A Tale of Two Trees: Modeling Apical and Basal Tree Contribution to L2/3 V1

Pyramidal Cell Orientation Selectivity

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Pyramidal neurons, a mainstay of cortical regions, receive a plethora of inputs from various

areas onto their morphologically distinct apical and basal trees. Both trees differentially

contribute to the somatic response, defining distinct anatomical and possibly functional sub-

units. To elucidate the contribution of each tree to the encoding of visual stimuli at the

somatic level, we modeled the response pattern of a L2/3 V1 pyramidal neuron to orientation-

tuned synaptic input. Towards this goal, we used a morphologically detailed computational

model of a single cell, validated against electrophysiological and two-photon imaging data.

Our simulations predict a synergistic effect of apical and basal trees on somatic action

potential generation: basal tree activity, in the form of either depolarization or dendritic

spiking, is necessary for producing somatic activity, despite the fact that most somatic spikes

are heavily driven by apical dendritic spikes. This model provides evidence for synergistic

computations taking place in the basal and apical trees of the L2/3 V1 neuron and emphasizes

the potential role of predictive and attentional feedback input in these cells.

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1. Introduction

In the primary visual cortex (V1), layer 2 and 3 (L2/3) pyramidal neurons conform to a stereotypical morphology characterized by separate apical and basal dendritic arbors. The apical tree consists of a thick apical trunk that extends into L1 and splits into an apical tuft. The basal tree consists of numerous dendritic segments, the majority of which sprout directly from the base of the soma, indicating that they can directly influence somatic output (Spruston, 2008). This compartmentalization of V1 neurons is also evident in the wiring diagram describing each tree: layer 4 (L4) and L2/3 pyramidal neurons synapse with basal dendrites of L2/3 pyramidal neurons, providing them with feedforward input (Coogan & Burkhalter, 1990; Larkum, 2013). The apical dendrites of the L2/3 pyramidal neurons instead receive feedback input from higher order areas of the cortex (such as the secondary and tertiary visual cortices, lateromedial area and prefrontal cortex (Coogan & Burkhalter, 1990; Larkum, 2013), as well as orientation-tuned thalamocortical input (Chen et al., 2013; Cruz-Martín et al., 2014; Jia et al., 2010; Roth et al., 2015). These anatomical and connectivity features of the apical and basal trees, complemented by their distinct biophysical properties (Cho et al., 2008), shape both the local dendritic processing and the integration of the two input streams at the soma.

Few studies have undertaken the technically difficult task of measuring single spine tuning properties in the mouse visual cortex (Chen et al., 2013; Iacaruso et al., 2017; Jia et al., 2010). These studies have shown that single spines of L2/3 pyramidal neurons exhibit orientation selectivity, with synaptic orientation preferences varying even on the same branch (Iacaruso et al., 2017; Jia et al., 2010). These findings raise the question of how dendritic integration of sparse synaptic inputs along the apical and basal trees shape orientation tuning in L2/3 V1

pyramidal neurons. In vitro studies have revealed multiple non-linear dendritic properties of

synaptic integration, including active dendritic spikes, the backpropagation of action

potentials, and dendritic properties of coincidence detection (Larkum, 2013; Palmer et al.,

2014; Shai et al., 2015; Smith et al., 2013). These dendritic non-linearities augment the

repertoire of single neuron computations (Poirazi et al., 2003; Poirazi & Papoutsi, 2020; Silver,

2010). Focusing on L2/3 V1 pyramidal neurons, it has been found that individual dendritic

segments of L2/3 neurons generate activity independently of the soma, making their exact

contribution to somatic output unclear (Palmer et al., 2014; Smith et al., 2013). This could

indicate that multiple dendritic segments need to be activated in order to generate a spike at

the soma, which would in turn further expand the space of possible neuronal computations.

In fact, new evidence from detailed computational modeling of V1 L2/3 pyramidal neurons

indicates strong, sparse inputs to the dendrites dictate somatic responses (Goetz et al., 2021).

Interestingly, although dendritic spikes in the apical tuft of L2/3 V1 pyramidal neurons may

influence orientation selectivity (Smith et al., 2013), this orientation selectivity is robust to

ablation of the apical tree (Park et al., 2019). Thus, in a realistic theory of visual processing, it

is essential to scrutinize the functional effect that synaptic inputs along the basal and apical

dendritic arbors have on the soma.

In this work, we used a detailed biophysical model of a L2/3 V1 pyramidal cell (Park et al.,

2019), to tease apart the relative contributions of apical vs. basal dendritic trees in somatic

orientation tuning. Given the evidence for sodium-dependent dendritic spikes in these

neurons and their role in orientation tuning (Smith et al., 2013), the dendrites of these

neurons, as well as their integrative properties, are clear targets for model experimentation

seeking to illuminate orientation tuning at the somatic level.

We start by examining the properties of dendritic non-linearities, which are found to be varied

along the apical and basal dendrites. By comparing the model to *in vivo* two-photon calcium

fluorescence imaging data, we verified that dendritic and somatic spontaneous activity

patterns in the model adequately reproduce experimental data.

We then simulate orientation tuning in our model and evaluate its responses to specific

combinations of dendritic disparity and synaptic distribution, finding that the output of the

model does not always match the expected orientation preference, instead deviating towards

orientation preferences more closely matching either the apical or basal mean orientation

preference.

We continue by evaluating the dependence of somatic spikes on dendritic voltage-gated

sodium conductances (VGSCs) by performing just-in-time interventions that nullify the VGSCs

on either the apical or basal tree prior to a somatic spike. This shows that the majority of

stimulus-driven spiking activity is dependent on sodium spikes generated in the apical tree,

revealing that apical and basal trees must contribute via different mechanisms to tuned

somatic output. We investigated this further by examining the role of both specific ionic

(VGSCs) and synaptic (AMPA/NMDA) mechanisms in shaping orientation tuning. Results

indicate that apical VGSCs are critical for proper neuronal tuning, while basal VGSCs do not

play a significant role. Conversely, AMPA and NMDA activity of both the apical and basal trees

is necessary for orientation tuning at the level of the soma. A potential reason for this is that

basal dendritic branches feature higher sodium spiking thresholds for similar electrotonic

length constants compared to apical branches, reducing their propensity to generate sodium

spikes compared to apical dendrites.

Our results provide evidence for a synergistic effect between the activity of the apical and

basal trees, with apical and basal synaptically-driven depolarizations alongside apical sodium

spikes playing a critical role in somatic spiking. This synergy manifests in the form of

temporally precise summation of apical and basal input, which together drive action potential

generation in this neuronal type.

2. Methods

2.1. Model Description

The single neuron model was derived from the L2/3 V1 pyramidal cell model of (Park et al.,

2019) created in the NEURON simulation environment (Hines & Carnevale, 2001). It is a

morphologically detailed reconstruction, featuring 43 apical dendritic segments, 7 basal

dendritic segments and an axon. As the model makes use of a variety of passive and active

mechanisms (*Supplementary Tables 1-4*), its electrophysiological properties were validated

against experimental data (Park et al., 2019). Neither the morphology nor the biophysical

properties of the source model were altered, barring specific ion channel or synaptic

mechanism conductance changes whenever specified.

The model features both excitatory and inhibitory synaptic input. Excitatory input was

subdivided into background-driven (noise) and stimulus-driven, the latter consisting of

orientation-tuned synapses. Inhibitory synapses are modeled as background-driven

(inhibitory noise) which alongside background-driven excitatory synapses (excitatory noise)

comprise the background-driven inputs. Stimulus-driven and background-driven synaptic

input to the neuron is distributed uniformly on the target compartments using a density of 2

synapses per μm (DeFelipe & Fariñas, 1992; Schuz & Palm, 1989). Activation of synapses

occurs via pseudo-randomly generated Poisson spike trains. Background-driven and stimulus-

driven synapses feature different Poisson activation frequencies (0.11 Hz and 0.3 Hz,

respectively) to produce the electrophysiologically-recorded firing rates (Adesnik et al., 2012;

Park et al., 2019).

We modulate the proportion of stimulus-driven excitatory synapses on the apical versus the

basal tree in various experiments in order to examine the effects of synaptic input distribution

on somatic output. In all cases, the amount of inhibition at the soma is 7% of the total

inhibitory synapses (Defelipe et al., 2002). The default synapse distribution was 40% apical to

60% basal ("biologically plausible model"), in adherence with experimental data (excitation:

40% apical, 60% basal; inhibition: 33% apical, 60% basal, 7% soma) (Defelipe et al., 2002). The

"inverse biologically plausible" model, conversely, features a 60% apical to 40% basal

distribution (excitation: 60% apical, 40% basal; inhibition: 55.8% apical, 37.2% basal, 7%

soma). The "even" model features a balanced distribution of synaptic input across both trees

(even: excitation: 50% apical, 50% basal; inhibition: 46.5% apical, 46.5% basal, 7% soma). The

synaptic distribution of the "even" and "inverse biologically relevant" models are not derived

by experimental data, yet they allow for the investigation of apical and basal tree interactions

under conditions of balanced or imbalanced synaptic input. We also make use of other

distributions biased in either the apical or basal direction (referring to stimulus-driven

excitatory input; from 10% basal – 90% apical to 90% basal – 10% apical, in steps of 10%).

Similar to above, background-driven synapses closely follow the stimulus-driven synapse

distribution in these cases, except for background inhibition, which always features 7%

allocated to the soma.

For the results shown In *Supplementary Figure 5,* we kept the inhibitory and excitatory

background inputs fixed at the levels of the "even" distribution, while exclusively varying the

distribution of the stimulus-driven synapses (e.g. from 10% basal – 90% apical to 90% basal –

10% apical, in steps of 10%). This allows us to observe the impact of the full spectrum of

synaptic distributions on somatic output unobstructed by the effects of changes in the

distribution of background inputs. For clarity, we note that the 60% basal – 40% apical

condition of this set of simulations ("60:40" condition) is *not* the same as the "biologically

plausible" model, as it features a different distribution of background inputs.

In our model, orientation preference was assigned on two levels: synaptic (input/single

synapse) and dendritic (whole tree). Single synapse orientation preference is modeled as a

synaptic weight vector biased towards a predefined orientation, which factors into the

activation function of orientation-tuned synapses (Chen et al., 2013). Each synapse is assigned

an orientation preference, with values being 0° to 360° in steps of 10°. This distribution of

orientation preferences to the stimulus-driven synapses at the dendritic tree level followed a

Gaussian probability density function wrapped around the unit circle:

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$$f(s) = \frac{1}{W\sqrt{2\pi}} \sum_{k=-1}^{1} e^{-\frac{(s-b+2\pi k)^2}{2W^2}}$$

where W is the standard deviation (30°; (Chen et al., 2013)), s is the synaptic orientation

preference and b is the mean of the distribution (apical or basal tree mean orientation

preference, ranging from 0° to 90° in steps of 10°). The parameter k controls the wrapping

around the unit circle, and takes the values [-1, -0.5, 0, 0.5, 1]. As a result, the mean

orientation preference of each dendritic tree is biased towards the chosen mean orientation

preference. We evaluated somatic/neuronal orientation preference using the approach

described in section 2.3.3.

2.2. Biophysical Manipulations

The following manipulations aimed to evaluate somatic output under different sets of conditions:

(1) <u>Sodium (channel) blockage</u>: selectively nullify (set to zero) VGSCs (g_{Na}) in a designated compartment (e.g. apical tree, specific basal dendritic segments, or soma).

(2) <u>Synaptic modulation</u>: selectively alter synaptic conductances (g_{AMPA}/g_{NMDA}) in a designated compartment or dendritic tree.

(3) <u>Input manipulation</u>: selectively deactivate specific types of input (excitatory stimulus-driven and background-driven).

2.3. Simulation Protocols

All simulations were performed on our High-Performance Computing Cluster (Rocks 7.0) with 624 cores and 3.328 TB of shared RAM under CentOS 7.4 OS (Papadopoulos et al., 2003), through the NEURON simulation environment (Hines & Carnevale, 2001). Model neuron output measurements were obtained at a sampling rate of 10 KHz (dt = 0.1 ms).

The following simulation protocols were implemented for validation and simulation purposes:

(1) <u>Iterative paired-pulse protocol (I3P)</u>: Used for the evaluation of dendritic non-linearities. Following synaptic silencing and g_{Na} nullification in the entire model neuron excluding the dendritic segment under investigation, a number (1-200, with step 1) of evenly spaced (uniformly distributed) excitatory synapses are synchronously activated (2 pulses at 50 Hz). Voltage at the midpoint of each dendritic segment was recorded to evaluate dendritic non-linear behavior caused by sodium and NMDA spikes.

(2) <u>Stimulation protocol</u>: the model is simulated for 2500 ms (2.5 s), with oriented stimulus

presentation onset at 500 ms (0.5 s), lasting until the end of the simulation (duration 2 s).

(3) <u>Orientation tuning validation protocol</u>: Using the above stimulation protocol, we simulate

10 model "neurons" that feature different synaptic location patterns, defining 10 distinct

neuron IDs. For each one of these neuron IDs, we use 10 different permutations, with

different instances of the input spike trains, resulting in 10 different *simulation IDs* for each

neuron ID. Stimuli of 18 different orientations (0° to 170° in steps of 10°) are then presented

in separate simulations. The tuning curve is extrapolated by averaging firing rates across

simulation IDs for each neuron ID. The overall tuning properties (width, OSI) were then

computed through averaging responses across neuron IDs for each stimulus orientation.

Neuron IDs with an OSI of less than 0.2 or a tuning width greater than 80° (as determined by

a Gaussian fit) are excluded. If more than 20% of neuron IDs do not meet this criterion, the

specific parameter combination is considered to not result in orientation tuning. No neurons

were excluded based on the preceding criteria in our simulations.

(4) <u>Orientation disparity protocol:</u> To evaluate the contribution of the apical and basal trees

in the somatic orientation preference, we keep the mean apical tree orientation preference

fixed at 0°, and set the mean basal tree orientation preference to any one of 10 different

values (0° to 90° in steps of 10°) (see also section 2.1). Thus, we introduce orientation tuning

disparity between the two trees, ranging from 0° to 90°. Akin to the orientation tuning

protocol, we independently generate 10 different neuron IDs with 10 different simulation IDs

for each one and present stimuli as above (section 2.3.3). Orientation preference is then

calculated as described above (section 2.3.3). If more than 20% of the neuron IDs are rejected

(OSI < 0.2 or tuning width > 80°) for a given degree of orientation tuning disparity, then the

model as a whole is deemed to exhibit no orientation tuning for that level of disparity.

In order to determine whether the somatic response follows the mean apical or basal tree

preferred orientation most closely, we compute the *expected* orientation preference of each

neuron ID; that is, the orientation preference assuming a linear summation of the synaptic

inputs. Specifically, we record the assigned orientation preference of all synapses per neuron

ID, and then we use Euler's Formula to calculate the expected orientation preference for this

neuron ID:

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$$E_d = arctan\left(\frac{\sum_i sin(T_i^A) + \sum_j sin(T_j^B)}{\sum_i cos(T_i^A) + \sum_j cos(T_j^B)}\right)$$

Where E_d is the expected orientation preference for a degree of disparity d, i and j are the

numbers of stimulus-driven apical and basal synapses respectively, and T^A and T^B are lists of

synaptic orientation preferences for the apical and basal trees, respectively.

In this particular protocol, a remapping of preferences is required in specific cases (e.g. 180°

→ 0°), since we are not simulating direction selectivity, projecting orientation preferences

into the [-90°, 90°) interval.

(5) <u>Spontaneous activity protocol:</u> Similarly to the orientation tuning validation protocol, we

use 50 different neuron IDs with 100 different simulation IDs for each one. Simulations

feature no stimulus-driven synaptic activity, leaving only background-driven synapses active.

(6) <u>Ionic intervention protocol:</u> Using 50 neuron IDs with 10 different simulation IDs for each

one, prior knowledge of action potential occurrence is obtained via the stimulation protocol

(see section 2.3.2; presented stimulus: 0°, corresponding also to the mean apical and basal

tree orientation preference), using the 'biological plausible' synaptic distribution (section 2.1).

This protocol proceeds similarly to the stimulation protocol (with the notable exception of

somatic g_{Na} being set to 0 to prevent backpropagation) until a time before point t_i, which is

the time of somatic spike occurrence. At t_i – 3 ms, g_{Na} is set to 0 on either the apical or basal

tree dendrites in two separate simulations, until t_i + 1 ms. This manipulation selectively

eliminates the contribution of apical or basal dendritic sodium conductances (seen either as

depolarization or as dendritic spikes) to the somatic spike. This protocol is repeated for all

somatic spikes and the ensuing voltage traces are analyzed to ascertain whether the somatic

spikes survived the manipulation. Spikes rendered extinct only when intervening on the apical

tree are labeled as "apically-driven", while those rendered extinct when intervening on the

basal tree are labeled as "basally-driven". The spikes rendered extinct by either intervention

are labeled as "cooperative". We observed no cases of somatic spikes failing to become

extinct in at least one of the two interventions while using this protocol.

(7) Synaptic intervention protocol: This protocol proceeds similarly to the ionic intervention

protocol (note that somatic g_{Na} is not set to 0 in this protocol), with two main differences: (a)

the intervention reduces stim-driven AMPA and NMDA synapse conductances on either the

apical or basal tree by a specific percentage (e.g. 50%, 90% etc.), rather than setting g_{Na} to 0,

(b) the window of the intervention changes from [t_i – 3 ms until t_i + 1 ms] to [t_i – 30 ms until

t_i +10 ms] (10x increase in window size). This was done to compensate for the relatively low

activation frequencies of stim-driven synapses (0.3 Hz).

2.4 Two-Photon Microscopy

(1) Animals in Experiments: All experimental protocols were approved by Brigham and

Women's Hospital (BWH) Institution Animal Care and Use Committee. Male and female wild-

type (C57BL/6) mice were purchased from The Jackson Laboratory and bred for experiments.

A total of 2 mice were used in these experiments.

(2) Chronic Window Implantation and Sparse Labeling: Viral vectors were purchased from

Addgene. For viral injections and chronic window implantation, 6-10 week old wild type

C57BL/6 mice (both male and female) were anesthetized with isoflurane (1-1.5%). The depth

of anesthesia was assessed via monitoring breathing rates. Meloxicam (2.5mg/kg) and

Buprenorphine (0.1mg/kg) were administered subcutaneously every 12 hours following the

surgical procedure for 2 days. A headpost was implanted on the skull and a 3mm diameter

craniotomy was made over the visual cortex of the left hemisphere. The craniotomy was

centered 2.7mm lateral to the midline and 1.5mm posterior to the bregma. To sparsely label

pyramidal neurons with GCaMP6s, a mix (~90 nl per site, up to 2 sites) of diluted CamKII-CRE

(AAV1, diluted 80,000X) and flex-GCaMP6s (AAV5, diluted up to 2X) was injected slowly over

5 minutes per penetration using a Drummond Nanoject III. Up to two penetrations ~0.5mm

apart on average were performed per craniotomy. After the viral injection, a round coverslip

was fitted to the craniotomy and sealed with vetbond and dental cement. Most chronic

windows in our hands remain clear for 2-3 months. 2-photon imaging was performed on week

3-4 following viral injection, at which time GCaMP6s expression was optimal.

(3) In vivo calcium imaging of sparsely labeled neurons: Three to four weeks after the

injection, each GCaMP6s expressing mouse was sedated with Fentanyl (0.5 mg kg^{-1}) and

Dexmedetomidine (0.5 mg kg^{-1}) (Mrsic-Flogel et al., 2007) for imaging experiments. A stable

level of anesthesia was confirmed by stable breathing rate and lack of movement. Neuronal

and dendritic calcium activity was recorded from L2/3 pyramidal neurons in V1. We chose

neurons with 1 to 3 primary basal dendritic segments clearly visible up to their primary

branching point on the same plane with the soma. Apical dendrites were not present in the

focal plane during recording. Mice were in the dark during the imaging process and their

spontaneous firing rate was recorded. Images were acquired at ~7.3 frames/sec using an

Ultima IV microscope in spiral scanning mode with a 25X 1.0 NA Nikon objective (5-30mW

laser power, 900 nm). Recordings lasted for 283 s (2750 frames). The size of the recording

FoV was 103 μm^2, and the pixel size was 0.201 μm^2. 12 pyramidal neurons were imaged, and

11 of these were selected for analysis. The final neuron was not selected due to an insufficient

number of basal dendritic segments being positively identified as belonging to said neuron

following visual inspection of a z-stack image.

2.5. Data Analysis

Data analysis of both modeling and two-photon imaging experiments was performed through

Python 3.8+, using publicly available libraries as well as custom-made scripts and functions.

The latter include:

(1) <u>Dendritic Spike Detection</u>: Dendritic sodium spikes were identified via a detection

algorithm that identified depolarizations exceeding a -20 mV threshold. For the identification

of NMDA spikes in the I3P experiments (see section 2.3.1) we located the inflection points

(points where the second derivative is zero) of the voltage trace after the second pulse. As

the NMDA spike exhibits a voltage plateau, 2 or more inflection points indicate an NMDA

spike. As our voltage measurements are not continuous and the I3P protocol does not feature

noise, it is unlikely to encounter a measurement point at which the second derivative is

exactly zero. Thus, we assume an inflection point exists at some recording point P_n if and only

if $(P_{n-1} \cdot P_{n+1}) < 0$, where P_{n-1} and P_{n+1} refer to existing points in the second derivative of the

voltage trace, immediately preceding and following the theorized inflection point.

(2) <u>Quantification of Dendritic Nonlinearities</u>: Using data from the I3P experiments, we

calculated the maximum amplitude of dendritic segment depolarization per synapse count. A

linear input-output curve is then generated, extrapolating from the response of the dendritic

segment to a single synaptic input (unitary response). This curve is compared to the actual

response of the dendritic segment to increasing input in an "Expected vs. Actual" plot, thus

coarsely characterizing the integration mode of the dendritic segment. The integration mode

of the dendritic segments was quantified in detail using the Nonlinearity Relative to Linear

Extrapolation (NRLE) metric (Behabadi et al., 2012), which is defined as the maximum ratio of

actual to expected neuronal output signal. An NRLE value of less than 1 denotes a sub-linear

dendritic segment, NRLE of exactly 1 indicates a linear dendritic segment and NRLE values

over 1 characterize a dendritic segment as supra-linear. The non-linearity threshold of all

dendritic segments was also calculated, measured as the minimum number of synapses

required to elicit the corresponding electrogenic event (sodium or NMDA dendritic spike).

(3) Two-Photon Imaging Data Analysis: Recordings are motion corrected via the suite2p

software (Pachitariu et al., 2016) to accurately extract fluorescence values from dendritic

segments. In the case of somatic activity data, we use the original video, as motion correction

is not necessary at the level of magnification used. Using the Fiji ImageJ software (Schindelin

et al., 2012), ROIs are manually defined, using a maximum intensity z-stack as a reference

image. Dendritic segments are identified as belonging to the cell of interest via visual

inspection of a z-stack recording. Each dendritic segment is broken up into multiple ROIs. Only

one ROI is defined for the soma. We completely avoid areas where there is an overlap of the

cell of interest with nearby dendrites/axons of cells from outside the field of view. Dendritic

ROIs are slightly wider than the dendritic segments themselves, to partially account for

misplacement of the ROI or uncorrected movement (drift) of the dendrite. 4 equal-sized

circular ROIs are also defined in areas where no visible dendritic segments or somata can be

found to calculate the background fluorescence level.

(4) Calcium Trace Processing: Raw calcium fluorescence data is normalized via a moving

window average of the fluorescence trace (window size of 20 s), subtracting F_0 from it

(defined as the average of the 20% lowest fluorescence values of said moving 20s window),

then dividing by the same value (F0), finally producing a normalized $\Delta F/F_0$ value. These $\Delta F/F_0$

values were significantly different from the background $\Delta F/F_0$ as calculated from the 4 circular

reference ROIs (cross-correlation test; $p < 0.05$; no manually-defined ROIs were excluded

through this process).

(5) <u>Calcium Event Detection:</u> Calcium events are detected via a simple dual-threshold

approach: a calcium event is registered when the $\Delta F/F$ value exceeds the 93.5^{th} percentile at

the same time as the derivative of the $\Delta F/F$ value exceeds the 67th percentile. These percentile

values were calculated via trial-and-error, attempting to match the expected firing frequency

of a neuron (soma) under spontaneous activity conditions (approximately 0.2 Hz (Haider et

al., 2013)). To ensure that the algorithm returns true events rather than random noise, we

performed an Event-Triggered Average (ETA) analysis. We extracted all of the detected event

waveforms, centered them on the time of detection, and averaged across waveforms. We did

this separately for somatic and dendritic events (*Figure 2C)*, as well as for each animal

separately (*Supplementary Figures 4A, 4B*). Calcium event-triggered averaging of the

processed fluorescence signal yields the expected temporal profile, indicating that these

events are not likely to represent noise.

(6) <u>Generation of "fluorescence-like" traces from model voltage traces:</u> 50 neuron IDs were

simulated with 100 simulation IDs each, using the "biologically plausible" configuration under

the spontaneous activity protocol, in order to closely match the properties of the real neurons

under spontaneous activity conditions. Next, we ran a spike detection algorithm to find all

somatic and dendritic spiking events. We binarized the traces, using 1 for spike peaks and 0

for everything else. We then convolved these binarized traces with a GcaMP6s kernel function

of our own design:

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$$F(t) = -s \left(e^{-\frac{t-t_0}{t_r}} - e^{-\frac{t-t_0}{t_d}} \right)$$

Where t is time (in ms), s is a scaling factor (s ≈ 0.5714), t0 is the time of the event, t_r is the

rise time (t_r = 74 ms) and t_d is the decay time (t_d ≈ 401.45 ms). The output of the function is

unitless, representing fluorescence. With these parameters, our kernel function has an actual

rise time constant of 194.225 ms, with output values decaying by ~52.11% within 0.6 s

(*Supplementary Figure 2A*). These values closely match experimentally observed values for

GCaMP6s (Chen et al., 2013) (*Supplementary Figure 2B*). The ensuing fluorescence traces

were then downsampled to a dt of 0.138 s (~7.25 Hz) from the original dt of 0.1 ms (10 KHz),

to emulate the sampling rate of two-photon calcium fluorescence imaging. The transformed

traces were then processed by the same pipeline as true calcium fluorescence data (see points

4 and 5 above).

(7) Electrotonic length constant calculation: The electrotonic length constants (mentioned

also in text as "electrotonic constants") of all dendritic segments were calculated using the

formula from van Elburg & van Ooyen (2010):

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$$\lambda_i = \sqrt{\frac{b_i r_m}{2 r_a}}$$

376   Where i refers to the dendritic segment, λ_i is the electrotonic length constant, b_i is the radius

377   of the branch, r_m is the specific membrane resistance and r_a is the intracellular resistivity.

3. Results

3.1. Varied Dendritic Non-Linearities in the Model Neuron

To ensure the validity of our modeling approach, we first examined dendritic integration by characterizing the sodium spike properties of all dendritic segments separately (*Figure 1A*), wherein a specific number of synapses is synchronously activated by two pulses at 50 Hz while all other compartments are modeled as passive (I3P; see *Methods* 2.3.1), and the response of each dendritic segment is assessed. Results show that the model can produce realistic dendritic spikes (*Supplementary Figures 1C, D*), while the properties of the spiking events vary between dendritic segments (*Figures 1B, C, D, Supplementary Figures 1A, B*). Both apical and basal dendritic segments exhibit a large degree of variation in their spiking thresholds (*Figures 1E, F;* threshold synapse count mean and standard deviations: basal 12.286±6.649 sodium, 51.429±13.179 NMDA; apical 14.442±13.825 sodium, 40.721±23.675 NMDA). Dendritic segments with high sodium spiking thresholds also tended to exhibit higher NMDA spiking thresholds. To quantitatively characterize the nonlinear behavior of each dendritic segment we used the Nonlinearity Relative to Linear Extrapolation (NRLE) metric (Behabadi et al., 2012), which is an estimate of the degree of non-linear dendritic integration within a specific branch (see *Methods,* section 2.5.2). Results indicate that all dendritic segments of the model neuron exhibit non-linear input-output relations (*Figure 1D,* NRLE=1 denotes linear, NRLE<1 sublinear, and NRLE>1 supra-linear), in line with experimental observations (Smith et al., 2013; Wilson et al., 2016). In addition, the NRLE values are relatively widely dispersed (values ranging from >1 to ~3.5), providing a large dynamic range of possible dendrite-level computations. The above results show that there is diversity in apical and basal dendritic integration that can be exploited to modulate orientation tuning. These differences are

potentially, at least in part, the result of differences in the morphological and

electrophysiological features of these dendritic segments, as we found that sodium spiking

propensity seems to be correlated with the electrotonic length constant for each dendritic

segment (*Figure 6C*).

Figure 1. Dendritic Properties of a L2/3 V1 Neuron Model*. A. Graphical representation of the iterative paired-*

pulse protocol. Green dots represent allocated synapses. Synapses were activated only within the designated

(red) dendritic segment, with all other segments being passive. We record the voltage response of the dendritic

segment with NMDA channels disabled or enabled, acquiring information on dendritic sodium and NMDA spikes,

respectively. B. Example of an "expected vs. actual plot", showing the response (blue) of a single basal dendritic

segment to 1 to 200 simultaneous synaptic inputs compared to a linear (red). C. Same as B, but for an apical

dendritic segment. D. NRLE value histogram for all apical (red) and basal (black) dendritic segments. E. Bar plot

of per-dendrite nonlinearity thresholds for all basal dendritic segments, sorted by the sodium spike thresholds.

Note that NMDA spike thresholds partially follow the sodium spike threshold increase. F. As E, for all apical

dendritic segments.

3.2. Model Neuron Matches *in vivo* Spontaneous Activity Patterns

Having validated the non-linear behavior of each dendritic segment, we now need to

adequately assess background-driven activity in our model. To assess this, we compare L2/3

V1 pyramidal neurons that were either imaged *in vivo* using two-photon calcium fluorescence

imaging, or simulated in our computational model. Specifically, simultaneous recordings from

basal dendritic segments and somata of L2/3 V1 pyramidal neurons were performed in

anesthetized mice (spontaneous activity of 11 neurons from 2 mice, see *Methods*, section 2.5;

Figure 2A). These imaging data were compared to modeling data generated using the

spontaneous activity protocol, whereby simulations feature only background activity, i.e.

noise, in the absence of a visual stimulus (see *Methods*, section 2.3). The resulting voltage

traces from the soma and basal dendritic segments were converted into approximations of

two-photon fluorescence traces (*Figure 2B* and *Supplementary Figure 2*, see *Methods*, section

2.5.6 for details) to allow a direct comparison with our *in vivo* calcium imaging data that were

collected under similar conditions (*Figures 2A - F*).

Significant fluorescence increase "events" were identified for both datasets using a custom

detection algorithm (see *Methods,* section 2.5.5). To ensure that the algorithm detects true

events rather than noise in the two-photon signal, we performed an event-triggered

averaging (ETA) analysis of trace segments centered on the event onset timepoint (see

Methods, section 2.5.5; 1131 dendritic events, 96 somatic events across all 11 neurons and

their respective dendritic ROIs). The results of this analysis indicate that random noise is

unlikely to be the cause of the detected events, as there is a marked increase in fluorescence

intensity at the time of event onset for all detected events (*Figure 2D*, *Supplementary Figures*

3, 4) which persists over time.

Next, in order to infer the relationship between basal and somatic events, we looked at their

relative timing differences. Using dendritic events as the starting point, we paired each

dendritic event with the most temporally proximal somatic event, and computed the time

difference between the two for each event pair (measured in imaging frames, each lasting

approximately 138 ms). Using the same approach, we measured the fraction of somatic-

dendritic event pairs in our model and compared them to the *in vivo* data. Event pairs

featuring a temporal distance of zero (occurring within the same frame) were excluded as

they are most likely to represent backpropagating events (excluded event pairs: model:

219,291/245,717 event pairs, 89.25%; *in vivo*: 29,143/237,572 event pairs, 12.27%; difference

attributed to a lack of noise-related fluctuations in the model data compared with the *in vivo*

data). The resulting distributions (*Figures 2E, F*) show that in both the model and *in vivo* data

local dendritic events can be generated independently of somatic events. Paired events

separated by more than 4 recording frames (~552 ms) are highly unlikely to be the result of

backpropagation, even when considering the slow kinetics of the calcium sensor. Moreover,

the somatic events that are preceded by basal dendritic events are similar in proportion

between datasets (Figures 2E, 2F red bins: 48.79% model, 48.12% actual when including all

event pairs with a non-zero timing difference; 24.3% model, 27.8% actual when including only

event pairs with an absolute timing difference >4 recording frames). This analysis indicates

that a notable number of detected somatic events are *not* closely preceded by any detected

basal dendritic events, suggesting that other events, perhaps arising in the apical tree or not

visible basal dendritic segments, contribute to somatic activity in such cases. It is important

to note that the analyses herein are inherently subject to the noisy nature of the recordings,

the limitations of the slow kinetics of the calcium indicator, the low temporal resolution of

the recording (frame time of ~138 ms), as well as the possibility of undetected events.

To mitigate the risk that the *in vivo* findings are the result of noise or random chance, we

constructed a null distribution of timing differences for the *in vivo* data by independently

circularly shuffling the time-series data from the somatic and various dendritic segment ROIs

1000 times, and then re-computed the timing differences. We then compared the actual data

with the null distribution data using the two-sample Kolmogorov-Smirnov (KS) test, finding

that the two distributions of timing differences were significantly different (*Supplementary*

Figure 4C; test statistic ~0.105, p << 0.05), whereas a similar analysis of completely randomly

generated event timings would results in identical (not statistically different) distributions for

event pair timing differences.

In conclusion, we evaluated whether the background-driven activity of the model adequately

approximates the dendro-somatic firing patterns seen in *in vivo* two-photon recordings

(*Figure 2A*) of spontaneous activity in mice (*Figure 2C*). Both model and *in vivo* data feature

"decoupled" dendritic events (*Figures 2E, 2F*), whereby dendritic events occur far in advance

of the most temporally proximal somatic event. These events are not likely to be the result of

noise, as event-triggered averaging analysis yields a distinct waveform and temporal profile

(*Figure 2D, Supplementary Figures 3, 4A, 4B*). In addition, comparing the null distribution with

the true data using the KS test, we find that the two are different (*Supplementary Figure 4C*).

Thus, we can conclude that the model adequately reproduces *in vivo* spontaneous activity

patterns.

Figure 2. Comparing the Model Output with Two-Photon Imaging Data. A: Image from one of the neurons in the dataset. Note the separate, multiple ROIs per dendritic segment, as well as the 4 circular ROIs used for background fluorescence calculation. B: Example traces from the model data conversion pipeline. Voltage spiking data is binarized and convolved with a GcaMP6s kernel function (Supplementary Figure 2). Resulting data is downsampled to two-photon sampling rates (~7.25 Hz) from the original sampling rate (10 KHz), and processed

exactly like in vivo *two-photon fluorescence data. Red vertical lines show the timing of the original spikes. C:*

Example somatic (top) and distal dendritic (bottom) ΔF/F₀ traces from the neuron in (A). Black dots at the origin

of colored dashed lines denote detected events. Note that somatic events (green dashed lines/arrows) are not

necessarily coincident with dendritic events (red dashed lines/arrows). D: Event-triggered averages for all

somatic (blue) and basal dendritic segment ROIs (1 dt ≃ 0.138 s; error bars: Standard error of the mean). E: Event

pair timing difference histogram for the two-photon data. 48.12% of event pairs feature basal dendritic events

that are decoupled from the soma (red; dendritic events occur at least 1 dt before the onset of a somatic spike),

with 27.8% being decoupled if the 4 bins (~552 ms) closest to zero are excluded (excluded data marked as lightly

shaded bars). F: Event pair timing difference histogram for the converted model data. 48.79% of event pairs

feature basal dendritic events that are decoupled from the soma (red; dendritic events occur at least 1 dt before

the onset of a somatic spike), with 24.3% being decoupled if the 4 bins (~552 ms) closest to zero are excluded

(excluded data marked as lightly shaded bars). Note that in both D and E, the zero-difference bin was removed,

as it is assumed to contain the majority of backpropagating events. The middle two bars in F have been truncated

for visualization purposes; the corresponding values are (left to right) 0.15958 and 0.16533, respectively.

3.3. Model Neuron Features Robust Orientation Tuning

We have now shown that the model can accurately reproduce spontaneous activity patterns

seen *in vivo*. The following experiments simulate stimulus-driven activity and orientation

selectivity. To do so, we implement the orientation tuning validation protocol (*Figure 3C;* see

Methods, section 2.3.3) for the biologically plausible model (see *Methods*, section 2.1), and

derive the orientation tuning curve of the neuron (preferred orientation firing rate: 1.56±0.26

Hz, orthogonal orientation firing rate: 0.22±0.12 Hz, OSI 0.78±0.03, tuning width 43±1.45°;

Figure 3A).

Next, we tried to disentangle the effects of the apical and basal trees by introducing dendritic

disparity (*Figure 3B;* see *Methods,* section 2.3.4). To do so, we make use of the disparity

protocol (see *Methods,* section 2.3.4; *Figure 3B*) in which the apical and basal trees are tuned

to different orientations: the apical tree features synaptic orientation preferences sampled

from a Gaussian distribution with a mean of 0°, whereas the basal tree uses the same method,

but with a different mean for its own distribution (0° - 90°, in steps of 10°). The expected

orientation preference of a neuron is calculated using Euler's Formula (see *Methods,* section

2.3.4).

We observe that under biologically plausible conditions, the model neuron appears to exhibit

a bias towards the preferred orientation of the basal tree – an example of *basal dominance*

(*Figure 3D*). Note also that the model did not exhibit adequate tuning beyond a dendritic

disparity of 40° (20% or more of neurons showed OSI<0.2 or tuning width >80°; see *Methods*,

2.3.3) To assess whether this is a feature of the model in general, or if it is the result of the

particular synaptic distribution used in the biologically plausible model, we repeated the

experiment using two different synaptic distribution profiles – the "inverse biologically

plausible" and "even" models (see *Methods,* section 2.1). We find that, indeed, the bias

displayed by the neuron appears to be decreased in the "even" model, and is clearly reversed

in the "inverse biologically plausible" model. Similar to the "biologically plausible" model,

these two models also did not feature adequate tuning past a certain dendritic disparity, 50°

in these cases. These results indicate that neuronal orientation preference is shaped, at least

in part, by the synaptic distribution along the two dendritic trees, as well as their mean

orientation preferences. In addition, orientation tuning persists up to a disparity of 40° in the

biologically plausible model, and up to 50° in the other models.

Figure 3. Neuronal Orientation Preference is Robust and Resists Dendritic Tuning Disparity. *A. Orientation*

tuning curve for the "biologically plausible" model. B: Example configuration of a model neuron set to a disparity

of 40°. Right-side polar plots display the distribution of synaptic orientation preferences for the indicated

dendritic segments. Left-side polar plots display the distribution of synaptic orientation preferences for the apical

(top) and basal (bottom) dendritic trees. C: Example voltage traces recorded at the soma of a model neuron from

(A). Note the clearly different responses between the preferred orientation (0°) and all others. D: Orientation

preference plot for the "biologically plausible" model featuring various degrees of disparity. Grey dashed line

denotes the expected orientation preference per degree of disparity. Black and red arrows point in the direction

of basal and apical dominance, respectively. E: As D, but for the "even" model. Note that dominance has shifted

from the basal domain towards the apical domain. F: As D, but for the "inverted biologically plausible" model.

This configuration clearly features apical dominance. Error bars: Standard error of the mean, for all panels.

To fully characterize the effect of synaptic distribution on the contribution of each tree to

neuronal tuning, we allocated different percentages (10%, 20% etc., up to 90%) of the total

synapses (including all stimulus-driven, background-driven, excitatory and inhibitory) on a

given tree (see *Methods,* sections 2.1 and 2.3.4), in a manner similar to how we generated

the "inverse biologically plausible" model. As seen above (*Figure 3D, E, F*) somatic orientation

preferences above the expectation curve produced by Euler's Formula indicate basal tree

dominance, whereas those below the expectation line indicate apical tree dominance. We

observe that the neuron does not always follow the expected orientation preference,

indicating that the actual combined effect of apical and basal synaptic inputs is not linear. As

such, apical and basal stim-driven inputs appear to cooperate in a non-linear fashion to

produce somatic output.

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3.4. Dendritic Sodium Spikes Alone are Insufficient to Explain Somatic Activity

The previous results show that orientation tuning is the result of the interaction between

apical and basal trees, but were not enough to elucidate the exact contribution of each tree.

In order to advance our understanding of the relative roles of each tree, we need to selectively

intervene and interrupt specific aspects of their operation, judging their role by the effect of

the intervention. The most obvious target for such an intervention is the sodium spiking

activity of each dendritic tree, controlling for the type of input received (stimulus-driven vs

background-driven). As such, we used the ionic intervention protocol (see *Methods* 2.3.6) in

order to remove either the apical or basal sodium component of somatic activity.

We start by using the "biologically plausible" model with no alterations ("Control

background") and a preferred (0°) stimulus (case #1). In this condition, both stimulus-driven

and background-driven synapses are active, and most synapses (60%; both stimulus-driven

and background-driven) are allocated on the basal tree. We perform the intervention

protocol for every recorded somatic spike under these conditions, and categorize these spikes

in terms of dendritic sodium conductance dependence: "apically-driven" (dependent on the

presence of apical sodium conductances), "basally-driven (dependent on the presence of

basal sodium conductances) or "cooperative" (dependent on the presence of both apical and

basal sodium conductances) (*Figure 5A*). We find that the majority of somatic spikes were

dependent on apical sodium conductance (apically-driven 70.18%, basally-driven 12.28%,

cooperative 17.54%; *Figure 5B)*, which is surprising, considering the proximity of the basal

tree to the soma, as well as the nature and relevance of its feedforward inputs. These somatic

events are both stimulus-driven and background-driven, so in order to separate the two, we

change the orientation of the stimulus being presented to 90°, rendering the vast majority of

stim-driven synapses inactive, while keeping background-driven synapses active as normal

(case #2). We find that the percentages change drastically compared to the previous condition

(apically-driven 33.33%, basally-driven 8.33%, cooperative 58.33%; *Figure 5D)*, indicating that

the dependence of spiking activity on dendritic sodium conductance differs between cases of

stimulus-driven and background-driven activity. In other words, the causality of spike

generation differs between cases of stimulus-induced spiking versus spontaneous activity-

induced spiking - at least in terms of their dependence on dendritic sodium conductances.

Importantly, however, we also need to eliminate the possibility of background-driven

synapses being the main determinant of somatic spiking. Although background-driven

synapses feature low activation frequency (0.11 Hz vs 0.3 Hz for stimulus-driven), they are

also vastly more numerous compared to stimulus-driven synapses (75% versus 25% of total

synapses). To equalize the potential contribution of background-driven synapses to apical and

basal tree output, we introduce a new condition ("Evenly distributed background") whereby

the model is configured in the same manner as in case #1 (i.e. "biologically plausible" model),

but we change the distribution of background-driven synapses to be the same on the apical

and basal trees (i.e. excitatory: 50% apical, 50% basal; inhibitory: 46.5% apical, 46.5% basal,

7% soma). In this way, we ensure that the normally increased number of background-driven

synapses on the basal tree is no longer a factor. Presenting a stimulus of the preferred

orientation (0°) under these conditions (case #3), we find that the resulting percentages

(apically-driven 85.53%, basally-driven 6.58%, cooperative 7.89%; *Figure 5C)* heavily resemble

those seen in case #1, indicating that the change in background-driven synapse distribution

did not substantially affect spike generation overall. By changing the orientation of the

presented stimulus to 90°, we can evaluate whether the same holds for the background-

driven spikes only (case #4). This does not appear to be the case, however, as the results

(apically-driven 95.45%, basally-driven 4.55%, cooperative 0.00%; *Figure 5E)* do not resemble

those seen in case #2. This implies that although spike generation overall is not substantially

affected, the minority of spikes that are background-driven *are* affected, becoming reliant

upon apical sodium conductances and completely losing their prior cooperatively-driven

nature. One possible explanation for these results is that basal dendritic segments do not

contribute significantly via sodium spikes, but also via AMPA- and NMDA-driven

depolarizations - a possibility that we will explore further.

To further investigate how synaptic distribution impacts somatic spike dependence on

dendritic sodium conductance, we repeated the experiments above, but this time using a

variety of synaptic distribution profiles whereby the percentage of synapses on the basal tree

varies from 10% to 90% in steps of 10% (see *Methods,* section 2.1), with background-driven

synapses either following the stimulus-driven synapse distribution (control case) or being

evenly distributed (evenly distributed noise case). In the control case with a preferred

stimulus, we observe that as the basal tree receives more stimulus-driven synapses, the

dependence of somatic spikes on basal sodium conductance increases (*Supplementary Figure*

5A). In addition, the dependence of somatic spikes on apical sodium conductance never

reaches zero, even in the most extreme bias regime (*Supplementary Figure 5A*). Inversely, as

the apical tree receives more and more synapses, the dependence of somatic spikes on apical

sodium conductance increases, whereas the corresponding dependence on basal sodium

conductance actually reaches zero (*Supplementary Figure 5A*). A similar trend can be seen in

the control case using a non-preferred stimulus (*Supplementary Figure 5C*). Conversely, the

evenly distributed noise case does not feature any strong trends (*Supplementary Figure 5B*),

as the relative abundance of background-driven inputs overshadows the effect of the

relatively few stimulus-driven inputs (75% of excitatory input is background-driven, with the

remaining 25% being stimulus-driven (Chen et al., 2013; Park et al., 2019). Presentation of a

non-preferred stimulus results in identical results from all model configurations in this case,

as background-driven synapses are evenly distributed in all configurations, and the non-

preferred stimulus does not lead to activation of stim-driven synapses (*Supplementary Figure*

5D).

Figure 4. Neuronal Output Relies Disproportionately on Apical Sodium Conductance. A: Diagram describing the ionic intervention protocol, with example traces on the right. Shaded areas on the traces denote the pre-spike (grey) and post-spike (red) intervention time windows. B: Somatic spike dependence on dendritic sodium conductance for the "biologically plausible" model with a stimulus of 0° (preferred orientation). C: Same as B, but background synapses follow the "even" model (evenly distributed), while stimulus-driven synapses follow the "biologically plausible" (60% basal, 40% apical) model. D: Somatic spike dependence on dendritic sodium conductance for the "biologically plausible" model with a stimulus of 90° (orthogonal/non-preferred orientation). E: Same as D, but background synapses follow the "even" model (evenly distributed), while stimulus-driven synapses follow the "biologically plausible" (60% basal, 40% apical) model.

3.5. Apical and Basal Dendrites Influence Somatic Tuning via Different Mechanisms

The previous results contain a counter-intuitive observation: although we would expect both apical and basal dendritic segments to contribute to somatic output, it appears that the contribution of the apical tree overshadows the basal tree significantly, at least as far as sodium conductance is concerned. Naturally, the basal tree, being the bearer of the majority of feedforward inputs, is very unlikely to have a minor role in the function of the neuron, and orientation tuning in particular. As such, it is reasonable to assume that it exerts its influence via a different mechanism. To assess the veracity of this assumption, we performed a series of experiments aimed at evaluating the impact of different ionic (see *Methods,* sections 2.2.1 and 2.3.6) and synaptic (see *Methods,* sections 2.2.2, 2.2.3 and 2.3.7) mechanisms on neuronal orientation tuning, using a stimulation protocol (see *Methods,* section 2.3.2) with the biologically plausible model.

First, we assess the tuning characteristics of the neuron when stimulus-driven input only is removed either from the apical (apical target) or from the basal tree (basal target) (see *Methods,* section 2.2.3), and compare it with the control case (in which both trees receive input as normal). We find that the neuron remains tuned in both cases, albeit with significantly decreased OSI values (t-test: control vs basal target: p = 0.0002; control vs apical target: p = 0.008) (*Figure 5A*).

Next, we use a synaptic intervention protocol to selectively reduce synaptic (AMPA and NMDA) weights by 50% on either the apical (apical target) or basal tree (basal target) for a brief time only, and only when a somatic spike is about to occur (see *Methods,* section 2.3.7), comparing with the control condition where no intervention takes place. Results show that tuning is completely abolished when basal AMPA and NMDA weights are reduced (*Figure 5B*),

whereas tuning persists, albeit significantly affected, in the apical intervention case (t-test;

control vs apical target: $p \approx 0.00004$). We also perform a similar experiment, reducing AMPA

and NMDA weights by 90% throughout the simulation for either the apical (apical target) or

basal tree (basal target) (see *Methods,* section 2.2.2), and comparing with the control case.

We find that the neuron remains tuned in these cases, albeit with significantly decreased OSI

values (t-test: control vs basal target: $p = 0.0011$; control vs apical target: $p = 0.0168$)

(*Supplementary Figure 6*). These results indicate that both apical and basal synaptic (AMPA

and NMDA) activity is necessary for orientation tuning.

We then move on to assessing the impact of sodium conductance nullification on orientation

tuning. We use an ionic intervention protocol to selectively nullify sodium conductance on

either the apical (apical target) or basal tree (basal target) for a brief time only, and only when

a somatic spike is about to occur (see *Methods,* section 2.3.6), again comparing with the

control condition where no intervention takes place. We discover that precise loss of apical

sodium conductances completely abolishes orientation tuning in the neuron (*Figure 5C*), but

precise loss of basal sodium conductances does not affect orientation tuning (t-test; control

vs basal target: $p = 0.3473$). The next experiment is similar, and less temporally precise – we

completely nullify sodium conductance on either the apical (apical target) or basal tree (basal

target) throughout the experiment, comparing with the control where no change in

conductance has taken place (see *Methods,* section 2.2.1). Unsurprisingly, the results are

nearly identical - total loss of apical sodium conductances completely abolishes orientation

tuning in the neuron (*Supplementary Figure 6*), but total loss of basal sodium conductances

does not affect orientation tuning (t-test; control vs basal target: $p = 0.6767$).

From these experiments we can surmise that apical sodium conductance is necessary for both

spike generation in general (*Figure 4, Supplementary Figure 5)* as well as for orientation tuning

in particular (*Figure 5C, Supplementary Figure 6)*. Conversely, AMPA and NMDA activity is

necessary for both apical and basal trees (*Figures 5A, 5B, Supplementary Figure 6)*. This means

that the apical tree most likely exerts its influence on somatic spiking via AMPA- and NMDA-

driven sodium spikes, whereas the basal tree mostly utilizes AMPA- and NMDA-driven

depolarizations, rather than sodium spikes.

Figure 5. Orientation Tuning Critically Depends on Apical Sodium Conductance and AMPA/NMDA Activity. *A:*

Orientation tuning curves (left) and mean OSI values (right) for the experiment where stimulus-driven synapses

are completely absent on either the apical ("apical target"; red) or basal ("basal target"; green) tree. B: Selective

reduction of stim-driven synaptic (AMPA/NMDA) weights by 50% 30 ms prior to somatic spike occurrence (lasting

up to 10 ms after the pre-recorded somatic spike timing), on either the apical (apical target; red) or basal (basal

target; green) tree. C: Nullification of sodium conductances 3 ms prior to somatic spike occurrence (lasting up to

1 ms after the pre-recorded somatic spike timing), on either the apical (apical target; red) or basal (basal target;

green) tree. Error bars: Standard error of the mean, for all panels. "n/a" denotes a lack of orientation tuning for

that experimental setup. Stars denote statistical significance at a significance level (α) of 0.05 (*), 0.01 (**) or

0.001 (***), whereas "n.s." stands for "no statistically significant difference", i.e. p-value ≥ α.

3.6. Morphological and Electrophysiological Properties Influence Dendritic Behavior

It is now clear that apical and basal dendritic segments exert causal influence on somatic spike

generation via separate mechanisms. In *Figure 1*, we showed the wide array of dendritic non-

linear behaviors present in the dendritic segments of the neuron. Here, we investigate the

relationship between the sodium spiking threshold of dendritic segments (henceforth

"threshold") and specific morphological and electrophysiological properties, attempting to

link the physiology of dendritic segments with their contribution to somatic spiking.

We first look into the diameter of dendritic segments, finding that there is a rough linear

relationship between diameter and threshold (*Figure 6A*). Outliers, marked as "atypical",

feature small diameters and very short lengths (lengths: 16.2±10.5 μm; diameters:

0.711±0.027 μm) but relatively high sodium spiking thresholds. Next, we look into distance

from the soma rather than diameter. Given the morphology of the neuron (see *Figure 6D* for

an illustration), basal dendritic segments are clearly much closer to the soma compared with

the majority of apical dendritic segments, located far from the soma, on the apical tuft. We

find that there is no clear relationship between distance and threshold (*Figure 6B*). The last

parameter we will look into is the electrotonic length constant (see *Methods,* section 2.5.7).

In order to avoid comparing basal dendritic segments with apical trunk or atypical dendritic

segments that feature extremely high thresholds, we excluded all such dendritic segments,

as well as any others matching their electrotonic length constant values, from further analysis

(*Figure 6C*; trunk segments marked in teal, atypical dendritic segments marked in green, grey

area denotes excluded dendritic segments). After excluding these dendritic segments, we

performed a linear regression for the remaining apical and basal dendritic segments,

generating one fit line for each (*Figure 6C;* basal dendritic segments: red dashed line, apical

dendritic segments: black dashed line). Inspection of the regression fit lines reveals that basal

dendritic segments feature higher sodium spiking thresholds than apical dendritic segments

for the same electrotonic length constant values, offering a potential explanation for the large

role apical sodium conductances play in shaping somatic output.

Finally, we can combine our observations of the contributions of apical and basal dendritic

segments to the operation of the neuron. We predict that neuronal output in L2/3 V1

pyramidal cells depends on the timely interaction of appropriate apical and basal inputs to

the soma (i.e. apical sodium spikes coinciding with basal depolarizations and/or spikes), a

phenomenon we dub *Bilateral Input Coincidence (BIC)* (*Figure 6D*): rather than solely relying

on inputs from one tree alone, L2/3 V1 neurons rely on the cooperation between apical and

basal inputs to produce their orientation-tuned output.

Figure 6. Neuronal Output as a Function of Physiologically Distinct Dendrite Outputs. *A: Sodium spiking*

threshold for basal (black), apical (red) and atypical (morphological outliers, typically featuring unusually high

thresholds; green) dendritic segments as a function of their diameter. B: As A, but for dendritic segment distance

from the soma (measured from their midpoint) rather than diameter. C: As A, but for the electrotonic length

constant rather than distance to the soma. Trunk segments (proximal and distal trunk) marked in teal. Red and

black dashed lines indicate lines fit using linear regression for the basal (red) and apical (black) dendritic

segments, respectively, excluding all dendritic segments within the area marked in grey. Note that for similar

values of the electrotonic length constant, the basal dendritic segments feature a higher sodium spiking

threshold. D: Schematic representation of Bilateral Input Coincidence (BIC). Somatic spiking is achieved when

apical sodium spikes coincide with basal depolarizations and/or spikes. Basal spikes combined with apical

depolarizations or depolarizations from both trees are generally unable to elicit somatic spiking.

4. Discussion

In this work, we have delineated the dendritic constituents of neuronal computation in L2/3

V1 pyramidal cells using a detailed computational model. We showed that the model exhibits

a wide range of non-linear behavior on the dendritic segment level (*Figure 1*), and that it can

accurately model the output of real neurons under spontaneous conditions (*Figure 2*).

Inspecting the behavior of the neuron under conditions of tuning disparity, we find that it

remains robustly tuned (OSI < 0.2 and tuning width > 80°) for non-extreme dendritic disparity

conditions, although the exact orientation preference does not necessarily always match the

expected one (*Figure 3*), indicating that apical and basal inputs exhibit non-linear

cooperativity. This is seemingly contradicted by our following finding, where we show that

the dependence of somatic sodium spikes on basal sodium conductances is low (*Figure 4*)

under biologically plausible conditions, or even zero in other synaptic distribution regimes

that heavily favor the apical tree (*Supplementary Figure 5*). The contradiction is resolved,

however, by taking into account the subsequent finding that orientation tuning requires

apical sodium channel conductances, but does not necessarily require basal sodium channel

conductances (*Figure 5*). An explanation for this observation might be that basal dendritic

segments feature morphological and electrophysiological properties that endow them with

higher sodium spiking thresholds, which renders basal sodium channels much less effective

at eliciting somatic responses compared to those located on the apical tree (*Figure 6C*). On

the other hand, AMPA and NMDA conductances are necessary in both dendritic trees for

orientation tuning. This indicates that depolarizations from the basal tree play a key role in

shaping neuronal output, even when basal sodium spikes do not. As such, neuronal output

must be a synergistic phenomenon mostly between apically-based sodium spikes and basally-

based depolarizations or sodium spikes. Our results thus indicate that a form of *intra-tree*

dendritic cooperativity - a synergistic effect between apical and basal dendritic segments that

exhibit activity in relative synchrony - might be present in these neurons. This is evidenced by

the existence of extinction-prone "cooperative" somatic spikes that are lost when either of

their dendritic components (apical or basal, be that depolarization or spiking) is lost (*Figure*

4). Additional evidence is provided by the previously mentioned experiments in which

removing basal inputs altogether silences the neuron, despite plentiful apical tree activity, as

well as through experiments in which removal or suppression (weight reduction) of basal

stimulus-driven synapses (i.e. only background-driven synapses are fully active) significantly

decreases OSI values in the model neuron (*Figures 5A, 5B*).

Finally, driven by our observations we hypothesize that somatic output is determined through

Bilateral Input Coincidence, a subcategory of coincidence detection (*Figure 6D*). The basal tree

receives visual feedforward input, representing the information therein as a series of

hyperpolarizations, depolarizations, and occasional dendritic spikes. Thus, visual input creates

a basal "backdrop" of depolarizations that represents visual information. At the same time,

predictive and attentional signals from higher-order cortical areas reach the apical tree of the

neuron, causing the generation of dendritic spikes that propagate to the soma and are

temporally summed with any concurrent basal depolarization. In the event that visual input

is non-existent, or of a non-preferred orientation, the depolarization is minimal to none. Thus,

the apical dendritic spike will not be significantly augmented through summation and will fail

to produce a somatic response in the majority of cases. If there is visual input, however,

especially of an orientation matching the preferred orientation of the basal tree, the

aforementioned "backdrop" will include multiple subthreshold depolarizations, perhaps even

dendritic spikes. The apical dendritic spikes are thus likely to be temporally summed with

these depolarizations and generate a somatic spike. As such, even though most somatic spikes

will be generated through an apical tree dendritic spike, the cases in which this is possible in

the first place will be dictated by the backdrop of depolarizations (and/or spikes) provided by

the basal tree. On the face of it, this finding apparently contradicts the findings of Park et al.

(2019), where *in vivo* micro-ablation via laser of the apical tree in L2/3 mouse V1 pyramidal

neurons did not abolish orientation selectivity, which remained essentially unchanged

following a recovery period of ~ 1 day. This apparent disparity can be potentially explained by

postulating the operation of cellular homeostatic mechanisms that might alter the

dependence of somatic spiking on basal inputs.

Our hypothesis also suggests that L2/3 V1 pyramidal cells may perform salient feature

extraction from visual input compatible with predictive coding (Rao & Ballard, 1999). Our

hypothesis suggests that when top-down signals reflecting attentional or predictive processes

reach the apical tree, they "highlight" appropriate parts of the signal "landscape" generated

by the visual input reaching the basal trees. This could result in a neuron that is activated only

when specific features are present in its receptive field, facilitating the processing of salient

stimuli. Such phenomena have been observed experimentally as a result of attention (Simons

& Chabris, 1999).

However, multiple questions still remain unanswered. First of all, the exact nature of the

attentional and predictive signals received by the apical tree remains to be deciphered.

Secondly, the formation of the visual "backdrop" as a result of basal tree depolarization merits

further study. Encoding of visual information using mostly subthreshold depolarizations in the

presence of noise is an interesting problem, possibly resolved through the effects of intra-

tree dendritic cooperativity, where apical spikes sharpen responses to stimulus-driven input

rather than noise. In addition, a number of differences emerge when comparing the visual

system across different model species. In rodents, orientation-selective cells akin to V1 simple

cells have been discovered in the thalamus itself (Scholl et al., 2013), and direct

thalamocortical projections to L1 of V1 have also been observed (Roth et al., 2016). In

addition, rodent V1 organization follows a dispersed salt-and-pepper motif, unlike the highly

structured organization typical of higher mammals like primates and cats (Ohki & Reid, 2007).

These differences are important to take into account when trying to infer general rules of

function from information derived from rules of information processing in different animal

models. Finally, perhaps the most interesting unanswered question is whether these

computations take place elsewhere in the cortex in addition to the visual system. Predictive

coding as a means of stimulus compression has already been proposed as a way to simplify

visual perception (Rao & Ballard, 1999), and this might also be the case for other sensory or

cognitive tasks. Regardless, further investigation is required in order to unravel the Gordian

knot that is visual perception.

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7. Supplementary Information

Compartment Type	Passive/Active Mechanisms	Synaptic Mechanisms
Soma	Hodgkin/Huxley voltage-gated Na$^+$ channels	GABA$_A$ (background-driven)
Basal Dendrites	Hodgkin/Huxley voltage-gated K$^+$ channels Muscarinic voltage-gated K$^+$ channels A-Type voltage-gated K$^+$ channels	AMPA (background-driven) NMDA (background-driven) GABA$_A$ (background-driven)
Apical Dendrites	T-Type Ca^{++} channels High voltage activated (HVA) Ca^{++} channels Calcium-dependent K$^+$ channels Active ATP Ca^{++} pumps	AMPA (stimulus-driven) NMDA (stimulus-driven) GABA$_A$ (stimulus-driven)

Supplementary Table 1. Outline of passive, active and synaptic mechanisms present in the model neuron.

Conductance (mS/cm^2)	Soma	Apical	Basal
g$_{Na}$	0.505	0.303	0.303
g$_{Kdr}$	0.05	1.5*10^{-3}	1.5*10^{-3}
g$_{Km}$	2.8*10^{-3}	1.27*10^{-3}	1.27*10^{-3}
g$_A$	5.4	Diameter ≤ 0.8 μm: 108 Diameter > 0.8 μm: 10.8	Diameter ≤ 0.8 μm: 108 Diameter > 0.8 μm: 10.8
g$_T$	0.03	x ≤ 260 μm: 0.029*sin(0.009*x+0.88) x > 260μm: 0.012	0.03+6*10^{-5}*x
g$_{HVA}$	0.05*10^{-3}	x ≤ 260 μm: 0.049*10^{-3}*sin(0.009*x+0.88) x > 260μm: 0.02*10^{-3}	0.05*10^{-3}+10^{-7}*x
g$_{KCa}$	2.1*10^{-3}	2.1*10^{-3}	2.1*10^{-3}

Supplementary Table 2. Outline of membrane mechanism conductances (not synaptic). Reproduced from (Park et al., 2019).

	Model	Cho et al., 2010
RMP, mV	-79	-78.56 ± 1.34
IR, MΩ	123.6	125.2 ± 8.2
τ, ms	17.3	16 ± 0.7
AP amplitude, mV	66.1	67.8 ± 1.8
AP threshold, mV	-41.8	-37.7 ± 1.3
AHP, mV	17.9	13.3 ± 0.5
P-T time, ms	38.6	55.3 ± 2.7
AP adaptation	1.16	1.18 ± 0.02

Supplementary Table 3. Outline of model electrophysiological properties. RMP: resting membrane potential, IR: Input Resistance measured at hyperpolarizing current (-0.04 nA), AP: action potential, AHP: after

	Conductance (nS)	τ_1, ms	τ_2, ms
NMDA	1.15	2	30
AMPA	0.84	0.1	2.5
GABA$_A$	1.25	0.2	1.4

998 ***Supplementary Table 4.*** *Outline of synaptic mechanism conductances and time constants. Reproduced from*
999 (Park et al., 2019).

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Supplementary Figure 1. *Validation of dendritic non-linear properties. A: Expected vs Actual plot for all basal*
dendrites undergoing the I3P protocol for 1-200 synapses. Blue line represents the dendrite in Figure 1B (basal
dendrite #0). B: Same as A, but for all apical dendrites. Blue line represents dendrite in Figure 1C (apical dendrite
#0). C: Example traces from the I3P protocol for NMDA spike generation from basal dendrite #0. Traces are 10
synapses apart from each other. D: Same as C, but for apical dendrite #42 (oblique branch). Dendrites in C and D
chosen to show examples of NMDA spiking.

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Supplementary Figure 2. Properties of the GcaMP6s kernel. A: Illustration of the output of the kernel function
for a single event (AP) occurring at $t_0 = 500$ ms. (marked as a red line). B: Response intensity as a function of the
number of events (APs) within a 250ms window. Compare with (Chen et al., 2013).

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1014   ***Supplementary Figure 3.*** *Waveforms of detected calcium fluorescence events. Somatic waveforms in blue, basal*
1015   *dendritic waveforms in red. Spike onset detected at t = 5 dt. A: Waveforms from all detected events from both*
1016   *mice. B: Waveforms from all detected events in mouse 1. C: Waveforms from all detected events from mouse 2.*
1017   *Note: 1 dt is equal to approximately 0.138 s, and exactly equal to the microscope frame time.*

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Supplementary Figure 4. *A: Event-triggered averages for somatic (blue) and basal dendrite (red) events in mouse*
1. B: Event-triggered averages for somatic (blue) and basal dendrite (red) events in mouse 2. Note the difference
in fluorescence intensity compared to mouse 1. Error bars: Standard error of the mean. C: Comparison of the true
(two-photon) data distribution (blue) with the null distribution (light red). Distributions are different (K.S. two-
sample test; test value ~0.105, p << 0.05). Note that the middle bar for the true data has been truncated (actual
value 0.12267).

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Supplementary Figure 5. *Results of ionic intervention experiments for all synaptic distribution profiles (from [10%*
basal : 90% apical] to [90% basal : 10% apical]). A: Somatic spike dependence on dendritic sodium conductance
when background input distribution follows the same distribution as the stim-driven inputs and the stimulus is
set to 0° (preferred orientation). B: Same as A, but background synapses are evenly distributed on both dendritic
trees rather than following the same distribution as the stim-driven inputs. C: Same as A, but the stimulus being
presented is set to 90° (orthogonal/non-preferred orientation). D: Same as C, but background synapses are evenly
distributed on both dendritic trees rather than following the same distribution as the stim-driven inputs. Note
how somatic spike dependence on apical sodium conductance decreases when biasing synapse distribution
towards the basal tree without ever reaching zero (A, left to right, apical). However, somatic spike dependence
on basal sodium conductance decreases and reaches zero when biasing synapse distribution towards the apical
tree (A, right to left, basal). A difference is also evident in sodium conductance dependence between stimulus
orientations (A vs C), revealing that stimulus-driven spikes do not feature the same dependencies as background-
driven spikes. Finally, the role of apical tree sodium conductance is amplified when background-driven synapses
are evenly distributed (A vs B and C vs D).

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A

B

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Supplementary Figure 6. *Aggregated results from Figure 5 and other experiments. A: Orientation tuning curves*
for the control condition and all ionic and synaptic mechanism manipulations. Note that the apical sodium
intervention (brown) and apical sodium blockage (orange), as well as the basal 50% (synaptic) intervention
(black) curves do not represent adequate orientation tuning. B: OSI values for the control condition and all ionic
and synaptic mechanism manipulations. Again, the apical sodium intervention (brown), apical sodium blockage
(orange) and basal synaptic intervention (black) curves did not feature adequate tuning, and thus have no OSI
values. Stars denote statistical significance at a significance level (α) of 0.05 (*), 0.01 (**) or 0.001 (***), whereas
"n.s." stands for "no statistically significant difference", i.e. p-value ≥ α.

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