

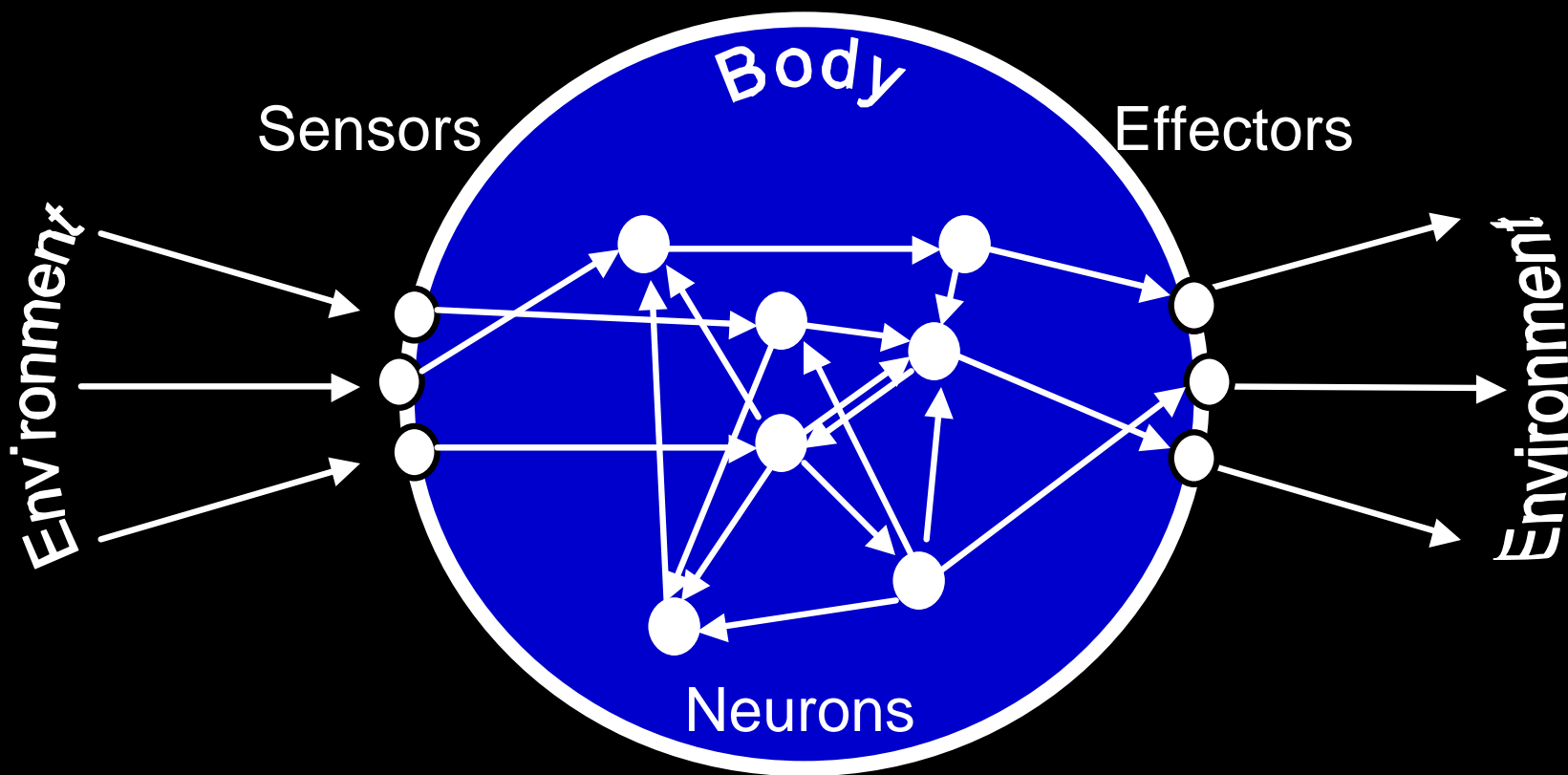
Evolution as the blind engineer: wiring minimization in the brain

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Optimization is a powerful theoretical tool for understanding brain design

- Evolutionary theory: survival of the fittest
- Maximize fitness to predict animal design
- Fitness \sim functionality – cost
- Minimize cost for given functionality

Brain as a neuronal network



Network functionality is captured by neuronal connectivity

Evolutionary cost of wiring

- Signal delay and attenuation
- Metabolic requirements
- Space constraints
- Guidance defects in development

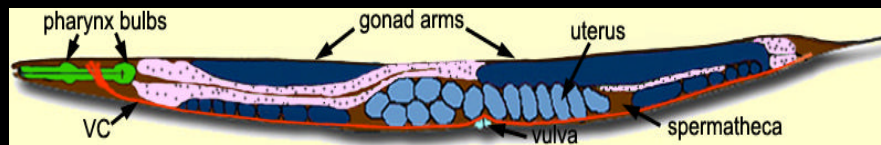
Wiring cost grows with the distance between
connected neurons

For given functionality minimize wiring length

C. elegans as Model System

Anterior

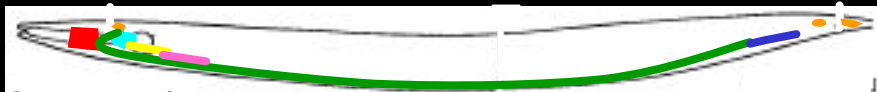
Posterior



1mm

Nervous system

A



P

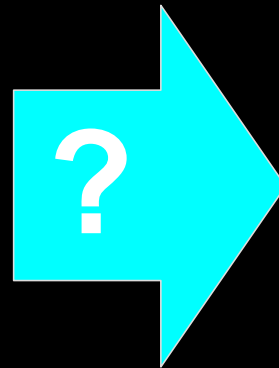
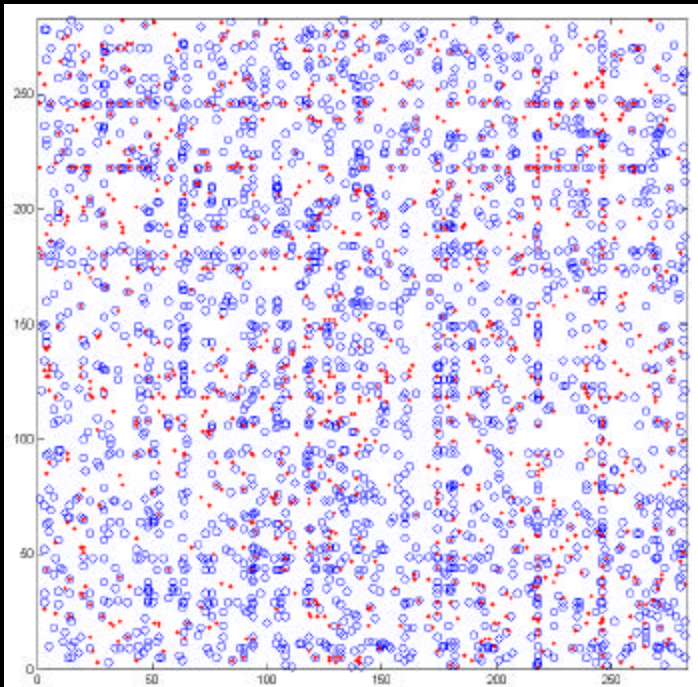
- Well documented
 - Wiring diagram
 - Neuronal map
- Simple system
 - 302 neurons
 - 11 ganglia
- One-dimensional problem

Can wiring minimization predict neuronal placement?

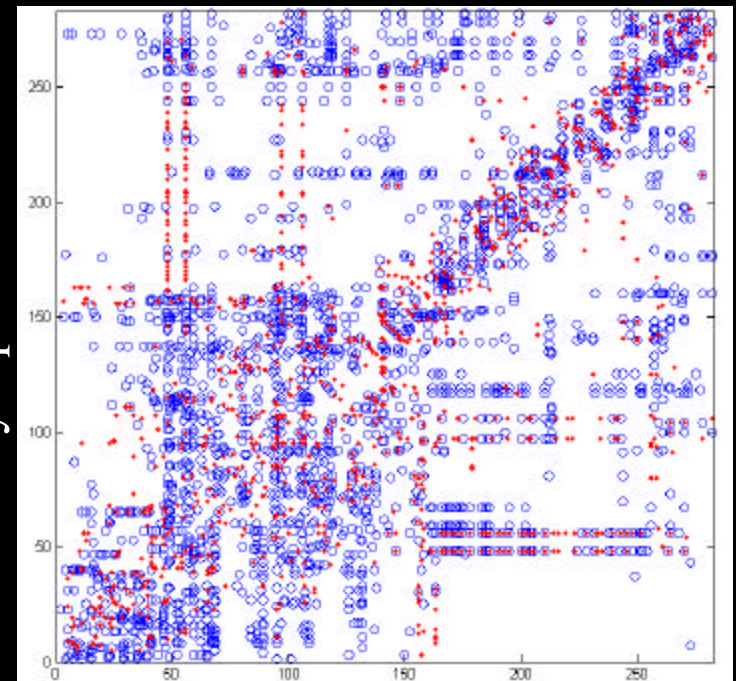
From the wiring diagram...

To the actual placement...

Pre-synaptic Neuron



Pre-synaptic Neuron



Post-synaptic Neuron

Post-synaptic Neuron

- Chemical synapse
- Electrical synapse



Quadratic Cost Function

$$E = \underbrace{\left[\frac{1}{2} \sum_{i,j} A_{ij} (r_i - r_j)^2 \right]}_{\text{Internal wiring cost}} + \underbrace{\left[\sum_{k,l} B_{kl} (r_k - f_l)^2 \right]}_{\text{External constraints}}$$

r_i = position of neuron i

f_l = position of sensor/effector l

A_{ij} = neuron i to neuron j
connection matrix

B_{kl} = neuron k to sensor/effector l
connection matrix

For symmetrized A , rewrite into matrix form...

$$E = [r^T \underbrace{(D_A - A)}_L r] + [r^T D_B r - 2r^T Bf + \text{const}]$$

Laplacian of A

$$D_{Aij} = \mathbf{d}_{ij} \sum_p A_{ip}$$

$$D_{Bij} = \mathbf{d}_{ij} \sum_p B_{ip}$$

Optimal placement coordinates:

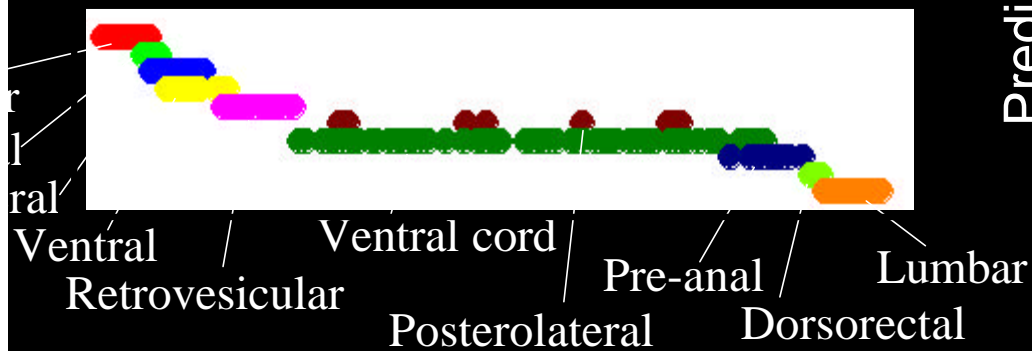
$$r = (L + D_B)^{-1} Bf$$

Actual vs. Predicted Neuron Positions

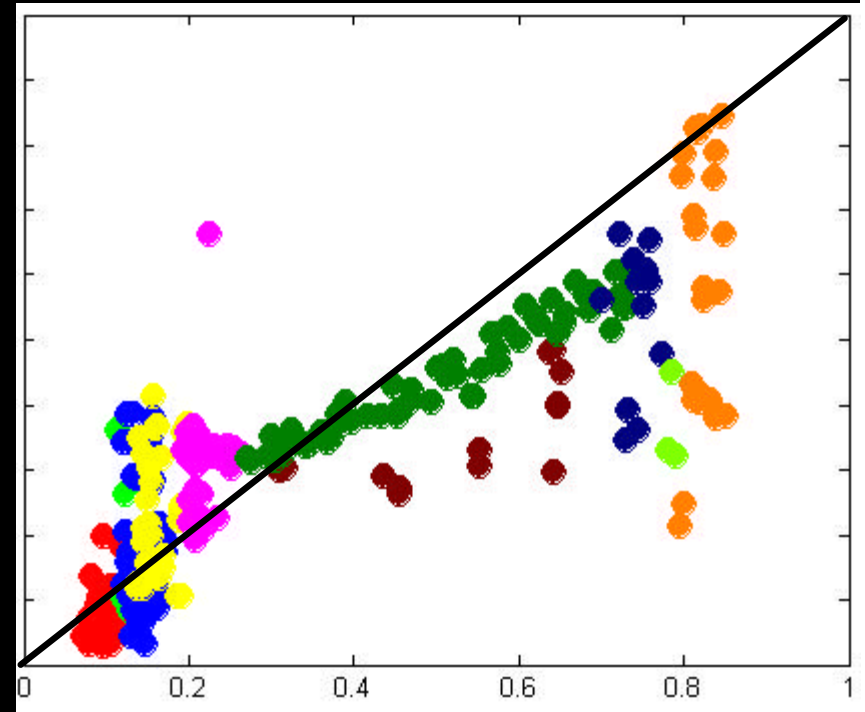
Predicted



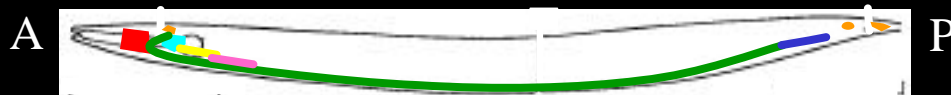
Actual



Predicted Position



Actual Position



Wiring minimization is reasonable but not perfect

Why is not wiring minimization prediction perfect?

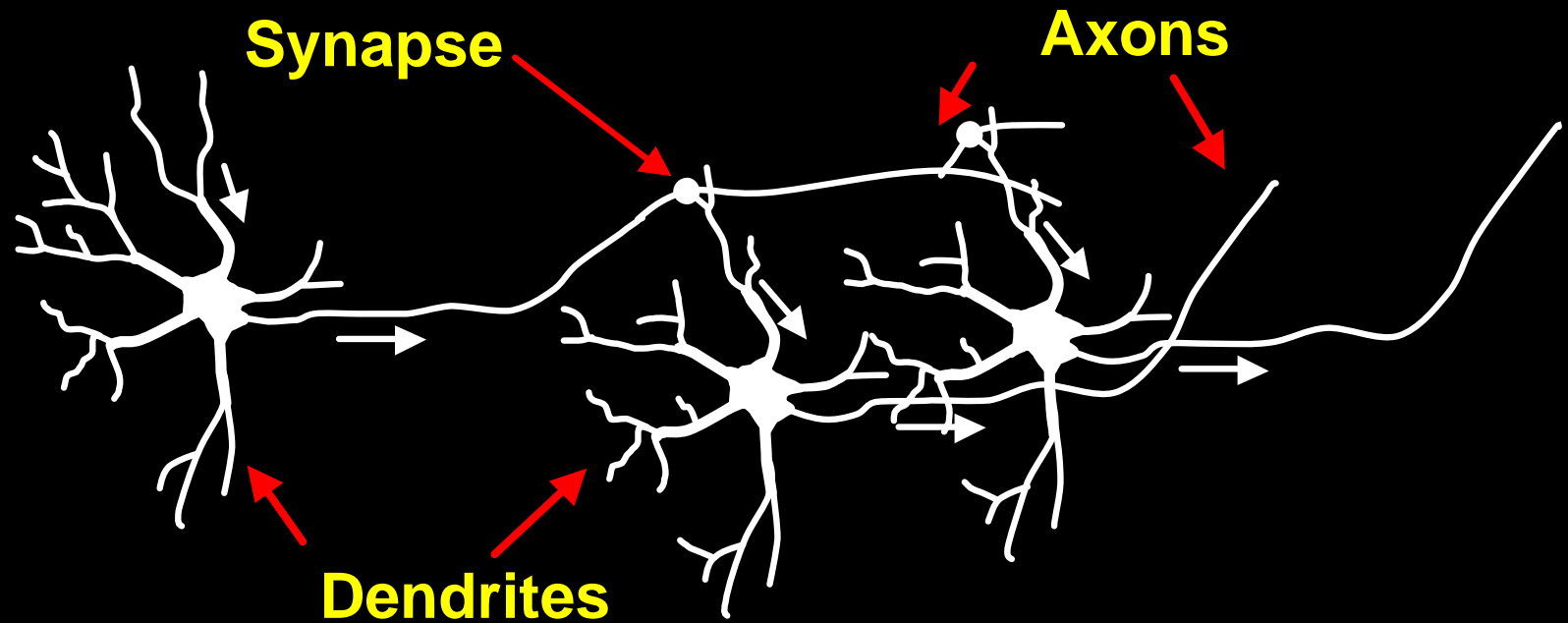
- Nervous system may be sub-optimal
- Other constraints may be important
(e.g. development)
- Quadratic cost function may be incorrect
- Routing optimization may affect placement

Routing or neuronal shape

point neurons



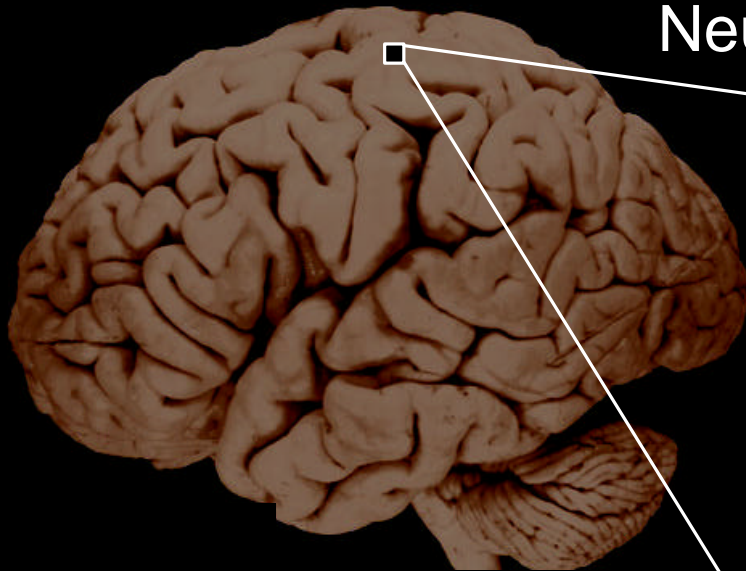
actual neurons



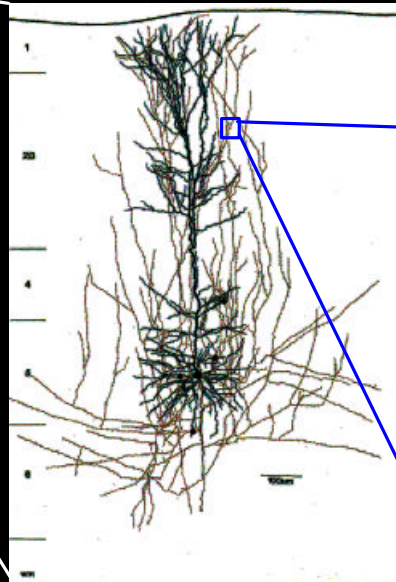
Big brains - large numbers

Brain $\sim 10^{11}$ neurons

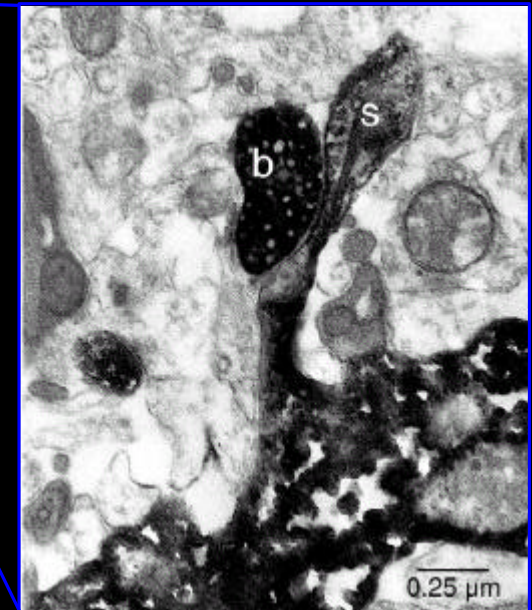
Neuron $\sim 10^4$ synapses



10cm



1mm



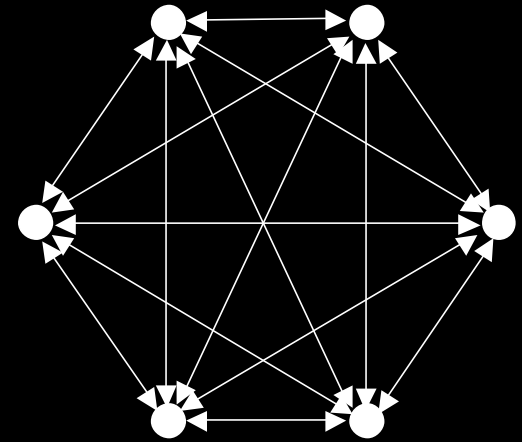
Synapse

1 μ m

Assembling the wiring diagram will take many years

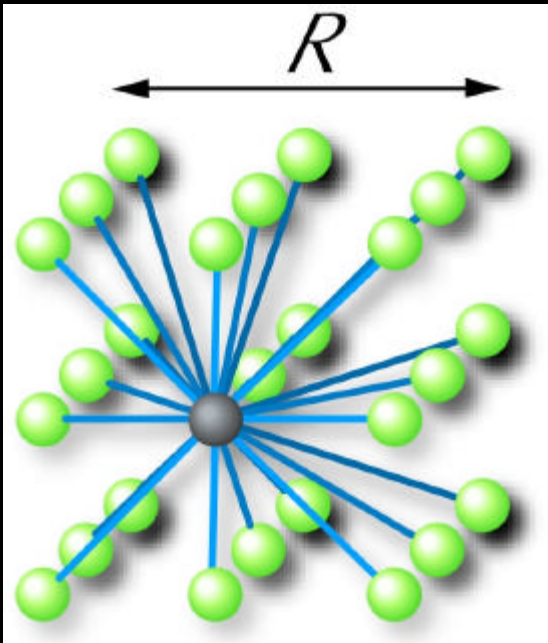
Routing problem

- Network of N neurons
- Fully connected (all-to-all)
- Fixed wire diameter, d



Find wiring design minimizing
network volume

Design I: Point-to-point axons



Number of neurons: N

Wire diameter: d

Axon length per neuron: $l \approx NR$

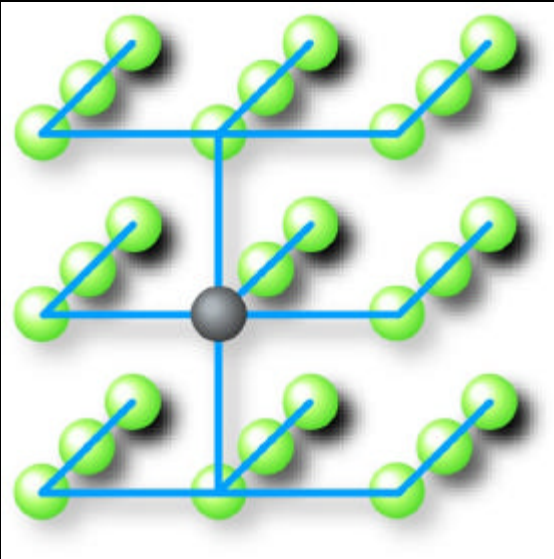
Total wiring volume: $R^3 \approx Nld^2$

\Rightarrow Network size: $R \approx Nd$

Mouse cortical column (1mm^3): $N=10^5$, $d=0.3\mu\text{m} \Rightarrow$

$\Rightarrow R=3\text{cm}$ ☹️

Design II: Branching axons (multi-pin nets)



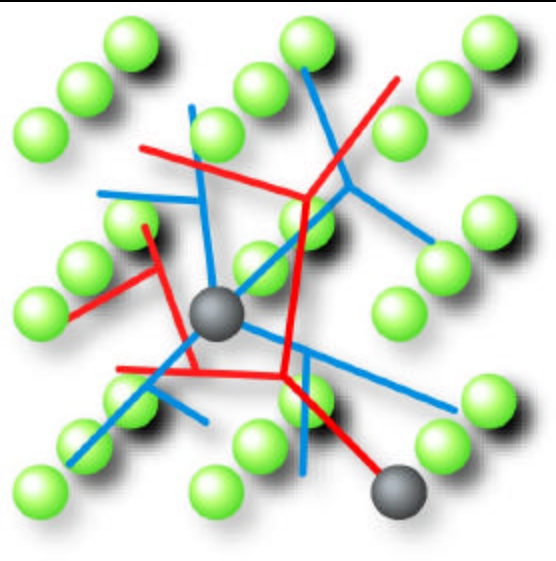
Inter-neuron distance: $R / N^{1/3}$

Axon length per neuron: $l = R N^{2/3}$

Total wiring volume: $R^3 \propto N l d^2$

\Rightarrow Network size: $R \propto N^{5/6} d$

Cortical column: $N=10^5$ $d=0.3\mu\text{m}$ $\Rightarrow R=4.4\text{mm}$ ☹️



Design III: Branching axons and dendrites

Total number of voxels: R^3 / d^3

Number of voxels containing axon: l/d

Fraction of voxels containing axon: ld^2 / R^3

Fraction of voxels containing dendrite: ld^2 / R^3

Number of voxels containing axon and dendrite: $l^2d / R^3 \sim 1$

Total wiring volume: $R^3 \square Nld^2$

Network size: $R \square N^{2/3} d$

Cortical column: $N=10^5$ $d=0.3; 1\mu\text{m} \Rightarrow R=1.6\text{mm}$ ☹️

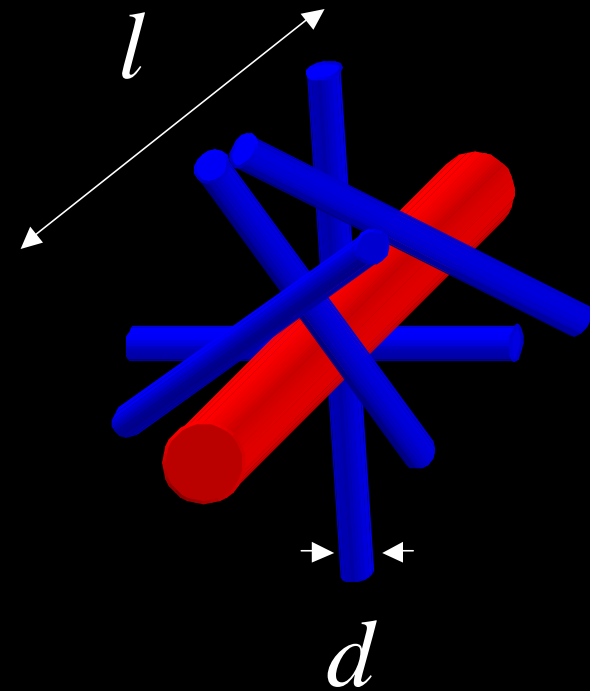
Is it possible to improve on Design III?

In Design III, dendrite length can be found...

$$\left. \begin{array}{l} R^3 \propto Nld^2 \\ R \propto N^{2/3}d \end{array} \right\} \Rightarrow l \sim Nd$$

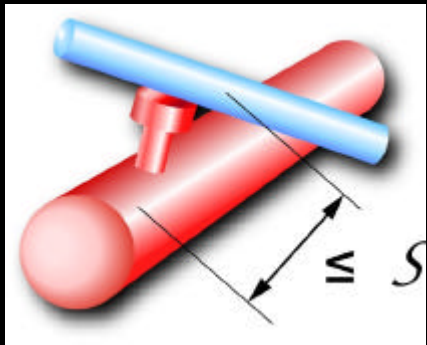
...to be smallest possible:

$$L \gtrsim Nd$$



Design III cannot be improved if dendrites are smooth

Design IV: Branching axons and spiny dendrites



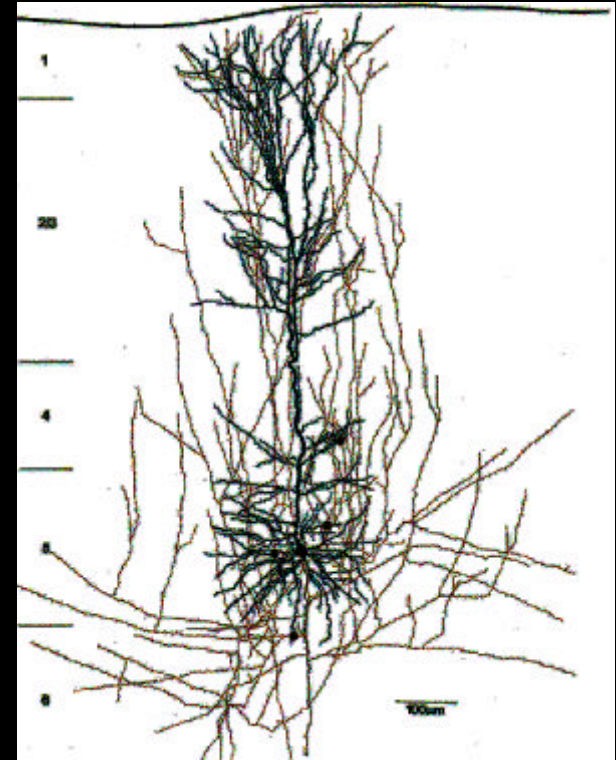
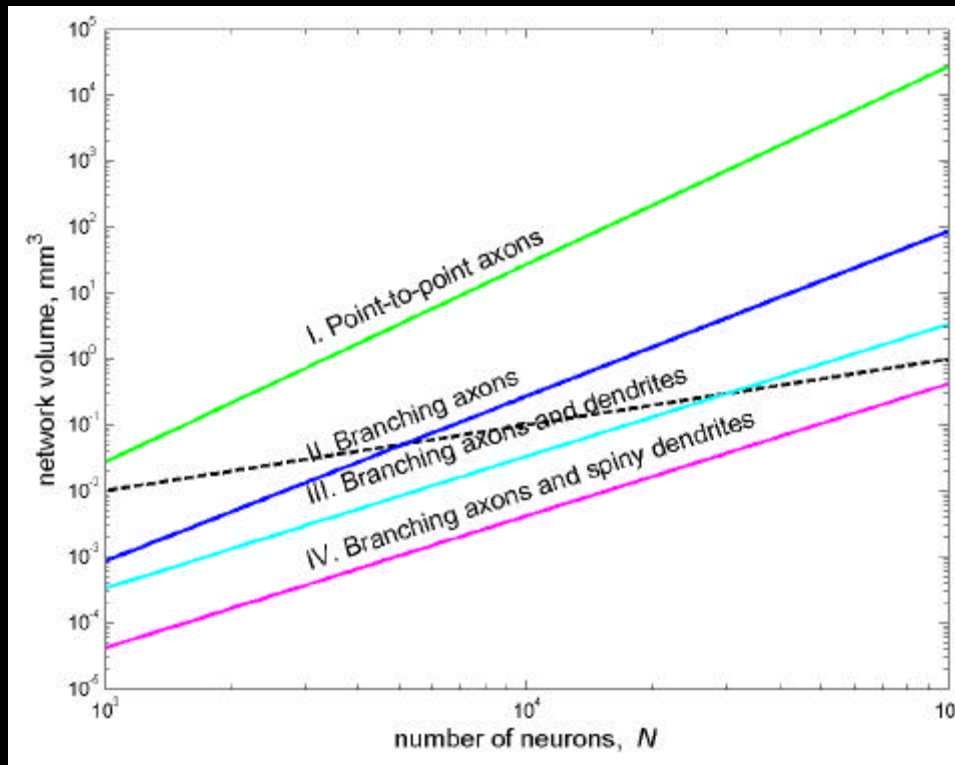
Number of voxels containing axon and dendrite: $l^2 s / R^3 \sim 1$ } \Rightarrow
 Total wiring volume: $R^3 \propto N l d^2$ }

\Rightarrow Network size: $R \propto N^{2/3} d^{4/3} / s^{1/3}$

Cortical column: $N=10^5$ $d=0.3; 1\mu\text{m}$ $s=2.5\mu\text{m}$ \Rightarrow

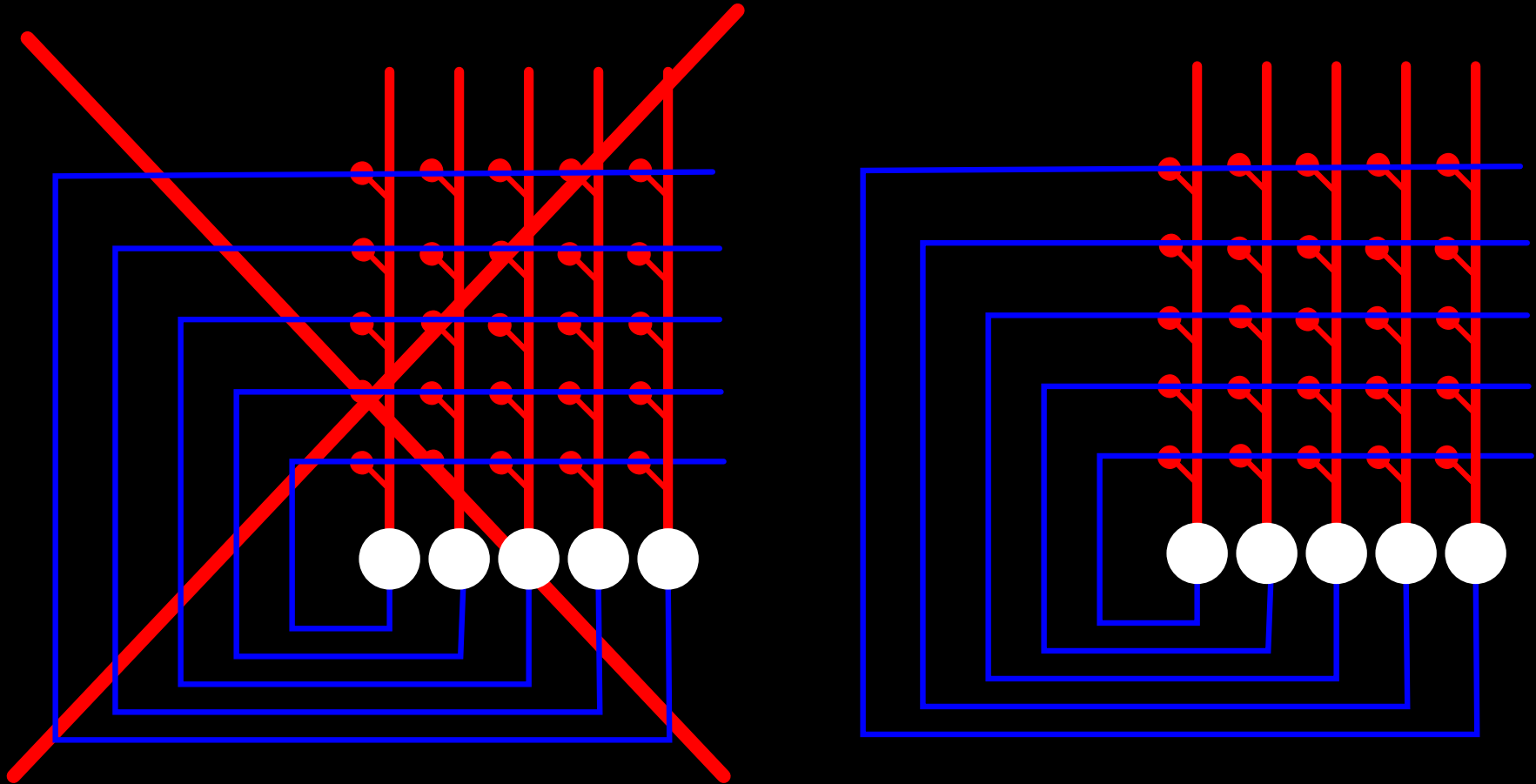
$\Rightarrow R=0.8\text{mm}$ ☺

Network volume for various wiring designs



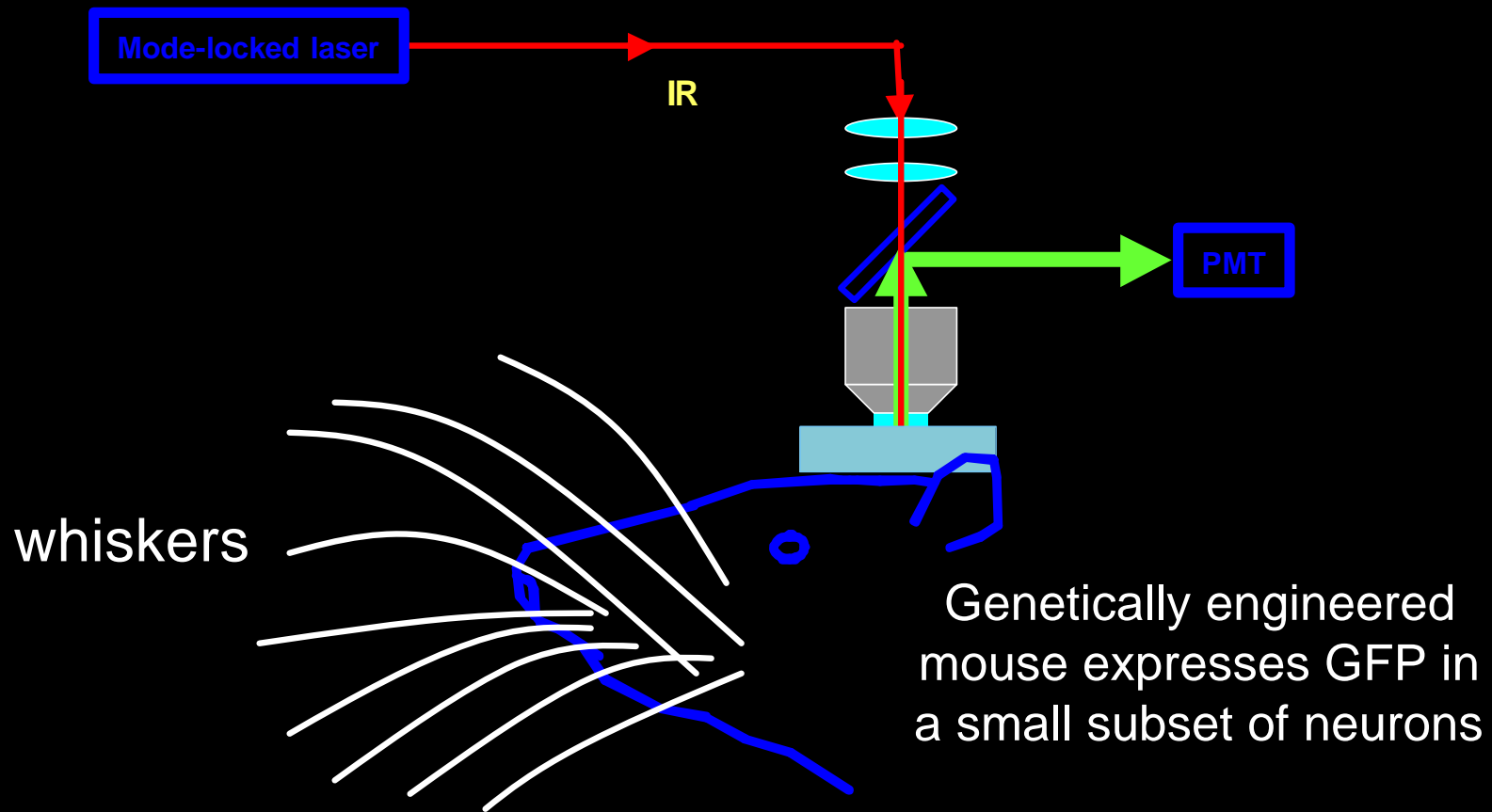
Neuronal shape is a routing solution implementing high inter-connectivity

Cortical architecture is optimized for high inter-connectivity



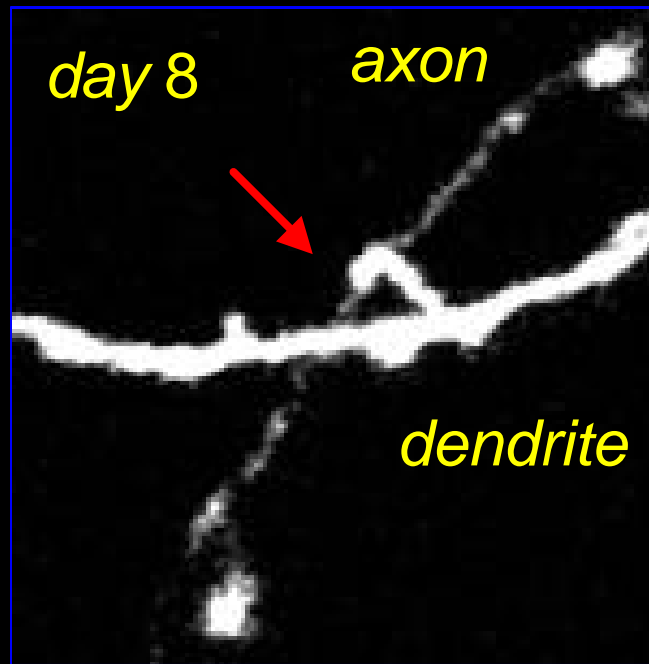
Synapse re-arrangement is potential memory mechanism with high information storage capacity (Stepanyants, Hof, Chklovskii, 2002)

Experiments on synapse re-arrangement



Two-photon microscope provides *in vivo* images with single-synapse resolution

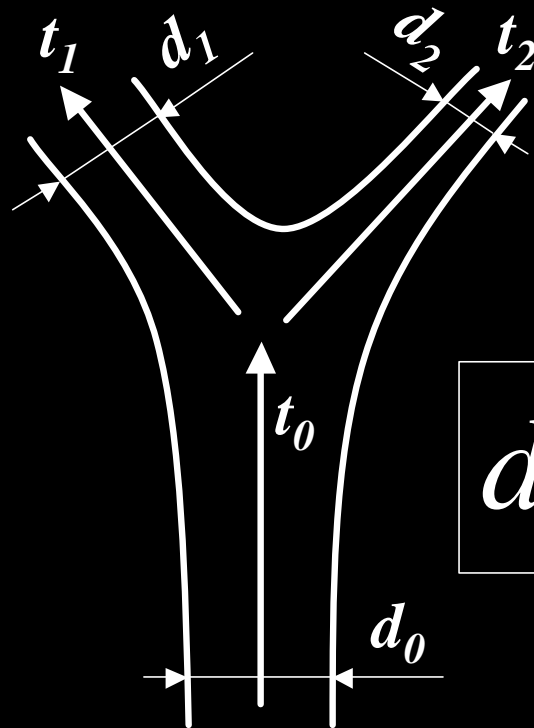
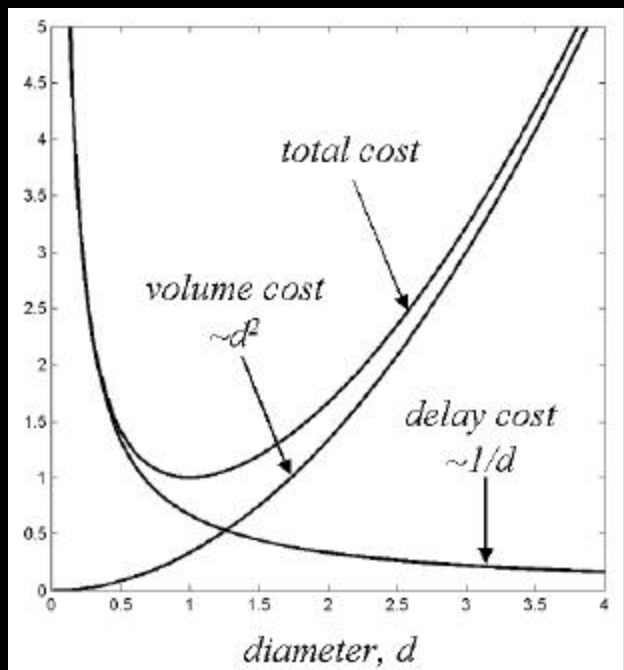
Spine remodeling indicates synapse re-arrangement *in vivo*



2mm

Trachtenberg, ..., Svoboda, 2002

What determines axon (dendrite) diameter?



$$d_0^3 = d_1^3 + d_2^3$$

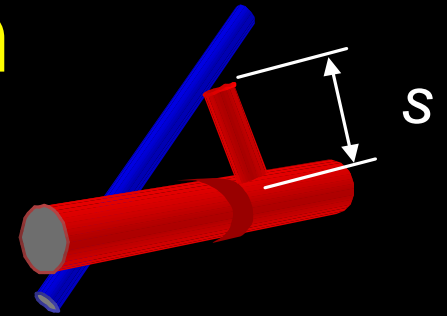
Axon diameter minimizes the combined cost of wiring volume and conduction delays

Summary

Wiring minimization is a key factor determining brain architecture

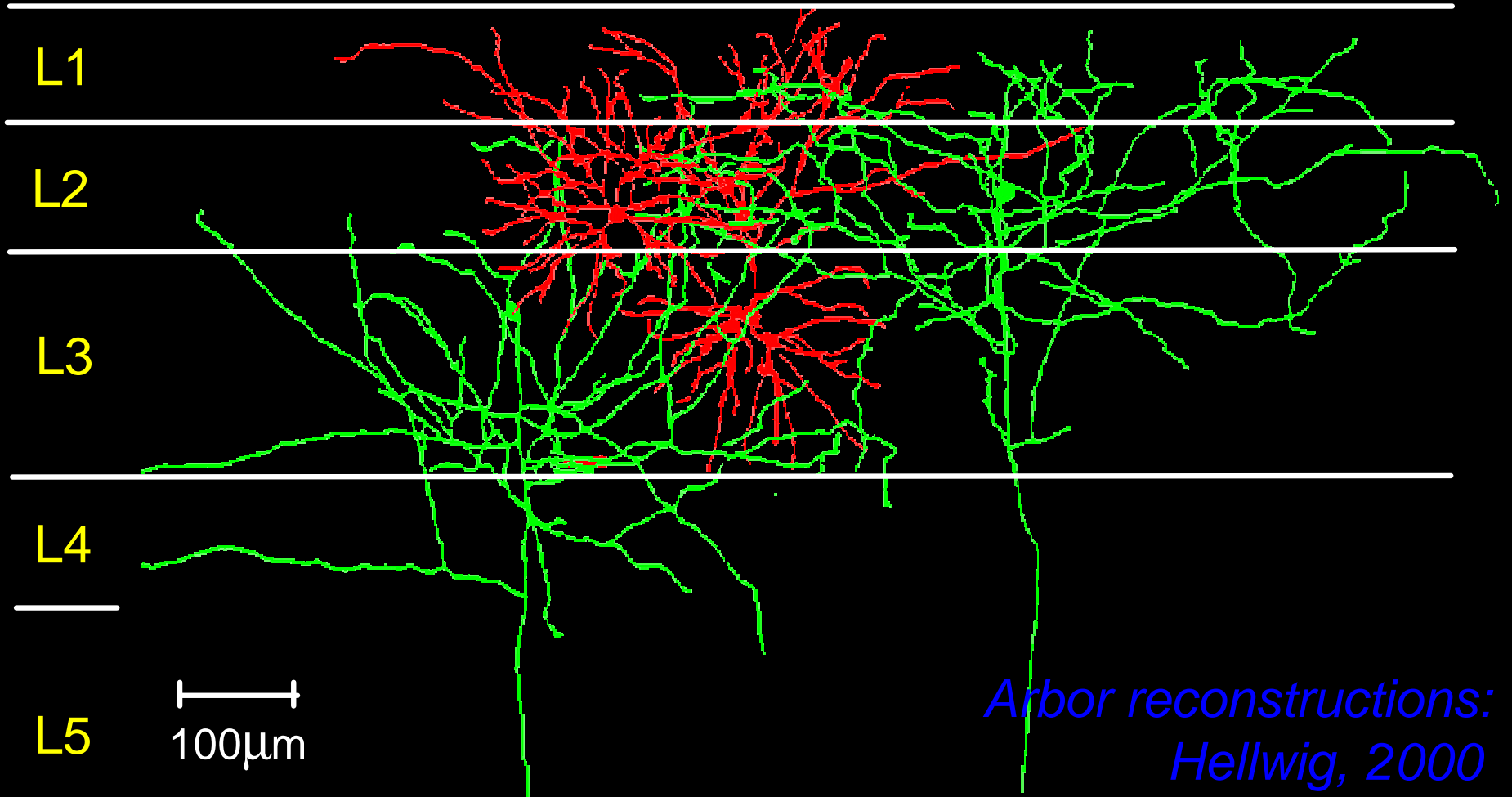
Complexity of neuronal networks poses challenging wiring minimization problems

Potential synapse is a location where axon comes within a spine length of a dendrites



- Potential synapse is a necessary (but not sufficient) condition for an actual synapse
- Potential synaptic connectivity is more stable than actual
- Potential synaptic connectivity can be evaluated geometrically

90% potential connectivity neighborhood



“Potential” definition of a cortical column

What is the correct cost function?

Biology: $\text{Min}\{V\} \rightarrow \text{Min}\{C=V-\lambda\log N\}$

Physics: $\text{Min}\{E\} \rightarrow \text{Min}\{F=E-TS\}$

Constrained optimization is a powerful tool
for building a theory of brain function

Acknowledgments



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