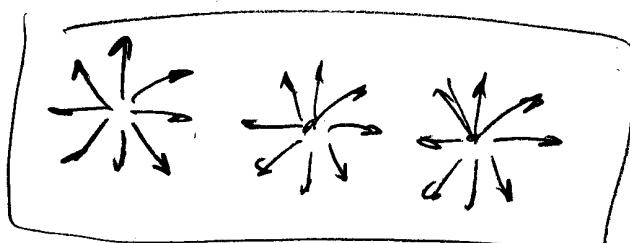
Georgopoulos (1) - bad



- M-I
- population vector

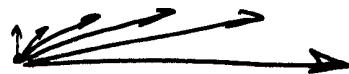
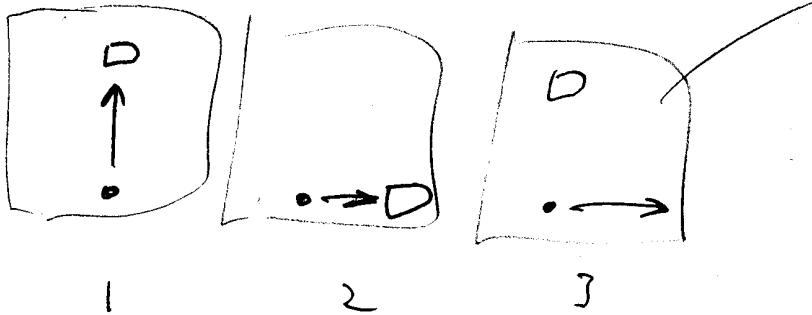


Camihihi



→ *differences suggest that neurons are probably coding muscle pulling direction, not direction in space

Georgopoulos (2) - good

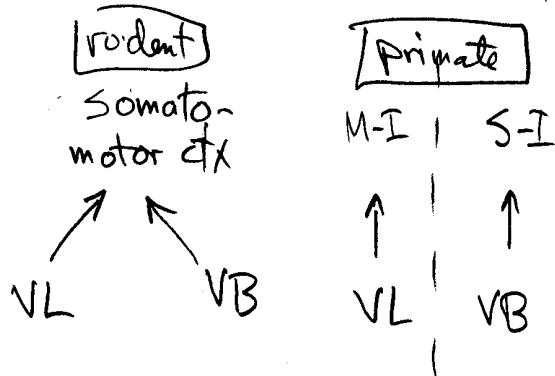


- like mental rotation
- think of as higher level programming leaking through into primary mot.
- rotation vs. ample: ↗ ↙

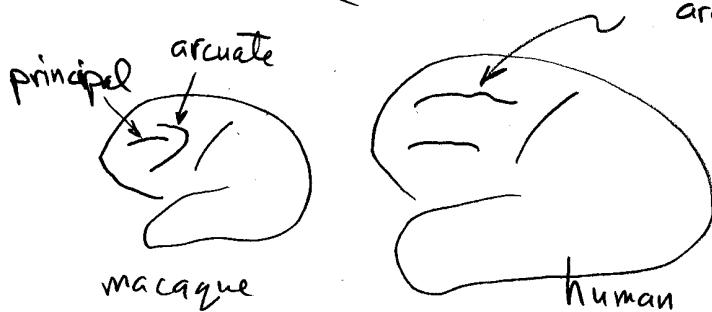
Motor Ctx (parietal) Stim

- muscles vs. movements
- = EMG of 'everything' is hard
- Graziano
 - [long stim trains
Position in extra personal space
cf. frog spinal cord!]
- Stepniewska/Kaas

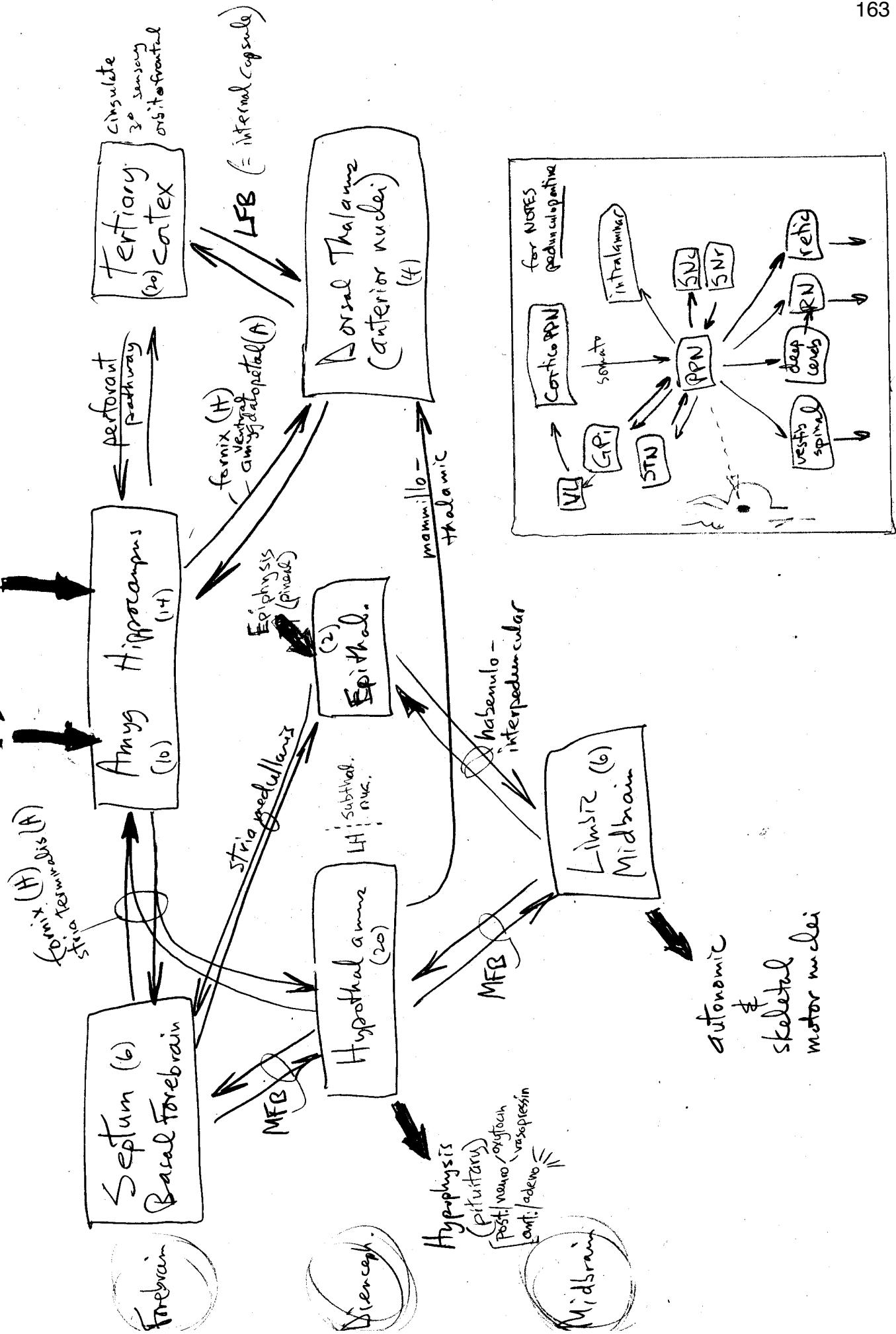
- rodent vs. primate
 - ↳ major diff thalamic afferents



- sulci (human: principal + top arcuate)

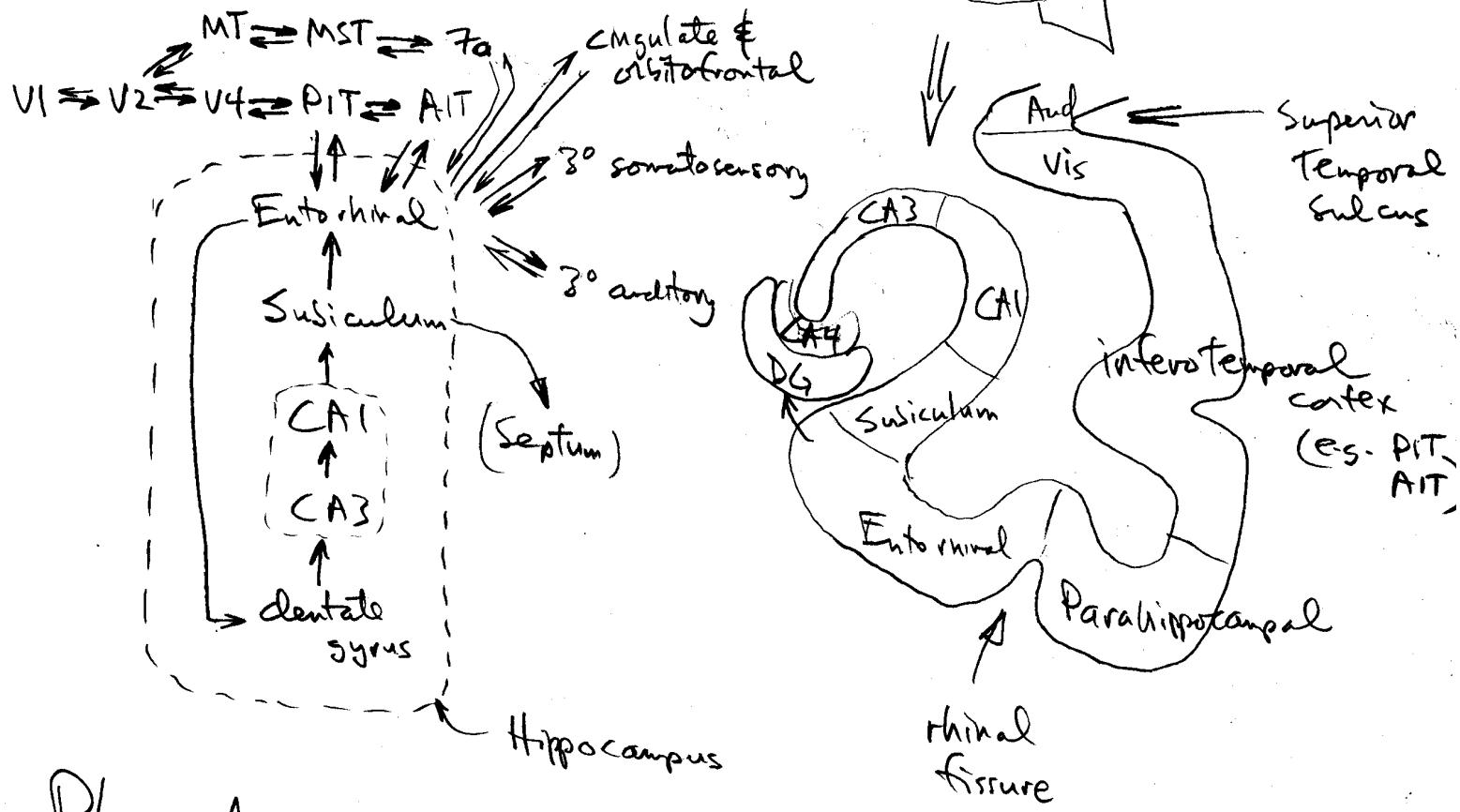


BASIC LIMbic CIRCUITS
 olf. sp. affil. → AFB → olf. sp. fil.
 (non-ossified organ)
 (accessory)
 s.c. nuc. → taste buds
 parabrach. nuc. → pyram. A
 (main)



Hippocampus

Anatomy



Physiology

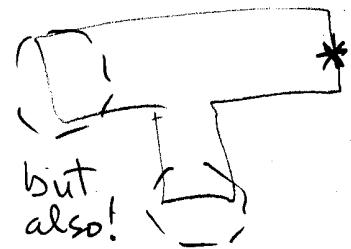
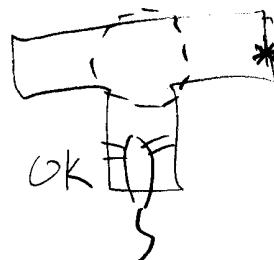
place cells in CA fields

- history of discovery of place cells
- modern study of place cells
 - radial arm maze ← distal cues to place works in dark, too
 - cue-card enclosure ← cue card
 - spatial deficits (Morris water maze) scales up to larger enclosure less firing in different shape
 - barrier ← clear opaque
 - no head direction
- head direction cells in "post subiculum"
 - no place fields
 - incredible inertial guidance
 - distal cues to head direction

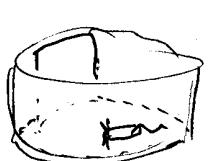
Place / Headdir / Grid

1) James Olds

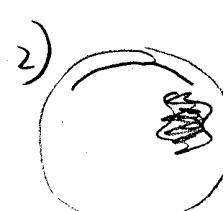
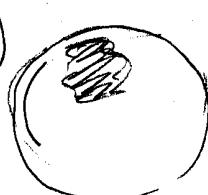
missed discovery
theory-driven observation



2) Basic place cell



→



3) Water maze



vs.

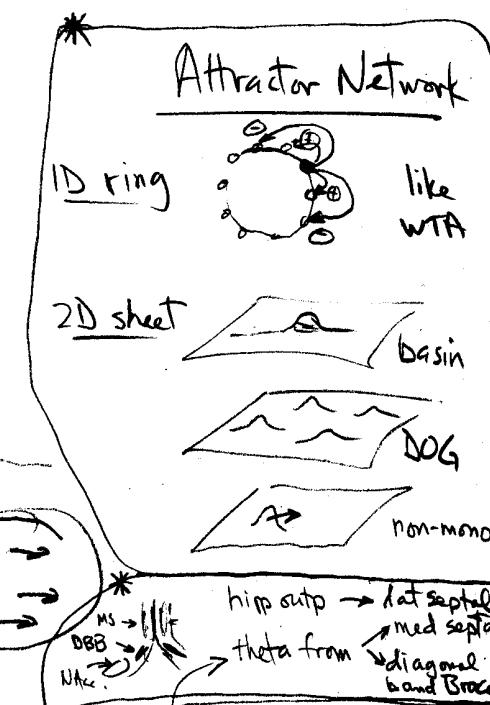
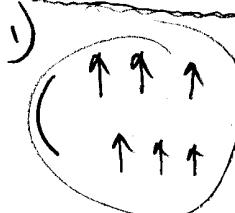
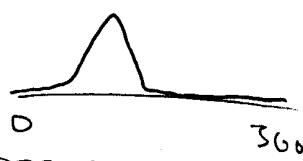


4) radial arm maze

local
distal

4b) move cue card in view → reset!

5) Head dir cell



6) Grid Cell

- med entorhinal cx

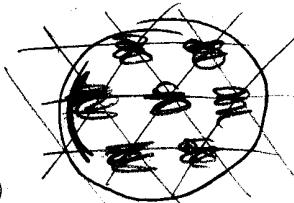
- size: deeper → larger

shape $\square \rightarrow \circ$

3) spiral

2 rooms

barriers



etc

Theta phase precess

outside place field
→ in phase

crossing place field

→ more and more
in advance of
theta
→ antiphase in middle
→ in phase learning

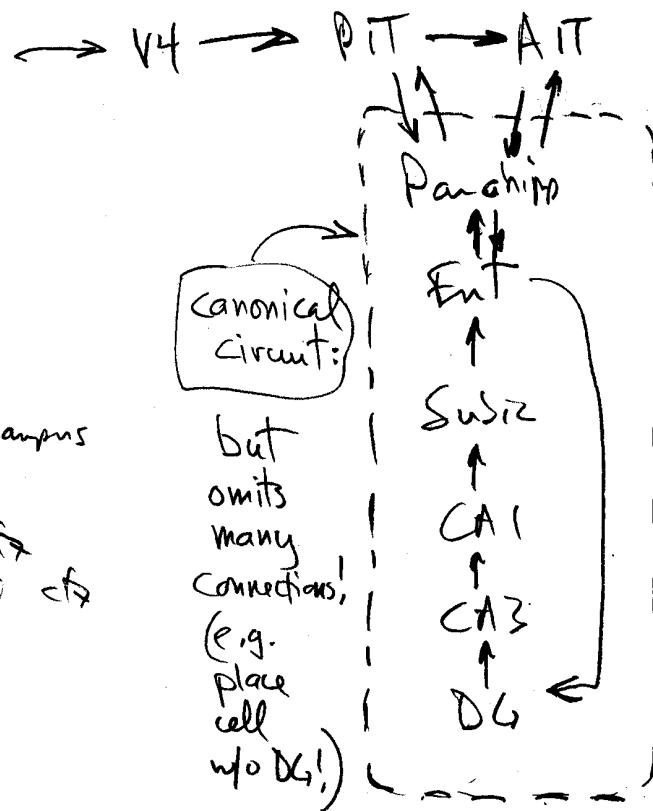
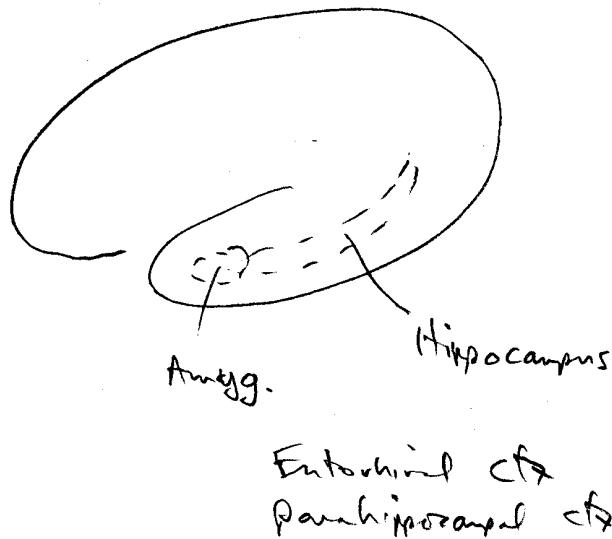
N.B.
can have
place fields
w/o theta

7) Elaboration

head fixed

VR navigation: place cells OFF! → reg. vestib./real movement

H.M.

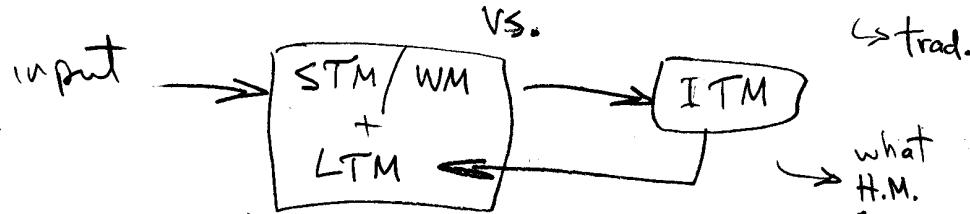


Evidence

- no new memories
- retrograde memory loss for a year
- preserved 'motor' learning
 - [maze \rightarrow explains animal exps]
 - [abstract \rightarrow Tower of Hanoi]

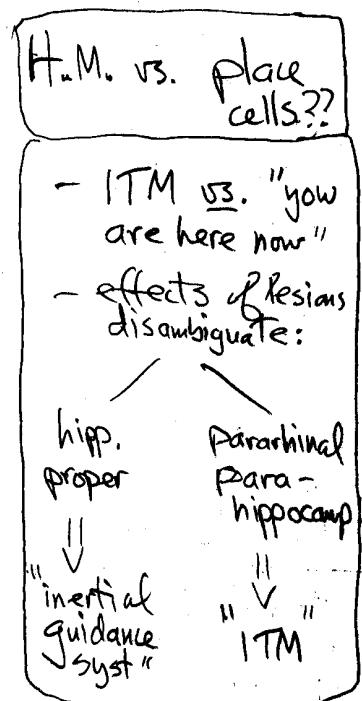
Interpretations

1. explicit/conscious memory/declarative



2. procedural/motor, separate

↳ New pathways: striatum cerebellum!



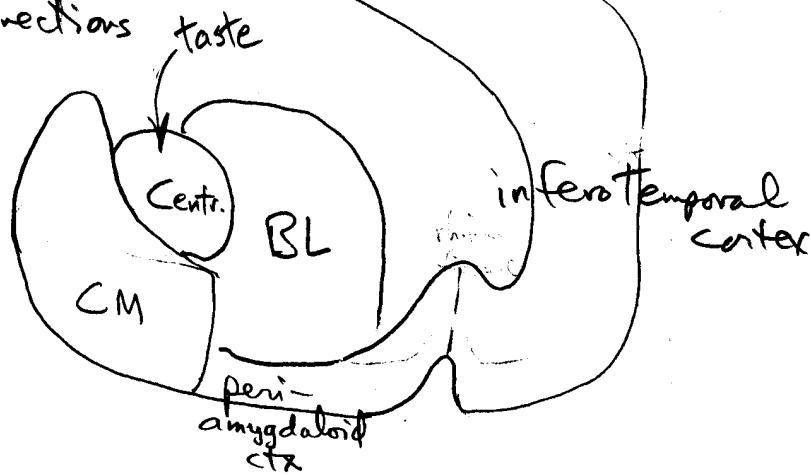
Amygdala

Anatomy

- inputs from 30 sensory areas

- outputs to 30 sensory areas

- limbic connections taste



3^o Somatosensory

(e.g. 5)

3^o Aud

(e.g. Tpt)

3^o Visual

(e.g. PIT)



Physiology

- Kliver-Bucy ^{Amyg} IT

- limbic connections

- ↳ "violence"

- ↳ "not very good wine" cells ☺

- Mishkin & cross-modal matching task deficits

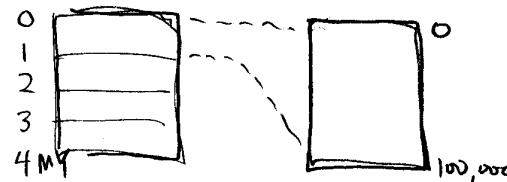
Origin of LANGUAGE

2 pre-adaptations

- primate evolution



- paleoanthropology



- ape language

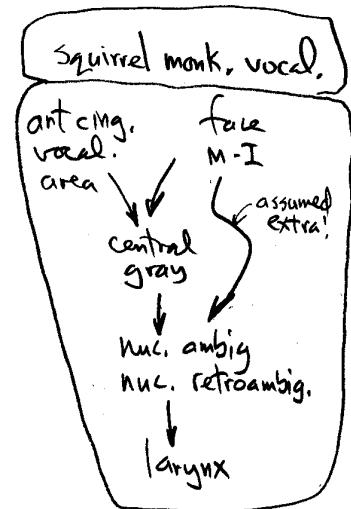
Washoe
Sherman/Austin
Kanzi, Alex

- Gallup mirror test

#1

- birdsong vs. calls, sexual selection, whales

- hominin vocal theory



- main distal sense

blind

primates
bats
electric
snakes
cotton
platypus

- "picture theory" (Bz, IT)

- word recog. (Peter pic expts)

- Lakoff / Fauconnier / Jackendoff

#2

- fictive scene comprehension

- film

- [Polysemy → "line" tiny context (cf. alanine)
anaphora → "context is key"]

[Anaphora → "that fictive scene comprehension stuff"]

DNA, protein self-assembly

