



# WHAT IS A FLAGSHIP OF ENTERPRISE AND TECHNOLOGY?

## Future and Emerging Technology Flagships (FET)

Are ambitious large-scale, informatics-driven, research Initiatives that aim to achieve a visionary goal.

The scientific advance should provide a strong and broad Basis for future **technological innovation** and economic Exploitation in a variety of areas, as well as **novel benefits for society**.

The research is collaborative, internally non-competitive, inter- and trans-disciplinary, driven by a commonly agreed road-map

**BLUE BRAIN PROJECT + NEUROIMAGING COMMUNITY**





# THE IMPACT OF INFORMATION & COMPUTING TECHNOLOGY

## VON NEUMANN MACHINES

MOORE'S LAW

ENERGY LIMITATIONS

INTERNET

DATABASE MANAGEMENT

CLOUD ENVIRONMENT

DATABASE QUERYING & ADDRESSING

REAL-TIME VISUALISATION

SUPERCOMPUTING

BEYOND EXASCALE

BANDWIDTH & ROUTING [**HTML5, Cisco**]

DISTRIBUTED [**Oracle**]

SECURITY [**Amazon, Dropbox, iCloud**]

LOCAL [**Google**] vs REMOTE [**EPFL**]

FOR SUPERCOMPUTING [**IBM, CRAY**]

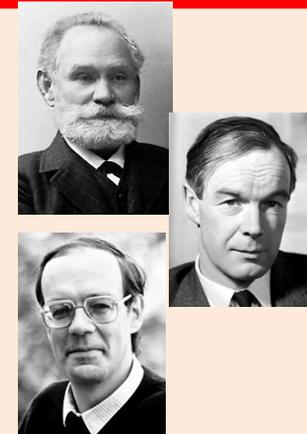
**NEUROMORPHIC COMPUTING**

# EUROPEAN NOBEL CONTRIBUTIONS TO NEUROSCIENCE



## NEUROANATOMY NEUROPHYSIOLOGY

Golgi & Ramon y Cajal  
Pavlov  
Barany  
Sherrington & Adrian  
Von Bekesy  
Eccles, Hodgkin & Huxley, Weisel  
Neher & Sakmann



## NEUROPHARMACOLOGY

Dale & Loewi  
Katz & von Euler  
Black



## NEUROIMAGING

Carlsson  
Josephson  
Hounsfield  
Mansfield



## NEUROSURGERY

Moniz

## SOCIAL & COGNITIVE NEUROSCIENCE

Von Frisch, Lorenz & Tinbergen  
Kahneman

## GENETICS

*Crick & Wilkins*

*Sanger*

*Sulston*

*Jerne, Milstein & Kohler*

*Levi-Montalcini*

*Evans*



## IMMUNOLOGY

## CELL BIOLOGY



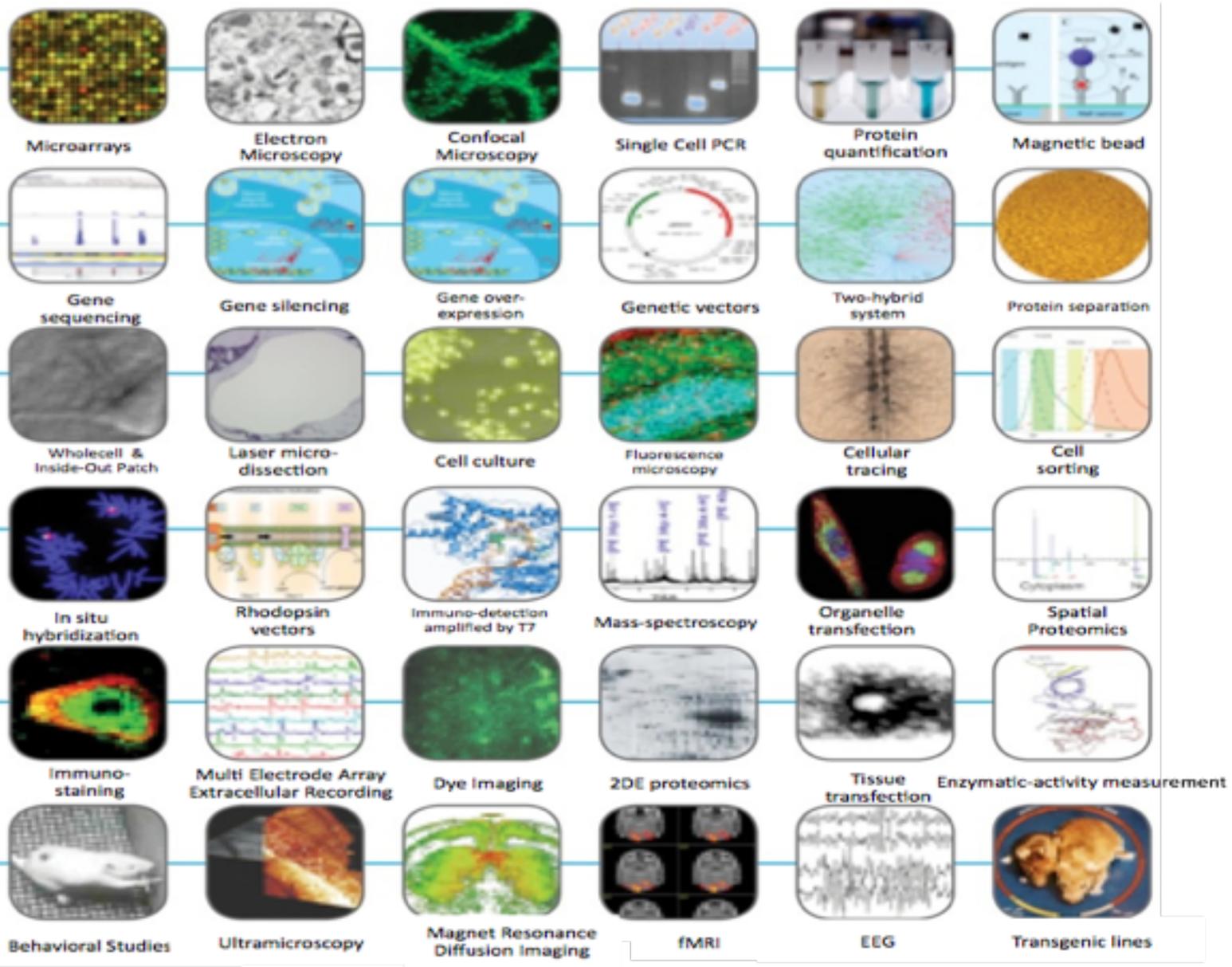
# NEUROSCIENCE METHODS

subcellular resolution

cellular resolution

tissue resolution

whole brain scale





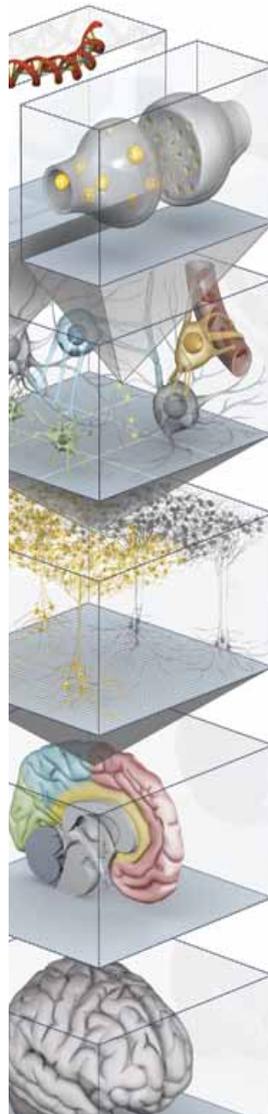
# MOTIVATION 1 – DATA DELUGE - FEDERATION & INTEGRATION

## The facts:

1. Data and knowledge is **growing** exponentially
2. Data and knowledge are increasingly **fragmented**
3. Benefits for society seem to be **decreasing**
4. Economic burden increasing to **unsustainable** levels

## The causes:

1. No integration plan
2. No data curation plan
3. No plan to link across levels
4. No plan to transfer knowledge from animal model to human
5. No plan to go beyond traditional classification of diseases



### Molecular

A century of research with the first insight under a microscope into a digital facet: component molecules assemble a cell through the essential properties of the transmission of chemical signals.

### Cellular

A brain-in-a-box has to capture neurons and non-neurons, including the shapes of their dendrites that receive and...

### Circuits

A model of the network between different regions among neighboring cells furnish clues to the complex brain disorders of autism and schizophrenia.

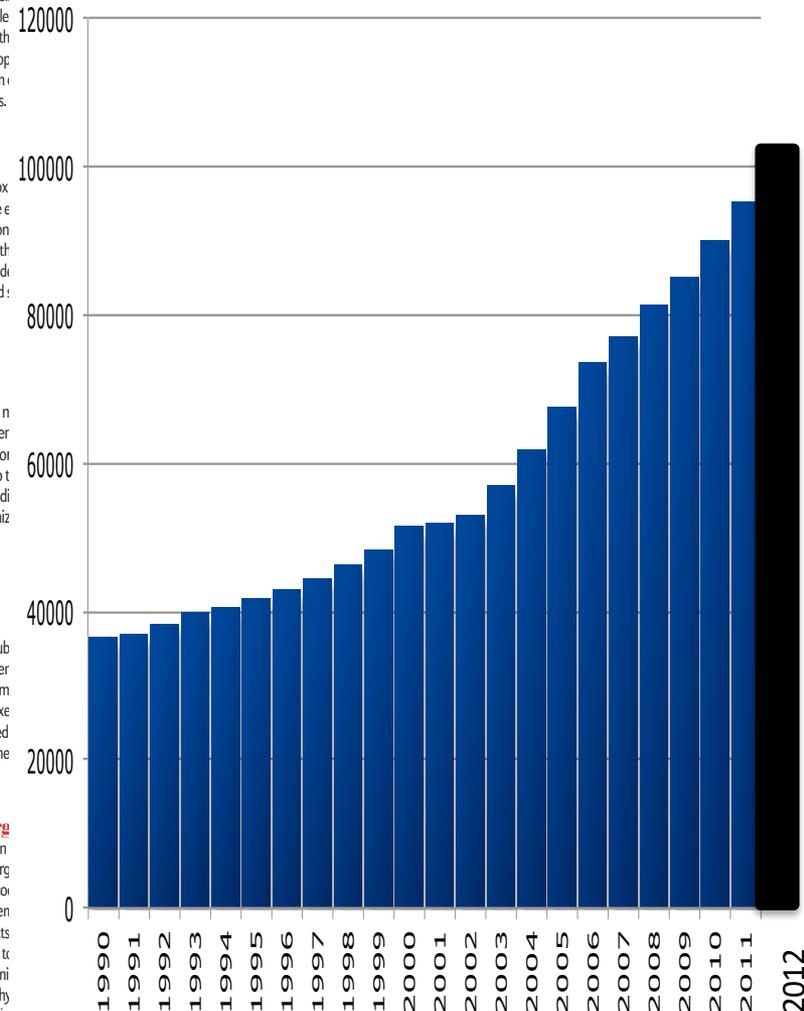
### Regions

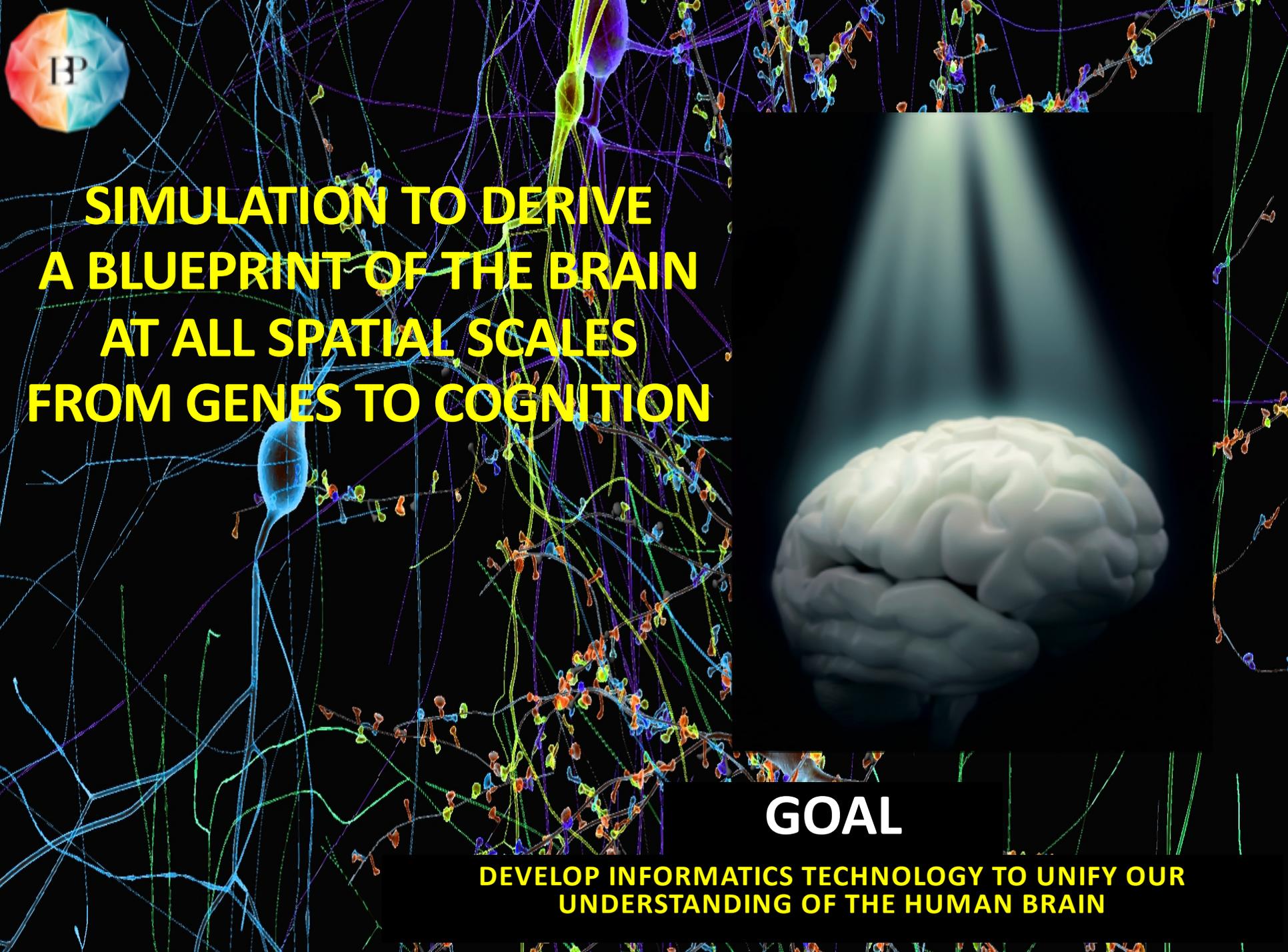
Major neural subregions: the amygdala (near the hippocampus) in the frontal lobes (executive functions) can be inspected and interact with one another.

### Whole Org

An in silico brain for the actual organ: the computer simulation of the virtual system mimics the effects as scientists do through "knockout" a gene in mice to avoid the lengthy...

## Number of Peer Reviewed Publications on the brain /yr





**SIMULATION TO DERIVE  
A BLUEPRINT OF THE BRAIN  
AT ALL SPATIAL SCALES  
FROM GENES TO COGNITION**

**GOAL**

**DEVELOP INFORMATICS TECHNOLOGY TO UNIFY OUR  
UNDERSTANDING OF THE HUMAN BRAIN**



# CONFRONTING PARADIGMS

## **CARTESIAN MODEL (TOP DOWN)**

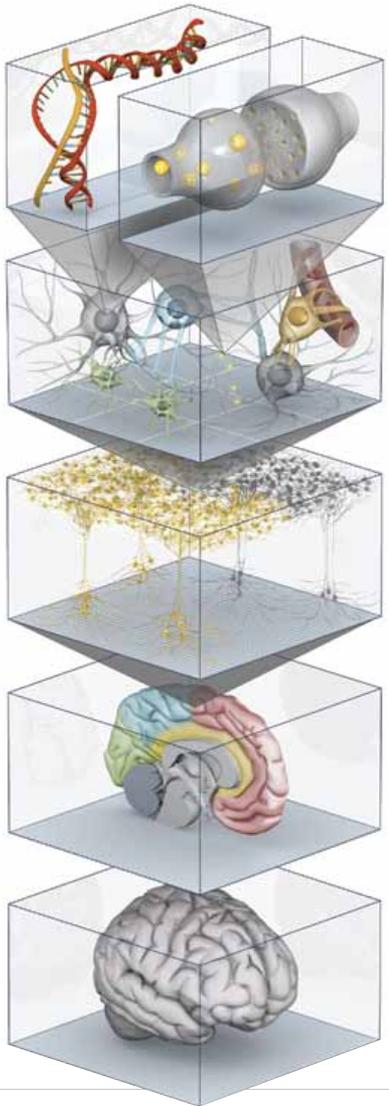
Mentally generated hypothesis  
Mathematically expressed in a model  
Confrontation with “relevant” data  
Parameterisation and optimisation of model  
**Correlations (non-causal)**

## **SIMULATION MODEL (BOTTOM UP)**

Multimodal and multivariate data  
Exhaustive mining to demonstrate coherent models  
Exploration of these mathematical models as generated hypotheses  
Investigation of hypotheses – clinical, mechanistic, prognostic, therapeutic  
**Knowledge (& causes)**

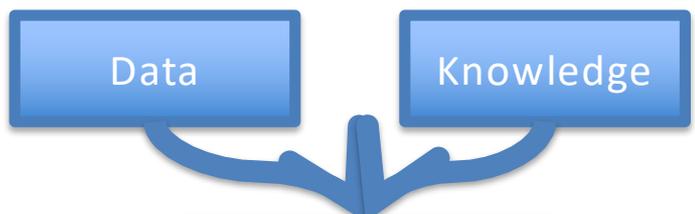
# BRAIN SIMULATION

## PROVIDE A THEORETICAL FRAMEWORK FOR HUMAN BRAIN ORGANISATION AND FUNCTION



1. Increase the value of all past and future experiments
2. Gather and organize all fragments of data and knowledge on the brain
3. Provide open accessible brain atlases
4. Fill knowledge gaps using novel ICT tools
5. Generate strategically selected missing data that will not be generated otherwise
6. Prioritize, optimize and accelerate biological experiments

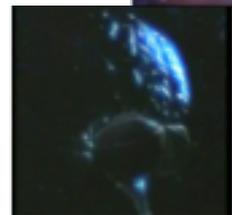
# Simulation-Based Brain Research



Cognition



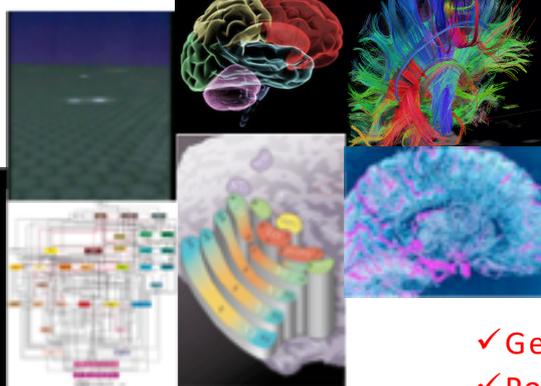
Whole brain



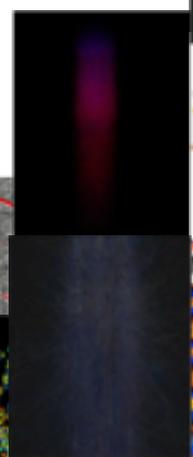
macrocircuits



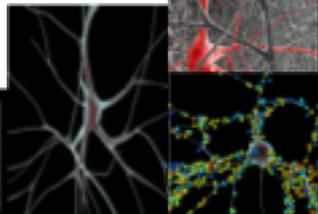
mesocircuits



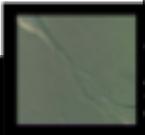
microcircuits



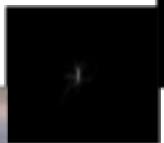
synapses



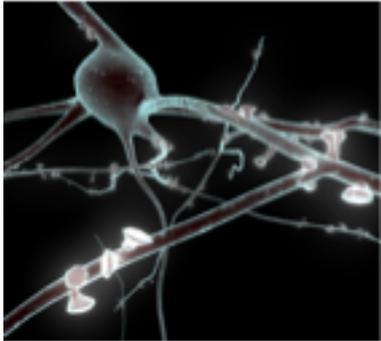
neurons



molecules



- ✓ Genomics
- ✓ Post-Genomics...
- ✓ Systems Biology...
- ✓ Integrative Biology...
- ✓ Informatics...
- ✓ Neuroscience...
- ✓ Mathematical Modeling...
- ✓ Supercomputer Simulation ...
- ✓ Visualization...
- ✓ Analysis...

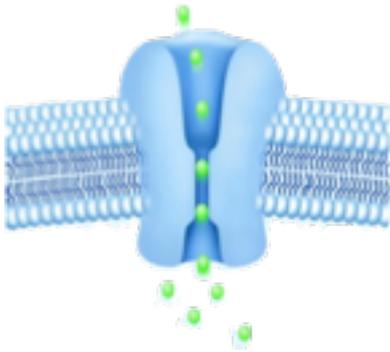


## NEURONS

10,000 neurons

~ 4,000,000

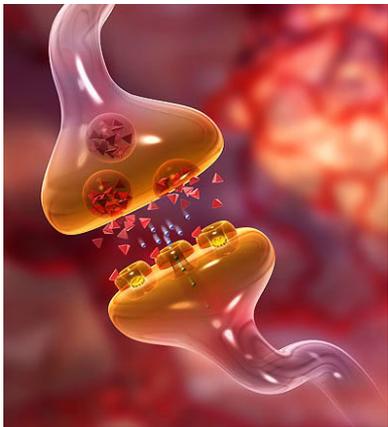
Electrical compartments  
(Rall Equations)



## ION CHANNELS

80,000,000 Ion Channels

(Hodgkin-Huxley Equations)

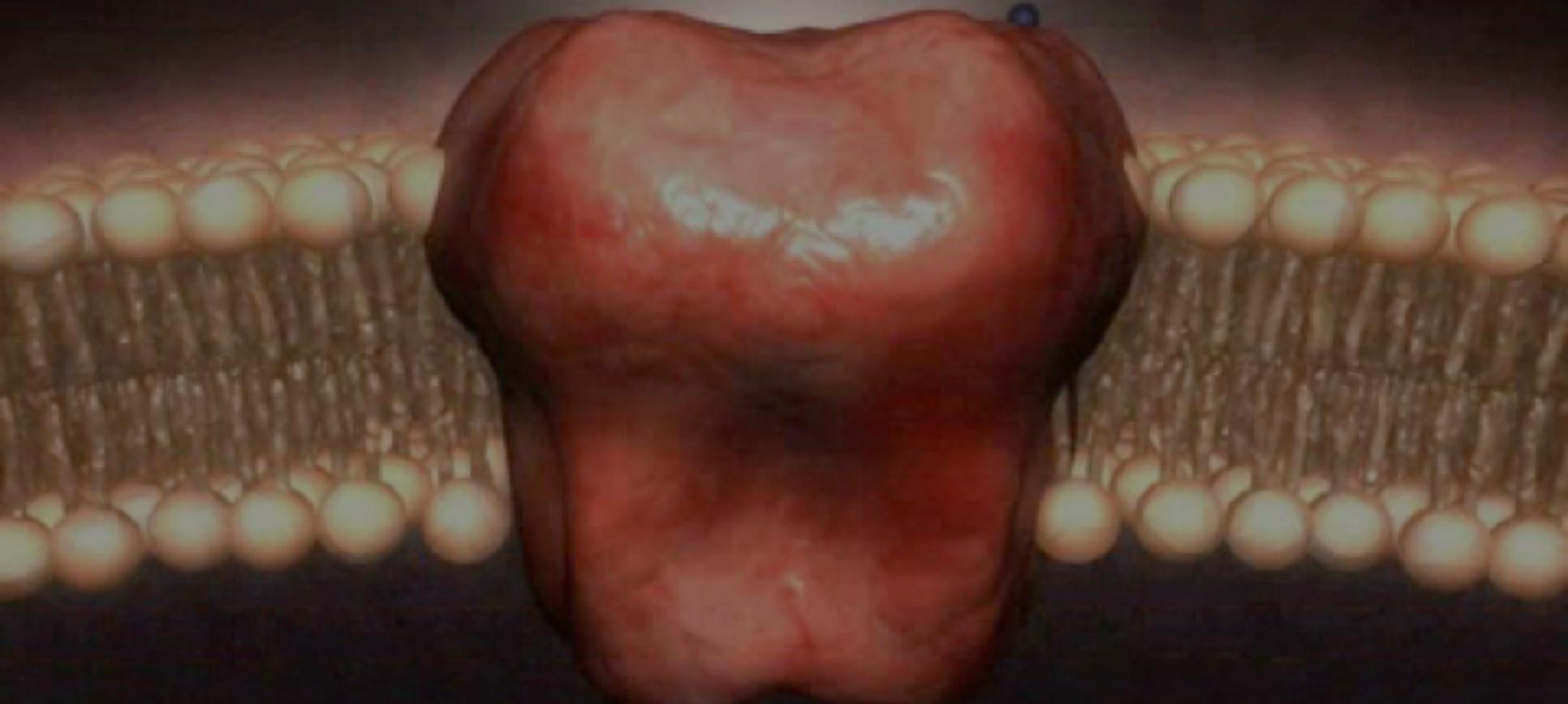


## SYNAPSES

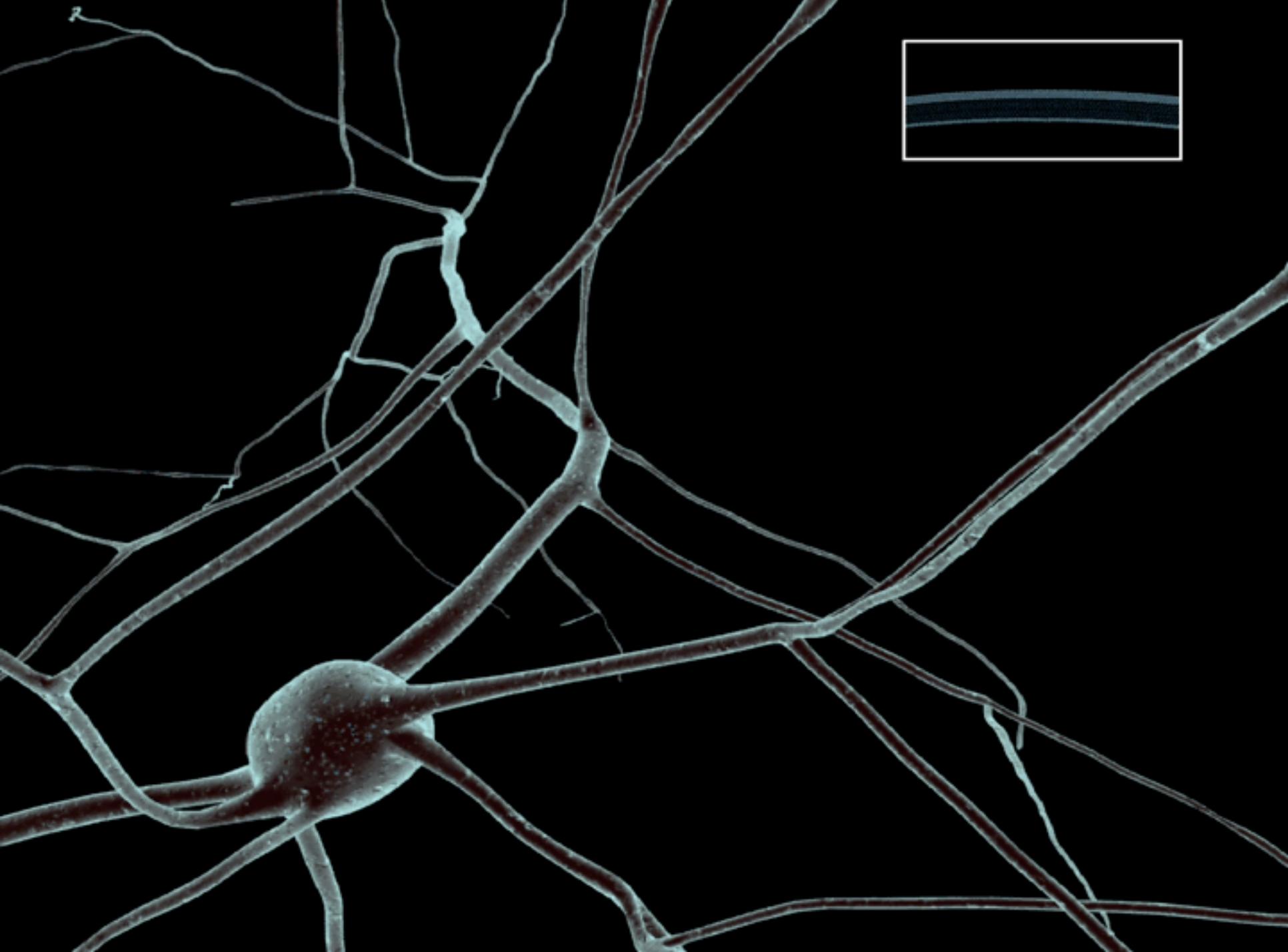
10,000,000 Synapses

(Tsodyks-Markram Equations)

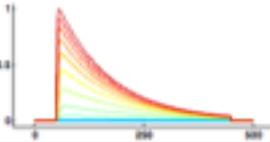
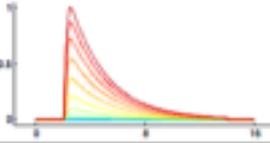
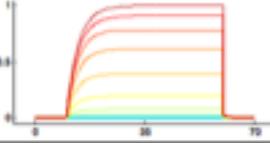
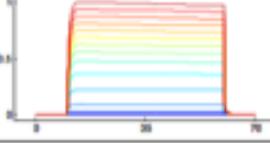
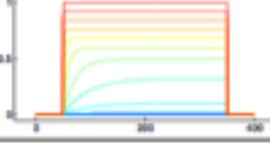
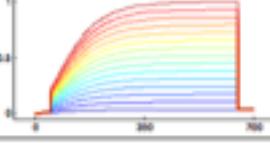
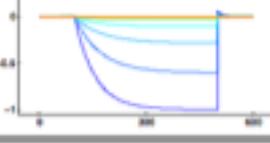
# Combination of ion channels determines the electrical phenotype

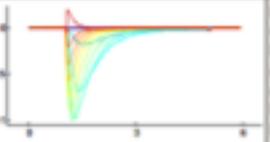
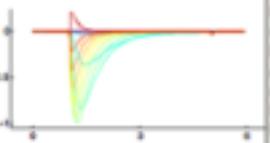
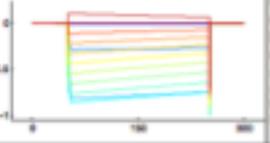
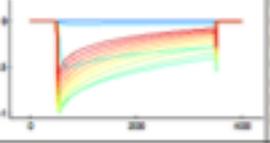
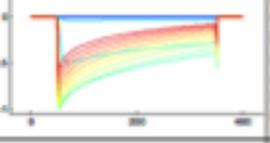
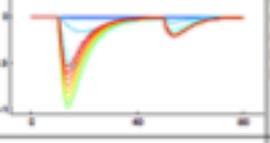


370 Ion Channel genes Expressed in the brain



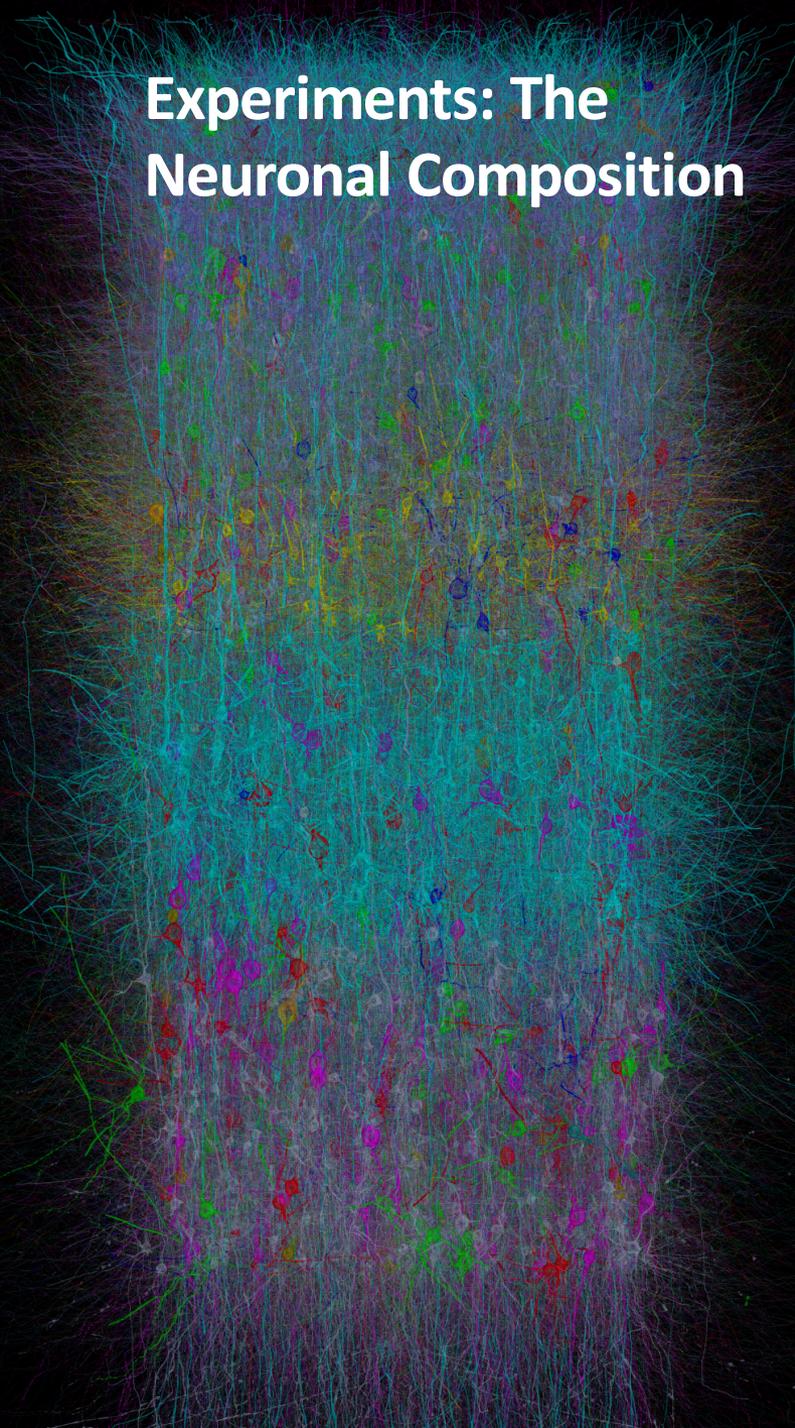
# HH Ion Channel Models

Ion channels	Kinetics	Parameters
Slow potassium (Kp) Ref: Details:		Ion = K <sup>+</sup> Rev = -85mV g = m/h $m_{\infty} = 1/(1 + \exp(-(v+1)/12))$ $m_t = \text{if } (v < -50) :$ $(1.25 + 175.03 * \exp(-v * -0.026)) / \text{qt}$ else : $(1) * 25 + 13 * \exp(-v * 0.026) / \text{qt}$ $h_{\infty} = 1/(1 + \exp((v+54)/11))$ $h_t = (360 + (1010 + 24 * (v + 55)) * \exp(-(v+75)/48) ^ 2) / \text{qt}$
Fast potassium (Kf) Ref: Details:		Ion = K <sup>+</sup> Rev = -85mV g = m/h $m_{\infty} = 1/(1 + \exp(-(v+0)/19))$ $m_t = (0.34 + 0.92 * \exp(-(v+71)/59) ^ 2) / \text{qt}$ $h_{\infty} = 1/(1 + \exp(-(v+66)/10))$ $h_t = (8 + 49 * \exp(-(v+73)/23) ^ 2) / \text{qt}$
Kv3.1 potassium (Kv3.1) Ref: Details:		Ion = K <sup>+</sup> Rev = -85mV g = m $m_{\infty} = 1/(1 + \exp((v - 18.700) / (1 - 9.700)))$ $m_t = 0.2 * 20.00 / (1 + \exp((v - 44.560) / (1 - 44.140)))$
Delayed potassium (Kd) Ref: Details:		Ion = K <sup>+</sup> Rev = -85mV g = m/h $m_{\infty} = 1 - 1/(1 + \exp((v - 43)/8))$ $m_t = 0.6$ $h_{\infty} = 1/(1 + \exp((v - 67)/7.3))$ $h_t = 1500$
m current (Im) Ref: Details:		Ion = K <sup>+</sup> Rev = -85mV g = m $m_{\infty} = m_p / (m_{\infty} + m_p) \quad m_{\infty} = 1.0 / (m_{\infty} + m_p)$ $m_p = 3.3e-3 * \exp(2.5 * 0.04 * (v - 35))$ $m_{\infty} = 3.3e-3 * \exp(-2.5 * 0.04 * (v - 35))$
Small conductance (Sk) Ref: Details:		Ion = K <sup>+</sup> Rev = -85mV g = m $m_{\infty} = 1.0 / (1.0 + (0.00043 / ca) ^ 4.6)$ $m_t = 1.0$
Hyperpolarization activated (h) Ref: Details:		Ion = K <sup>+</sup> Rev = -85mV g = m $m_{\infty} = m_p / (m_{\infty} + m_p) \quad m_{\infty} = 1.0 / (m_{\infty} + m_p)$ $m_p = 0.001 * 6.43 * (v + 154.9) / (\exp((v + 154.9) / 11.9) - 1)$ $m_{\infty} = 0.001 * 193 * \exp(v / 53.1)$

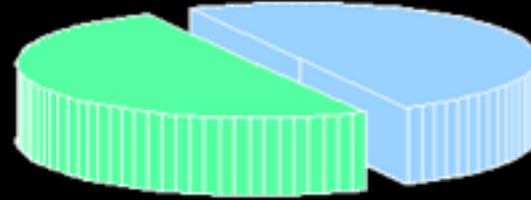
Ion channels	Kinetics	Parameters
Fast sodium (NaFa) Ref: Details:		Ion = K <sup>+</sup> Rev = -85mV g = m/h $m_{\infty} = m_p / (m_{\infty} + m_p) \quad m_{\infty} = 1.0 / (m_{\infty} + m_p)$ $h_{\infty} = h_p / (h_{\infty} + h_p) \quad h_t = 1.0 / (h_{\infty} + h_p)$ $m_p = 0.182 * (v - 38) / (1 - \exp(-(v - 38) / 6))$ $m_{\infty} = 0.124 * (v - 38) / (1 - \exp(-(v - 38) / 6))$ $h_p = (-0.015 * (v - 66)) / (1 - \exp((v - 66) / 6))$ $h_{\infty} = (-0.015 * (v - 66)) / (1 - \exp((v - 66) / 6))$
Fast sodium (NaFs) Ref: Details:		Ion = K <sup>+</sup> Rev = -85mV g = m/h $m_{\infty} = m_p / (m_{\infty} + m_p) \quad m_{\infty} = 1.0 / (m_{\infty} + m_p)$ $h_{\infty} = h_p / (h_{\infty} + h_p) \quad h_t = 1.0 / (h_{\infty} + h_p)$ $m_p = 0.182 * (v - 32) / (1 - \exp(-(v - 32) / 6))$ $m_{\infty} = 0.124 * (v - 32) / (1 - \exp(-(v - 32) / 6))$ $h_p = (-0.015 * (v - 60)) / (1 - \exp((v - 60) / 6))$ $h_{\infty} = (-0.015 * (v - 60)) / (1 - \exp((v - 60) / 6))$
Persistent sodium (Nap) Ref: Details:		Ion = K <sup>+</sup> Rev = -85mV g = m/h $m_{\infty} = 1.0 / (1 + \exp((v - 52.6) / 4.6))$ $h_{\infty} = 1.0 / (1 + \exp((v - 48.8) / 10))$ $m_p = 6.0 / (m_{\infty} + m_p) \quad h_t = 1.0 / (h_{\infty} + h_p)$ $m_{\infty} = 0.182 * (v - 38) / (1 - \exp(-(v - 38) / 6))$ $m_p = 0.124 * (v - 38) / (1 - \exp(-(v - 38) / 6))$ $h_p = 0.0065 * (v + 17) / (1 - \exp((v + 17) / 4.6))$ $h_{\infty} = 0.0065 * (v + 17) / (1 - \exp((v + 17) / 4.6))$
Generic Calcium (Ca) Ref: Details:		Ion = K <sup>+</sup> Rev = -85mV g = m/h $m_{\infty} = m_p / (m_{\infty} + m_p) \quad m_{\infty} = 1.0 / (m_{\infty} + m_p)$ $h_{\infty} = h_p / (h_{\infty} + h_p) \quad h_t = 1.0 / (h_{\infty} + h_p)$ $m_p = (0.055 * (27 - v)) / (\exp((27 - v) / 3.8) - 1)$ $m_{\infty} = (0.94 * \exp(-75 - v) / 17)$ $h_p = (0.00043 * \exp(-13 - v) / 50)$ $h_{\infty} = (0.0065 * \exp(-v - 15) / 28) + 1$
High voltage activated (CaHVA) Ref: Details:		Ion = K <sup>+</sup> Rev = -85mV g = m/h $m_{\infty} = m_p / (m_{\infty} + m_p) \quad m_{\infty} = 1.0 / (m_{\infty} + m_p)$ $h_{\infty} = h_p / (h_{\infty} + h_p) \quad h_t = 1.0 / (h_{\infty} + h_p)$ $m_p = (0.055 * (27 - v)) / (\exp((27 - v) / 3.8) - 1)$ $m_{\infty} = (0.94 * \exp(-75 - v) / 17)$ $h_p = (0.00043 * \exp(-13 - v) / 50)$ $h_{\infty} = (0.0065 * \exp(-v - 15) / 28) + 1$
Low voltage activated (CaLVA) Ref: Details:		Ion = K <sup>+</sup> Rev = -85mV g = m/h $m_{\infty} = 1.0 / (1 + \exp((v - 30.0) / 6))$ $m_t = (5.0 + 20.0 / (1 + \exp((v - 25.0) / 5))) / \text{qt}$ $h_{\infty} = 1.0 / (1 + \exp((v - 40.0) / 6.4))$ $h_t = (20.0 + 50.0 / (1 + \exp((v - 40.0) / 7))) / \text{qt}$
Stochastic potassium (K) Ref: Details:		



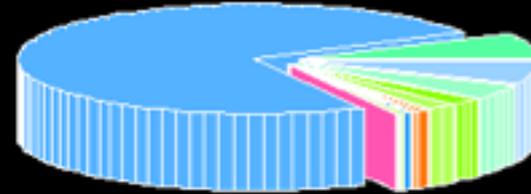
# Experiments: The Neuronal Composition



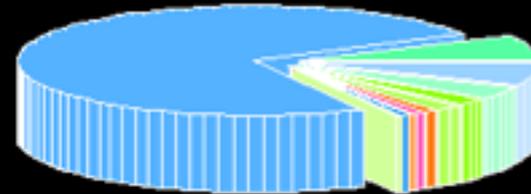
Layer 1



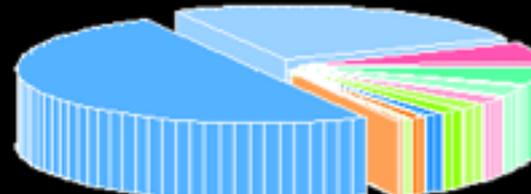
Layer 2



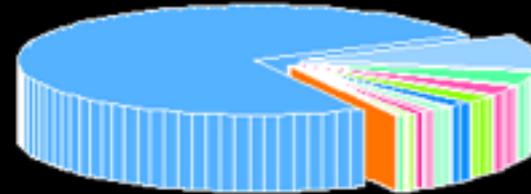
Layer 3



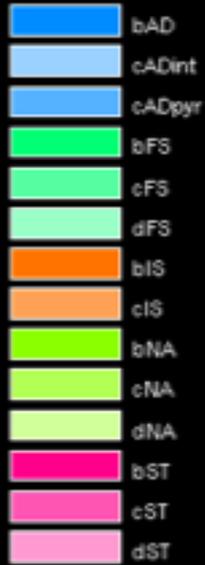
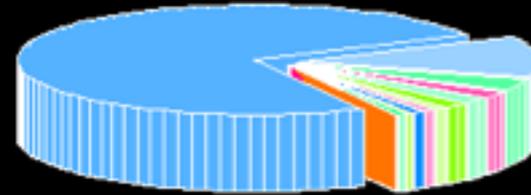
Layer 4



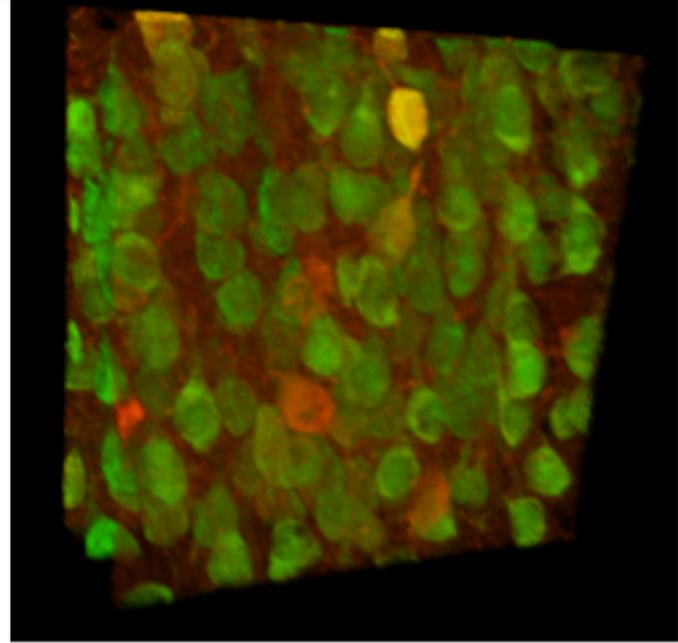
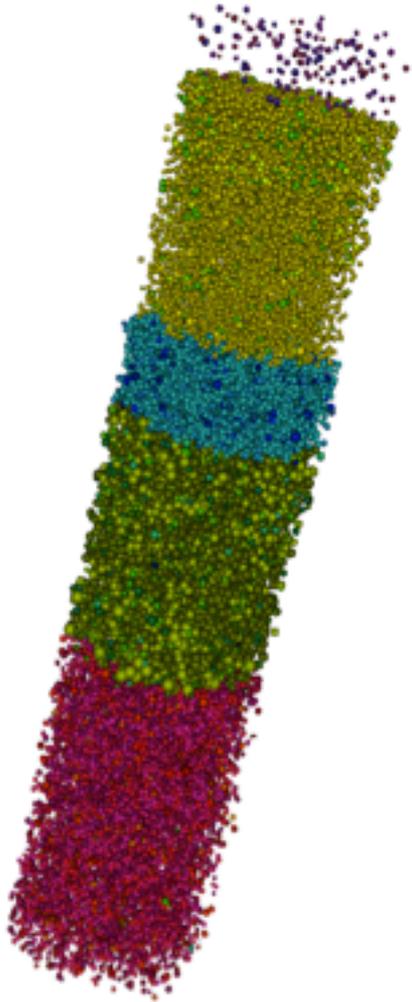
Layer 5



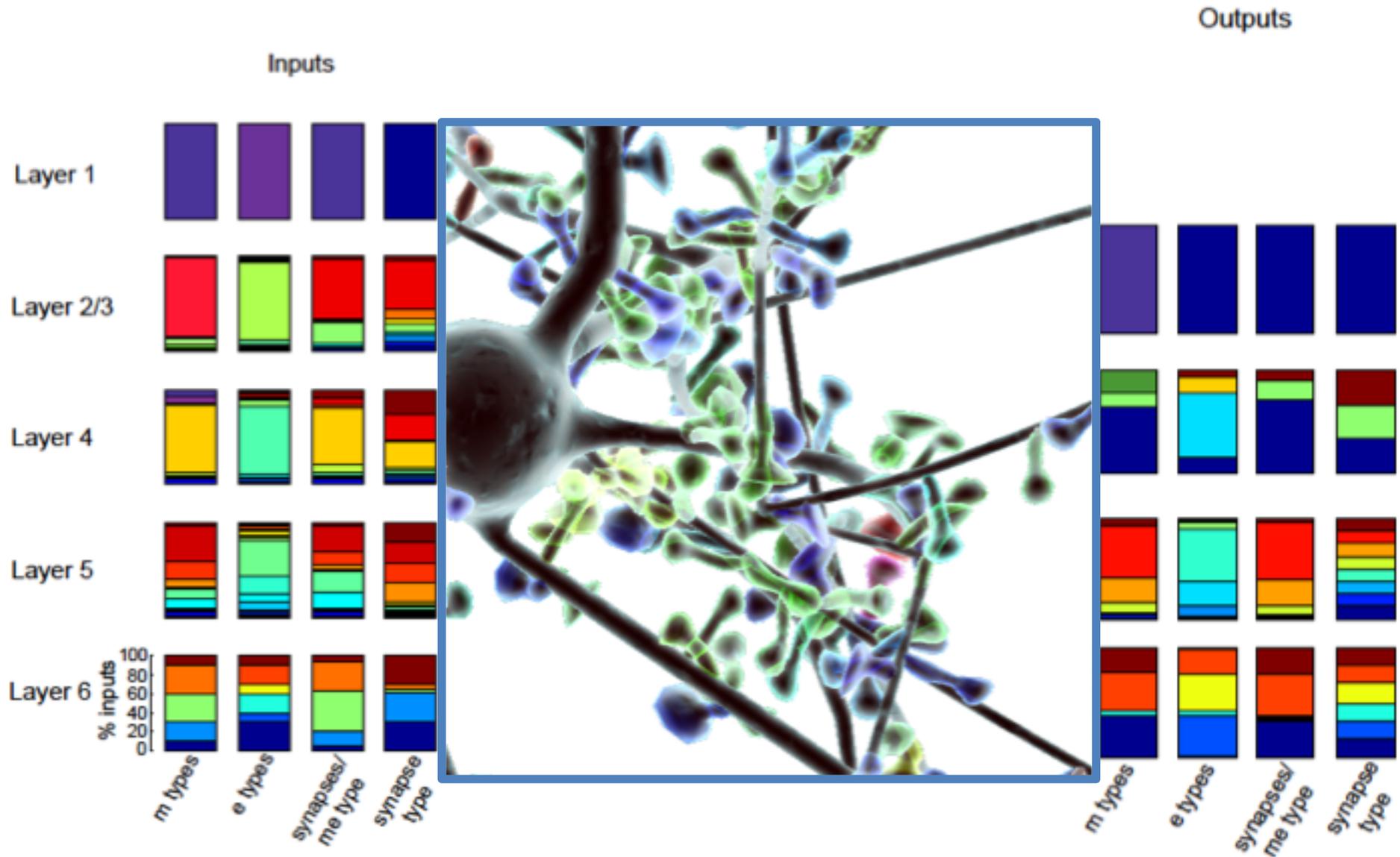
Layer 6



# Neuronal Densities and E:I Proportions

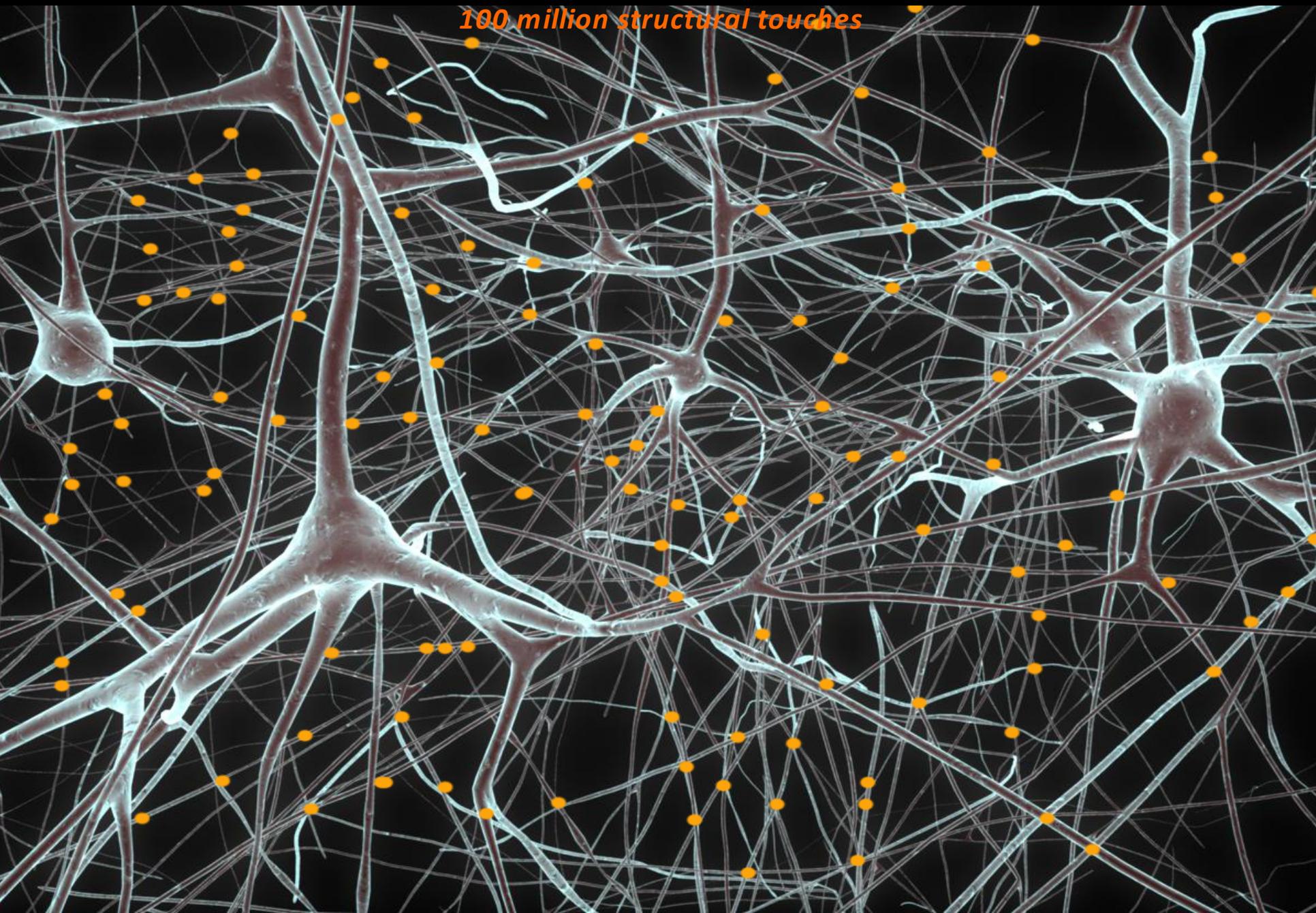


# Predicting extrinsic connectivity



# Building: the structural connectome

*100 million structural touches*



# Building: the functional connectome

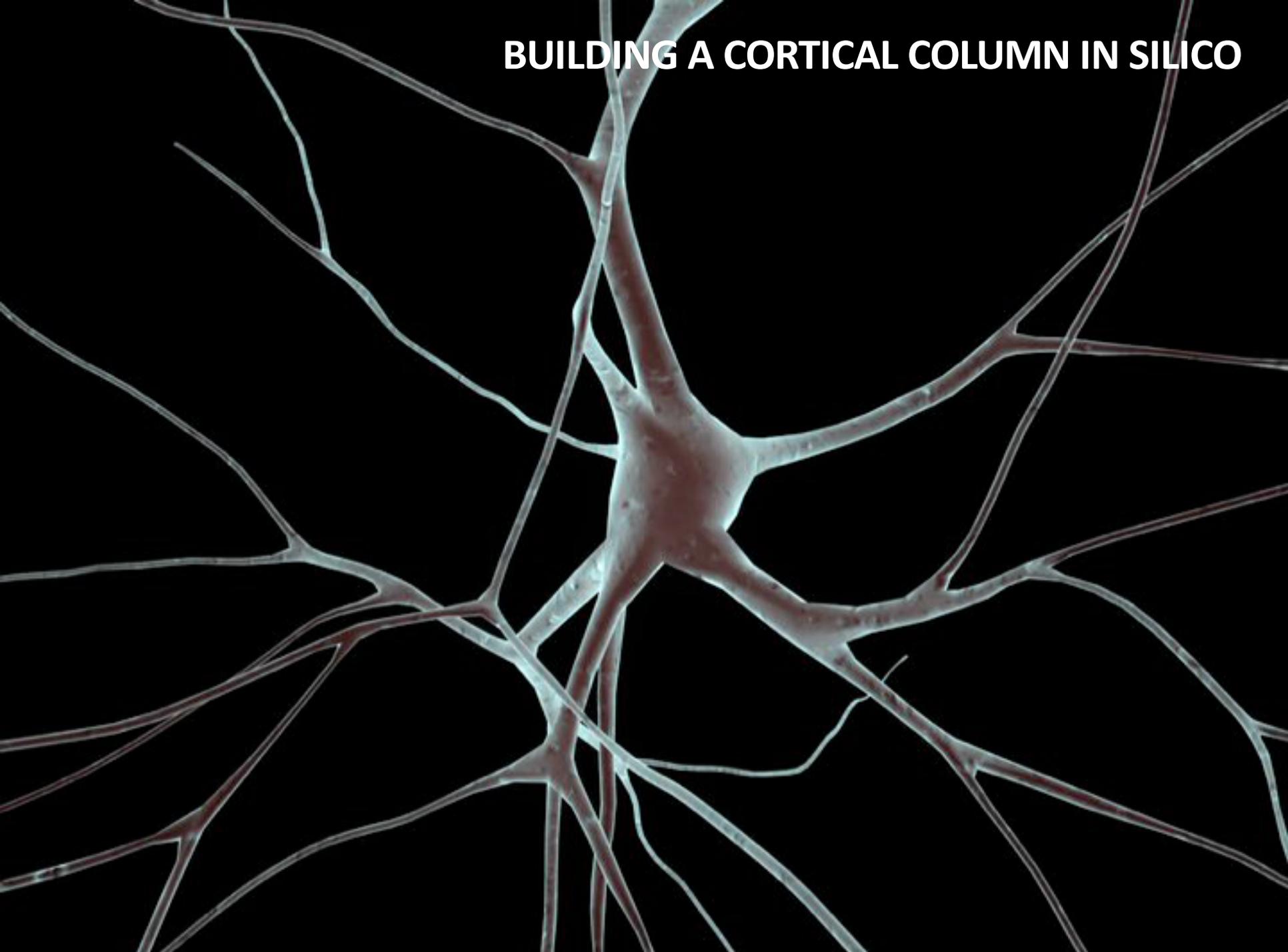
*Apply the Neuron-Neuron Connection Probability Matrix*



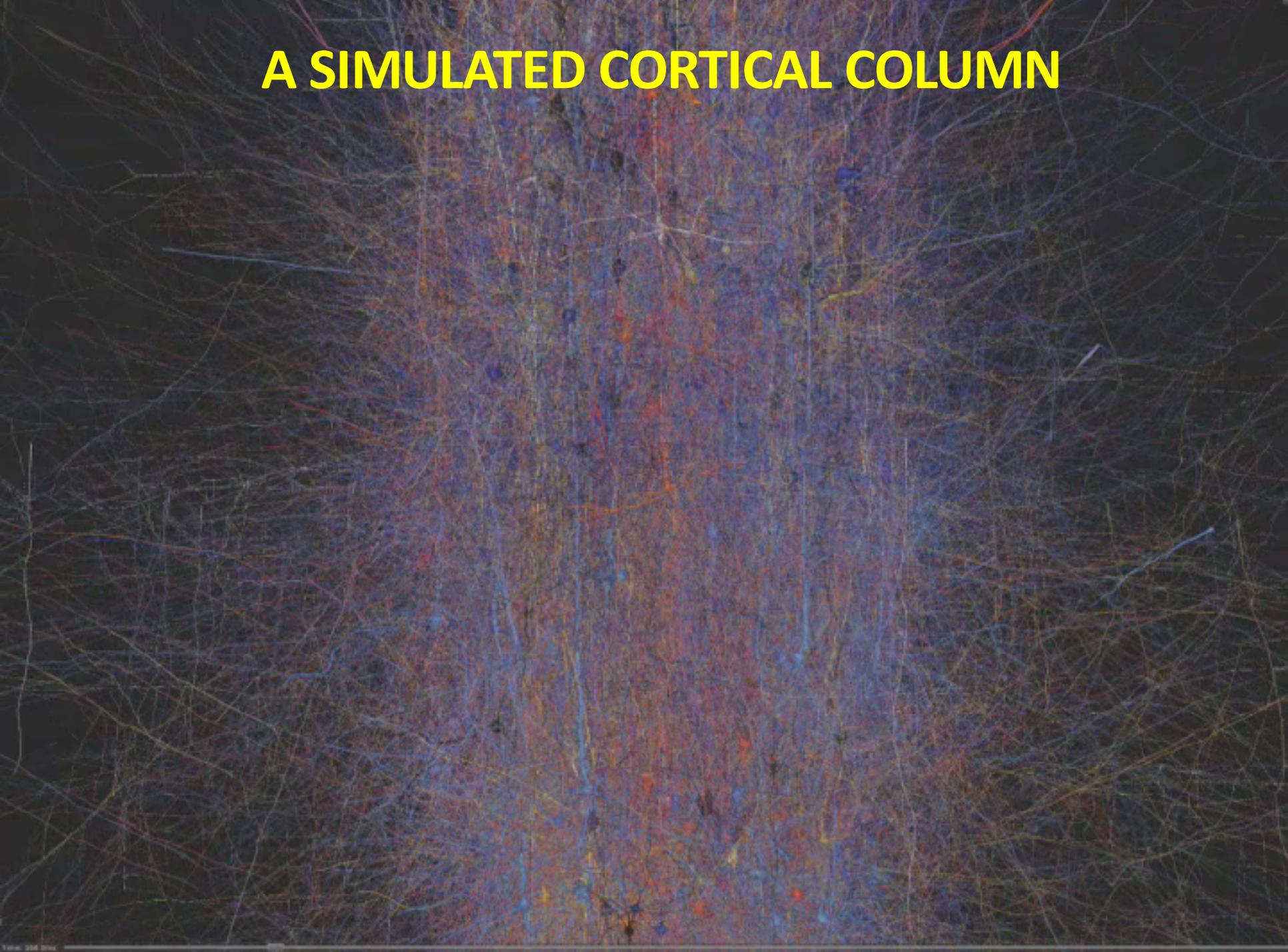
$P_{connect}$

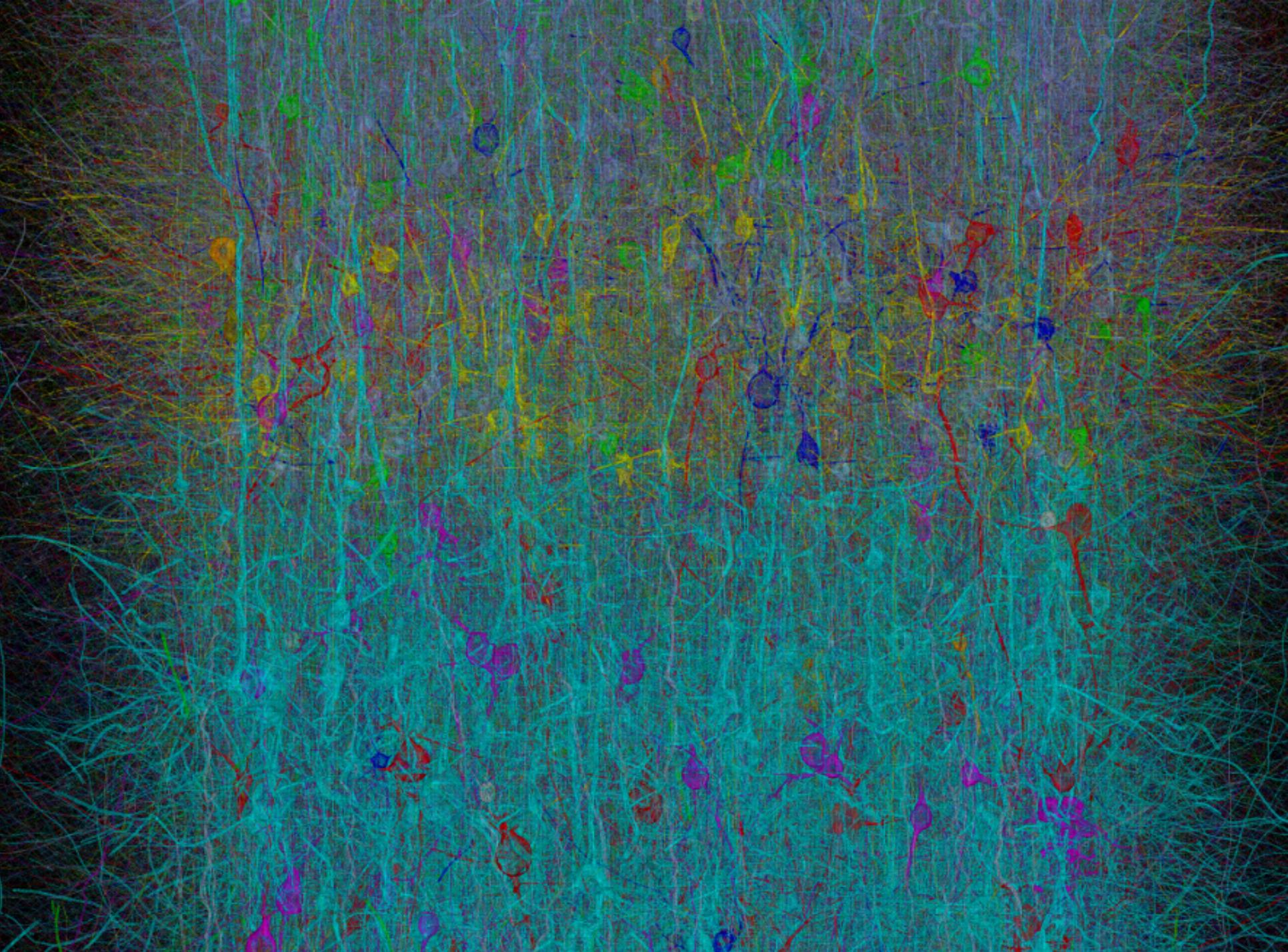
*10 million synapses*

# BUILDING A CORTICAL COLUMN IN SILICO

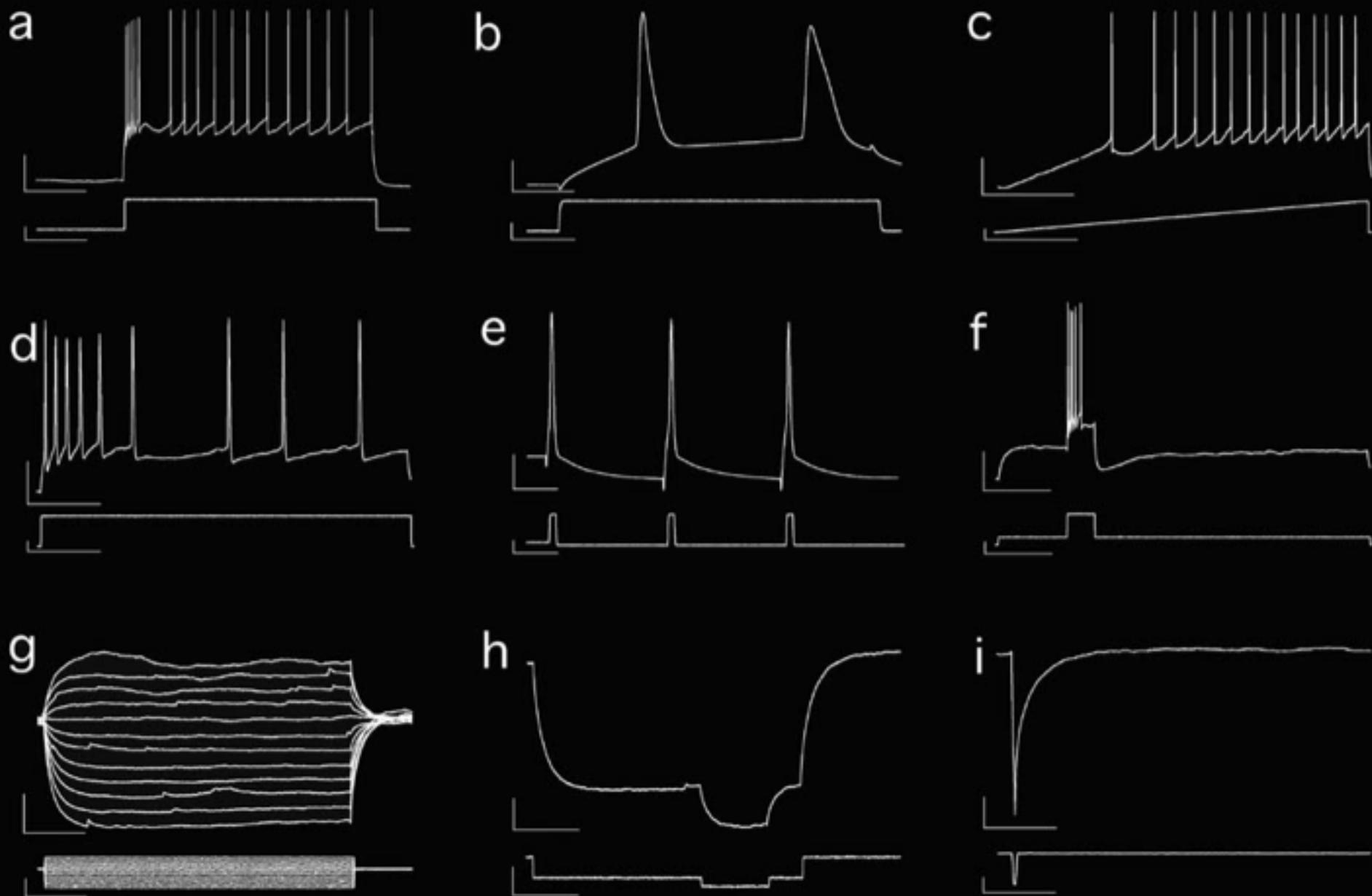


# A SIMULATED CORTICAL COLUMN

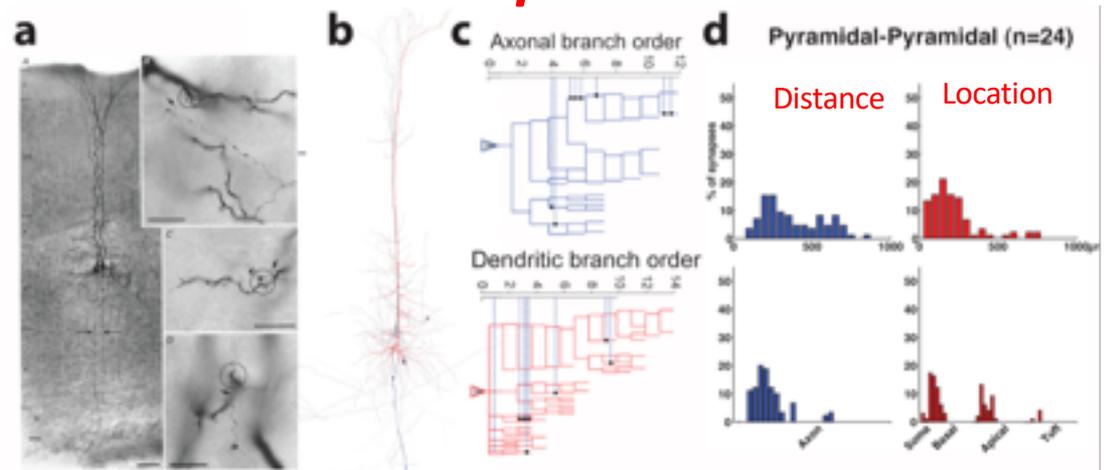
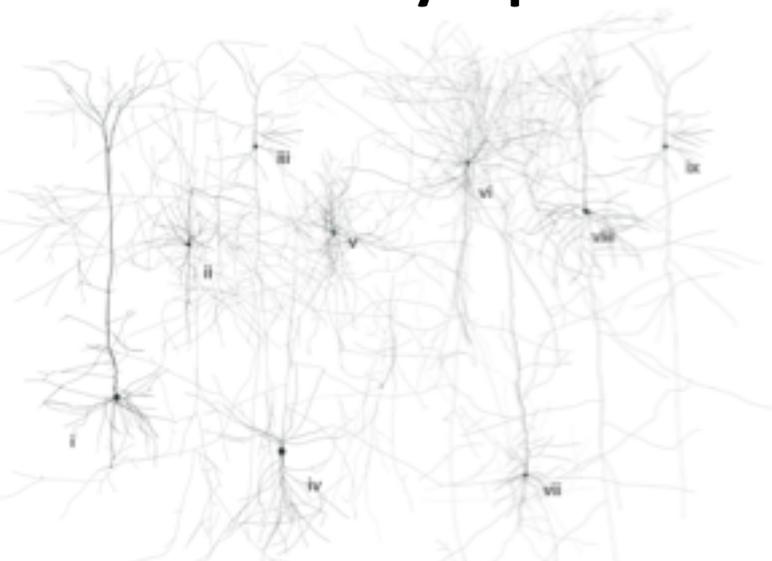




# The Neurons: probing the eTypes

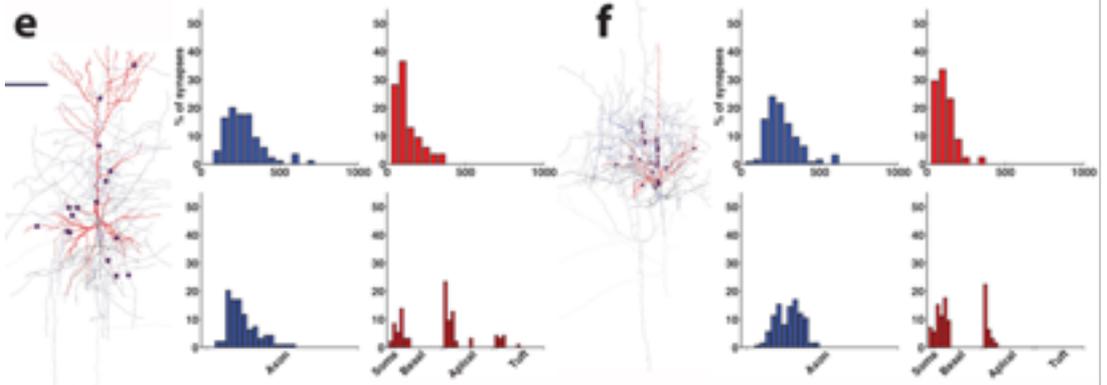


# Synaptic Numbers & Locations - *Experiments*



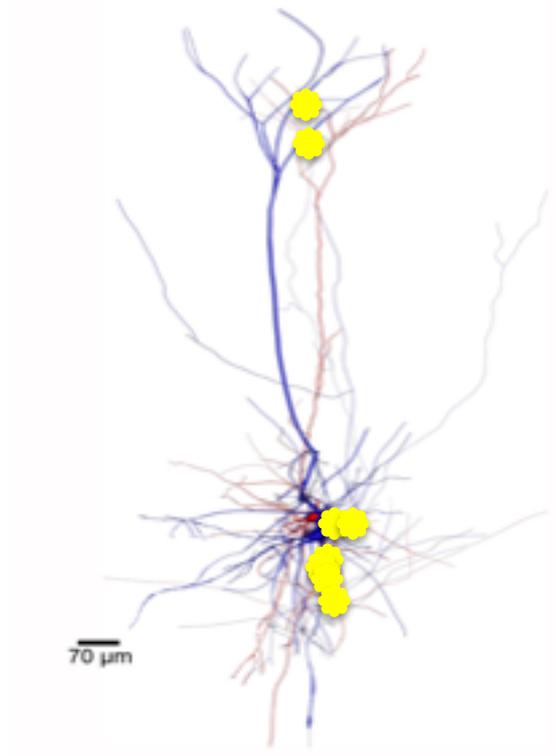
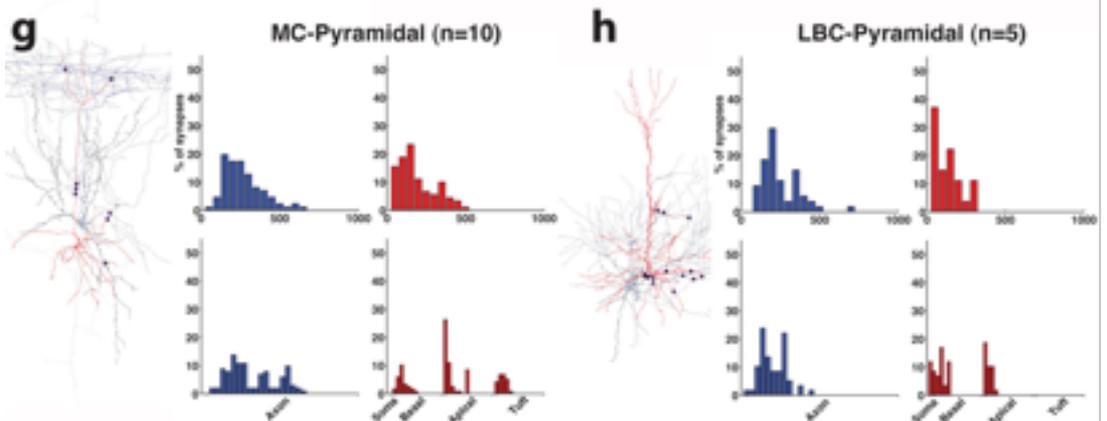
BTC-Pyramidal (n=6)

SBC-Pyramidal (n=6)

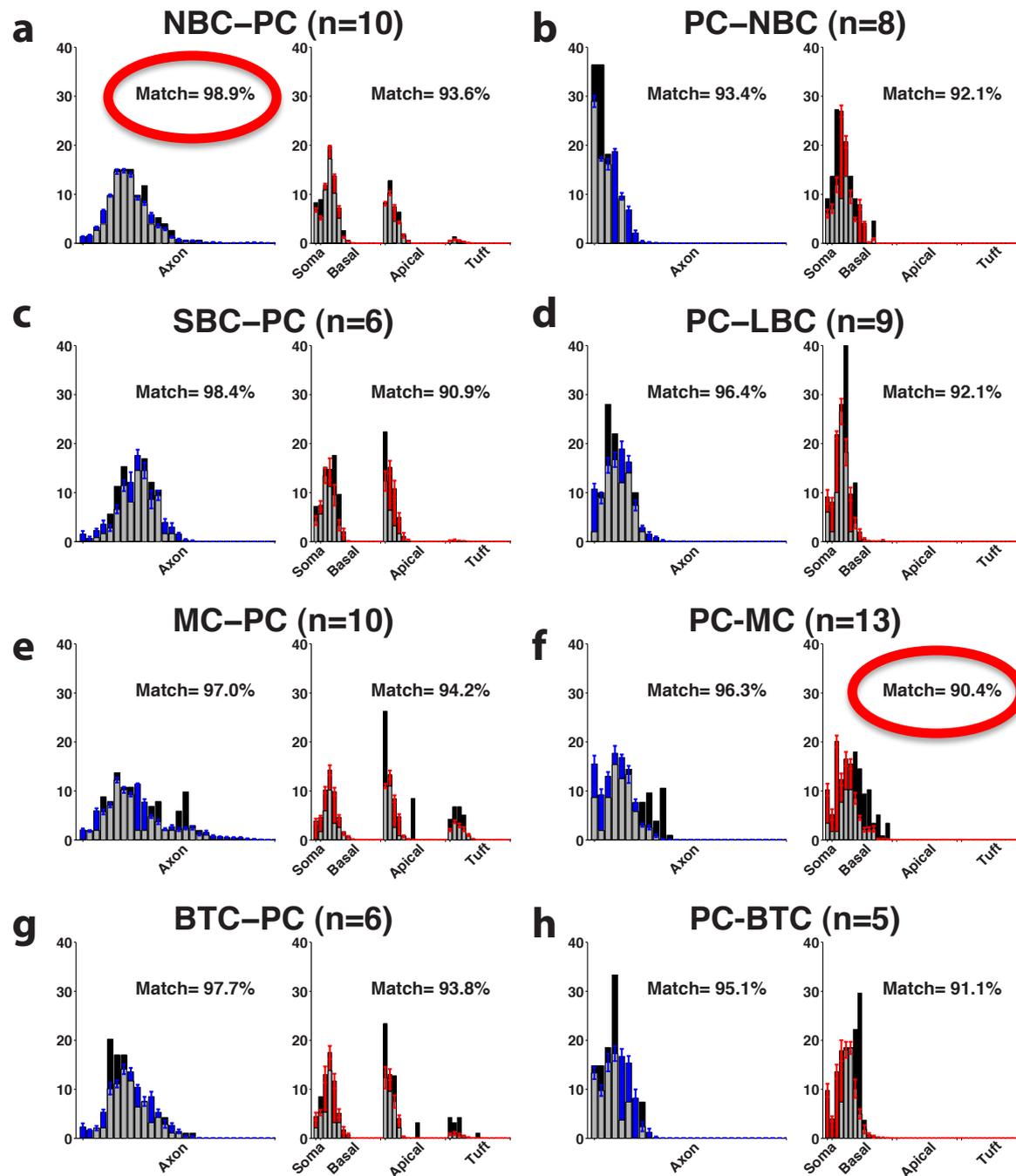


MC-Pyramidal (n=10)

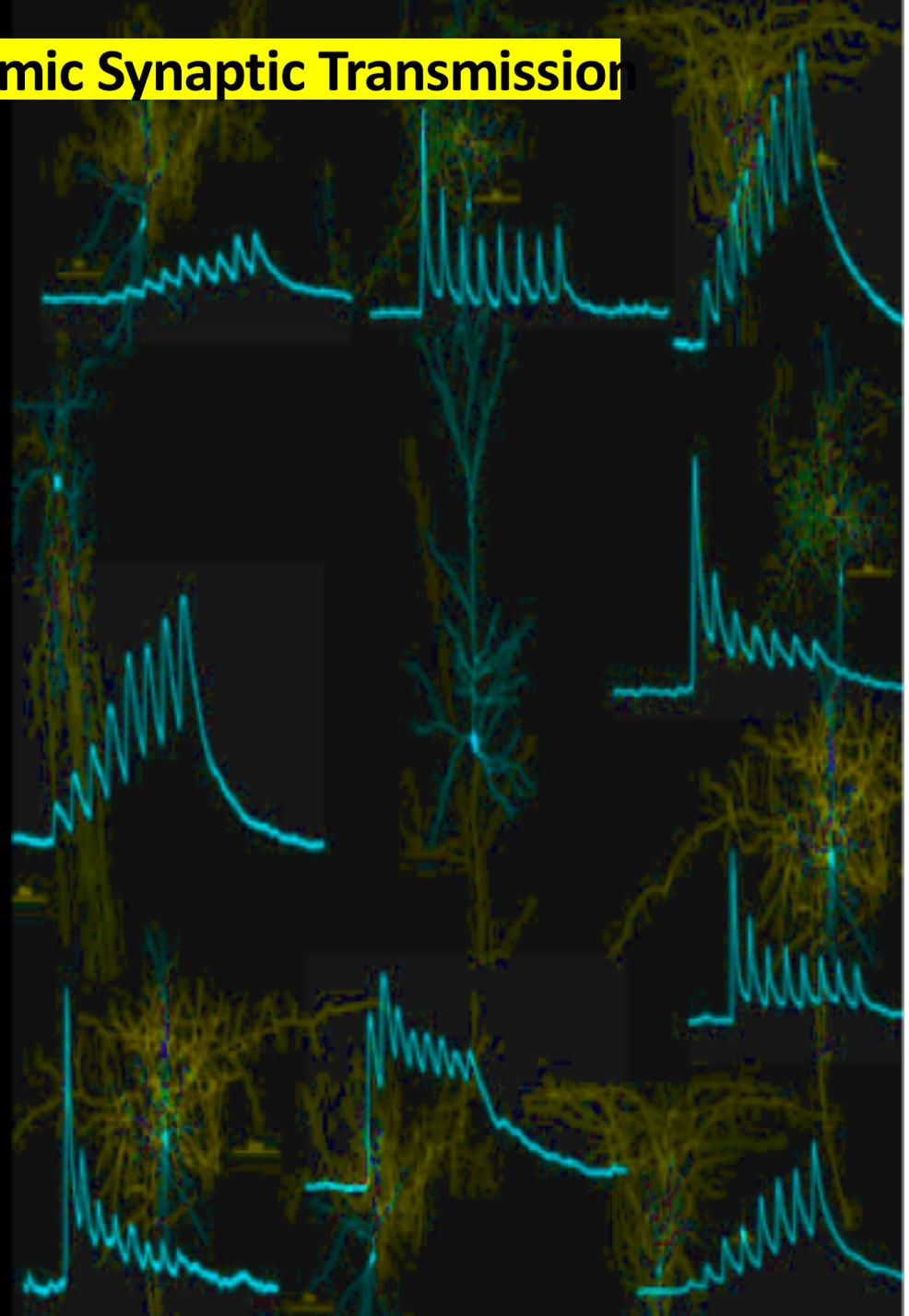
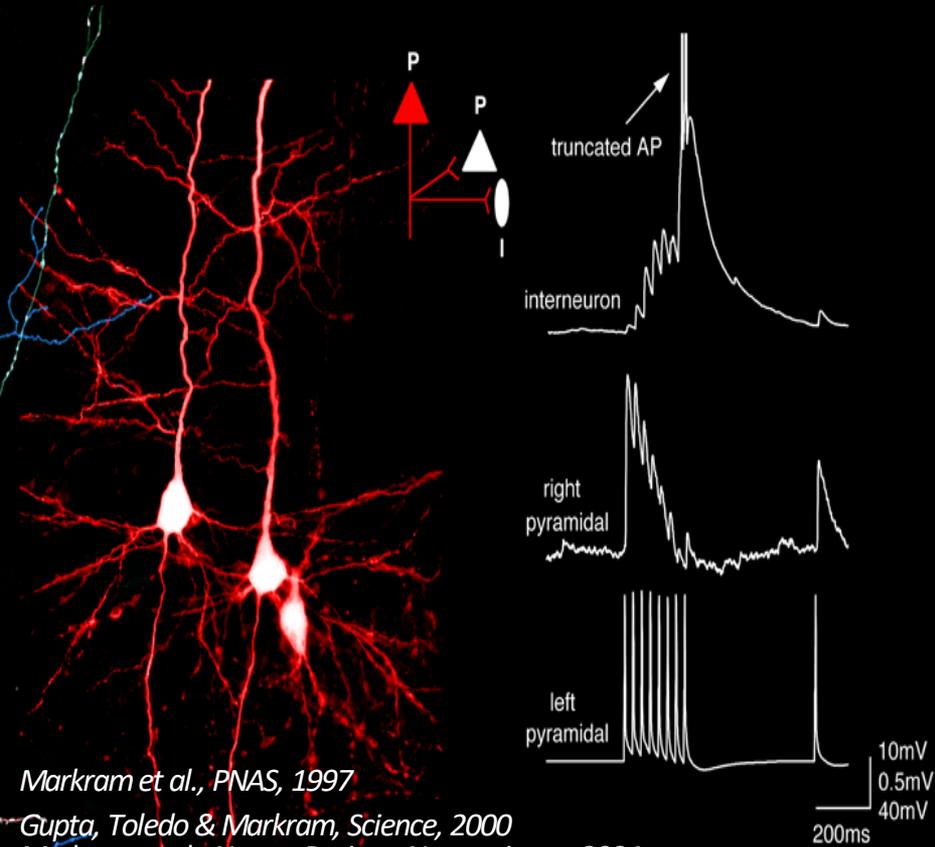
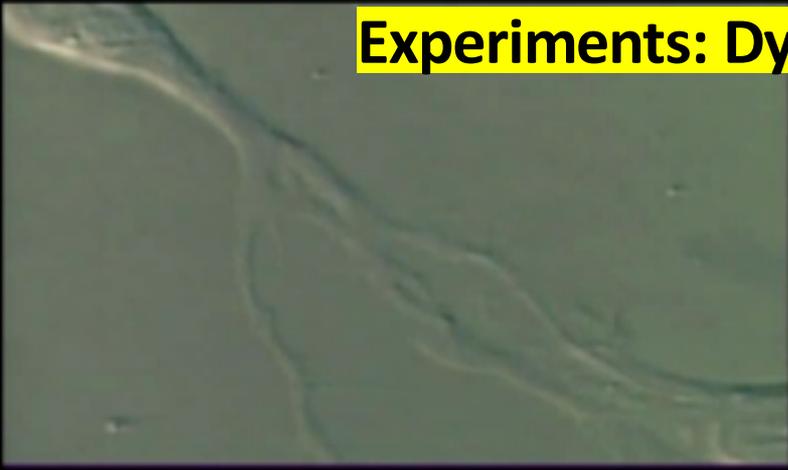
LBC-Pyramidal (n=5)



# Discovery: Synaptic Positions in the Model MATCH those in Experiments



# Experiments: Dynamic Synaptic Transmission



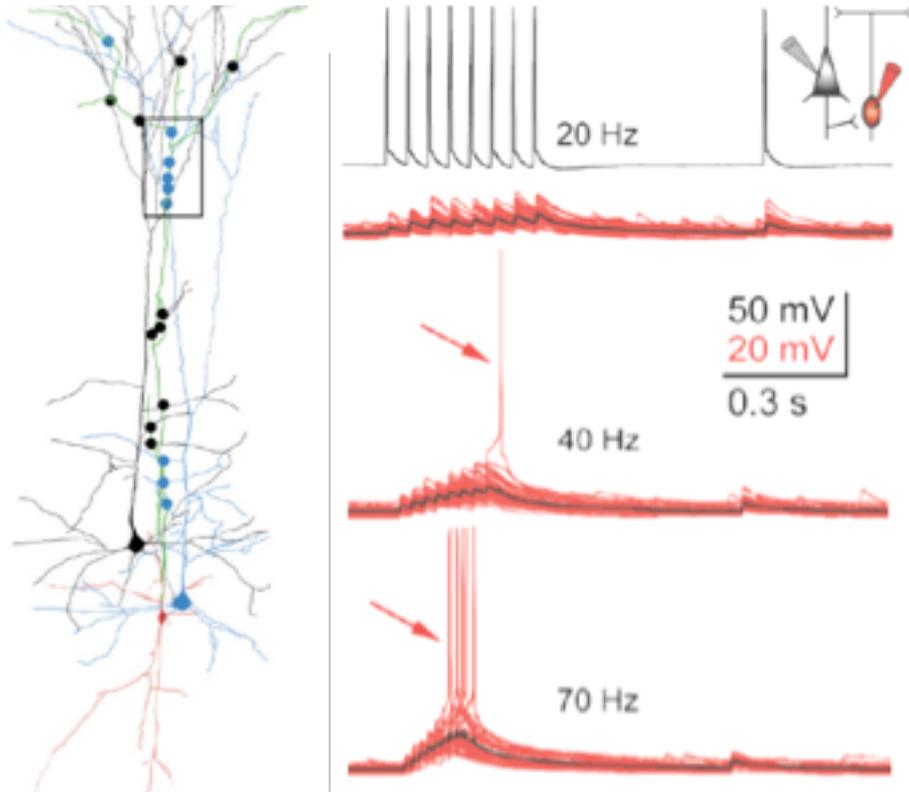
Markram et al., PNAS, 1997

Gupta, Toledo & Markram, Science, 2000

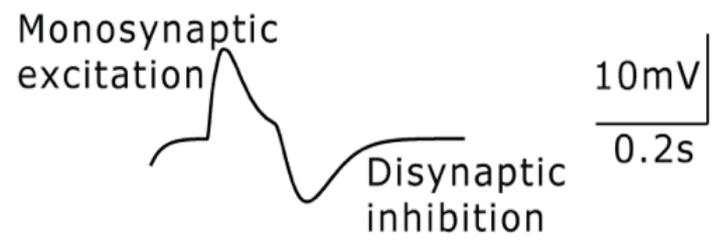
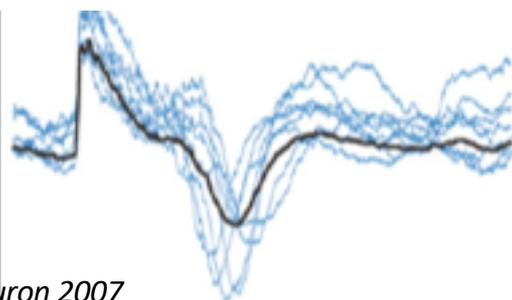
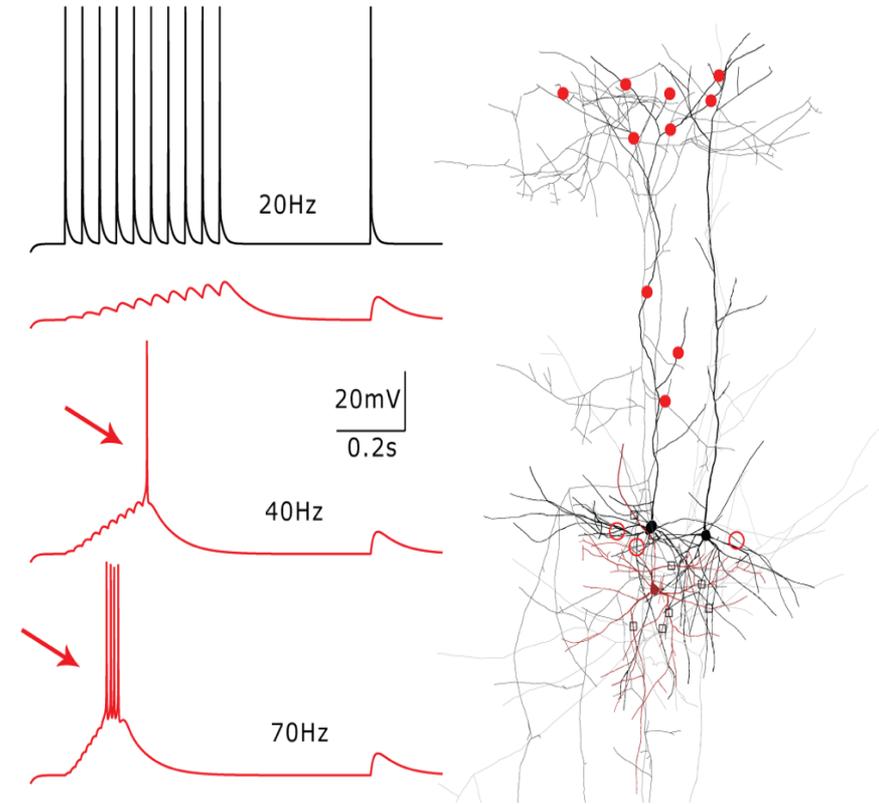
Markram et al., Nature Reviews Neuroscience, 2004

# In silico synaptic recordings match in vitro synaptic recordings

In vitro

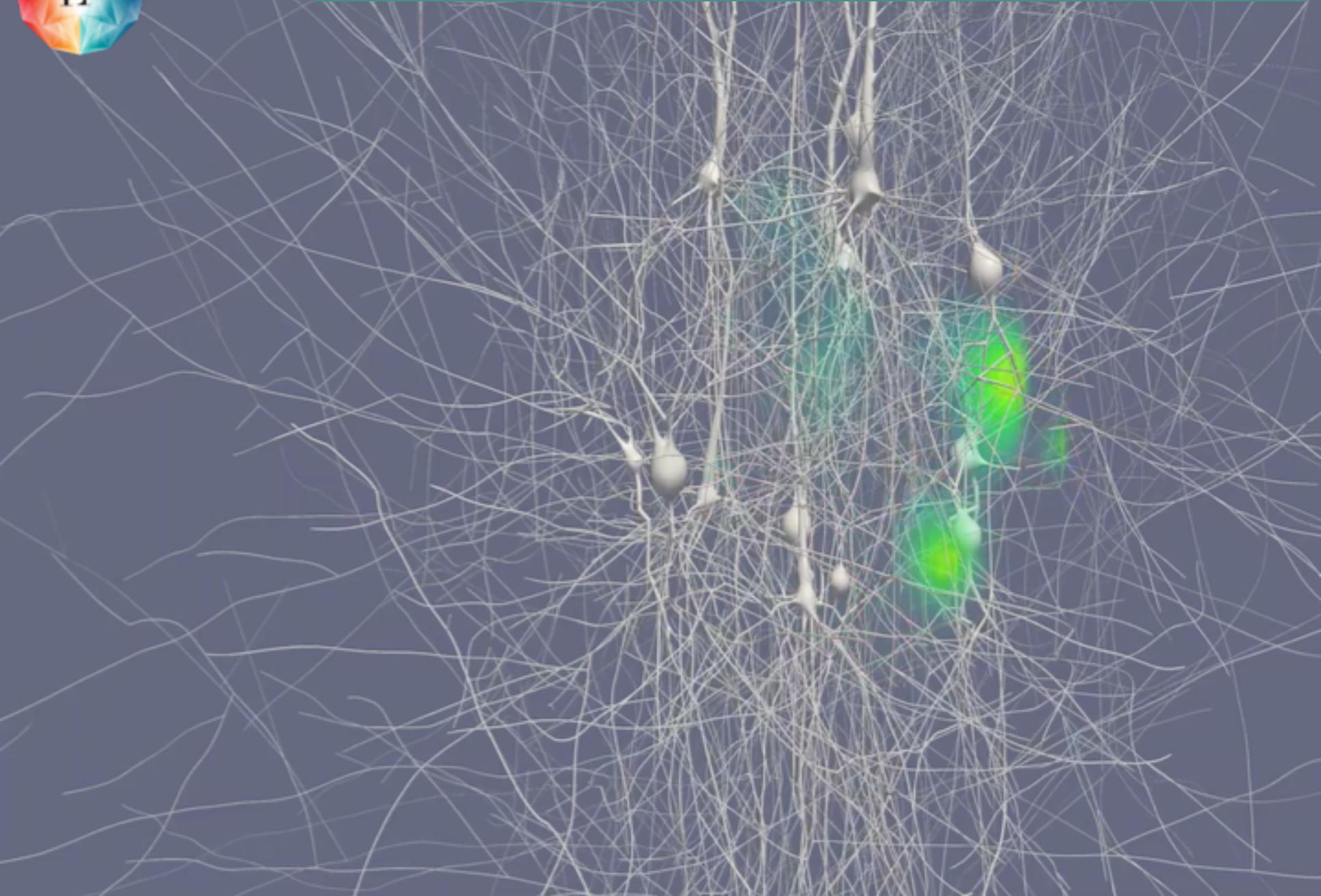


In silico

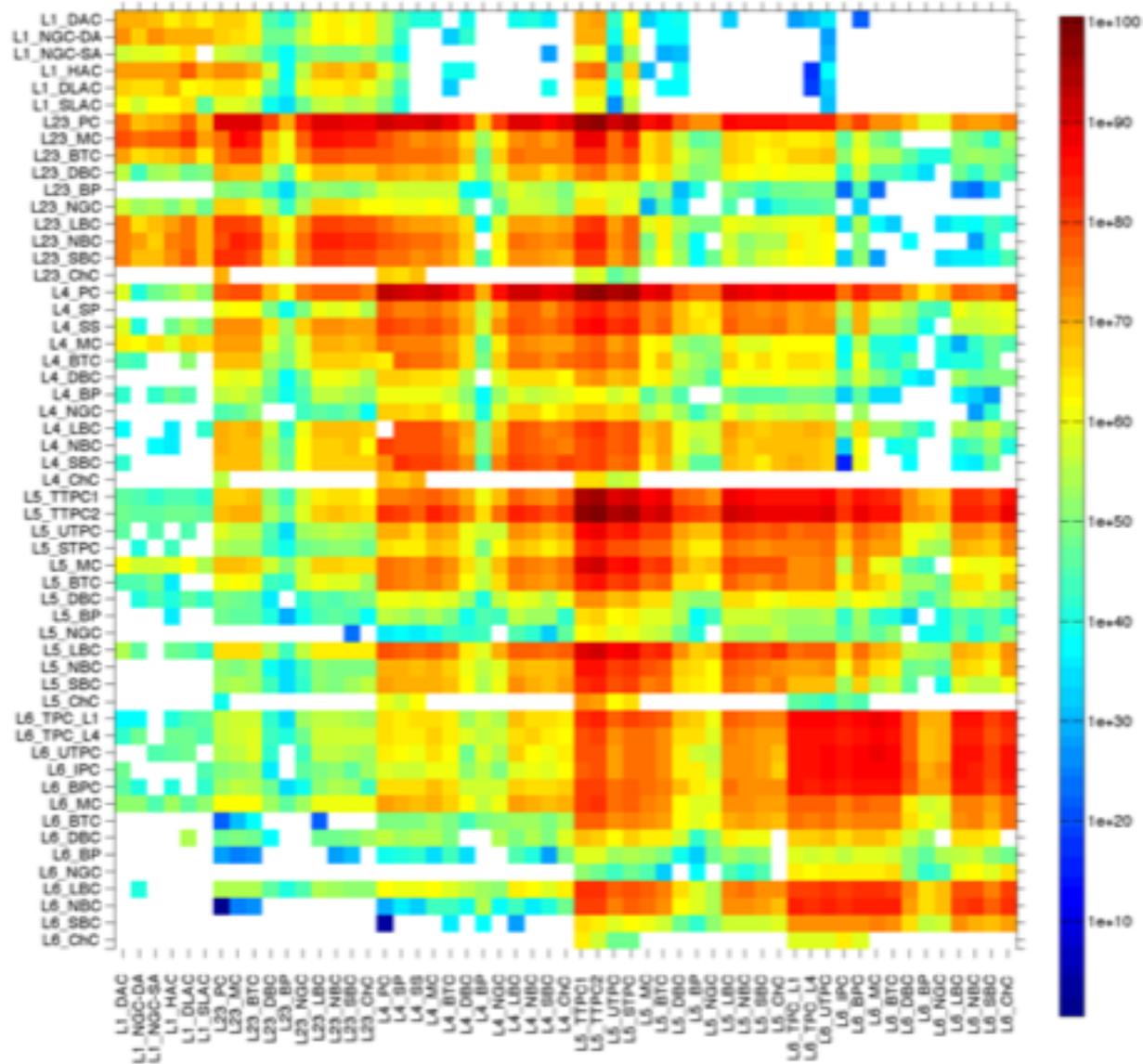


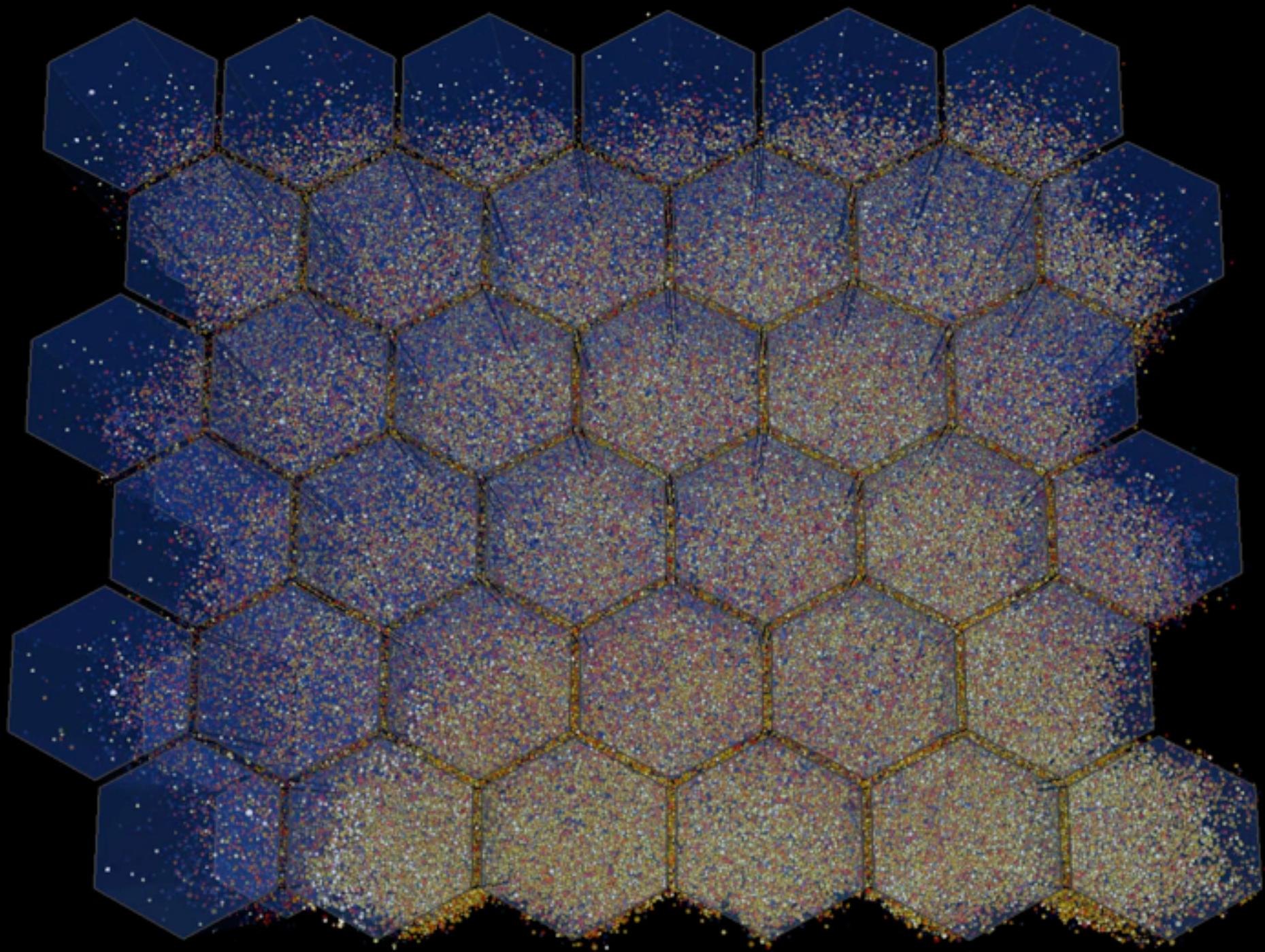


# MODELLED LFPs IN A CORTICAL COLUMN



# Normalized by number of neurons per mType





# NEURONS

$$C_m \frac{dV_m}{dt} = \frac{E_m - V_m}{R_m} + I_{channels} + I_{synapses}$$

$$+ \frac{2(V_{m_{i+1}} - V_m)}{R_{m_{i+1}} + R_m} + \frac{2(V_{m_{i-1}} - V_m)}{R_{m_{i-1}} + R_m}$$

10,000 neurons  
 ~ 4,000,000  
 Electrical compartments  
 (Rall Equations)

# ION CHANNELS

$$\frac{dm}{dt} = \alpha_m(V_m)(1-m) - \beta_m(V_m)m$$

$$\frac{dh}{dt} = \alpha_h(V_m)(1-h) - \beta_h(V_m)h$$

$$I_{channel} = m^4 h g_{channel}(V_m - E_{channel})$$

80,000,000 Ion Channels  
 (Hodgkin-Huxley Equations)

# SYNAPSES

$$\frac{dx}{dt} = \frac{x}{\tau_{rec}} - ux d(t - t_{sp})$$

$$\frac{dy}{dt} = -\frac{y}{\tau_i} - ux d(t - t_{sp})$$

$$\frac{dz}{dt} = \frac{y}{\tau_i} - \frac{z}{\tau_{rec}}$$

$$\frac{du}{dt} = \frac{u}{\tau_{inact}} + U(1-u)d(t - t_{sp})$$

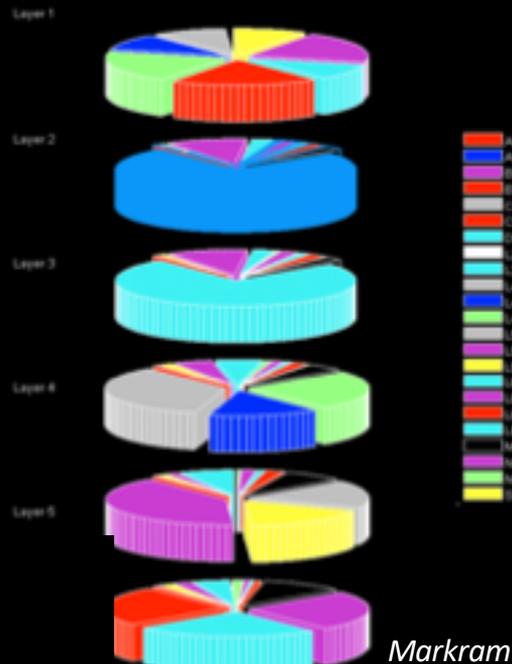
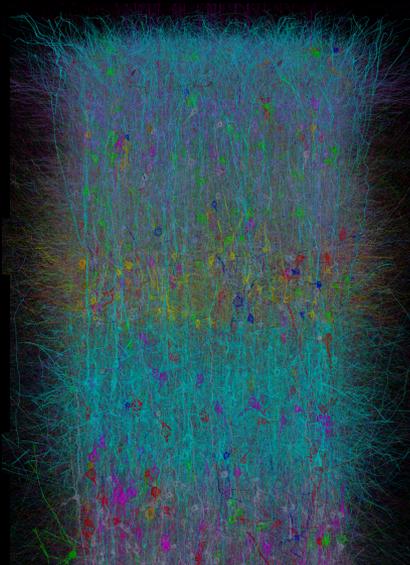
$$I_{synapse}(t) = \sum_j A_{ij} y_j(t)$$

10,000,000 Synapses  
 (Tsodyks-Markram Equations)

MODELS ARE DATA DRIVEN

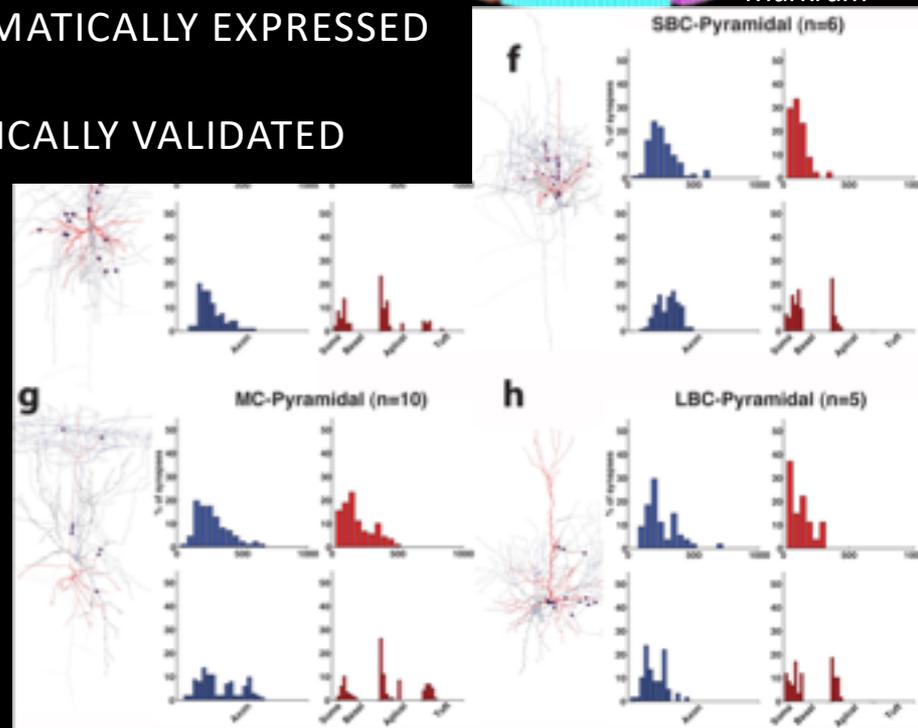
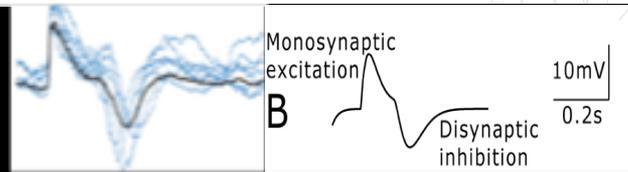
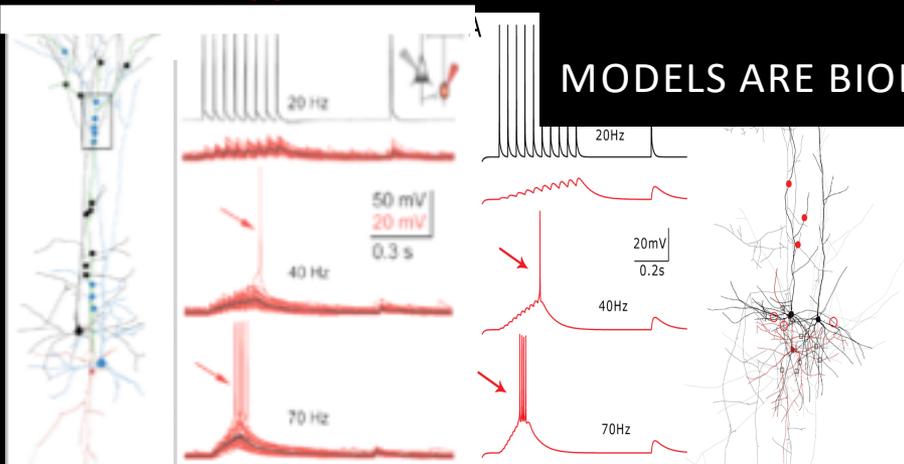
MODELS ARE MATHEMATICALLY EXPRESSED

MODELS ARE BIOLOGICALLY VALIDATED



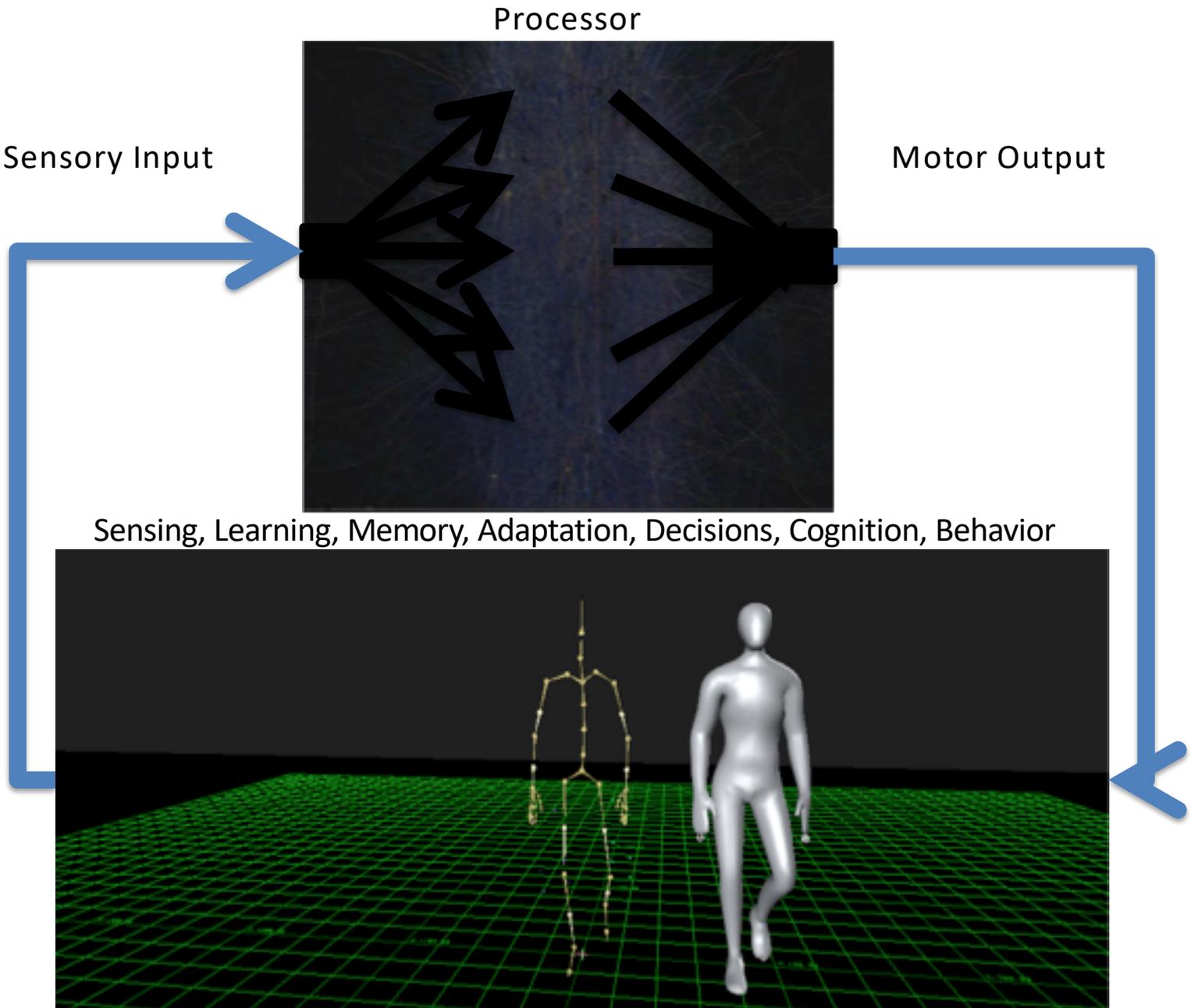
Markram

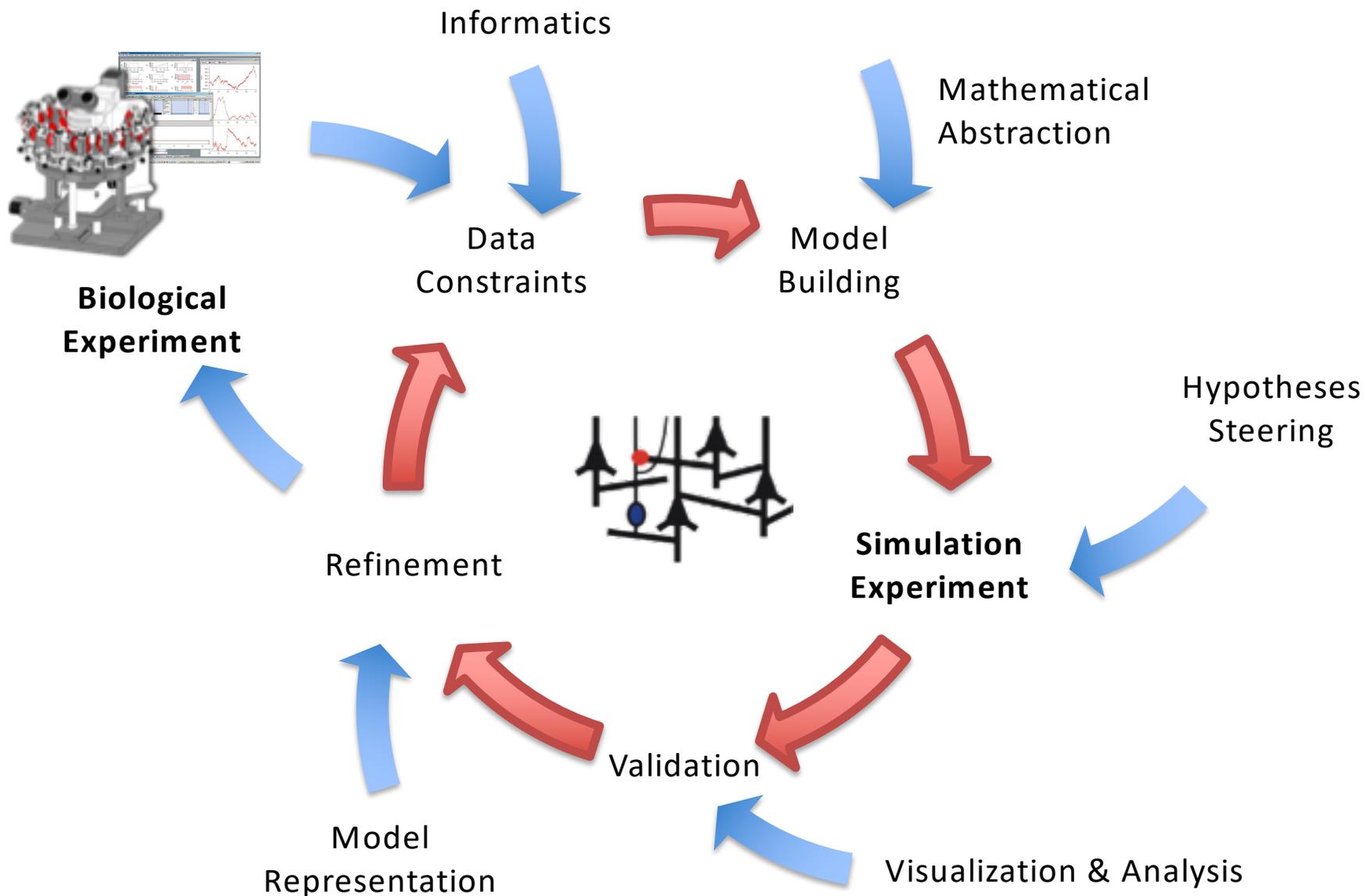
In vitro



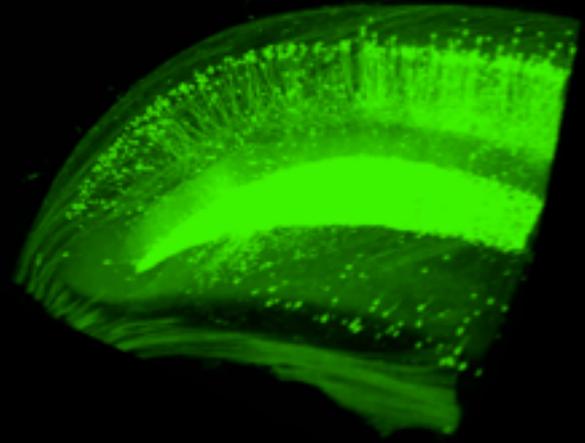
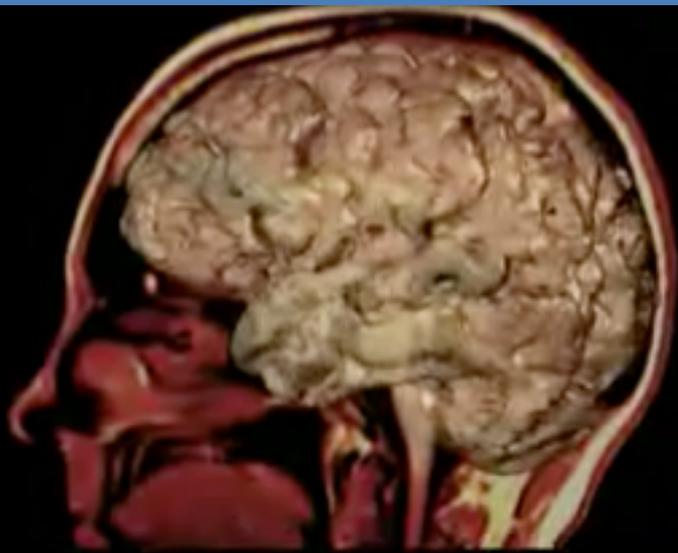


# NEURAL COMPUTATION : CHAIN OF EVENTS LEADING TO COGNITION







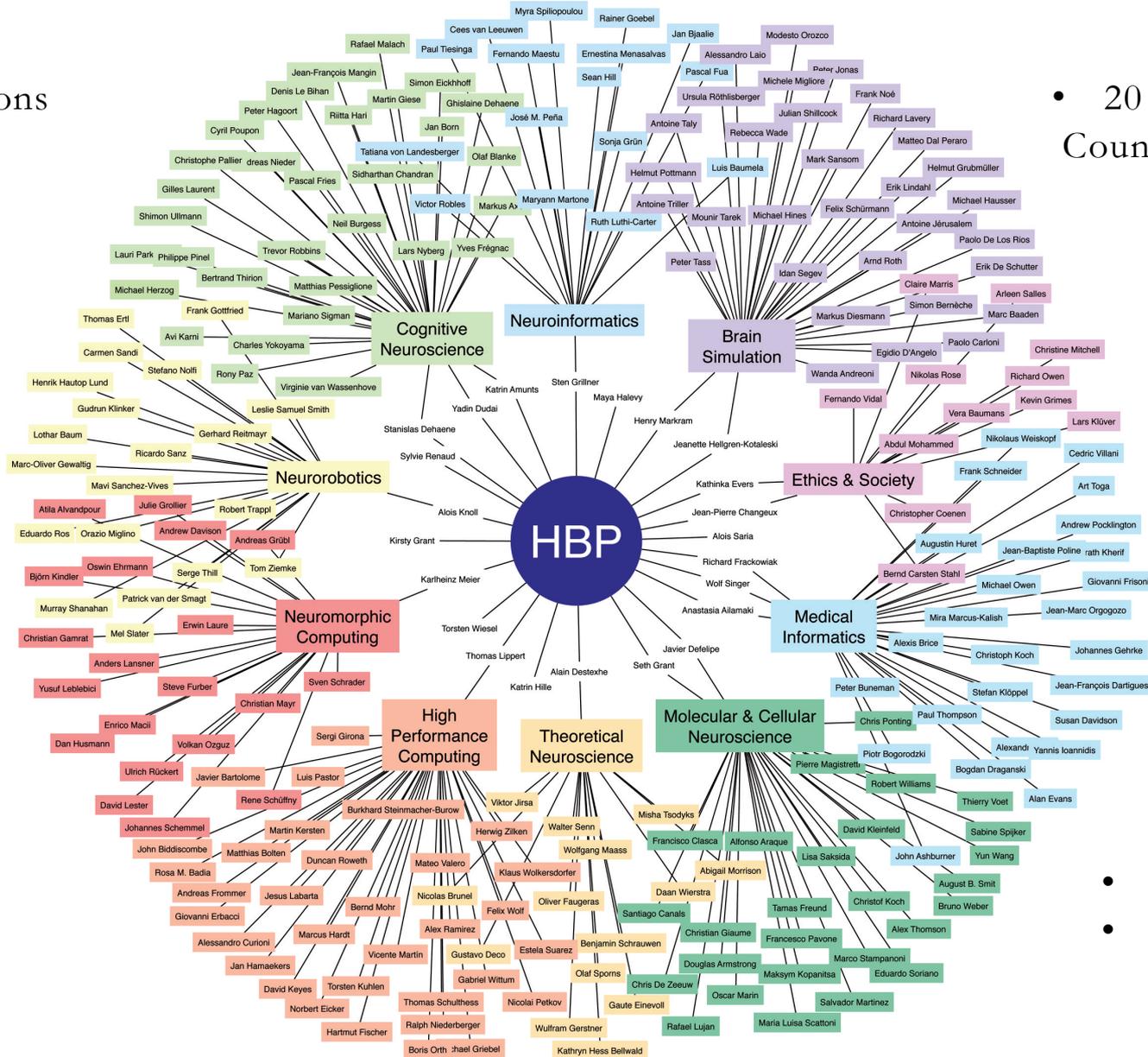




# THE CONSORTIUM

- 80 Institutions
- 150 PIs
- 2000 PhDs

- 20 European Countries



- USA
- Israel
- Japan
- China
- Canada

- 50% Core
- 50% Grants