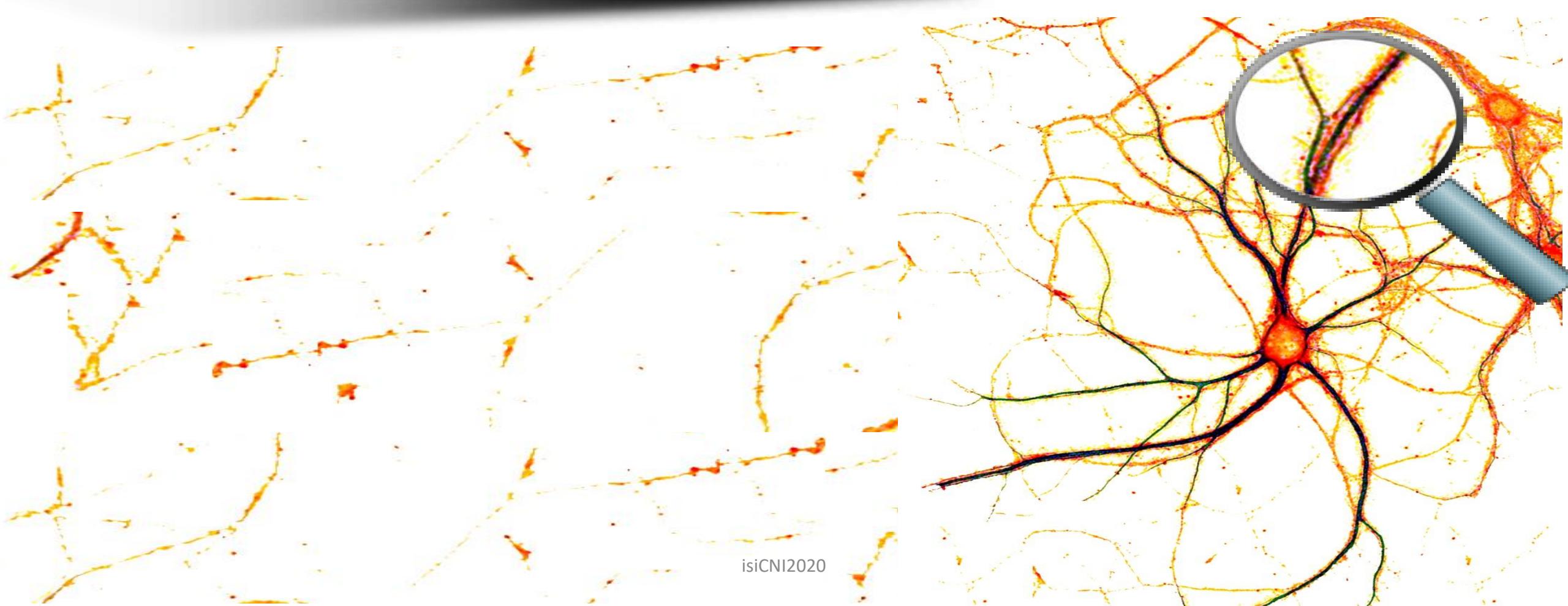


Dendritic computations

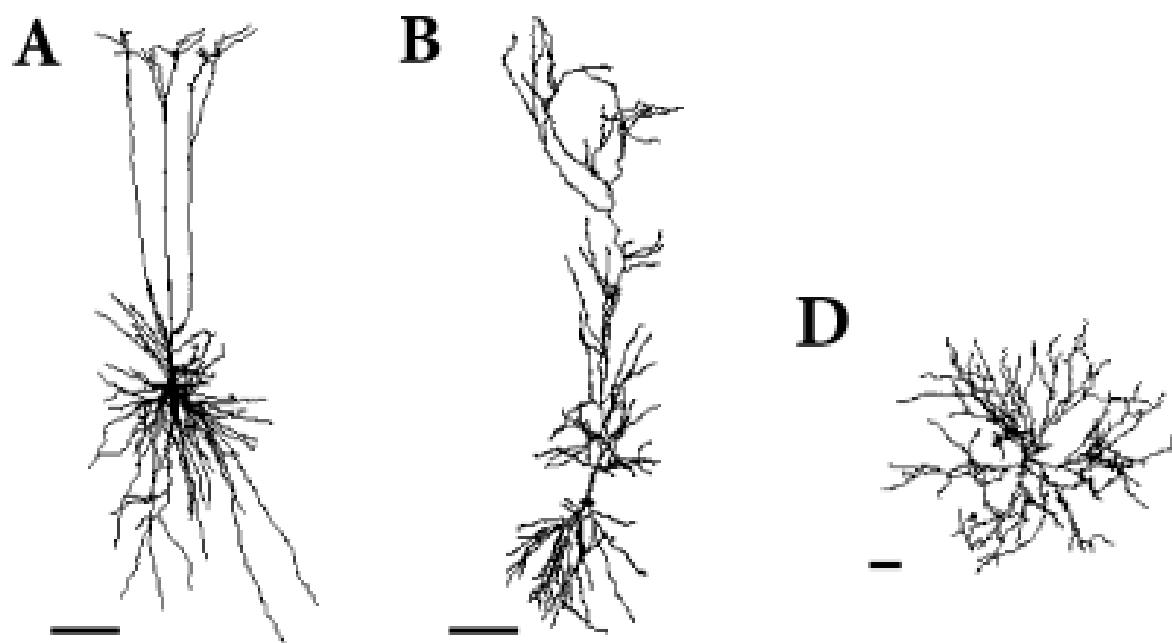
Athanasia Papoutsi, Institute of Molecular Biology and Biotechnology (IMBB), Foundation for Research and Technology-Hellas (FORTH)



Neurons are (also) characterized by their dendritic tree

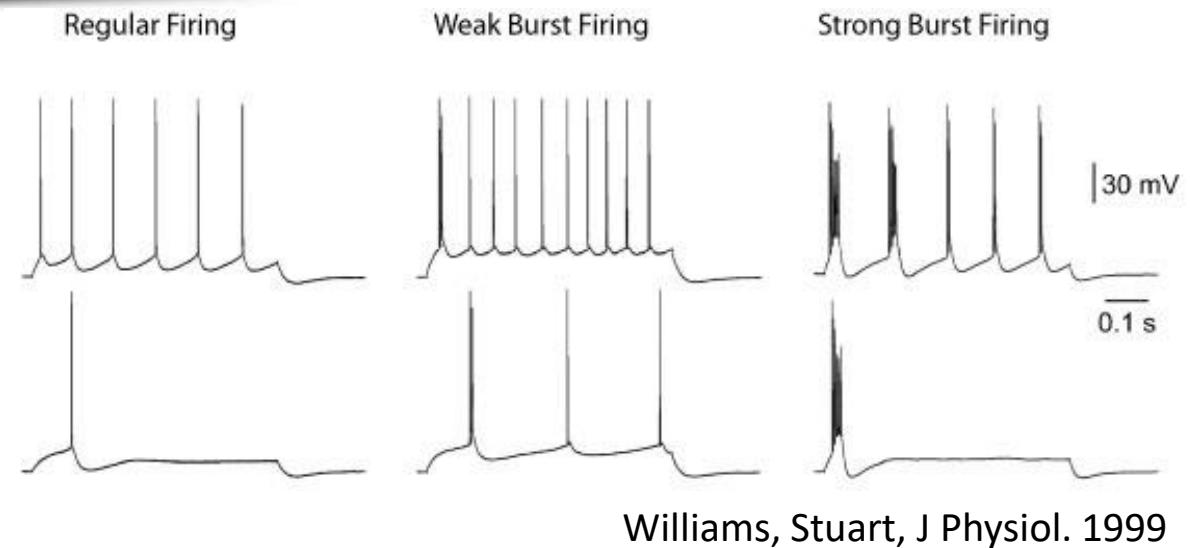
Neuron Types:

1. Classification based on Excitatory or Inhibitory neurotransmission.
2. Classification using gene expression
3. Classification using spiking activity
4. Classification by anatomical features

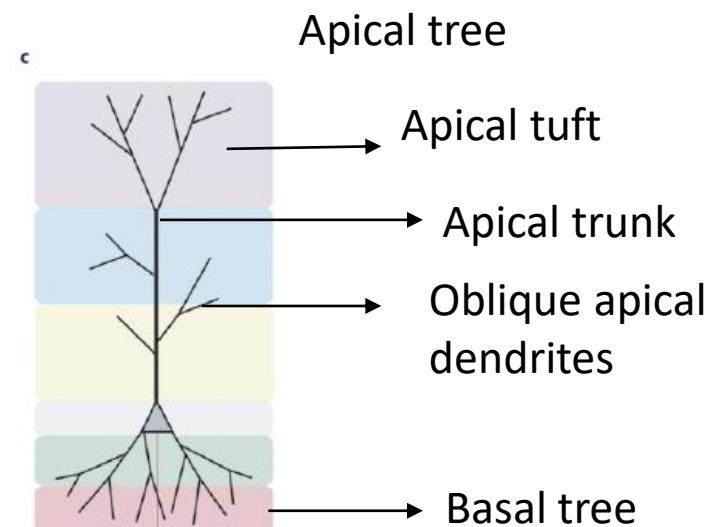


<http://NeuroMorpho.org>

isiCNI2020

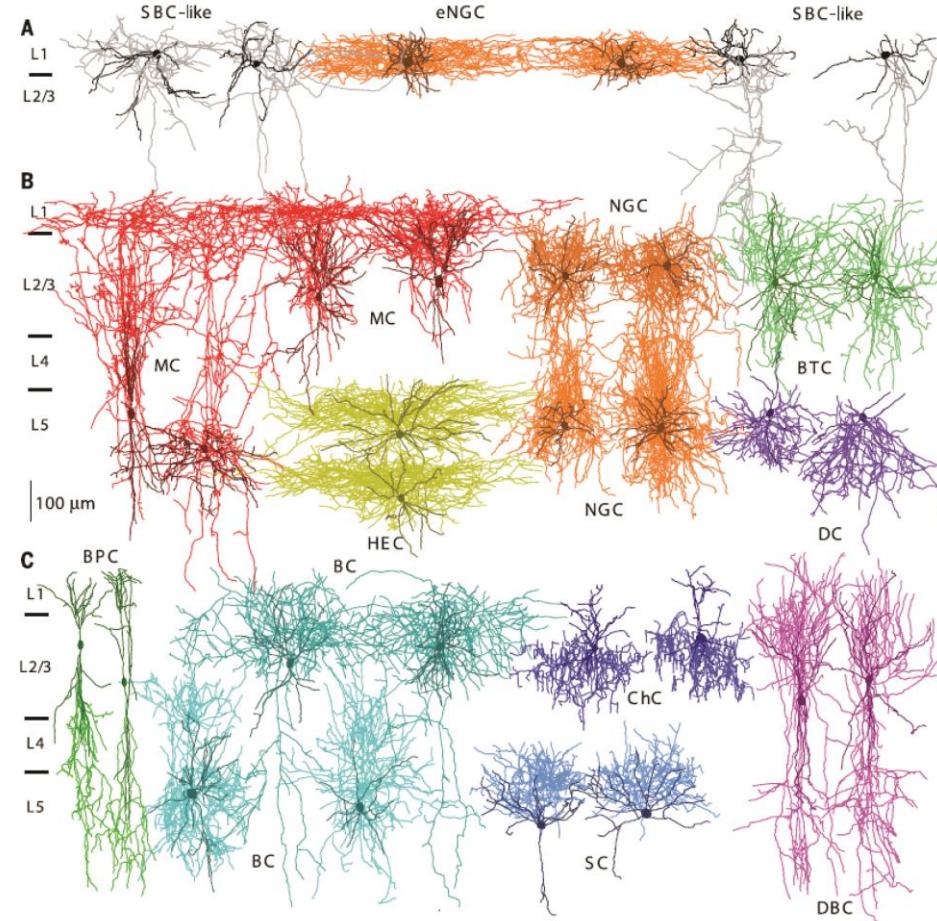


Williams, Stuart, J Physiol. 1999



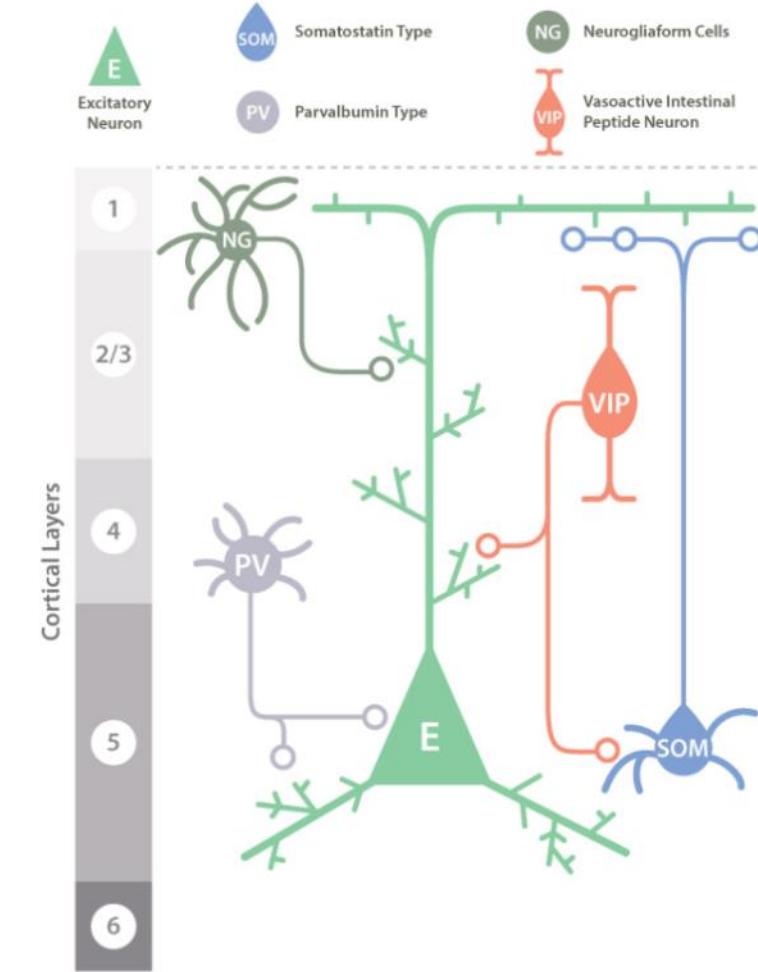
Spruston, Nature Neuro. Rev. 2008

Inhibitory neurons are (also) characterized by their dendritic tree



Jiang et al., Science 2015

isiCNI2020



© Knowing Neurons <http://knowingneurons.com>

Why model the details?

1. Suggest possible computational (functional) role for the modeled system (**predictions**).
2. Interpretation of **experimental** results - Gain insights into key biophysical parameters.
3. **Inspire** implementations in other fields.

Where can you find the details?

- Find Neuronal Morphologies in NeuroMorpho

<http://neuromorpho.org/>

- Find neuron's electrophysiological properties

<http://neuroelectro.org/>

- Allen Institute

<http://portal.brain-map.org/>

- Blue Brain Portal

<https://portal.bluebrain.epfl.ch/>

- Find models in ModelDB

<https://senselab.med.yale.edu/modeldb/>

- Choose your channels using
IonChannelGenealogy

<https://icg.neurotheory.ox.ac.uk/>

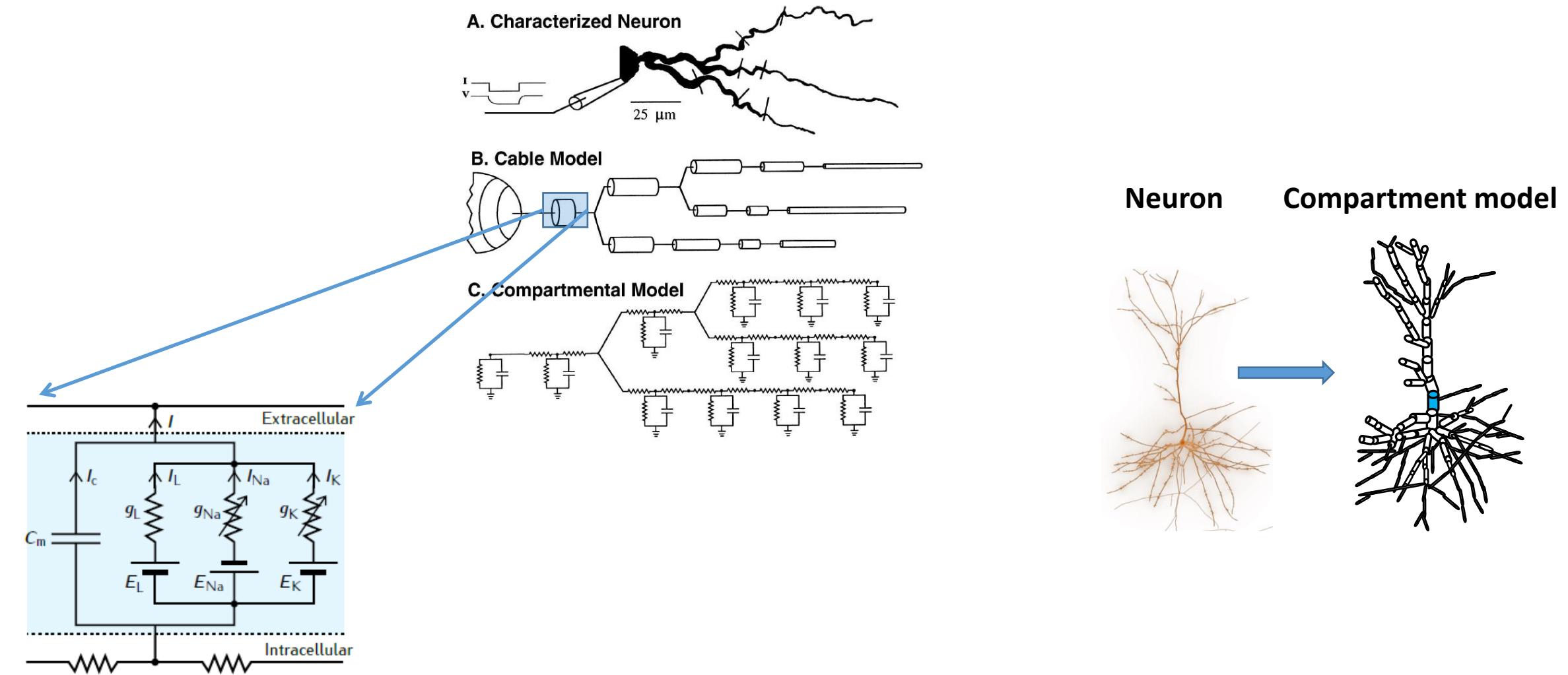
- Find Access to HPC (mostly for Parallel simulations) in...

<https://www.nsgportal.org/overview.html>

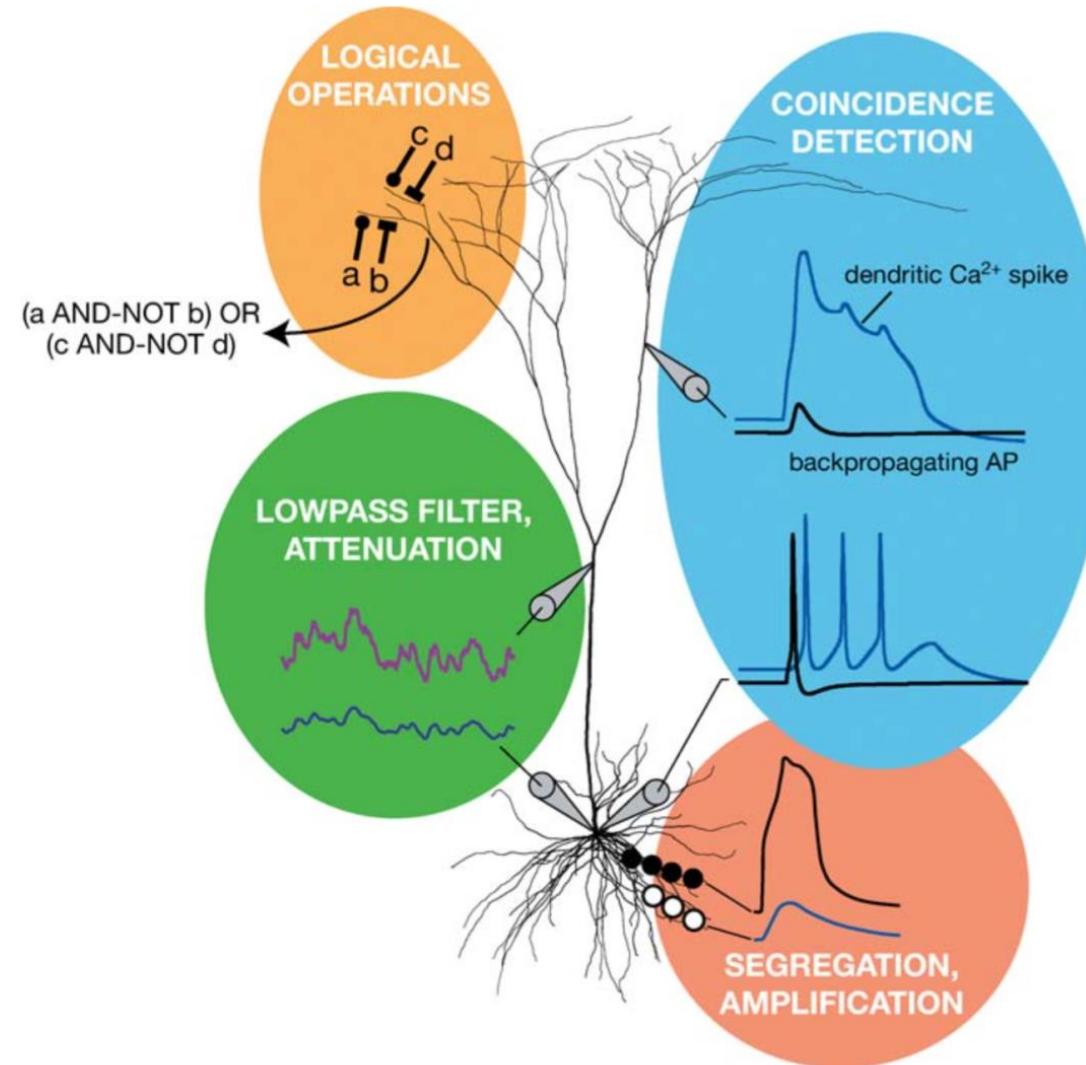
Cable theory

Cable theory is concerned with *how inputs propagate* to the soma or the axon initial segment, *how these inputs interact* with one another, and how the placement of an input on a dendritic tree affects its *functional importance* to the neuron.

Modeling dendrites: Compartmental Modeling



What can dendrites do?



The NEURON simulation environment

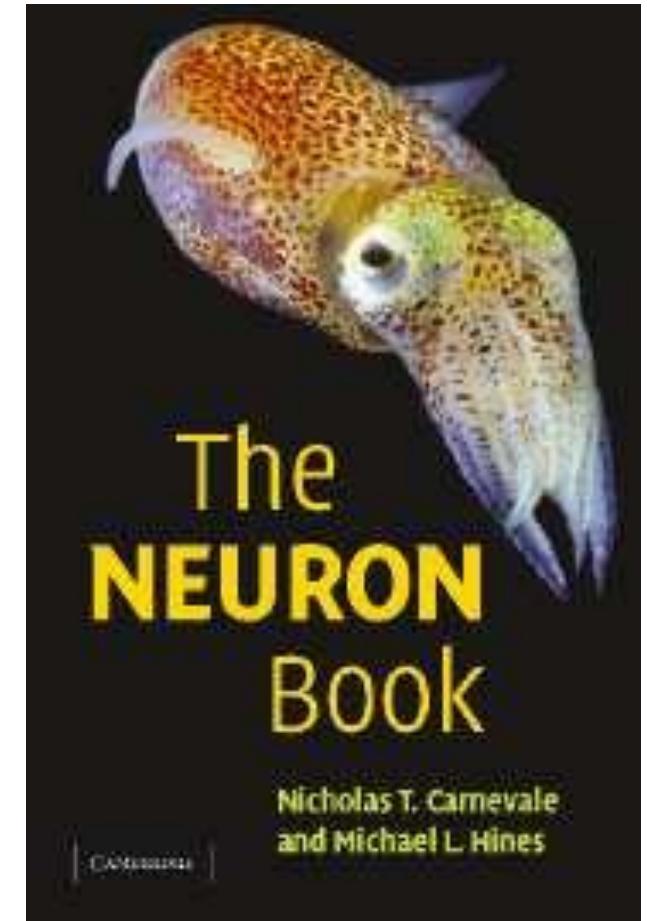
NEURON is a simulation environment for models of individual neurons and networks of neurons that are closely linked to *experimental data*.

Documentation:

https://www.neuron.yale.edu/neuron/static/new_doc/index.html

- *Neuron with Python*

from neuron import h, gui



The main features of NEURON

- Graphical User Interface (GUI)

- Two programming languages:

1. NMODL

Ion channel mechanisms (Hodgkin-Huxley-like kinetics, calcium-dependent kinetics, synaptic properties and others).

2. All other operations in NEURON are performed using scripts written in HOC.

The main features of NEURON

- Defining the **anatomical** and **biophysical** properties of models of neurons and neuronal networks,
- Controlling simulations etc.,

e.g.

```
soma=h.Section(name='soma')
soma.diam=10      #μm
soma.L=3.18    #μm
```

Distributed (or Density) Mechanisms in hoc

- Properties that are distributed over the cell surface (e.g. membrane capacitance, active and passive ionic conductances) or throughout the cytoplasm (e.g. buffers).

e.g.

```
soma=h.Section(name='soma')
```

```
soma.diam=10      #μm
```

```
soma.L=3.18  #μm
```

```
soma.cm=1      #membrane capacitance, μF/cm2
```

```
soma.Ra=100    #axial resistance, ohm cm
```

```
soma.insert('hh')  #HH channels
```

Point Processes in hoc

Synapse or electrode for passing current (current clamp or voltage clamp) is represented by a **point** source of current which is associated with a localized conductance.

Syntax:

```
varname = h.Classname(section_name(x))
```

```
varname.attribute = value
```

Point Processes in hoc

Example 1

Current clamp

```
ic = h.IClamp(soma(0.5))  
ic.del = 100 #ms  
ic.dur = 200 #ms  
ic.amp = 0.1 #nA
```

Example 2

Synapse

```
syn = h.ExpSyn(soma(0.5))  
syn.e = 0 #mV  
syn.tau=10 #ms
```

Example 3

NetStim

```
ns = h.NetStim(0.5)  
ns.start = 100 #ms  
ns.number = 2  
ns.interval = 10 #ms
```

Example 4

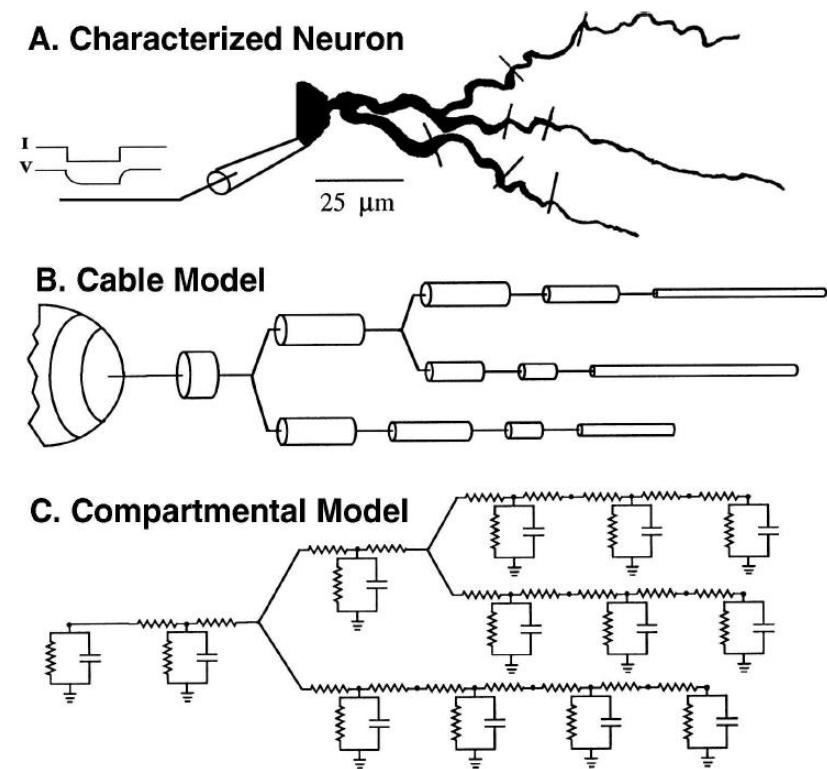
NetCon

```
nc = h.NetCon(source,target, [threshold, delay,  
weight])
```

Cable theory

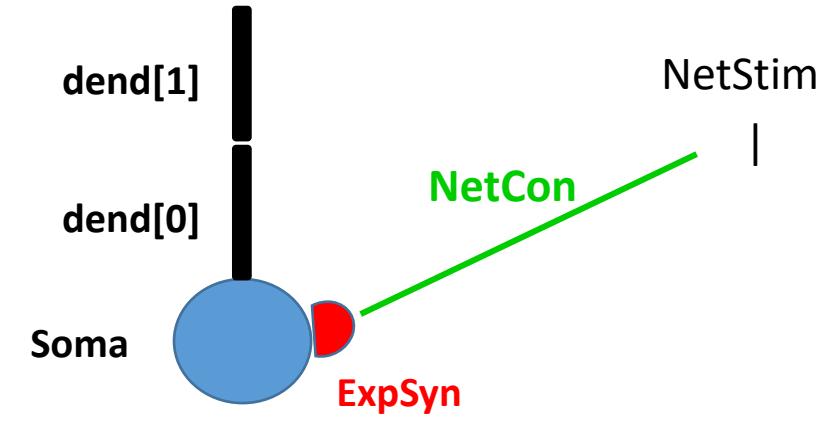
- Connecting compartments
`child.connect(parent, [0 or 1])`
e.g.
`dend0.connect(soma(0),0)`

Cable theory is concerned with ***how inputs propagate to the soma or the axon initial segment, how these inputs interact*** with one another, and how the placement of an input on a dendritic tree affects its ***functional importance*** to the neuron.



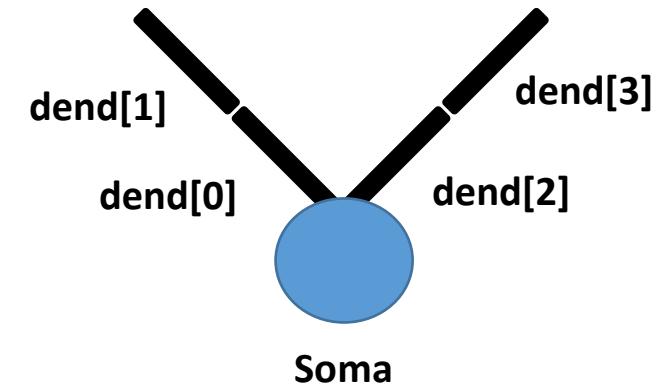
How inputs propagate / interact

Jupyter notebook: Exc_1a



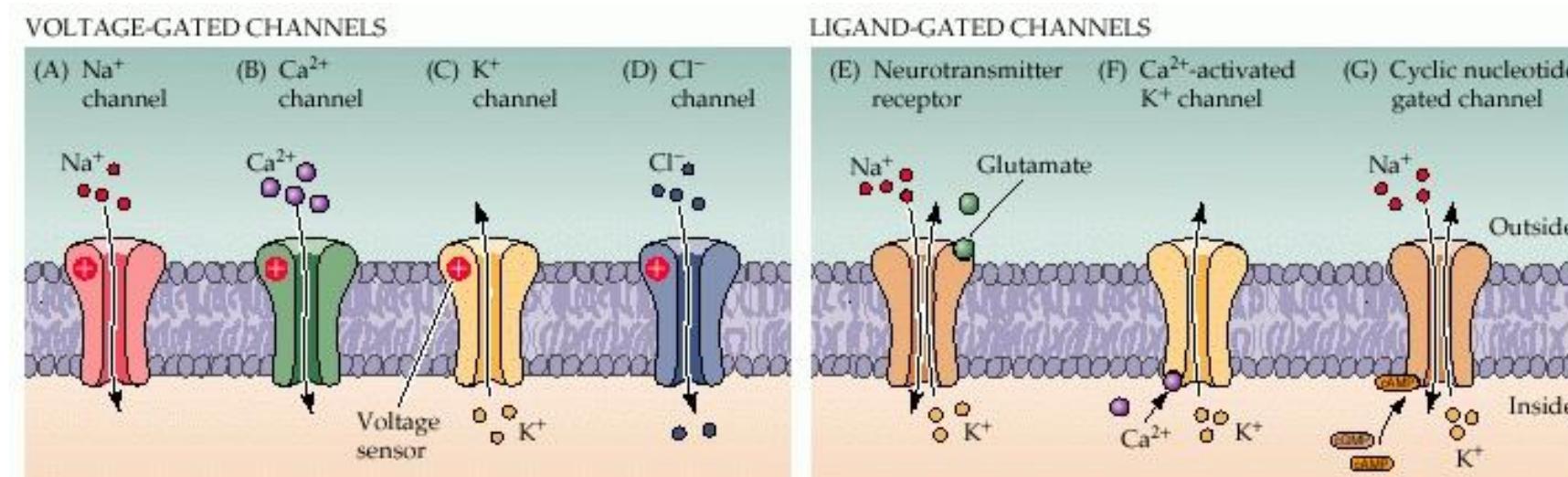
How inputs interact: Segregation

Jupyter notebook: Ex1_b



Active mechanisms

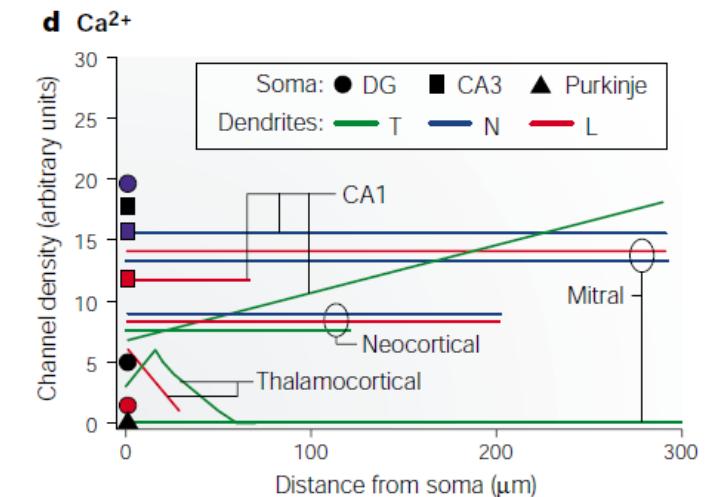
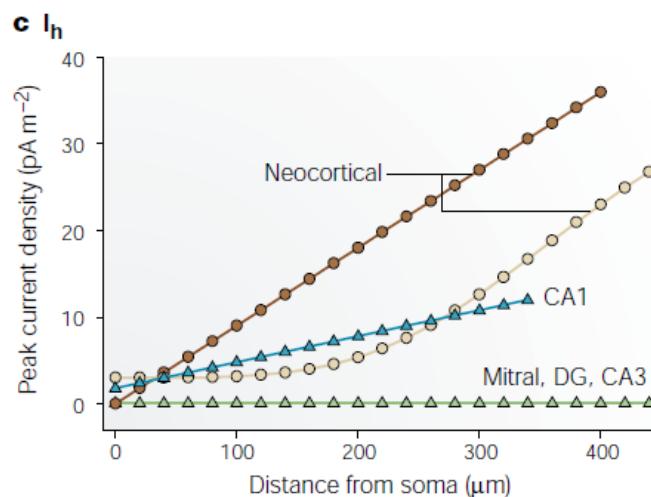
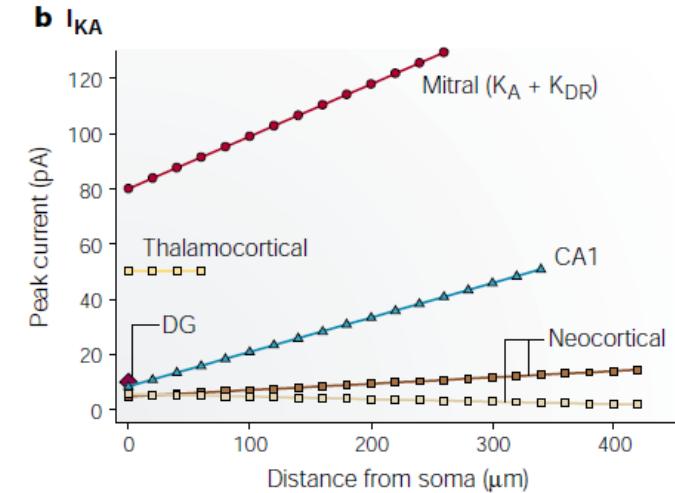
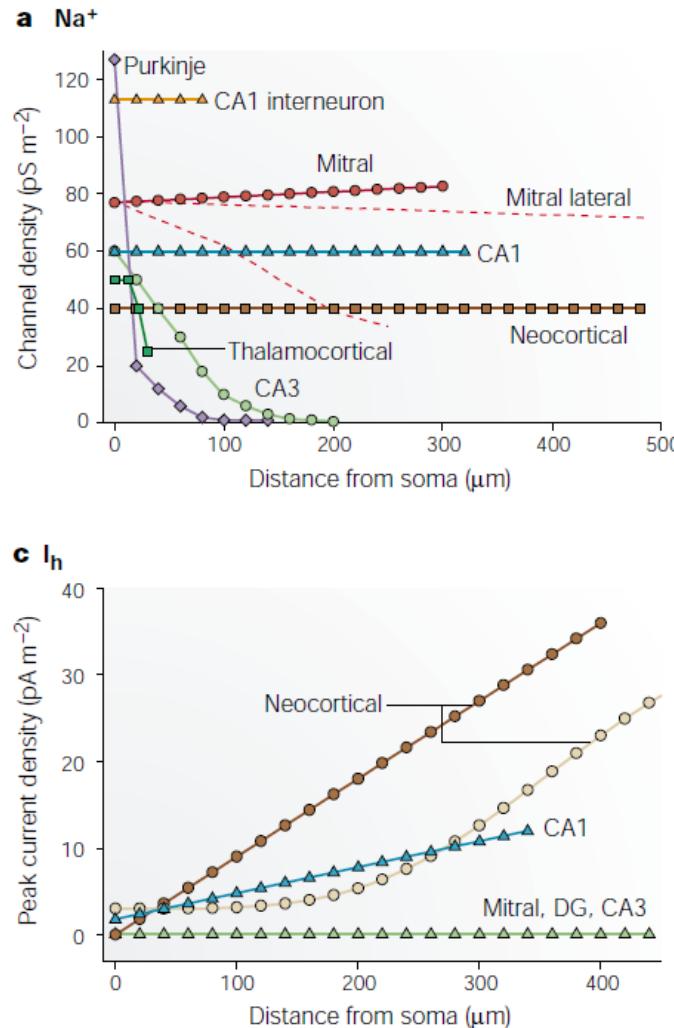
- Passive ion channels -- no change in permeability due to external factors
- Active ion channels -- permeability is dependent upon factors such as:
 - Membrane potential (e.g., voltage-dependent)
 - Ionic concentrations (e.g., calcium-dependent)
 - Ligands (e.g., neurotransmitters)



Are active mechanisms found in dendrites?

Voltage-gated ion channels
are prevalent in neuronal
dendrites

How do they contribute to
active processing?



Modelling active conductances in NEURON

- Custom ion channel models are made in separate files with the extension “.mod”
- They need to be compiled before the code is run (nrnivmodl / mknrndll)
- They must be inserted to each compartment separately (with separate parameters)

The NEURON block:

SUFFIX: the name of the mechanism

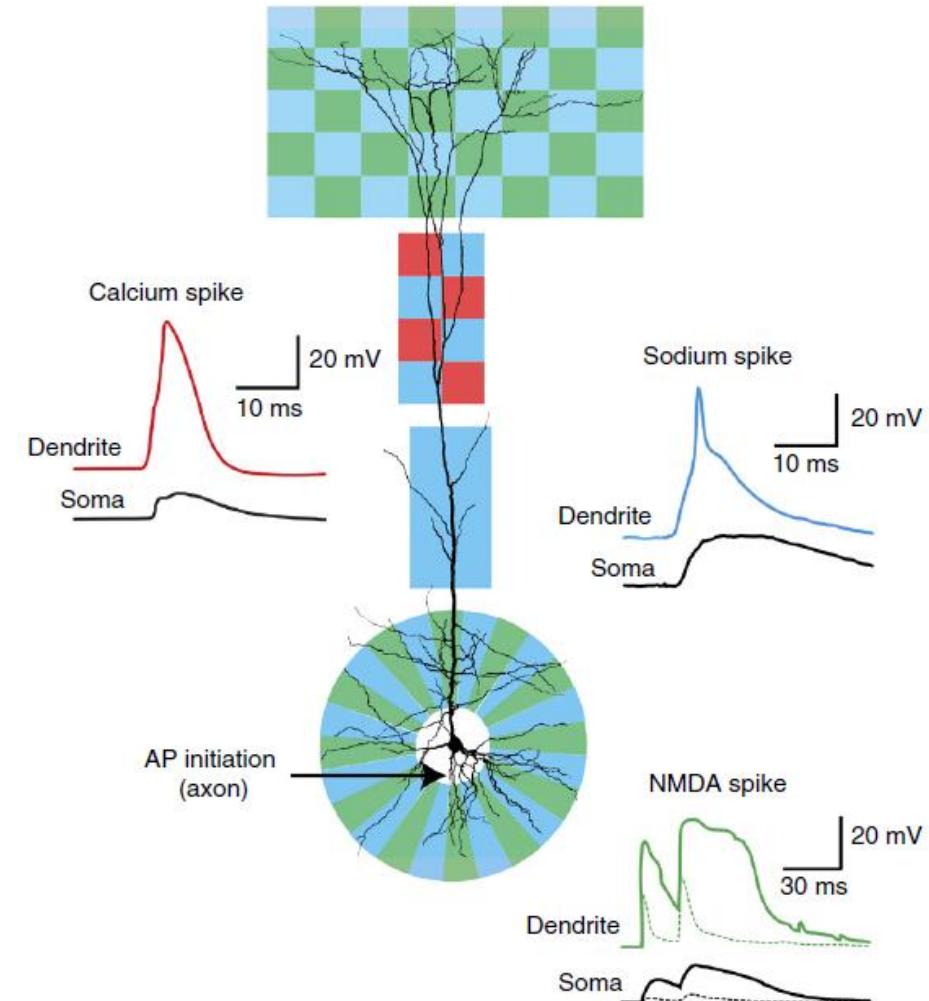
USEION / NONSPECIFIC_CURRENT statement: specifies which ionic species the model uses

RANGE variables: can be accessed from hoc / python code (e.g., gbar)

Functional roles of active mechanisms

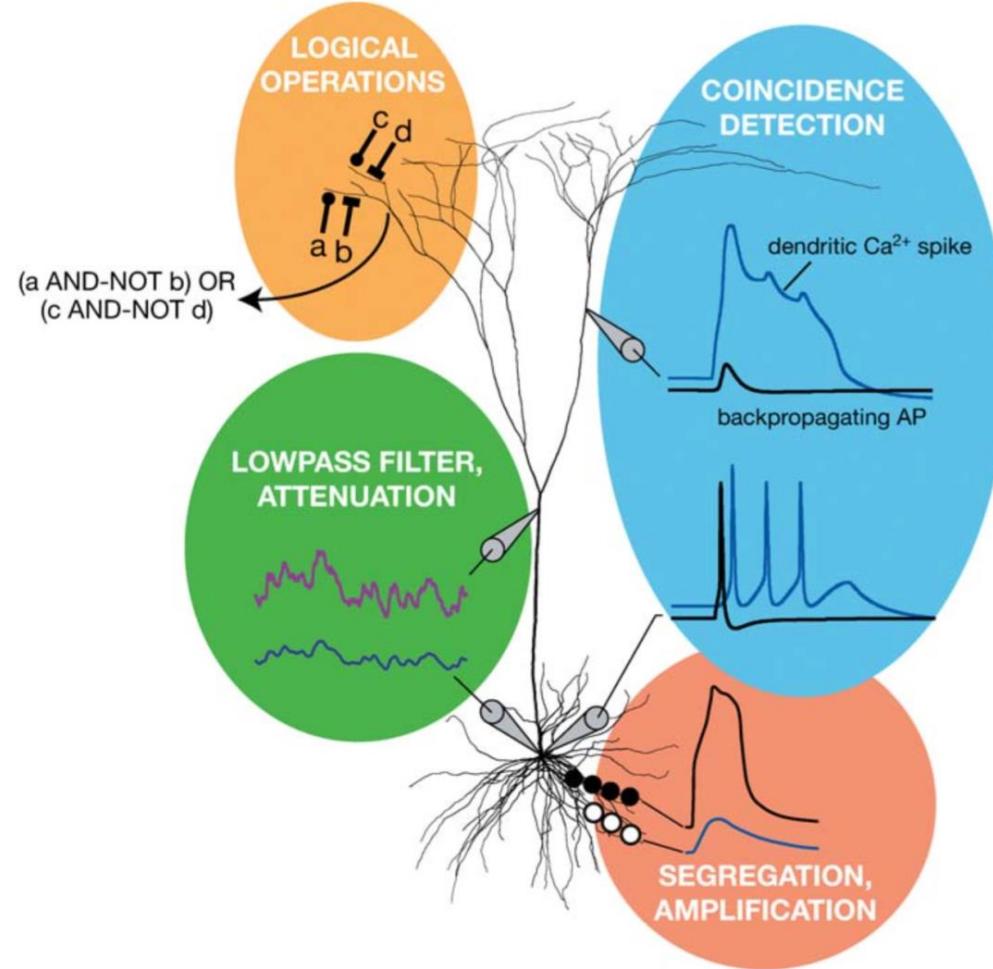
What active processes occur in dendrites?

What functional roles do they have?



Stuart, Spruston, Nature Reviews, 2015

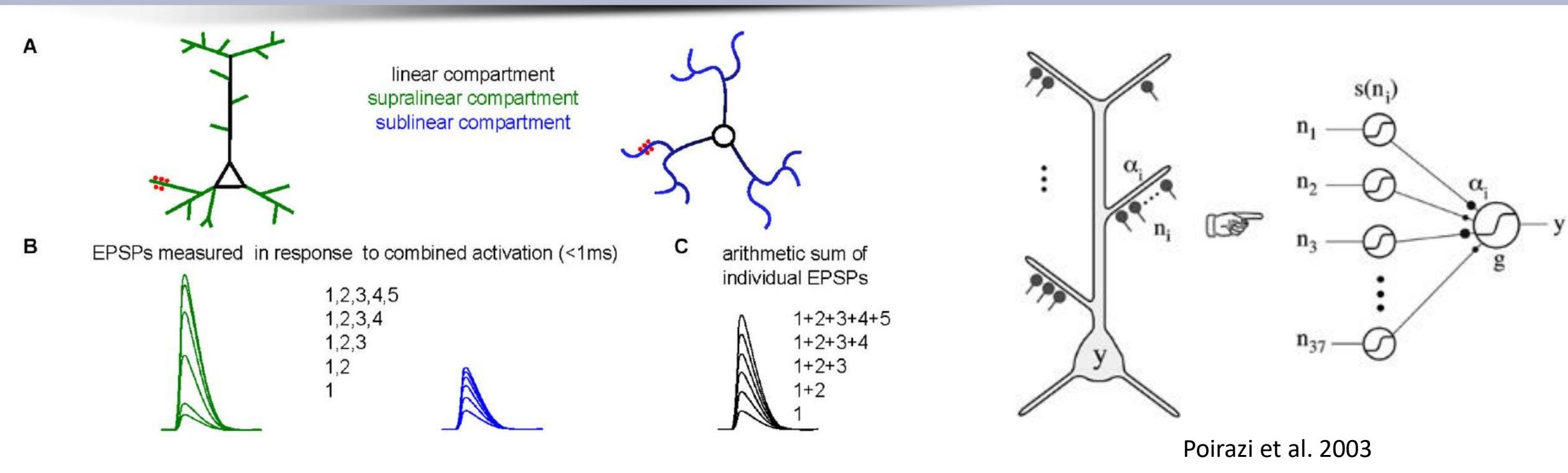
Functional roles of active mechanisms



How inputs interact: Active Integration

Jupyter notebook: Ex1c

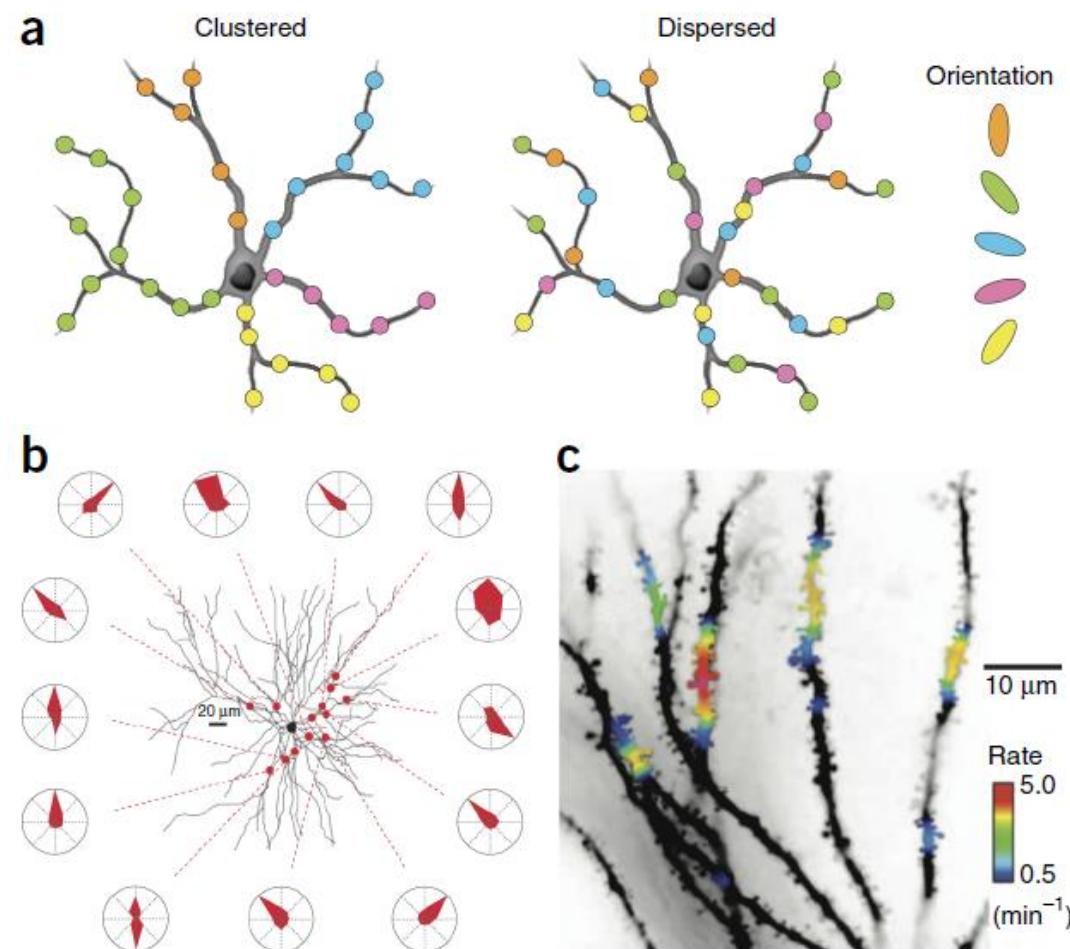
How inputs interact: Spatial Summation



Tran-Van-Minh et al. Front. Cell. Neuroscience, 2015

Which is the best strategy for maximum depolarization – scatter inputs to many dendrites or cluster them into one?

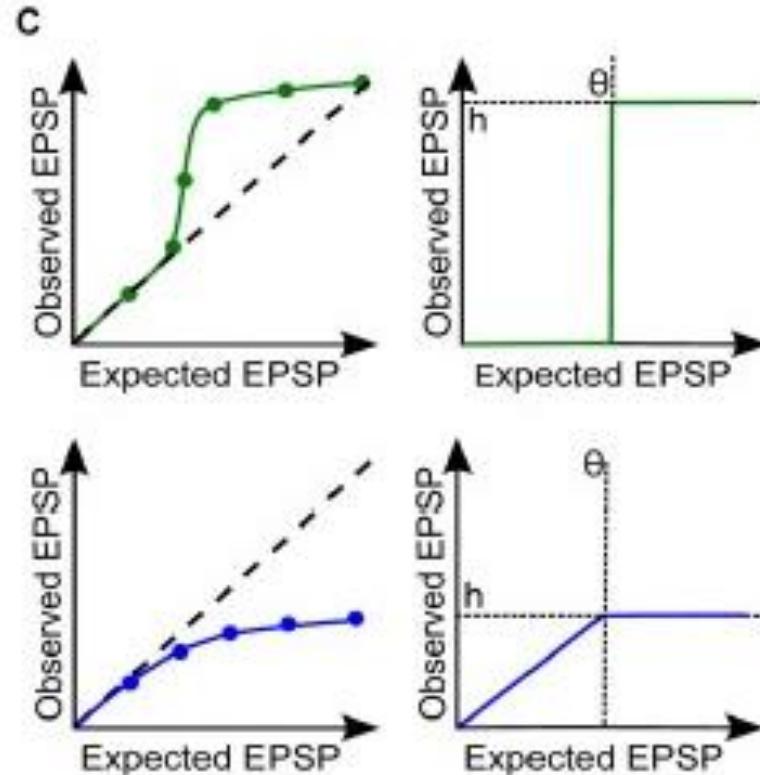
How inputs interact: Spatial Summation



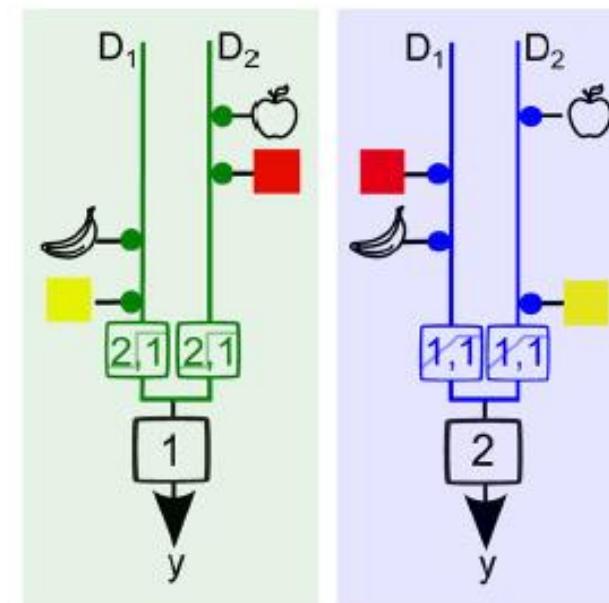
Jia et al. Nature 2010

Takahashi et al. Science 2012

Spatial Summation & Feature Binding problem

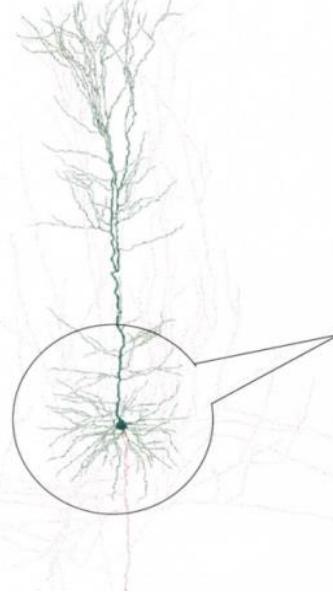


	Apple	Red Square	Banana	Yellow Square	D ₁	D ₂	D ₁	D ₂	y
Apple	1	1	0	0	1	0	1	1	1
Banana	0	0	1	1	0	1	1	1	1
Red Square	0	1	1	0	0	0	1	0	0
Yellow Square	1	0	0	1	0	0	0	1	0



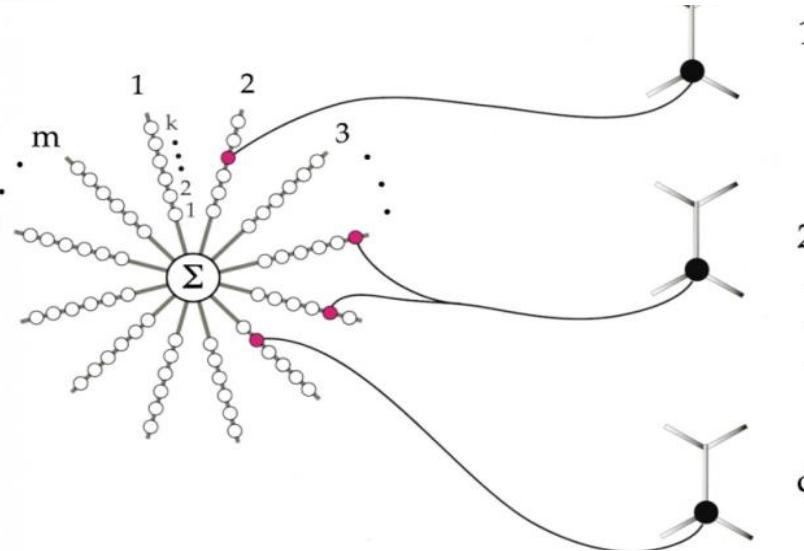
Tran-Van-Minh et al. Front. Cell. Neuroscience, 2015

Spatial Summation & Storage Capacity



Linear

Nonlinear



$$a_L(\mathbf{x}) = \sum_{j=1}^m \sum_{i=1}^k w_{ij} x_{ij}$$

$$a_N(\mathbf{x}) = \sum_{j=1}^m b \left(\sum_{i=1}^k w_{ij} x_{ij} \right).$$

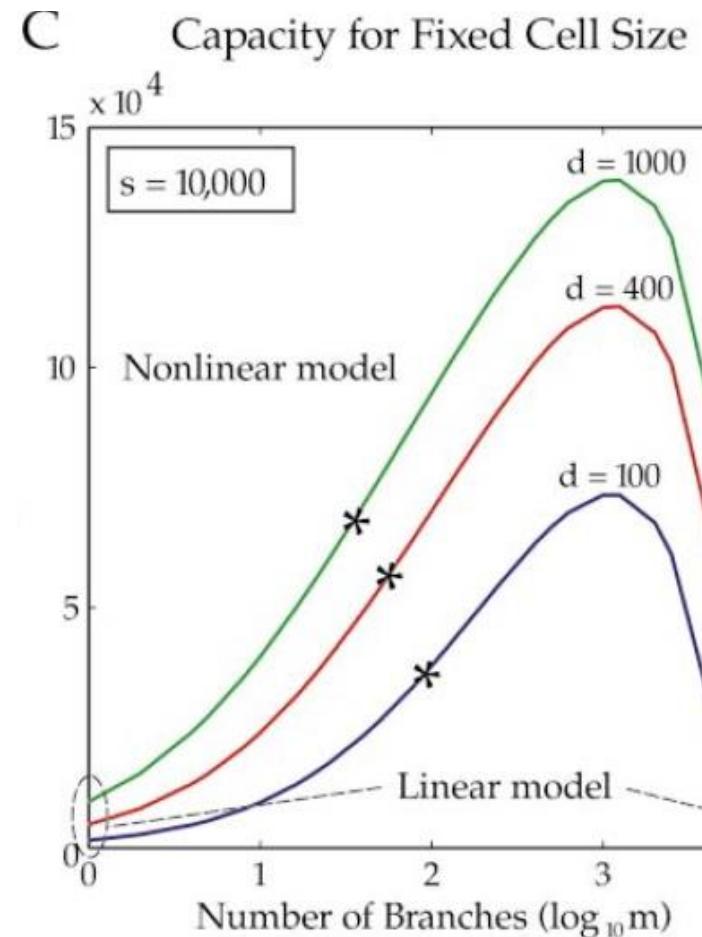
1
2
d

A nonlinear cell can distinguish between wiring configurations

	Linear Cell	Nonlinear Cell
Wiring Configurations		
(1)		$4x_1 + 3x_2 + 2x_3$
(2)		$b(2x_1 + x_2) + b(2x_1 + x_2) + b(x_2 + 2x_3)$
		$b(2x_1 + x_3) + b(x_1 + 2x_2) + b(x_1 + x_2 + x_3)$
Total number of distinct i/o functions	110	220

- m branches, where each branch contains k excitatory synaptic contacts.
- Each synapse is driven by one of d input lines.

Spatial Summation & Storage Capacity



$$B_L = 2 \log_2 \left(\frac{s + d - 1}{s} \right)$$

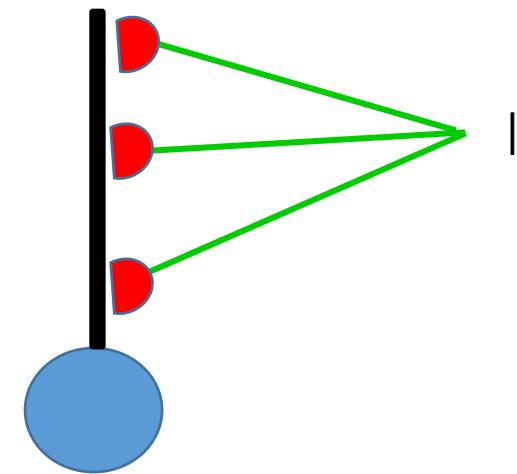
$$B_N = 2 \log_2 \left(\left(\frac{k + d - 1}{k} \right) + m - 1 \right)$$

$$s = m * k$$



How inputs interact: Temporal summation

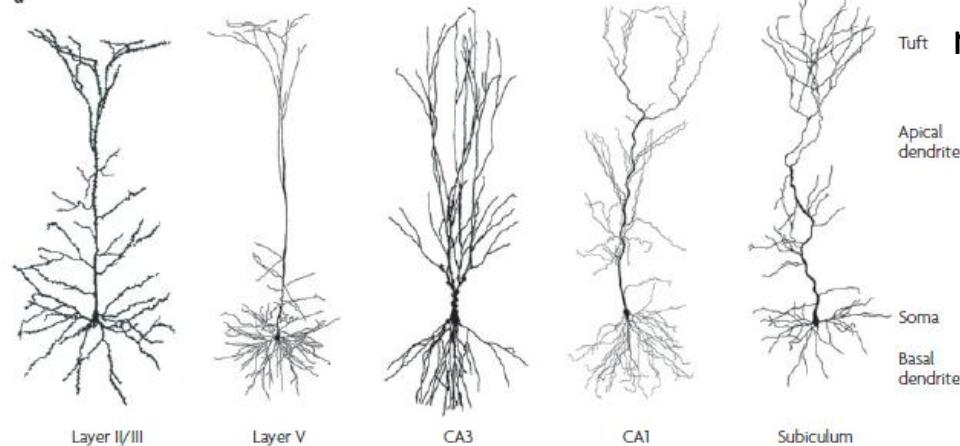
Jupyter notebook: Exc_2



Focusing on the cortex: The pyramidal neuron

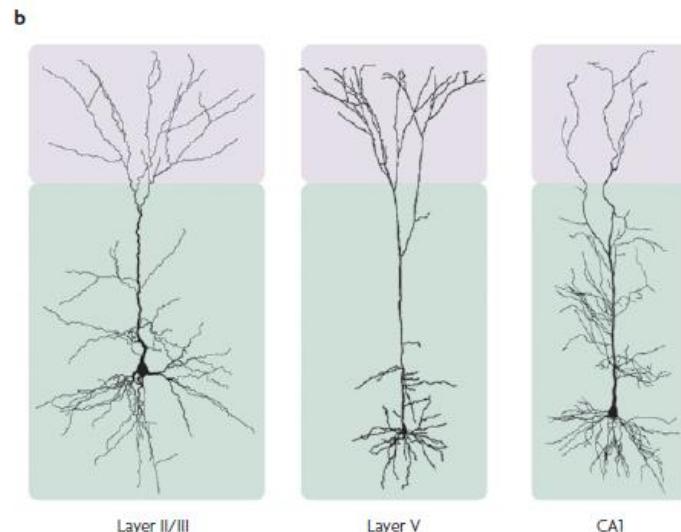
(also hippocampus and amygdala)

Spruston,
Nature Reviews,
2008

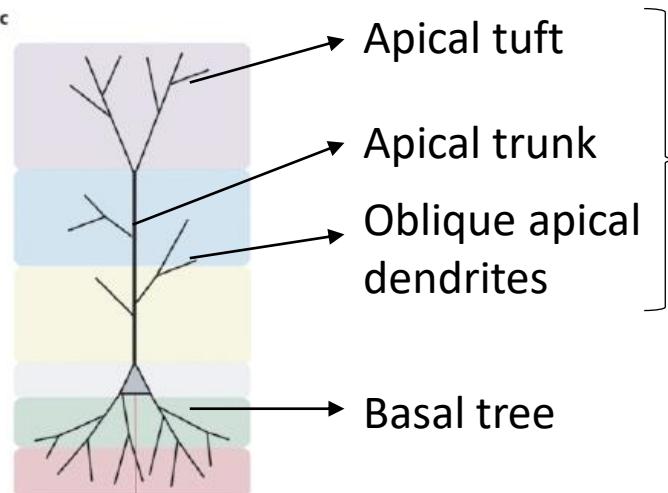


...display area and region related morphological variability and...

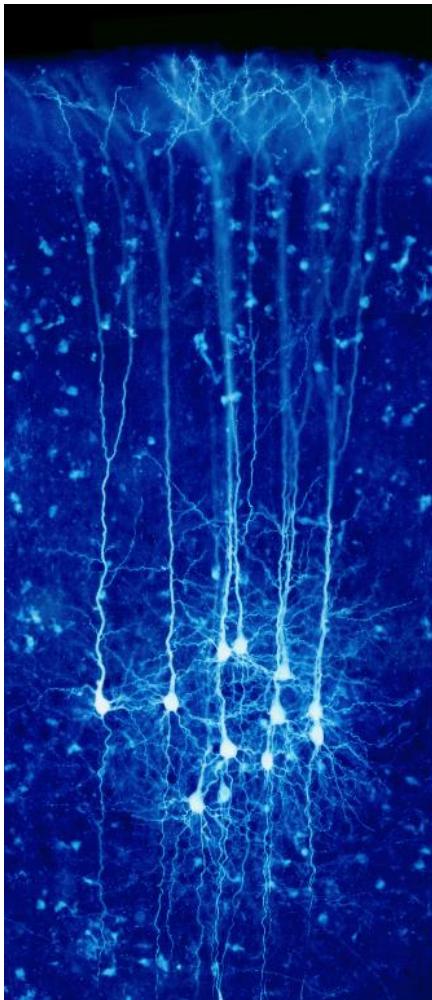
...dendritic domains that receive unique synaptic inputs.



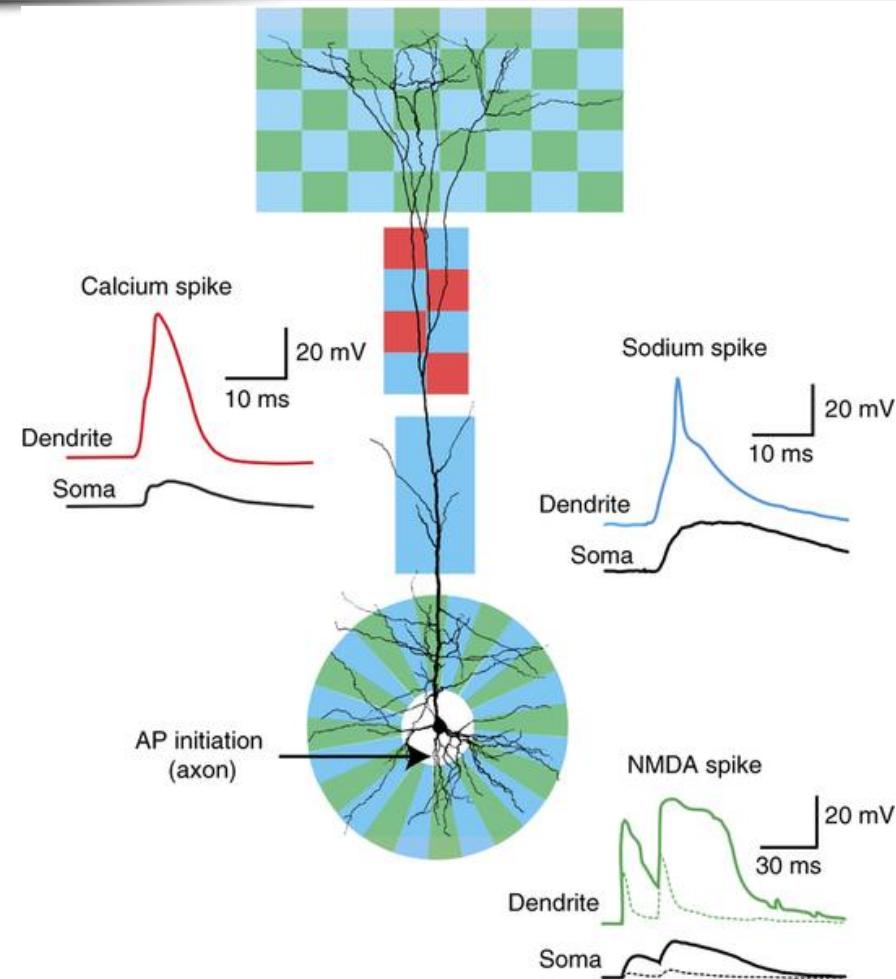
Apical tree



Focusing on the cortex: The pyramidal neuron



Human brain project

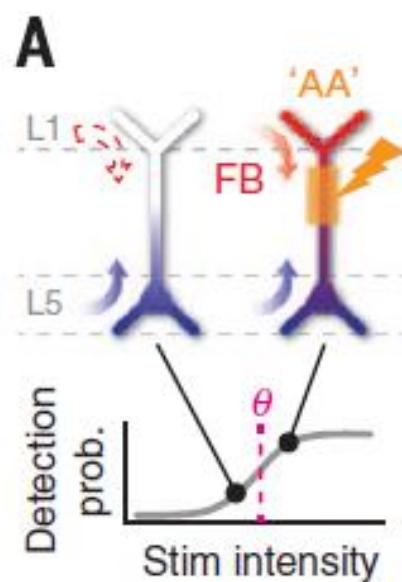
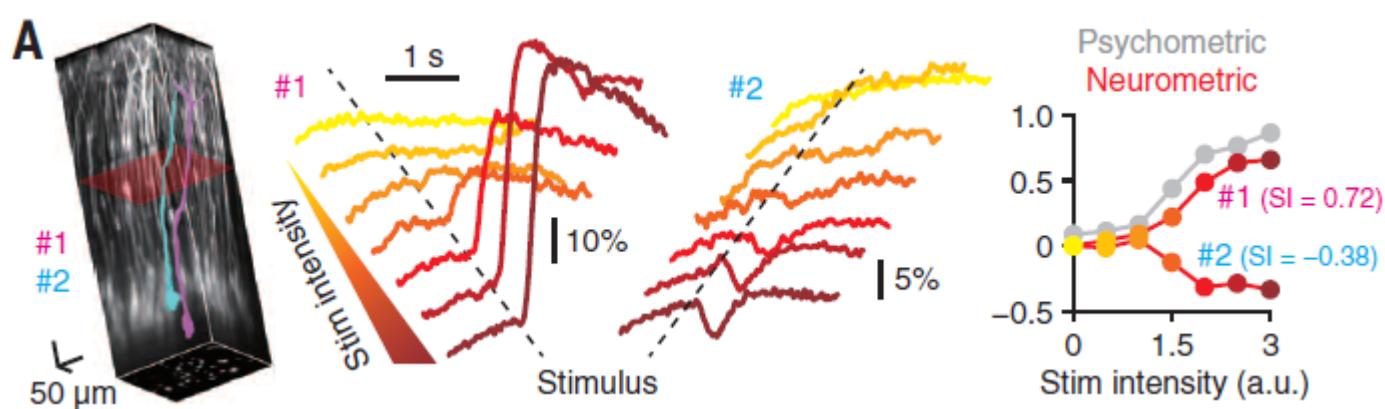


Stuart, Spruston, Nature Reviews, 2015

Functional importance: Coincidence detection

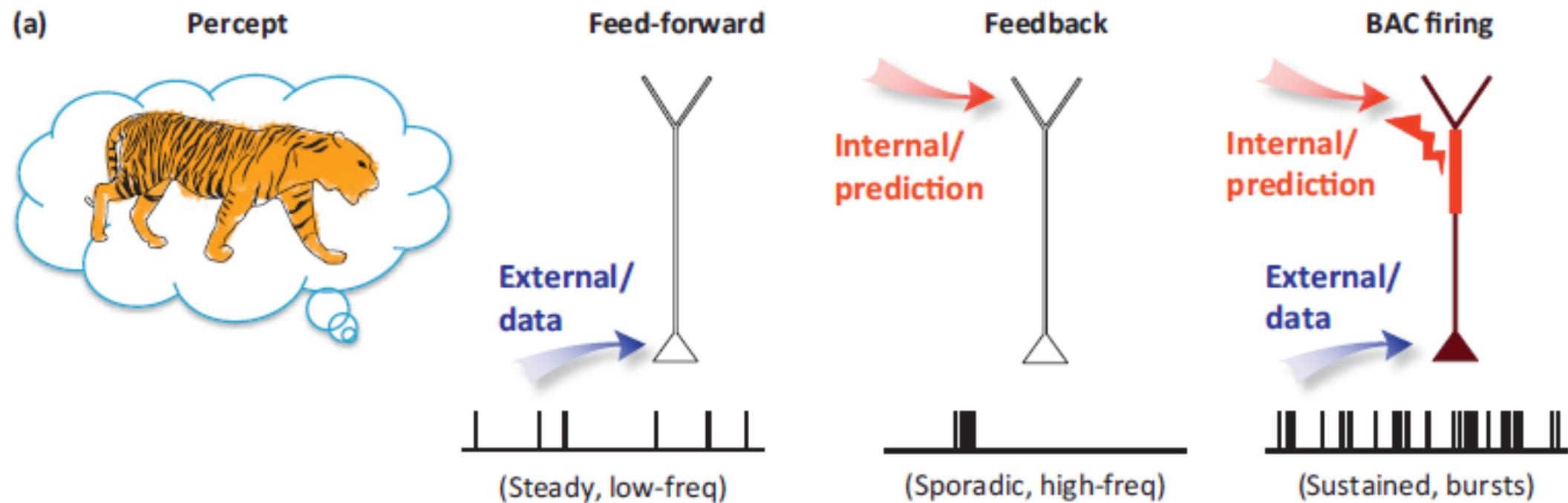
Jupyter notebook: Ex3 – BAC model

Functional importance: Perceptual threshold



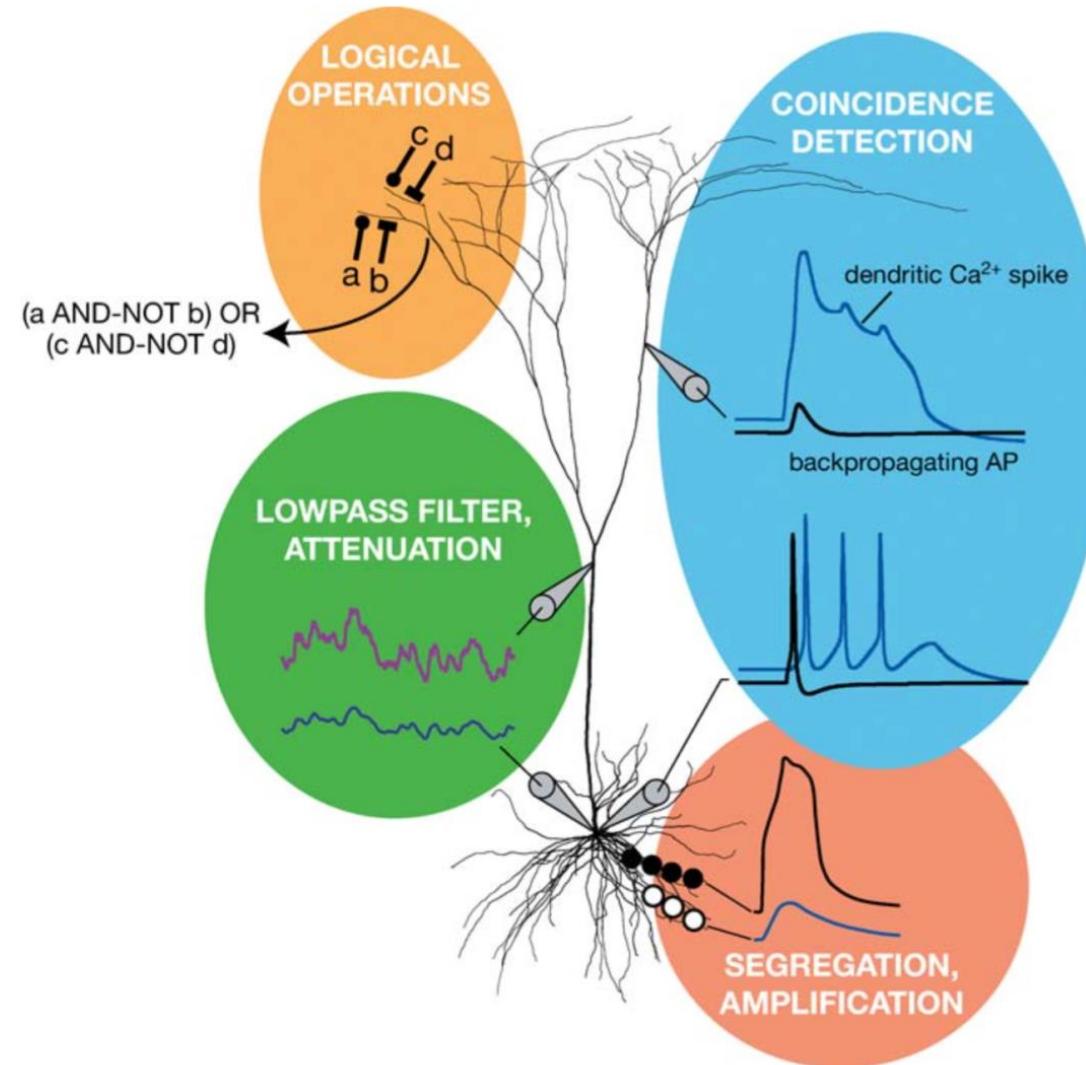
Takahashi et al. Science, 2016

Functional importance: Coincidence detection

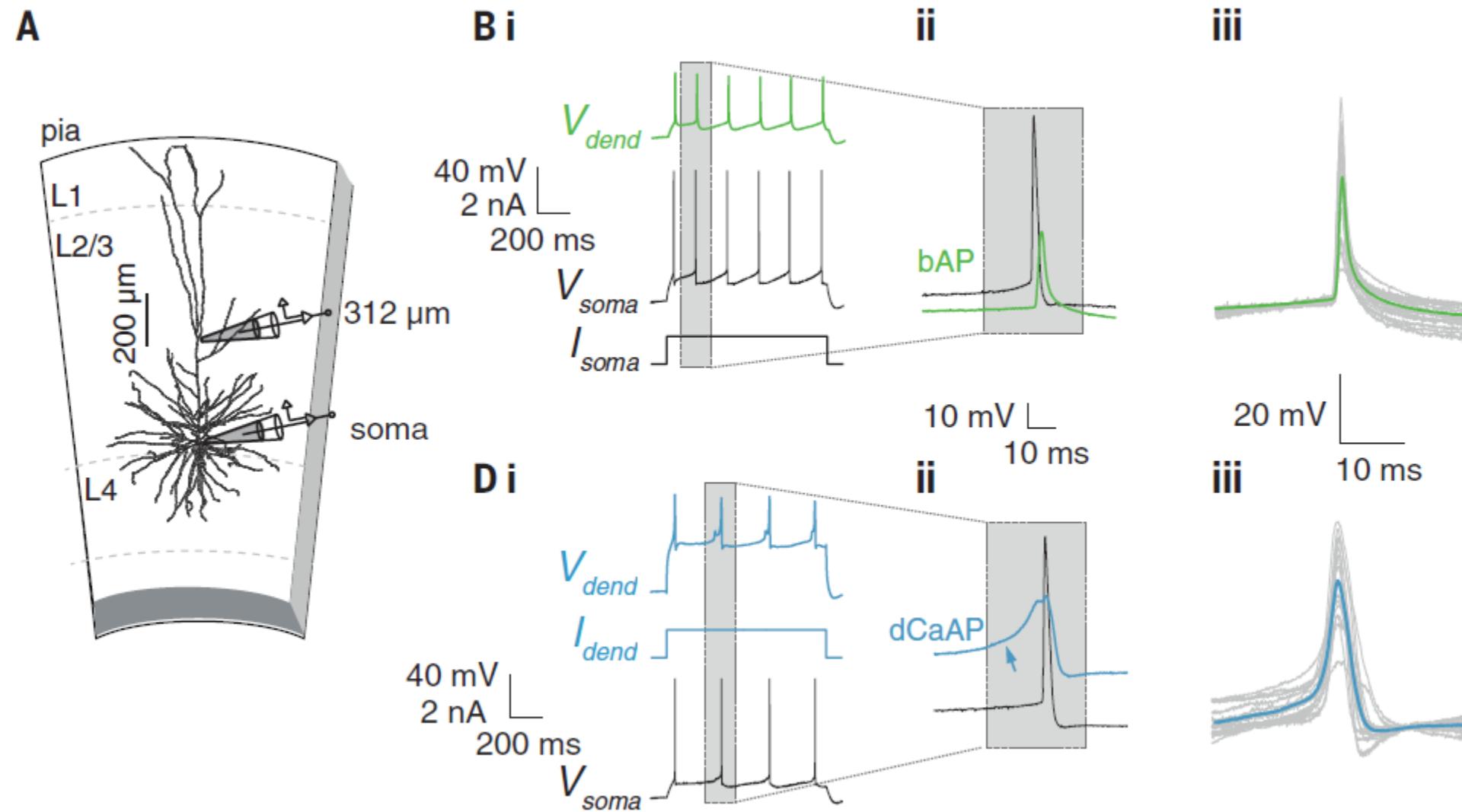


Larkum, TINS 2013

What can dendrites do?

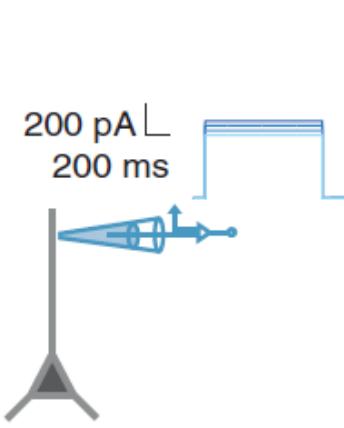


Dendritic computations in human neurons

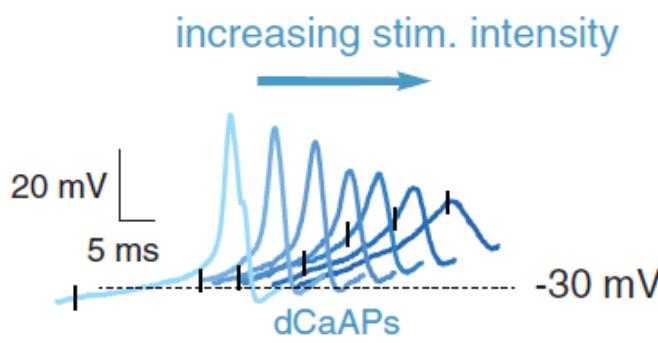


Dendritic computations in human neurons

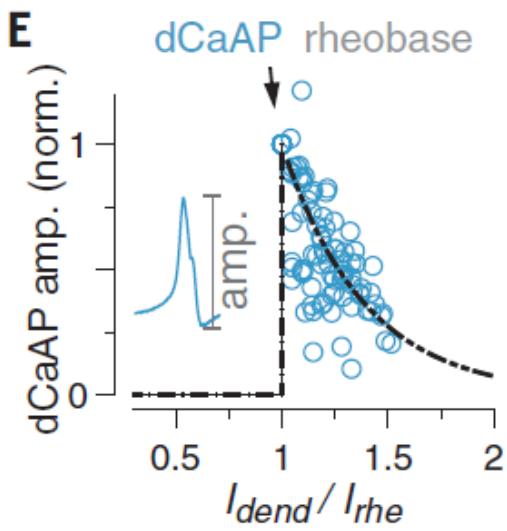
D i



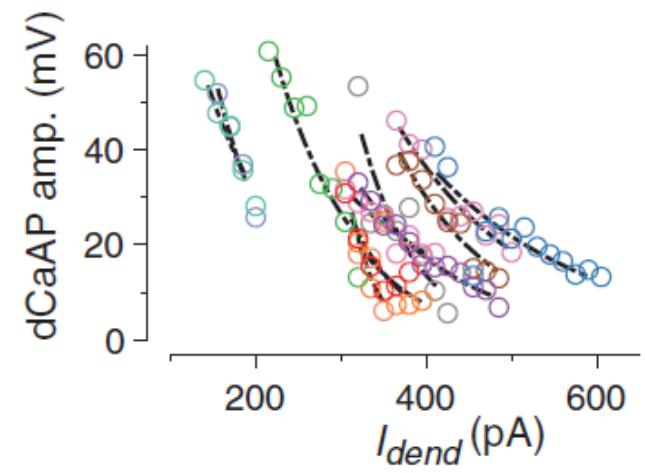
ii



E

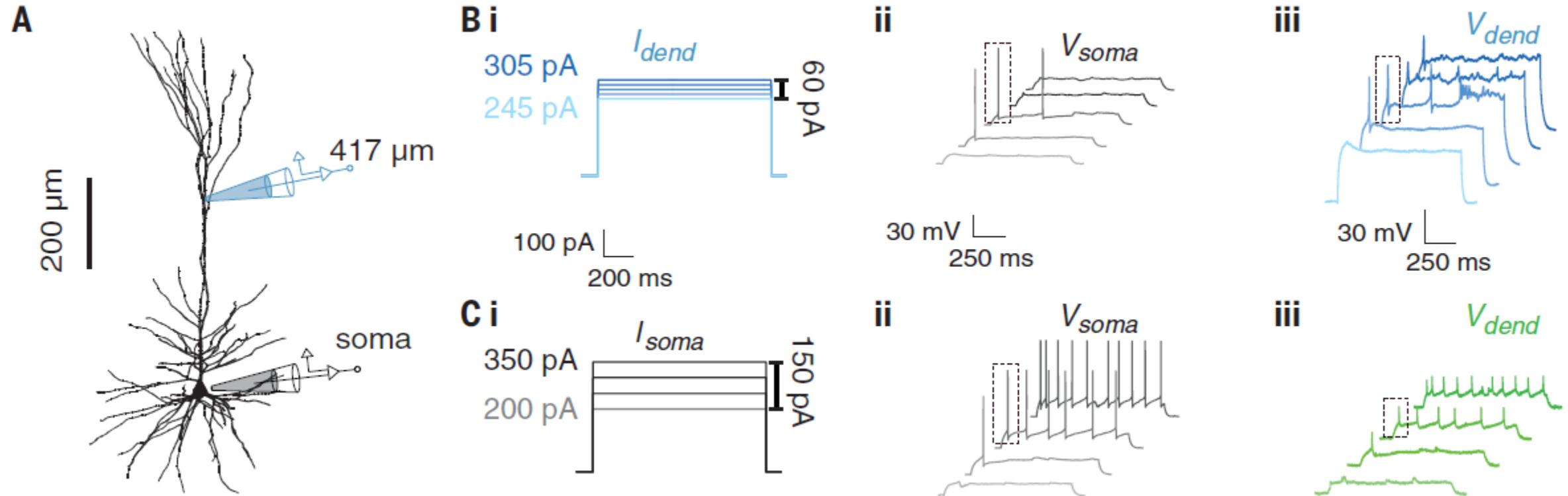


F



Gidon et al. Science 2020

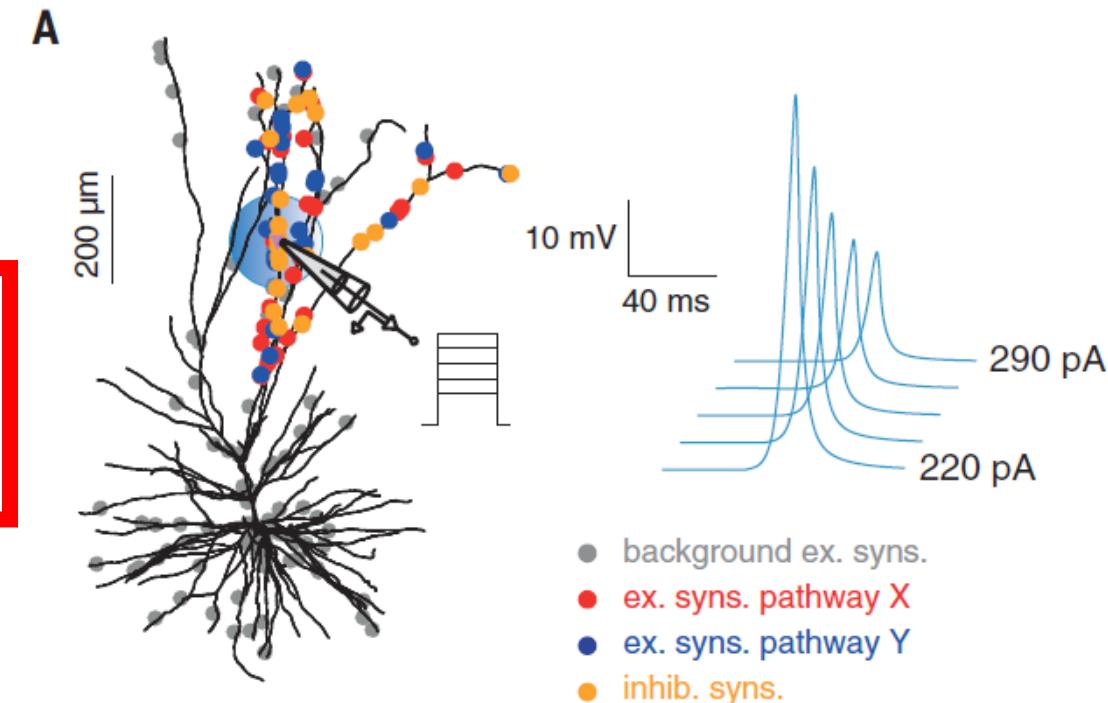
Dendritic computations in human neurons



Gidon et al. Science 2020

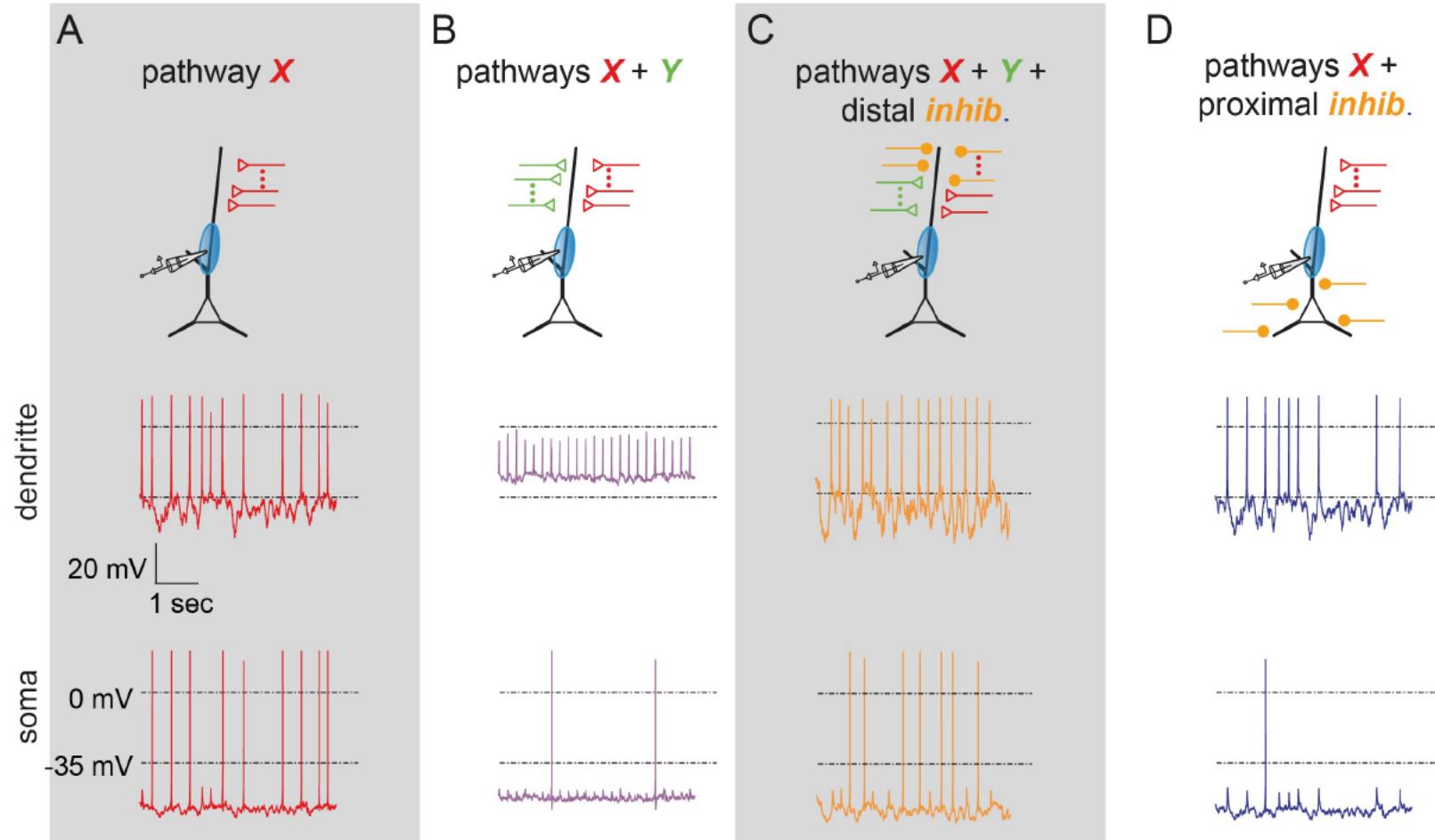
Dendritic computations in human neurons

Jupyter notebook: Human_dendrites



Gidon et al. Science 2020

Dendritic computations in human neurons



And why do I care how neurons do it?

Neuron
Perspective



Engineering a Less Artificial Intelligence

Fabian H. Sinz,^{1,2,7,*} Xaq Pitkow,^{6,7,8} Jacob Reimer,^{6,7} Matthias Bethge,^{2,3,4,5,7} and Andreas S. Tolias^{6,7,8,*}

¹Institute Bioinformatics and Medical Informatics (IBMI), University of Tübingen, Germany

²Bernstein Center for Computational Neuroscience, University of Tübingen, Germany

³Centre for Integrative Neuroscience, University of Tübingen, Germany

⁴Institute for Theoretical Physics, University of Tübingen, Germany

⁵Max Planck Institute for Biological Cybernetics, Tübingen, Germany

⁶Department of Neuroscience, Baylor College of Medicine, Houston, TX, USA

⁷Center for Neuroscience and Artificial Intelligence, BCM, Houston, TX, USA

⁸Department of Electrical and Computer Engineering, Rice University, Houston, TX, USA

*Correspondence: fabian.sinz@uni-tuebingen.de (F.H.S.), astolias@bcm.edu (A.S.T.)

<https://doi.org/10.1016/j.neuron.2019.08.034>

nature
neuroscience

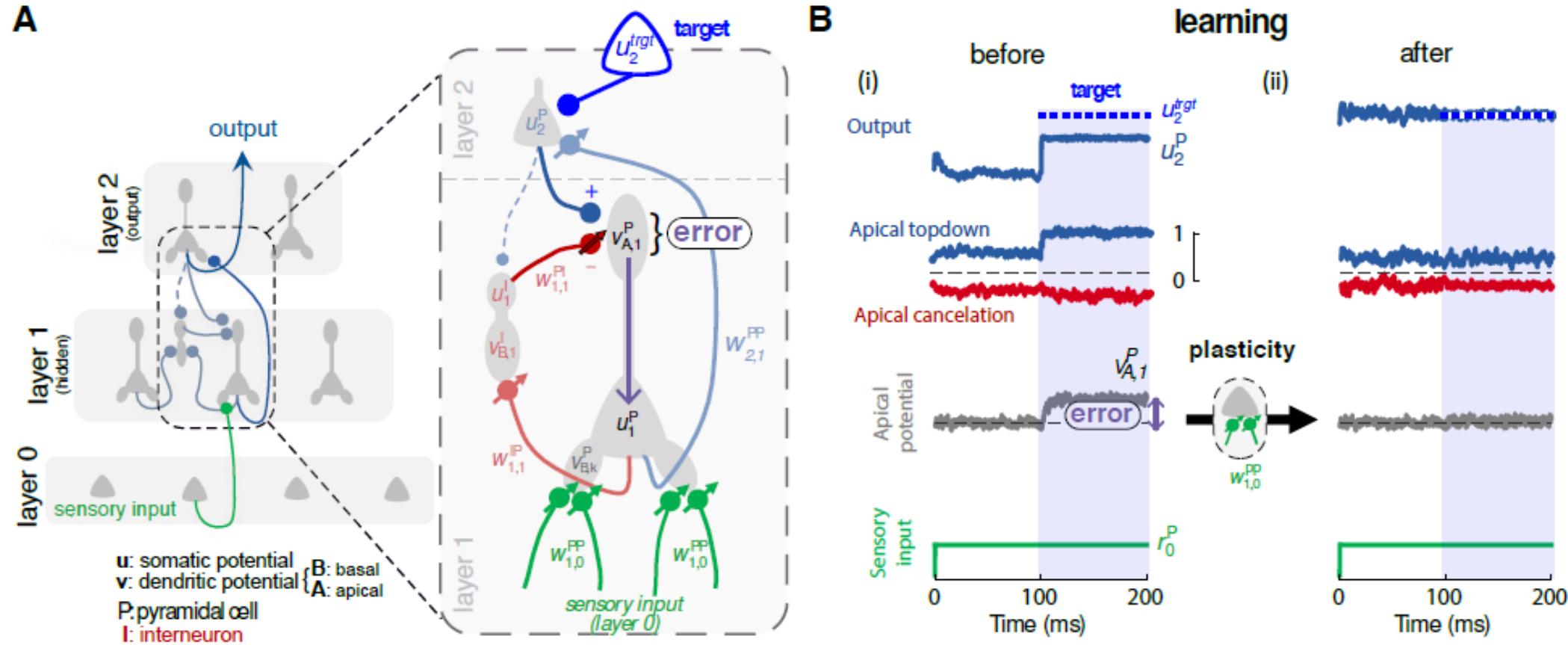
FOCUS | PERSPECTIVE

<https://doi.org/10.1038/s41593-019-0520-2>

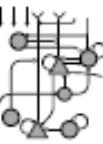
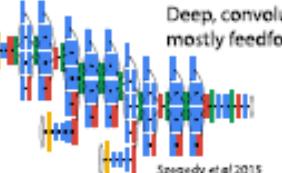
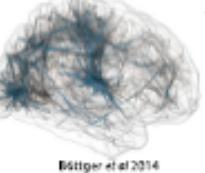
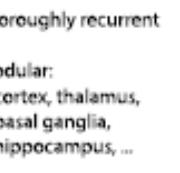
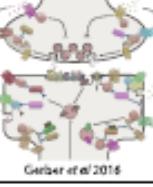
A deep learning framework for neuroscience

Blake A. Richards^{1,2,3,4,42*}, Timothy P. Lillicrap^{5,6,42}, Philippe Beaudoin⁷, Yoshua Bengio^{1,4,8}, Rafal Bogacz⁹, Amelia Christensen¹⁰, Claudia Clopath¹¹, Rui Ponte Costa^{12,13}, Archy de Berker⁷, Surya Ganguli^{14,15}, Colleen J. Gillon^{16,17}, Danijar Hafner¹⁸, Adam Kepecs²⁰, Nikolaus Kriegeskorte^{21,22}, Peter Latham²³, Grace W. Lindsay^{22,24}, Kenneth D. Miller^{22,24,25}, Richard Naud^{26,27}, Christopher C. Pack³, Panayiota Poirazi²⁸, Pieter Roelfsema²⁹, João Sacramento³⁰, Andrew Saxe³¹, Benjamin Scellier^{1,8}, Anna C. Schapiro³², Walter Senn¹³, Greg Wayne⁵, Daniel Yamins^{33,34,35}, Friedemann Zenke^{36,37}, Joel Zylberberg^{4,38,39}, Denis Therien^{1,42} and Konrad P. Kording^{4,40,41,42}

Dendritic cortical microcircuits approximate the backpropagation algorithm.

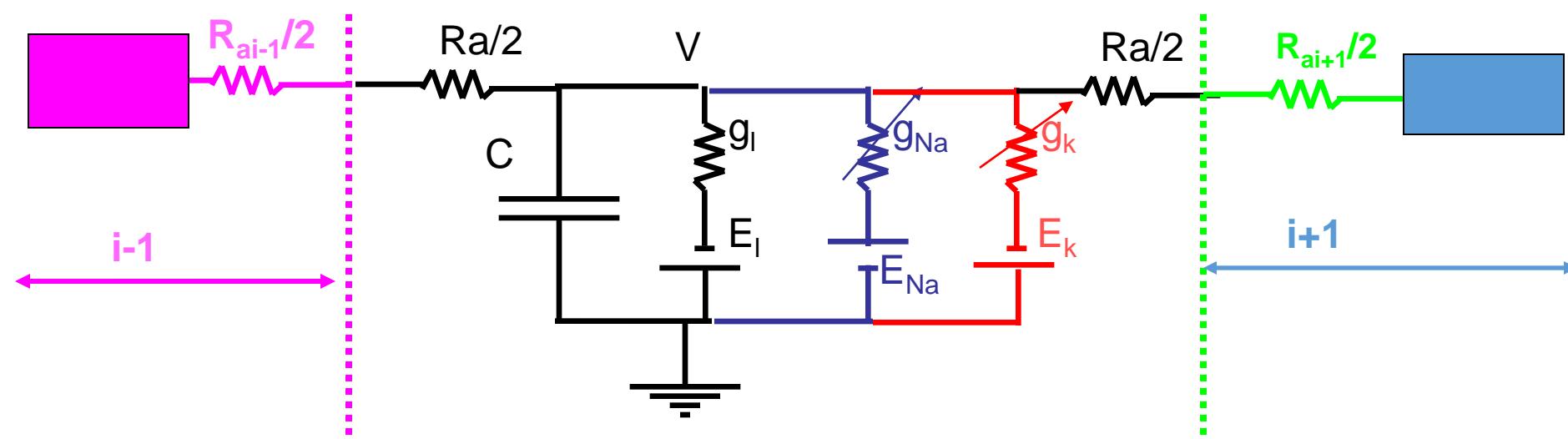


Sacramento et al. NIPS 2018

	Artificial networks	Biological networks
Connectivity	 Synaptic weights	 Synaptic weights
Nonlinearities	 ReLU, ELU, sigmoid, gating...	 Complex
Microcircuit	$\max \sum \odot$ Sum, Max Gated recurrent unit	 Cell types Canonical circuit
Macrocircuit	 Deep, convolutional, mostly feedforward <small>Szegedy et al 2015</small>	 Thoroughly recurrent  Modular: cortex, thalamus, basal ganglia, hippocampus, ... <small>Büttner et al 2014</small>
Learning	 Backpropagation	 Complex plasticity: timing-dependent, short-term, long-term <small>Geiger et al 2016</small>
Learning Strategy	 Big data <small>Photo by Susan Yin on Unsplash</small>	 Active learning, Curiosity-driven

Thank you!

Compartmental model



$$\frac{CdV}{dt} = g_l(E_l - V) + g_{Na}(E_{Na} - V) + g_K(E_K - V) + 2 * \frac{V_{i+1} - V_i}{R_{ai+1} + R_a} + 2 * \frac{V_{i-1} - V_i}{R_{ai-1} + R_a} + I_{inj}$$

$$I_{dCaAP} = -wK(v)(A-B),$$

$$A=1/(1+exp(-(t-t')/\tau_A)) \qquad\qquad K(v)=exp[\frac{-F\times(v-v_{th})}{\tau_K}]$$

$$B=1/(1+exp(-(t-(t'+\Delta t')/\tau_B)) \qquad\qquad F~=~1/(v_{th}-v_{rest})$$

