# A biophysical minimal model to investigate age-related changes in CA1 pyramidal cell electrical activity

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## Abstract

Aging is a physiological process that is still poorly understood, especially with respect to effects on the brain. There are open questions about aging that are difficult to answer with an experimental approach. Underlying challenges include the difficulty of recording *in vivo* single cell and network activity simultaneously with submillisecond resolution, and brain compensatory mechanisms triggered by genetic, pharmacologic, or behavioral manipulations. Mathematical modeling can help address some of these questions by allowing us to fix parameters that cannot be controlled experimentally and investigate neural activity under different conditions. We present a biophysical minimal model of CA1 pyramidal cells (PCs) based on general expressions for transmembrane ion transport derived from basic thermodynamical principles. The model allows directly varying the contribution of ion channels by changing their number. By analyzing the dynamics of the model, we find parameter ranges that reproduce the variability in electrical activity seen in PCs. In addition, increasing the L-type $Ca^{2+}$ channel expression in the model reproduces age-related changes in excitability that are qualitatively and quantitatively similar to those observed in PCs from aged animals.

# 1 Introduction

As we age, our brains undergo many changes [1, 2], but we understand relatively little about these and their effects on neural function. What does normal neurophysiological aging look like and what are the various stages? How does the electrical activity of neurons change and what are the biophysics underlying those changes? How do aging neurons respond to input from other cells? Answering these questions is not just fundamental to understanding aging as a neurophysiological process, but also to understanding how this process may be altered in age-related disorders of clinical importance such as Alzheimer's [3] and Parkinson's [4] disease.

Many aging studies have focused on the hippocampus, an area of the brain involved in learning, memory formation, and spatial processing [1, 2]. Aged rats [5–7] and humans [8] show impaired learning of hippocampal-dependent spatial tasks. Long-term potentiation (LTP), a proposed physiological substrate of memory formation, has been investigated in the hippocampus and its induction and maintenance shown to be impaired in aged rats [9, 10]. A short-term form of plasticity, frequency potentiation/facilitation (FP/FF), is also impaired in hippocampal pyramidal cells (PCs) from aged rats and correlates with learning deficits [11].

Plasticity changes and behavioral impairments may result in part from altered $Ca^{2+}$ signaling in aged neurons [1, 2]. Compared to CA1 PCs from young animals, PCs from aged animals show larger and longer post-burst afterhyperpolarizations (AHPs) [12–14]. AHPs are mediated by $Ca^{2+}$-dependent $K^+$ currents, which can act like brakes on the electrical activity of CA1 PCs [15, 16]. As a result, PCs show increased spike frequency adaptation and fire fewer action potentials (APs) in response to acute stimuli or during bursting activity [17–19]. Larger AHPs are associated with increased intracellular $Ca^{2+}$, mediated in part by $Ca^{2+}$ entry via L-type channels [14, 18, 20, 21]. Aged animals show increases in L-type channel expression and/or channel density at the plasma membrane [22–25]. Animals with higher $Ca^{2+}$ channel density perform poorly in spatial tasks [23], while blockers of L-type channels can restore learning and plasticity in older animals [26, 27].

It is not well understood how changes in ion channel gene expression and hippocampal PC excitability may affect neuron responsiveness and microcircuit output. In part, this is due to challenges inherent in performing the needed experiments. Single PCs are difficult to access in intact animals where hippocampal microcircuit function is preserved. It is also difficult to tease apart the influence of the many different neurophysiological factors that change during aging. Mathematical modeling provides means to understand more about the effects of aging on hippocampal cellular excitability by controlling factors we cannot control experimentally.

Our previous work shows that mathematical expressions for different passive and active ion transport mechanisms can be derived from first principles of thermodynamics [28, 29] using a common functional form [30]. This results in a realistic representation of ionic flow across the membrane, and allows the model to reproduce phenomena such as rectification of ion currents seen in recordings. We present a model that reproduces the diversity of firing patterns in CA1 PC recordings, including adaptive firing, stimulus-induced bursting, and spontaneous bursting [31]. In addition, we reproduce several electrophysiological characteristics of aging by varying the expression of $Ca^{2+}$ channels in the model, and make predictions about bursting activity in aged CA1 PCs, which to our knowledge has not been reported. We believe this model is ideal to further study the effects of various biophysical changes in CA1 PCs during aging, as well as potentially forming the basis for biophysical, yet computationally inexpensive, network models.

## 2 Methods

### 2.1 Model

To simulate the electrical activity of CA1 PCs, we used an extended version of a two-dimensional, biophysical model previously developed and characterized by two of the present authors [28–30]. The equations for the ionic currents are derived from first principles of thermodynamics. Previous modeling studies have shown that to reproduce firing behaviors such as spike frequency adaptation and bursting, the minimum number of variables is three [32, 33]. In particular, $Ca^{2+}$ dynamics are important for producing adaptation and burst firing in CA1 PCs (for review see [31]). The model dynamics are therefore described by three ordinary differential equations for the time-dependent changes in the transmembrane potential ($v$, in mV), the proportion of open $K^+$ channels ($w$ in [0,1]), and the intracellular $Ca^{2+}$ concentration ($c$, in $\mu$M), respectively [30]. Based on a well-known relationship between voltage-dependent activation of delayed rectifier $K^+$ channels and inactivation of $Na^+$ channels, $w$ also represents the proportion of inactivated $Na^+$ channels [34, 35].

It is assumed that the membrane potential changes due to currents produced by ions transported across the membrane. We take into account currents mediated by a voltage-gated, inactivating $Na^+$ channel ($I_{NaT}$), voltage- and $Ca^{2+}$-gated $K^+$channels ($I_{DK}$ and $I_{SK}$, respectively), and a $Na^+/K^+$-ATPase ($I_{NaK}$). The time-dependent change in membrane potential can be written as

$$C_m\partial_t v = I_F - I_{NaT}(v, w) - I_{CaL}(v, c) - I_{DK}(v, w) - I_{SK}(v, c) - I_{NaK}(v). \tag{1}$$

with $\partial_t$ representing the instantaneous change with respect to time. $C_m$ (pF) is a constant representing the change in the density of charge around the membrane with respect to voltage, typically referred to as membrane capacitance in models based on electrical circuits [36]. Based on recordings from rat CA1 PCs, it is assumed that $C_m = 25$ pF [37].

All the currents in Eq. (1) are modeled using the same generic functional form, a product

$$I_x = a_x G_x \varphi_x,$$

$x \in \{NaT, CaL, DK, SK, NaK\}$, where $a_x$ is a whole-cell current amplitude (pA), $G_x$ is a gating term (between 0 and 1), and $\varphi_x$ is an adimensional term describing the driving force for the transmembrane flux (Table 1). The terms $a_x = s_x N_x$, $x \in \{NaT, CaL, DK, SK, NaK\}$ are whole-cell current amplitudes with $s_x$ (pA) representing the current flowing through a single channel (or pump), and $N_x$ representing the number of membrane proteins mediating the current (e.g. number of $K^+$ channels). Of interest, $s_x$ is ~1 pA for most voltage-gated channels [38], and is ~5-10 pA for SK channels [39].

Assuming that none of the currents in the model exhibit rectification [30], which agrees with recordings of the included currents, the adimensional component of the transmembrane flux can be simplified and written as

$$\varphi_x(v) = 2\eta_x \sinh\left(\eta_x \frac{v - v_x}{2v_T}\right), \tag{2}$$

where $\eta_x$ represents the number of charges transported across the membrane in a single transport event, and $v_x$ is the reversal potential for the current, $x \in \{NaT, CaL, DK, SK, NaK\}$. For channels, $v_x$ is the Nernst potential for the ion [30]. Of note, $\eta_x = 1$ for $x \in \{DK, SK, NaK\}$ and

80 $\eta_{NaT} = -1$ [30], which gives

$$\varphi_x(v) = 2 \sinh \left( \frac{v - v_x}{2 v_T} \right),$$

81 for $x \in \{NaT, DK, SK, NaK\}$. For $Ca^{2+}$ channels, the total charge transported by one ion
82 crossing the membrane from the extracelular space is $\eta_{CaL} = -2$, so

$$\varphi_{Ca}(v) = 4 \sinh \left( \frac{v - v_{Ca}}{v_T} \right).$$

The driving force for flux is assumed to be the same for DK and SK channels. Therefore, the
label $K$ is used for both fluxes from here on. The thermal potential $v_T = kT/q$ (mV), where
$k$ is Boltzmann's constant (mJ/$^o$K), $T$ is the absolute temperature ($^o$K), and $q$ is the elementary
charge (Coulombs). The Boltzmann constant can be thought of as a scaling factor between
macroscopic (thermodynamic temperature) and microscopic (thermal energy) physics [40]. The
reversal potentials for the different currents depend on the Nernst potentials for each ion, as given
by

$$v_x = \frac{v_T}{z_x} \ln \left( \frac{[x]_o}{[x]_i} \right), \quad x \in \{Na, Ca, K\} \tag{3}$$

90 where $z_x$ is the ion valence and $[x]_o$ and $[x]_i$ are the ion concentrations outside and inside the cell,
91 respectively. The reversal potential for the $Na^+$/$K^+$-ATPase is given by $v_{NaK} = v_{ATP} + 3 v_{Na} - 2 v_K$
92 [30]. The Nernst potentials for $Na^+$ and $K^+$ are assumed to be constant, but $v_{Ca}$ varies because
93 the intracellular $Ca^{2+}$ concentration is a state variable in the model.

**Table 1:** Transport mechanisms included in the model

| Transport mechanism | Current | Amplitude ($a$) | Gating ($G$) | Flux $\varphi$ |
|---|---|---|---|---|
| Transient $Na^+$ channels | $I_{NaT}(v, w)$ | $a_{Na}$ | $S_m(v)(1-w)$ | $\varphi_{Na}(v)$ |
| L-type $Ca^{2+}$ channels | $I_{CaL}(v, c)$ | $a_{Ca}$ | $S_n(v)$ | $\varphi_{Ca}(v)$ |
| Delayed rectifier $K^+$ channels | $I_{DK}(v, w)$ | $a_{DK}$ | $w$ | $\varphi_K(v)$ |
| SK $Ca^{2+}$-dependent $K^+$ channels | $I_{SK}(v, c)$ | $a_{SK}$ | $H_{SK}(c)$ | $\varphi_K(v)$ |
| $Na^+$/$K^+$ pumps | $I_{NaK}(v)$ | $a_{NaK}$ | $1$ | $\varphi_{NaK}(v)$ |

All ion fluxes are given by a product of the form $I_x = a_x G_x \varphi_x$, where $a_x$, $G_x$, and $\varphi_x$ represent, respectively, the amplitude (normalized by membrane capacitance), gating, and driving force terms for the flux. $G_{NaK} = 1$ represents saturation of the $Na^+$/$K^+$ pumps. Note that inactivation of $Na^+$ channels is also represented by $w$ [34, 35], so the proportion of non-inactivated $Na^+$ channels is $1 - w$.

94 **Gating.** The auxiliary functions describing voltage-dependent activation are given by

$$S_j(v) = \frac{\exp \left( g_j \frac{v - v_j}{v_T} \right)}{1 + \exp \left( g_j \frac{v - v_j}{v_T} \right)}, \quad j \in \{m, n, w\}, \tag{4}$$

where $g_j$ controls the steepness of the activation curve for $Na^+$ ($m$), $Ca^{2+}$ ($n$), or DK ($w$) channels, and $v_j$ represents the half-activation voltage for those channels. The function

$$R_w(v) = r_w \left[ \exp \left( b_w g_w \frac{v - v_w}{v_T} \right) + \exp \left( (b_w - 1) g_w \frac{v - v_w}{v_T} \right) \right], \tag{5}$$

describes the voltage-dependence of the rate of activation of the DK channels. The parameters $r_w$ and $b_w$ represent the recovery rate and the asymmetry in the gating relative to voltage that biases the time constant for the gating process, respectively.

The dynamics for the proportion of activated DK channels, $w$, are assumed to be logistic,

$$\partial_t w = w(S_w(v) - w)R_w(v), \tag{6}$$

which yields better fits and is more consistent with the dynamics of activation in channel populations recorded in voltage-clamp experiments (e.g., see the activation curves in [41–43]).

The gating of the SK channel is not voltage-dependent, but instead depends on intracellular $Ca^{2+}$ binding. Its activation is modeled using a Hill equation that depends on the intracellular concentration of $Ca^{2+}$, as used to fit data from channel recordings [44]:

$$H_{SK}(c) = \frac{c^2}{c^2 + c_{SK}^2}, \tag{7}$$

where $c_{SK}$ represents the half-activation $Ca^{2+}$ concentration for the SK channels, with a reported value of 0.74 $\mu$M (740 nM) [39, 44].

For the dynamics of intracellular $Ca^{2+}$, we assume recovery toward a steady state $c_\infty$ at a rate $r_c$, with increments caused by the $Ca^{2+}$ current $I_{Ca}$ [35],

$$\partial_t c = r_c(c_\infty - c) - \tilde{k}_c I_{CaL}(v, c). \tag{8}$$

The term $\tilde{k}_c$ in equation (8) is a conversion factor ($\mu$M/pCoul) that accounts for the effect of $Ca^{2+}$ flux across the membrane on the intracellular $Ca^{2+}$ concentration.

The term $I_F$ represents a stimulus *forcing* the membrane; that is, current from an electrode, or time-dependent fluctuations from the local field potential. LFP activity is simulated by replacing the term $I_F$ with a time-dependent, Ornstein-Uhlenbeck (OU) process with amplitude $a_F(t)$ (pA). The mean is represented by $\mu_F$ (pA) (drift term) [45] given by [46]

$$a_F(t + \delta) = a_F(t) \left( 1 - \frac{\delta}{\tau_F} \right) + \left[ \mu_F \delta + \eta(t) \sqrt{d_{Stim}\delta} \right], \tag{9}$$

where $\delta$ is a small time step, $\tau_F$ is a relaxation time, and $\eta(t)$ is an independent white noise process with zero-mean and unit standard deviation. The mean is set close to the rheobase for the model PCs ($\sim$50 pA). The process has a variance $\sigma_F^2 = d_F\delta/2$ (pA), which means $d_F$ can be approximated if an estimation of the variance of the current $a_F$ is available [47, 48].

**Change of variables to obtain numerical solutions.** To simplify the numerics, we change variables

$$u = v/v_T,$$

and adjust all voltages accordingly as

$$u_l = v_l/v_T, \quad l \in \{NaT, CaL, K, NaK, m, n, w\}.$$

The new equation for the normalized voltage is

$$\partial_t u = \frac{\partial_t v}{v_T}.$$

To simplify the notation and reduce the number of operations during the numerical integration, we also reparametrize the amplitudes as

$$A_l = \frac{2a_l|\eta_l|}{v_T \, C_m}, \tag{10}$$

for $l \in \{NaT, CaL, K, NaK\}$, in units of 1/ms. Similarly, the activation functions for the Na$^+$, Ca$^{2+}$, and DK currents can also be rewritten respectively as

$$f_\infty(u) = \{1 + \exp[g_f(u_f - u)]\}^{-1}, \quad f \in \{m, n, w\}. \tag{11}$$

The result is a new equation of the form

$$\partial_t u = J_F - A_{NaT}(1 - w)m_\infty(u)\varphi_{Na}(u) - A_{CaL}n_\infty(u)\varphi_{Ca}(u, c) -$$

$$(A_{DK}w + A_{SK}H_{SK}(c))\varphi_K(u) - A_{NaK}\varphi_{NaK}(u). \tag{12}$$

The term $J_F$ (1/ms) is the input current $I_F$ (pA) divided by $v_T C_m$. After the change in variables and the normalization of the current amplitudes, equation (8) changes to

$$\partial_t c = r_c(c_\infty - c) - k_c A_{CaL} n_\infty(u)\varphi_{Ca}(u, c), \tag{13}$$

where $k_c = \tilde{k}_c v_T C_m$.

## 2.2 Parameters

The currents were modeled to fit as closely as possible the biophysical properties of those carried by channel variants expressed in mammalian neurons, and specifically CA1 PCs, where data are available. The DK current is based on that mediated by K$_v$2.1 channels, the predominant channel underlying the delayed rectifier current in rat hippocampal neurons [49]. The L-type Ca$^{2+}$ current is based on that carried by Ca$_v$1.2 (class C) channels, the predominant L-type channel isoform expressed in rat brain [50]. Additional details about the parameters can be found in Table 2.

Wherever possible, model parameters were taken from studies in rodent (mice and rat) hippocampal CA1 PCs. If data were not available, we obtained parameters from other types of mammalian cell, or from studies of mammalian ion channels in expression systems like *Xenopus* oocyte. Physical constants and other parameters we would not expect to vary, such as the intra- and extracellular concentrations of ions or the cellular capacitance, were fixed. Biophysical properties of the ion channels, such as their half-activation voltages, were also fixed. The parameters we varied were primarily those corresponding to maximum current amplitudes, which can change acutely due to modulation or channel phosphorylation [51, 52], or chronically due to changes in ion channel expression that occur with age [22, 24].

**Table 2:** Constants and parameters

| parameter | description | value | units | reference |
|---|---|---|---|---|
| $k$ | Boltzmann's constant | $1.381\mathrm{e}^{-20}$ | mJ/K | physical constant [38] |
| $q$ | elementary charge | $1.602\mathrm{e}^{-19}$ | C | physical constant [38] |
| $T$ | absolute temperature | 273.15 + 37 | K | adjusted to mammalian body temperature of 37 °C [38] |
| $a_{NaT}$ | amplitude of transient $Na^+$ current | 1000-2300 | pA | set to produce currents of ~2-7 nA, in range recorded in CA1 PCs from rats [53] and guinea pigs [54] |
| $a_{CaL}$ | amplitude of L-type $Ca^{2+}$ current | 25 or 50 | pA | set to produce currents of ~2-3 nA or ~5-6 nA as recorded in young and aged CA1 PCs, respectively [20] |
| $a_{DK}$ | amplitude of delayed rectifier $K^+$ current | 6000-8000 | pA | set to produce currents of ~6-9 nA, in range recorded from HEK cells expressing rat Kv2.1 and $I_K$ in hippocampal neurons [55] |
| $a_{SK}$ | amplitude of $Ca^{2+}$-dependent $K^+$ current | 300-1600 | pA | set to produce currents of ~100-800 pA, depending on $Ca^{2+}$ concentration, as recorded in SK-transfected cells [56] |
| $a_{NaK}$ | amplitude of $Na^+$/$K^+$-ATPase current | 10-23 | pA | set to produce currents of ~70-190 pA, similar to but on high end of range recorded in hippocampal PCs [57] |
| $\eta_x$ | charge moved across the membrane in a single transport event $x \in \{NaK, DK, SK\}$ | 1 | – | 1 net positive charge moving outward [30] |
| $\eta_{NaT}$ | charge transported across the membrane by $NaT$ channels | -1 | – | 1 net positive charge moving inward [30] |
| $\eta_{CaL}$ | charge transported across the membrane by Ca channels | -2 | – | 2 net positive charges moving inward [30] |
| $v_{Na}$ | Nernst potential for $Na^+$ | 60 | mV | in range reported for mammalian cells [58] |
| $[Ca]_o$ | extracellular $Ca^{2+}$ concentration (intracellular varies) | 1.5 | mM | in range reported for mammalian neural tissue [59] |
| $v_{Ca}$ | Nernst potential for $Ca^{2+}$ | variable; baseline ~128 | mV | in range reported for mammalian cells [58]; varies since intracellular $Ca^{2+}$ concentration is a model variable |
| $v_K$ | Nernst potential for $K^+$ | -89 | mV | in range reported for mammalian cells [58] |
| $v_{ATP}$ | Nernst potential for ATP | -420 | mV | value used in model of mammalian heart cells and based on fit to data [60] |
| $v_{NaK}$ | Nernst potential for $Na^+$/$K^+$-ATPase | -62 | mV | calculated based on the Nernst potentials for ATP, $Na^+$, and $K^+$, and a 3:2 stoichiometry, respectively [61]; $v_{NaK} = 3v_{Na} - 2v_K - v_{ATP}$ |

| parameter | description | value | units | reference |
|---|---|---|---|---|
| $r_w$ | rate of activation of delayed rectifier K$^+$ current | 1.0-1.8 | ms | fit so that the duration of the action potential is approximately 2 ms [62] |
| $s_w$ | asymmetry of time constant of delayed rectifier K$^+$ current | 0.3 | - | based on fit; if higher (0.5-0.7) APs are the wrong shape and do not ride on sufficient plateau potential compared to recordings |
| $v_m$ | half-activation potential of Na$^+$ current | -19 | mV | in range reported for transient Na$^+$ channels in CA1 PCs [63, 64] |
| $v_n$ | half-activation potential of Ca$^{2+}$ current | 3 | mV | in range recorded for high-voltage activated Ca$^{2+}$ currents in rat CA1 PCs [65]; see also recordings from oocytes [66] or HEK cells [67] expressing Ca$_v$1.2 channels |
| $v_w$ | half-activation potential of delayed rectifier K$^+$ current | -1 | mV | in range reported for rat Kv2.1 channels expressed in COS-1 cells [49] |
| $c_{SK}$ | half-activation Ca$^{2+}$ concentration for SK current | 7.4e$^{-4}$ | mM | based on recordings from oocytes expressing rat SK channel variant [44] |
| $g_m$ | activation slope of Na$^+$ current | 5.0 | - | in range reported for Na$^+$ current in mouse [68] and rat [69] CA1 neurons |
| $g_n$ | activation slope of Ca$^{2+}$ current | 5.0 | - | in range reported for Ca$_v$1.2 expressed in HEK cells [70] |
| $g_w$ | activation slope of delayed rectifier K$^+$ current | 3.8 | - | fit to data from rat brain delayed rectifier channels [71] |
| $c_\infty$ | minimum intracellular Ca$^{2+}$ concentration | 1e$^{-4}$ | mM | equivalent to 100 nM, approximate resting intracellular Ca$^{2+}$ concentration in rat CA1 PCs [17, 72, 73] |
| $r_c$ | intracellular Ca$^{2+}$ removal rate constant | 1e$^{-3}$ to 5e$^{-3}$ | ms$^{-1}$ | adjusted to produce Ca$^{2+}$ dynamics as recorded in rat CA1 PCs [72] |
| $k_c$ | conversion factor to calculate effect of Ca$^{2+}$ current on intracellular Ca$^{2+}$ concentration | 3e$^{-6}$ to 6e$^{-6}$ | mM | adjusted to produce Ca$^{2+}$ dynamics as recorded in rat CA1 PCs [72] |

145 By exploring the model through parameter variations, we were able to find parameter sets that
146 produced different firing patterns, such as adaptive firing, conditional bursting, and spontaneous
147 bursting. These firing patterns are described in more detail in the Results section, but the respective
148 parameter sets are included in Table 3 for ease of comparison.

**Table 3:** Parameters used to produce different firing patterns in the young PC model. See equation (10).

| parameter | adaptive firing | conditional bursting | spontaneous bursting |
|---|---|---|---|
| $a_{NaT}$ ($A_{NaT}$) | 1000 (2.99) | 1300 (3.89) | 2300 (6.88) |
| $a_{CaL}$ ($A_{CaL}$) | 25 (0.15) | 25 (0.15) | 25 (0.15) |
| $a_{DK}$ ($A_{KD}$) | 8000 (23.95) | 6000 (17.96) | 7000 (20.95) |
| $a_{SK}$ ($A_{SK}$) | 1400 (4.19) | 1600 (4.79) | 300 (0.90) |
| $a_{NaK}$ ($A_{NaK}$) | 10.0 (0.03) | 13 (0.04) | 23 (0.07) |
| $r_w$ | 1.0 | 1.8 | 1.1 |
| $r_c$ | 1e$^{-3}$ | 5e$^{-3}$ | 5e$^{-3}$ |
| $k_c$ | 3e$^{-6}$ | 6e$^{-6}$ | 6e$^{-6}$ |

Original amplitudes ($a_x$) are in pA. Reparametrized amplitudes ($A_x$) in parentheses are calculated by $2 * \frac{a_x}{v_T C_m}$ for $x \in \{NaT, DK, SK, NaK\}$ and $4 * \frac{a_{CaL}}{v_T C_m}$ for Ca$^{2+}$ channels. For all calculations, $v_T C_m$ = 668.171 mV pF.

## 2.3 Simulations

All code was written in Python 3.7.4 and run on MacBook Pro laptops with 2.9 GHz Intel Core i5 processors. Simulations were performed using functions from the Python library NumPy [74]. Figures were produced with the Python library Matplotlib [75]. OU processes were simulated using the pyprocess (pypi.org/project/pyprocess) module [76].

## 2.4 Resource availability

Resources generated by this study (code, figures, and manuscript files) are available via GitHub (github.com/emckiernan/agingCA1) and archived via Zenodo (doi.org/10.5281/zenodo.6788229). To facilitate reuse, resources are shared under open licenses (see the license file in our GitHub repository). To promote reproducibility, Python code is embedded in a Jupyter notebook [77] that explains the code, how to use it, and how to generate the figures herein, plus additional ones.

## 2.5 Study design

While aged cells display a number of biophysical changes, we focused on their Ca$^{2+}$ channel expression. Aged CA1 PCs show an increase in the number of functional transmembrane L-type Ca$^{2+}$ channels [22–24]. In particular, CA1 PCs from aged rats have increased expression of Ca$_v$1.2 at the plasma membrane [25]. With these results in mind, we decided to simulate one aspect of aging by changing the number of Ca$_v$1.2 channels in our model membrane. We asked the question, is a change in Ca$_v$1.2 expression sufficient to reproduce the various changes in excitability, such as increased spike frequency adaptation, observed experimentally in aged CA1 PCs? In addition, CA1 PCs are known to burst [31], but we are not aware of studies comparing the bursting patterns in young versus aged animals. Therefore, we used our model to also investigate the effects of altered Ca$_v$1.2 channel expression on bursting activity. In all the following simulations, once the models have been tuned to a specific firing pattern, young and aged model PCs (hereafter referred to as yPC and aPC, respectively) are identical with respect to every parameter except the maximum amplitude of their L-type Ca$^{2+}$ current, which is set to produce currents of ∼2-3 nA or

~5-6 nA to match the magnitude of currents seen in recordings of CA1 PCs from young and aged animals, respectively [20].

# 3   Results

CA1 PCs display diverse firing patterns, ranging from repetitive adaptive spiking to stimulus-induced or spontaneous bursting (for review see [31]). Thus, to represent these cells, our model must reproduce this diversity, as well as age-related effects on firing reported in the literature.

## 3.1   Modeling age-related changes in spike frequency adaptation

Many CA1 PCs respond to square-pulse current injection by firing several early spikes followed by adaptation which slows the frequency of firing [17, 18, 78]. To generate this firing pattern, we set the ionic currents to be the same amplitude range as observed in recordings of young adult CA1 PCs, with amplitudes for the $Na^+$ and $Ca^{2+}$ currents at ~2-3 nA, the DK current approximately double the inward cationic currents, and the SK current at ~400-700 pA, depending on the intracellular $Ca^{2+}$ concentration (see Tables 2 and 3). This balance of ionic currents successfully generates adaptive firing similar to recordings (Fig. 1, top panel; and additional figure in our Jupyter notebook).

**Figure 1:** Adaptive firing in the yPC (solid black traces) versus aPC (dashed red traces) in response to a 800 ms square-pulse stimulation (top panel). Corresponding $Ca^{2+}$ currents, intracellular $Ca^{2+}$ concentration, and SK currents are in the 2nd, 3rd, and 4th panels, respectively.

Studies show that adaptation is more pronounced in aged than young animal cells, leading to a shorter initial period of fast spiking, followed by fewer spikes or complete cessation of firing [17–19, 79]. To compare the young (yPC) and aged (aPC) model cells, all parameters were fixed except for the maximum amplitude of the L-type $Ca^{2+}$ current, which was set to produce currents of

$\sim$2-3 nA (young) or $\sim$5-6 nA (aged), based on recordings [20]. This difference in the Ca$^{2+}$ current causes the rate of firing in the first $\sim$100 ms (early firing) to decrease from 60 Hz in the yPC to 40 Hz in the aPC (Fig. 1). Firing for the remaining stimulation time (late firing) is also affected, decreasing from $\sim$6 Hz to $\sim$3 Hz in the yPC versus aPC, respectively. The effect and frequencies are similar to those seen in recordings of CA1 PCs in young and old rabbits [18].

Examining the Ca$^{2+}$ and SK dynamics during the response reveals the mechanisms underlying the stronger adaptation in the aPC (Fig. 1, 2nd through 4th panels). The first two spikes occur nearly simultaneously in the two model cells. However, the larger increase in intracellular Ca$^{2+}$ in the aPC induces a larger SK current, which in turn slows the cell's firing. The aPC falls behind the yPC by the third spike, and then slows its firing further as the response continues.

## 3.2   Modeling age-related changes in AHPs

AHP generation has been studied in CA1 PCs [80], particularly in the context of aging [17, 18, 79, 81]. To induce AHPs, we kept the same parameters as in the previous simulations. We then stimulated model PCs with a 100 ms square pulse of sufficient amplitude to generate a burst of 4 APs (Fig. 2). The AHPs produced under these conditions in the yPC have a peak amplitude of 3-4 mV, similar to recordings [14, 81]. The aPC required 30 pA more current than the yPC to fire the same number of spikes. However, the aPC fires earlier than the yPC due to its increased Ca$^{2+}$ current (Fig. 2, top panel inset). The aPC generates an AHP 1-2 mV larger than seen in the yPC (Fig. 2, 2nd panel), similar to the difference observed in recordings between young and aged cells [14, 17, 18]. In the model, this larger AHP is due to an increased accumulation of Ca$^{2+}$ in the aPC, which in turn produces a larger SK current (Fig. 2, 3rd and 4th panels).

## 3.3   Modeling age-related changes in burst firing

### 3.3.1   Bursting in response to stimulation

Some CA1 PCs fire bursts instead of trains of spikes [82], especially in certain developmental periods [83]. Burst firing can be generated in the model with several different parameter combinations. For the following simulations, we modified several of the current amplitudes and select kinetics, all within physiological limits (see Table 3). Under this parameter regime, model PCs are silent at rest but burst if stimulated (Fig. 3).

To explore the effects of aging on bursting, we fixed all parameters except for the maximum Ca$^{2+}$ current amplitude, as previously. We then stimulated the two model PCs with a series of square-pulse current injections of increasing amplitudes (50-170 pA) to compare their responses. Again, the larger Ca$^{2+}$ current in the aPC causes it to fire either sooner or nearly simultaneously with the yPC shortly after stimulus onset in all simulations. However, the relative timing of the PCs' firing after the first burst depends on the stimulus amplitude (Fig. 3).

At lower stimulation amplitudes (50 and 80 pA; first two panels), the aPC continues to burst sooner or nearly simultaneously, but always fires fewer spikes per burst than the yPC (2 versus 3, respectively). As the stimulation amplitude increases (110 pA; 3rd panel), the two PCs again fire nearly simultaneously at the onset. However, because the aPC fires fewer spikes per burst, it is able to recover sooner and burst before the yPC for four cycles. It is only towards the end of the stimulus that the larger AHP in the aPC eventually brings it into sync again with the yPC. Finally, at the highest stimulation amplitudes (140 and 170 pA; 4th and 5th panels), both PCs show stronger

**Figure 2:** Responses of the yPC (solid black traces) and aPC (dashed red traces) to 100 ms pulse (top panel). Current amplitude adjusted to produce 4 spikes in each cell (yPC: 112 pA; aPC: 142 pA). Voltage zoom in 2nd panel shows AHPs in detail. Corresponding Ca$^{2+}$ concentrations and SK currents are in the 3rd and 4th panels, respectively.

adaptation, but the yPC maintains something akin to bursting, while the aPC 'devolves' to a pattern more like tonic spiking. Depending on what is considered the relevant electrical event – the single spike or the burst – the aPC shows an increased number of events relative to the yPC (13 spikes versus 8 burst-like events, respectively), but the normal bursting pattern is lost.

### 3.3.2 Spontaneous bursting

A small percentage of CA1 PCs fire bursts in the absence of stimulation [84]. To generate this pattern, we again modifed several of the current amplitudes and select kinetics, all within physiological limits (see Table 3). Under this parameter regime, the yPC fires spontaneous bursts at a frequency of ∼1 Hz with 3 spikes per burst (Fig. 4, 3rd panel), similar to recordings [85].

Increasing the Ca$^{2+}$ current, as previously, changed the spontaneous firing pattern (Fig. 4). Additional interesting effects can be seen if we vary the DK current amplitude within the range previously used for simulations, 6000-8000 pA (6-8 nA). At the highest DK amplitude, the yPC bursts spontaneously but with only 2 spikes per burst (Fig. 4, 1st panel). The aPC, however, does not burst but spikes tonically at a frequency of ∼2 Hz. When the DK amplitude is reduced (7.5 nA), the yPC continues to burst, now with 3 spikes per burst, while the aPC still spikes tonically.

Reducing the DK amplitude further (7 nA or 6.5 nA), causes both PCs to burst spontaneously,

**Figure 3:** Bursting in the yPC (solid black traces) and aPC (dashed red traces) models in response to 800 ms current injections of varying stimulus amplitudes (indicated to the left of each panel). Parameters for the yPC: $a_{NaT}$ = 1300, $a_{CaL}$ = 25, $a_{DK}$ = 6000.0, $a_{SK}$ = 1600, $r_w$ = 1.8, $r_c$ = 5e$^{-3}$, $k_c$ = 6e$^{-6}$. All parameters for the aPC the same except $a_{CaL}$ = 50.

though the aPC always fires fewer spikes per burst. One can use our Jupyter notebook to explore further, and see that while the Ca$^{2+}$ current is larger for the aPC, the maximum accumulation of intracellular Ca$^{2+}$ is either similar or higher in the yPC due to its additional spiking. This causes a similar or larger SK current in the yPC, and results in a slower burst frequency.

Finally, reducing the DK amplitude to the lowest level (6.0 nA) removes more of the 'brake' on the yPC and causes it to spike at high frequency and then quickly block depolarize. The aPC, on the other hand, retains the spontaneous bursting pattern, now with more spikes per burst.

**Figure 4:** Comparison of spontaneous electrical activity in the yPC (solid black traces) and aPC (dashed red traces) for different levels of DK current amplitude. $a_{DK}$ decreasing from 8000 to 6000 pA (8 to 6 nA) in steps of 500 pA, as indicated. Parameters: $a_{NaT}$ = 2300, $a_{CaL}$ = 25, $A_{DK}$ = 7000, $a_{SK}$ = 300, $r_w$ = 1.1, $r_c$ = 5e$^{-2}$, $k_c$ = 6e$^{-6}$, $I_F$ = 0.0. All parameters for aPC the same except $a_{CaL}$ = 50.

## 3.4 Responses to local field potential forcing

Square-pulse stimulation is useful for examining the timing of PC responses, but is not a physio-logically realistic stimulus. Instead, to simulate local field potential (LFP) forcing onto CA1 PCs, we use an Ornstein-Uhlenbeck stochastic process [47, 48], as in Eq. 9. First, we reset the model with the parameters needed to produce adaptive firing, as in Section 3.1. In the yPC, LFP forcing produces repetitive, irregular firing at a frequency of ∼3 Hz on average (Fig. 5, top panel), which is similar to recordings of spontaneous firing in CA1 PCs [86], particularly in response to specific brain rhythms recorded in the surrounding field [87, 88].

In response to the exact same LFP forcing, the aPC with increased Ca$^{2+}$ current shows a similar irregular firing pattern, but slower frequency of ∼2 Hz (Fig. 5, 2nd panel; compare overlap in 3rd

**Figure 5:** Responses of the yPC (solid black traces) and aPC (dashed red traces) to the same LFP forcing while in adaptive firing mode. Top panel shows just the yPC response, the 2nd panel shows the aPC response, and the 3rd panel shows the overlap of the two traces. Parameters for the yPC and aPC are the same as in Fig. 1. LFP parameters: $\mu_F$=50.0 pA, $\sigma_F$=25.0 pA, $\tau_F$=1/2.0 for both models.

panel). The simulation also shows several time points when the two cells fire almost simultaneously, and then the yPC fires again while the aPC does not. This apparent 'spike failure' has been seen in recordings of PCs from aged animals [13].

Next, we set the parameters to produce conditional bursting, as previously. Under these conditions, LFP forcing in the yPC produced irregular burst firing at a frequency of ∼5Hz (i.e. theta frequency) with 2-4 spikes per burst (Fig. 6, top panel). This firing pattern is similar to spontaneous activity recorded in a subset of CA1 PCs known as phasic theta-ON cells, which preferentially burst during theta activity recorded from the surrounding field [87, 89]. Increased $Ca^{2+}$ current in the aPC changes the firing pattern (Fig. 6, 2nd panel). In response to LFP forcing, the aPC still fires irregular bursts, but with fewer spikes per burst (3 max) and a higher occurrence of 2-spike bursts than seen in the yPC. In addition, the aPC fires single APs amidst the bursts, which occurs half as frequently in the yPC under this parameter regime. Also, the timing of the bursts in the aPC is altered relative to the yPC (see 3rd panel overlap).

**Figure 6:** Responses of the conditionally bursting yPC (solid black traces) and aPC (dashed red traces) models to the same LFP forcing. Top panel shows just the yPC response, the 2nd panel shows the aPC response, and the 3rd panel shows the overlap of the two traces. Parameters for the yPC and aPC are the same as in Fig. 3. LFP parameters the same as in Fig. 5 except $\sigma_F$=20.0 pA.

## 4    Discussion

### 4.1    Cellular heterogeneity

Our three-dimensional, single-compartment model derived from first principles of thermodynamics can reproduce the diversity of firing patterns recorded in CA1 PCs. Moving between the different patterns was achieved primarily by changes to the relative expression of ion channels in the model. We did not systematically explore the full parameter space, but future work could include bifurcation analysis to determine boundaries for each firing pattern. The flexibility of the model could be useful for researchers to study the effects of PC heterogeneity on network function. Geiller and colleagues [90] write, "Until very recently, hippocampus models and theories were built on a view of homogenous population of principal cells" (pg. 6). However, there is heterogeneity in CA1 PCs, especially during different development stages (for review, see [31]). Lee and colleagues [91] write, "how the heterogeneous PCs integrate into the CA1 circuit remains unknown" (pg. 1129).

Experimentally, it is difficult to quantify how many PCs in a given network are displaying a specific firing pattern, and even harder, if not impossible, to manipulate these percentages. Furthermore, cells can transition between firing patterns [92], meaning percentages might fluctuate. With our minimal model, however, we could build small networks with different balances of adapting versus

bursting PCs, and explore how this affects network output. We could also model the progression of aging in the network by varying the percentage of PCs which have altered $Ca^{2+}$ channel density, or implement a whole spectrum of channel expression across the simulated network.

## 4.2   Aging and $Ca^{2+}$ channel expression

Our model can also reproduce changes in electrical activity seen in aged CA1 PCs, including larger AHPs [12–14] and increased adaptation [17–19], by increasing the L-type $Ca^{2+}$ current amplitude to a level similar to that found in recordings. The L-type channel was modeled after the $Ca_v1.2$ isoform based on work showing this is the primary contributor in rodent brain, responsible for ~70-80% of the L-type current [50, 93]. mRNA expression of *Cacnac1C* (the gene encoding $Ca_v1.2$) is increased in aged mice and rats [22, 94]. Increases in plasma membrane expression [25] and phosphorylation [95] of $Ca_v1.2$ channels have also been seen in aged rats. In addition, changes in *Cacna1c*/$Ca_v1.2$ expression are correlated with memory impairments [94, 96].

However, CA1 PCs also express the $Ca_v1.3$ isoform [97], which is responsible for ~20% of the total L-type current [50, 93]. Studies have found both increased mRNA [22] and protein [24] expression of $Ca_v1.3$ in aged rats, and this increased expression is correlated with memory impairment [98]. Knockout studies in mice indicate that it is this isoform, and not $Ca_v1.2$, which contributes to slow AHP generation [99], possibly via activation of co-localized SK channels [97]. While experimental studies have been complicated by a lack of pharmacological agents which can isolate currents carried by the different isoforms, it would be relatively simple with our model to study the contributions of these two channels. The primary difference between the two is a shift in the activation curve of the $Ca_v1.3$ channel to more hyperpolarized values, relative to $Ca_v1.2$ [66]. Changing the parameter $v_j$ in Eq. 4 would allow us to represent the different isoforms and explore how changes in the expression of each during aging might affect PC activity.

There are many cellular changes apart from $Ca^{2+}$ channel expression that occur during aging and could contribute to altered activity in PCs. Studies have implicated $Ca^{2+}$ release from intracellular stores as an important contributor, particularly to larger AHPs in aged animals (for reviews see [100, 101]). We did not explore the role of intracellular $Ca^{2+}$ stores in this study, nor many of the other cellular changes that surely contribute to the multifactorial process of aging. It is not our intention to suggest $Ca^{2+}$ channel expression is the only factor altering the activity of PCs in aged animals. Nevertheless, we do demonstrate that an increase in L-type $Ca^{2+}$ channels is *sufficient* to reproduce many of the changes in PC firing seen during aging. These results agree with a previous modeling study, which also found that an increase in L-type $Ca^{2+}$ conductance was sufficient to produce changes in adaptation and AHPs similar to those seen in aged CA1 PCs [102]. However, their model was mathematically complex, with 183 compartments and more than a dozen ionic currents – very different from the single-compartment, minimal model we present here. In addition, their study looked only at adaptation and AHPs, while our study goes further to investigate conditional bursting, spontaneous bursting, and responses to LFP stimulation.

## 4.3   Aging and excitability

Our simulations do show changes in the electrical activity of aged PCs, but do these changes represent decreased excitability? This question relates more broadly to how we think about excitability – the term is rarely clearly defined or used in a standardized way. In some studies, excitability is used to refer to a change in the firing rate of a cell over the course of an injected

current pulse, claiming that PCs with stronger adaptation are less excitable [18]. In our simulations under the adaptive firing parameter regime, the aPC did have stronger adaptation and fired fewer times during the stimulation period than the yPC. However, under some circumstances, this stronger adaptation occurred only after the aPC initially fired faster than the yPC (see for example Fig. 2 inset). Should we consider this decreased excitability?

If what concerns us with excitability is the activity of the cell over a given time period, then the results under the bursting parameter regimes are even less clear. The aPC always fired fewer APs per burst than the yPC, indicating something akin to stronger adaptation. However, if the 'event' we are considering is instead the burst, there are conditions under which the aPC fired a greater number of bursts in a given time period than the yPC. How should we interpret these results with respect to excitability? To our knowledge, there are very few experimental studies to date that have compared burst firing in young versus aged animal CA1 PCs, perhaps because of the relatively low percentage of cells with this firing pattern in certain developmental periods [83]. The only study of which we are aware compared spontaneous burst firing in young versus aged animal PCs during rest and exploratory behavior [103]. Unfortunately, there are several factors that make it difficult to compare our results with theirs. First, their recordings were extracellular. As such, no bursts from individual PCs are shown in the paper to compare to our simulations, which reproduce intracellular firing. Furthermore, the study focused on interspike intervals (ISIs), rather than the overall bursting pattern (e.g. number of bursts in a given period), which was our focus. Interestingly, Smith and colleagues [103] found that while ISIs were more left-skewed in aged animals when burst firing was recorded at rest, they found no difference between the age groups during behavior, suggesting that compensatory mechanisms work differently during behavior to adjust for changes in cellular excitability. Such compensatory mechanisms could be explored using our model, for example by adding simulated cholinergic input.

Other researchers use the term excitability to describe how easy it is to get a cell to fire in response to stimulation, referring to "propensity" [104] or "readiness" [105]. In this context, excitability could be measured by the rheobase, or minimum current which generates firing in a neuron, as done in some studies of aging in CA1 [106]. However, 'propensity' or 'readiness' could also be interpreted as how quickly a cell fires after stimulus onset. In the AHP simulations, we saw that the aPC required additional current to fire the same number of spikes as for the yPC. On the other hand, the aPC often fired sooner than the yPC. These effects were a result of the increased $Ca^{2+}$ current in aged cells – the larger $Ca^{2+}$ current depolarized the cells faster and caused them to fire sooner, but it also caused the SK current to be larger and consequently slowed firing. It is as if the aPCs were initially more excitable, but then 'burned out' more quickly than the yPCs.

Overall, we believe our results highlight the importance of moving away from vague, ill-defined terms like 'excitability' in favor of precise language that describes the effect of interest (e.g., PCs spiked faster, spiked sooner, etc.). We hope that, in a broader sense, our study will encourage neuroscientists, and particularly aging researchers, to reevaluate how they think and write about excitability.

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