Branching dendrites with resonant membrane: a “sum-over-trips” approach

S. Coombes · Y. Timofeeva · C.-M. Svensson · G. J. Lord · K. Josić · S. J. Cox · C. M. Colbert

Abstract Dendrites form the major components of neurons. They are complex branching structures that receive and process thousands of synaptic inputs from other neurons. It is well known that dendritic morphology plays an important role in the function of dendrites. Another important contribution to the response characteristics of a single neuron comes from the intrinsic resonant properties of dendritic membrane. In this paper we combine the effects of dendritic branching and resonant membrane dynamics by generalising the “sum-over-trips” approach (Abbott et al. in Biol Cybernetics 66, 49–60 1991). To illustrate how this formalism can shed light on the role of architecture and resonances in determining neuronal output we consider dual recording and reconstruction data from a rat CA1 hippocampal pyramidal cell. Specifically we explore the way in which an $I_h$ current contributes to a voltage overshoot at the soma.

Keywords Dendrites · Quasi-active membrane · “Sum-over-trips” · Cable theory

1 Introduction

The dendrites of neurons often exhibit elaborate branching structures, as so wonderfully described in the book Dendrites (Stuart et al. 1999). These branching projections act to transfer electrical activity between synapses and the soma. Indeed the electrical and branching properties of dendrites are known to play a critical role in integrating synaptic inputs and in determining whether action potentials are generated at either the soma or other hot spots within the dendritic tree (Mainen and Sejnowski 1996; van Ooyen et al. 2002). Much insight into the contribution of the electrical properties of dendrites to neuronal function has come from theoretical work first developed by Rall. In this regard Rall’s work on dendritic modelling is one of the more obvious success stories in the field of mathematical neuroscience. For a historical perspective on Rall’s contributions in this area we refer the reader to the book by Segev et al. (1995), and for a review of many of the mathematical techniques still being used today we recommend Tuckwell’s (1988) book. Although much of the more recent work on dendritic modelling has tended to focus on the dynamics of nonlinear voltage-gated ion conductances, reviewed in (Segev and London 2000), it is important to recognise that the passive properties of the dendritic tree provide the fundamental substrate for dendritic dynamics, as emphasised in the recent review article of London and Häusser (2005). In fact it would be fair to extend the statement that
passive electrical properties form the backbone of neuronal computation to also include the resonant properties of dendritic membrane. Many neurons exhibit resonances whereby subthreshold oscillatory behaviour is amplified for inputs at preferential frequencies. One illustrative example is that of hair cells from the sacculi of frogs, which are seismic receptors sensitive in the frequency range of 10–150 Hz. For these cells the ionic mechanism underlying resonance is known to involve both a Ca\(^{2+}\) and a Ca\(^{2+}\)-activated K\(^+\) current (Hudspeth and Lewis 1988). From a mathematical perspective Mauro et al. (1970) have shown that a linearisation of such channel kinetics, about rest, may adequately describe the observed resonant dynamics. In the terminology of electrical engineering the resulting linear system has a membrane impedance that displays resonant-like behaviour due to the additional presence of inductances. This extends the more usual ‘RC’ circuit description of passive membrane to the so-called quasi-active or ‘LRC’ case. Further work by Koch and Poggio (2005) showed how the response function for an arbitrary branched dendritic tree with quasi-active membrane could be calculated in the Laplace (frequency) domain. This approach generalised the original graphical calculus of Butz and Cowan (1974), valid for passive dendritic geometries. Later work by Abbott et al. (1991) showed how to calculate response functions for passive branched dendritic trees directly in the time-domain. The machinery to do this borrows heavily from the path-integral formalism for describing Brownian motion, and was used to give simple diagrammatic rules for obtaining the Green’s function for a passive tree (Abbott 1992). For a discussion of the computational advantages of this approach see Cao and Abbott (1993). As noted in (Abbott et al. 1991) the path integral approach relies on the superposition principle and so should also be applicable to quasi-active membranes, since these are described by linear equations. It is precisely this problem that we address and solve in this paper. We present our results in the “sum-over-trips” language of Abbott et al. (1991) and show how it may be interpreted in the language of ‘LRC’ circuits, i.e. circuits with a resistor, capacitor and inductance in parallel. To start with consider a generic ionic membrane current of the form

\[ I = I(V, w_1, \ldots, w_N), \]

(1)

where \( V \) is a voltage and the \( w_k \) are gating variables that satisfy

\[ \tau_k(V) \delta w_k = w_{k, \infty}(V) - w_k, \quad k = 1, \ldots, N. \]

(2)

It is traditional to write \( \tau_k(V) = (\alpha_k(V) + \beta_k(V))^{-1} \), where \( w_{k, \infty}(V) = \alpha_k(V) \tau_k(V) \). Now consider variations around some fixed point \( (V, w_1, \ldots, w_N) = (V_{ss}, w_{1, \infty}(V_{ss}), \ldots, w_{N, \infty}(V_{ss})) \), so that

\[ \delta I = \frac{\delta V}{R} + \sum_{k=1}^{N} \left( \frac{dI}{\delta w_k} \right)_{ss} \delta w_k, \]

(3)

where we introduce the resistance \( R \) defined by \( R^{-1} = \frac{\delta I}{\delta V}_{ss} \), and the subscript ss denotes that quantities are evaluated at steady state. Using (2) we can write the evolution of the perturbations in the gating variable as

\[ \left( \frac{d}{dt} + \alpha_k + \beta_k \right) \delta w_k = \left[ \frac{d\alpha_k}{dV} - w_{k, \infty} \frac{d(\alpha_k + \beta_k)}{dV} \right] \delta V. \]

(4)

We may now write (3) in the form

\[ \delta I = \frac{\delta V}{R} + \sum_{k=1}^{N} \delta I_k, \]

(5)

where

\[ \left( r_k + L_k \frac{d}{dt} \right) \delta I_k = \delta V. \]

(6)

Here

\[ r_k^{-1} = \tau_k \frac{\delta I}{\delta w_k} \left[ \frac{d\alpha_k}{dV} - w_{k, \infty} \frac{d(\alpha_k + \beta_k)}{dV} \right]_{ss}, \]

(7)

\[ L_k = \tau_k r_k. \]

(8)

Hence, for a small perturbation around the steady state, the current \( I \) responds as though the resistance \( R \) is in parallel with \( N \) impedance lines. Each of these is a resistance \( r_k \) that is itself in series with an inductance \( L_k \). Such inductive terms account for the oscillatory overshoot commonly seen in response to depolarising current steps or even after the firing of an action potential. Koch terms this form of equivalent linear membrane circuit quasi-active to distinguish it from a truly active (i.e. nonlinear) membrane (Koch 1984).
Now consider a general current balance equation in the form
\[
\frac{dV}{dt} = -g_L(V - V_L) - I + I_{\text{inj}},
\]
(9)

The linearised equations will be
\[
\frac{dV}{dt} = -\frac{V}{R} - \sum_{k=1}^{N} I_k + I_{\text{inj}}, \quad \frac{1}{R} = g_L + \frac{1}{R'}
\]
(10)
\[
L_k \frac{dI_k}{dt} = -r_k I_k + V.
\]
(11)

The steady state voltage satisfies \( I(V_{ss}, w_{1,\infty}(V_{ss}), \ldots, w_{N,\infty}(V_{ss})) + g_L(V_{ss} - V_L) = I_{\text{inj}} \). Introducing the Laplace transform (with spectral parameter \( \omega \))
\[
f(\omega) = \int_{0}^{\infty} dt e^{-\omega t} f(t),
\]
(12)
we find that \( V(\omega) = K(\omega) I_{\text{inj}}(\omega) \), where
\[
K(\omega) = \frac{\sum_{k=1}^{N} r_k + \omega L_k}{(C \omega + \tilde{R}^{-1})(\sum_{k=1}^{N} r_k + \omega L_k) + 1}.
\]
(13)

We identify \( K(\omega) \) as the impedance of the linearised system, and note that it is a ratio of two polynomials, with the denominator of order \( N + 1 \), and the numerator of order \( N \) (where \( N \) is the number of gating variables). For example, the linearisation of the Hodgkin–Huxley model generates a bandpass filter with optimal response around 67 Hz [see for example (Koch 1984)]. The range of validity of the reduction process is limited to a few millivolts around the resting potential. Later in Sect. 4 we will use this approach to calculate the equivalent 'LCR' circuit model for a membrane with a so-called \( I_h \) current.

We are now in a position to treat the dynamics of an infinite unbranched passive dendritic cable model supplemented by resonant currents of the type just described. In Sect. 3 we will show how the response of a tree can be built up in terms of this solution and an appropriate set of coefficients determined both by the geometry of the tree and the resonances on each segment of the tree. We take as our starting point the standard cable equation coupled to a set of resonant currents. Writing \( V = V(X, t), X \in \mathbb{R}, t \geq 0 \), as the dendritic voltage the resonant cable equation is
\[
\frac{dV}{dt} = -\frac{V}{\tau} + D \frac{\partial^2 V}{\partial X^2} - \frac{1}{C} \left[ \sum_{k=1}^{N} I_k - I_{\text{inj}} \right],
\]
(14)
\[
L_k \frac{dI_k}{dt} = -r_k I_k + V.
\]
Here \( D \) is the cable diffusion coefficient and \( \tau \) the (passive) cell membrane time constant. After Laplace transforming (14) we obtain the ODE
\[
-V_{XX} + \gamma^2(\omega) V = \frac{I_{\text{inj}} - \sum_{k} L_k I_k(X, t=0) + CV(X, t=0)}{CD},
\]
(15)
\[
\gamma^2(\omega) = \frac{1}{D} \left[ \frac{1}{\tau} + \omega + \frac{1}{C} \sum_{k} \frac{1}{r_k + \omega L_k} \right],
\]
(16)
where \( V = V(X, \omega) \) and \( I_{\text{inj}} = I_{\text{inj}}(X, \omega) \). Introducing a re-scaled space \( x = \gamma(\omega) X \) gives
\[
-V_{xx} + V = A,
\]
(17)
where
\[
A(x, \omega) = \frac{1}{CD\gamma^2(\omega)} \left[ I_{\text{inj}}(x/\gamma(\omega), \omega) - \sum_{k} L_k I_k(x/\gamma(\omega), t=0) \right] \frac{1}{r_k + \omega L_k} + CV(x/\gamma(\omega), t=0).
\]
(18)

The Green’s function of the operator \( (1 - d_{xx}) \) is simply \( H_{\infty}(x) = e^{-|x|/2} \), and we may write the general solution to (17) in the form
\[
V(x, \omega) = \int_{0}^{\infty} dy H_{\infty}(x - y) A(y, \omega).
\]
(19)

In original co-ordinates (and considering vanishing initial data) we have that
\[
V(X, \omega) = \int_{0}^{\infty} dy G_{\infty}(X - Y, \omega) I(Y, \omega),
\]
(20)
where \( I(X, \omega) = I_{\text{inj}}(X, \omega)/C \) and
\[
G_{\infty}(X, \omega) = \frac{H_{\infty}(\gamma(\omega) X)}{D\gamma(\omega)} = \frac{e^{-\gamma(\omega)|X|}}{2D\gamma(\omega)}.
\]
(21)

Performing the inverse Laplace transform gives
\[
V(X, t) = \int_{0}^{t} ds \int_{0}^{\infty} dy G_{\infty}(X - Y, t - s) I(Y, s),
\]
(22)
where \( G_{\infty}(X, t) \), is the inverse Laplace transform of \( G_{\infty}(X, \omega) \). Note that in the limit \( r_k \rightarrow \infty \) we recover the purely passive system \( (\gamma^2(\omega) = (1/\tau + \omega)/D) \) with Green’s function
\[
G_{\infty}(X, t) = \frac{e^{-t/\tau}}{\sqrt{4\pi D t}} e^{-X^2/(4Dt)} \Theta(t).
\]
(23)
Here \( \Theta(t) \) is the Heaviside step function.

Next we will see how to treat a branched network of connected dendritic segments, each possessing its own resonant dynamics.
3 Branching

A caricature of a branched dendritic tree with resonant membrane is depicted in Fig. 1. We define a node as point where branch segments touch (i.e. the vertices of the graph describing the tree). Nodes that do not have child nodes will be called terminal nodes. A finite segment of the tree, labelled by \( i \), is described by the dynamics

\[
\frac{\partial V_i}{\partial t} = -\frac{V_i}{\tau_i} + D_i \frac{\partial^2 V_i}{\partial x^2} - \frac{1}{C_i} \left[ \sum_k L_{k,i} I_{k,i} - I_{inj,i} \right],
\]

\[
L_{k,i} \frac{dI_{k,i}}{dt} = -r_{k,i} I_{k,i} + V_i, \quad 0 \leq X \leq L_i.
\]

We may then specify the dynamics on a tree by ensuring the appropriate boundary conditions at all nodes and terminals. These are (i) continuity of potential, and (ii) conservation of current. If we choose the coordinates on all of the radiating branches so that the node is at the point \( X = 0 \) then continuity of potential requires that

\[
V_i(0, t) = V_j(0, t),
\]

for all values of \( i \) and \( j \) corresponding to segments radiating from the node. Conservation of current gives

\[
\sum_j r_j \left. \frac{\partial V_j}{\partial x} \right|_{x=0} = 0.
\]

Here \( r_j \) is the axial resistance on segment \( j \) (in \( \Omega/cm \)), and the sum is over all \( j \) values corresponding to segments radiating from the node in question. At an open terminal we impose \( V_i(L_i, t) = 0 \) and at a closed end \( \left. \frac{\partial V_i(X, t)}{\partial X} \right|_{X=L_i} = 0 \). After Laplace transforming (24) and rescaling we have that

\[
-(V_i)_x + V_i = A_i, \quad 0 < x < L_i(\omega),
\]

where \( L_i(\omega) = \gamma_i(\omega)L_i \) and

\[
\gamma_i^2(\omega) = \frac{1}{D_i} \left[ \frac{1}{\tau_i} + \omega + \frac{1}{C_i} \sum_k r_{k,i} + \omega L_{k,i} \right]^{-1},
\]

with

\[
A_i(x, \omega) = \frac{1}{C_i D_i} \gamma_i^2(\omega) \left[ I_{inj,i}(x/\gamma_i(\omega), \omega) \right.
\]

\[
- \sum_k L_{k,i} I_{k,i}(x/\gamma_i(\omega), \omega), \quad t = 0 \]

\[
+ C_i V_i(x/\gamma_i(\omega), \omega), \quad t = 0 \right].
\]

We may write the general solution to (27) in the form

\[
V_i(x, \omega) = \sum_j \int_0^{L_j(\omega)} dy H_{ij}(x, y, \omega) A_j(y, \omega),
\]

where \( H_{ij}(x, y, \omega) \) satisfies

\[
[1 - d_{x,i}] H_{ij}(x, y, \omega) = \delta_{ij} \delta(x - y).
\]

In addition \( H_{ij}(x, y, \omega) \) must satisfy boundary conditions similar to those given above for \( V_i \), namely

\[
\sum_i z_i(\omega) \left. \frac{\partial H_{ij}(x, y, \omega)}{\partial x} \right|_{x=0} = 0, \quad \gamma_i(\omega) = H_{i0}(0, y, \omega),
\]

Also at an open terminal node we require

\[
H_{ij}(L_i(\omega), \omega) = 0,
\]

and at a closed terminal node

\[
\left. \frac{\partial H_{ij}(x, y, \omega)}{\partial x} \right|_{x=L_i(\omega)} = 0.
\]

As it stands our discussion, and indeed that in (Abbott et al. 1991), does not include the coupling of a tree to a soma. For passive trees one way to treat this is to consider equivalent cylinders connected to an isopotential soma, as in the work of Evans et al. (1992, 1995). To include this important special node we also treat the soma as an equipotential surface with a resistance \( R_s \), capacitance \( C_s \) and further include an inductive pathway described by \( L_s \) and \( I_s \). The membrane voltage in the soma \( V_s \) has to satisfy the following equations

\[
C_s \frac{dV_s}{dt} = -\frac{V_s}{R_s} + \sum_j \left. \frac{1}{r_j} \frac{\partial V_j}{\partial X} \right|_{X=0} - I_s,
\]

\[
L_s \frac{dI_s}{dt} = -r_s I_s + V_s,
\]

with \( V_j(t) = V_j(0, t) \) and \( j \in \Gamma \), where \( \Gamma \) is the set that indexes all of the branches attached to the soma. In this case \( H_{ij}(x, y, \omega) \) must satisfy the further boundary condition

\[
\sum_{k \in \Gamma} \gamma_k(\omega) \left. \frac{\partial H_{kj}(x, y, \omega)}{\partial x} \right|_{x=0} = 0, \quad \gamma_s(\omega) H_{ks}(0, y, \omega) = 0,
\]
where
\[ y(\omega) = \frac{C_s \omega}{R_s} + \frac{1}{r_s + \omega L_s}. \]  

We now seek a solution for \( H_{ij}(x, y, \omega) \) in terms of the known response function \( H_\infty \) and the labels \((i, j, x, y)\) and the frequency dependent parameters \( \gamma_k(\omega) \), where \( k \) indexes every segment in the tree. An infinite series expansion of this type that is consistent with (31) is
\[ H_{ij}(x, y, \omega) = \sum_{\text{trips}} A_{\text{trip}}(\omega) H_\infty(L_{\text{trip}}). \]  

Here \( L_{\text{trip}} = L_{\text{trip}}(i, j, x, y, \omega) \) is the length of a path along the tree that starts at point \( x = \gamma_i(\omega)X \) on branch \( i \) and ends at the point \( y = \gamma_j(\omega)Y \) on branch \( j \). Note that on intermediate branches between branches \( i \) and \( j \), labelled by \( k \) say, that distances are measured in terms of the scaled coordinates \( \gamma_k(\omega)X, X \in [0, L_i] \). We shall call these frequency dependent path lengths trips in analogy to the terminology used in the “sum-over-trips” formalism of Abbott et al. (1991). However, it is important to stress that in our case these trips depend upon the set of frequency dependent parameters \( \gamma_k(\omega) \) that capture the resonant properties of the tree. As in (Abbott et al. 1991) trips are constructed in accordance with the following rules:

1. A trip may start out from \( \gamma_i(\omega)X \) by travelling in either direction along segment \( i \), but it may subsequently change direction only at a node or a terminal. A trip may pass through the points \( \gamma_i(\omega)X \) and \( \gamma_j(\omega)Y \) but must begin at \( \gamma_i(\omega)X \) on segment \( i \) and end at \( \gamma_j(\omega)Y \) on segment \( j \).

2. When a trip arrives at a node, it may pass through the node to any other segment radiating from the node or it may reflect from the node back along the same segment on which it entered.

3. When it reaches a terminal, a trip always reflects back, reversing its direction.

Every trip generates a term in (40) with \( L_{\text{trip}} \) given by summing the lengths of all the steps taken along the course of the trip. For example the four primary trips \( L_{\text{trip}}(i, j, x, y, \omega) \) on a simple dendritic tree consisting of two segments are
\[ \gamma_i(\omega)(L_i - X) + \gamma_j(\omega)Y, \gamma_i(\omega)(L_i + X) + \gamma_j(\omega)Y, \gamma_i(\omega)(L_i - X) + \gamma_j(\omega)(2L_i - Y) \text{ and } \gamma_i(\omega)(L_i + X) + \gamma_j(\omega)(2L_i - Y) \]  

respectively. Note that all longer trips, even in a larger branched network, would consist only of constant additions to these four basic lengths. Hence, \( L_{\text{trip}}(i, j, x, y, \omega) \) is a linear function of \( \pm x \), as required for \( H_\infty(L_{\text{trip}}) \) to be a solution to (31).

In Appendix A we prove that (40) satisfies the required boundary conditions if the trip coefficients \( A_{\text{trip}}(\omega) \) are chosen according to the following rules:

1. From any starting point \( A_{\text{trip}}(\omega) = 1 \).
2. For every node at which the trip passes from an initial segment \( k \) to a different segment \( m \) \((m \neq k)\) \( A_{\text{trip}}(\omega) \) is multiplied by a factor \( 2p_m(\omega) \).
3. For every node at which the trip enters along segment \( k \) and then reflects off the node back along segment \( k \) \( A_{\text{trip}}(\omega) \) is multiplied by a factor \( 2p_k(\omega) - 1 \).
4. For every closed (open) terminal node \( A_{\text{trip}}(\omega) \) is multiplied by a factor +1 (−1).

Here the frequency dependent parameters \( p_k(\omega) \) are given as
\[ p_k(\omega) = \frac{z_k(\omega)}{\sum_m z_m(\omega)}. \]  

For a node describing the soma this coefficient takes the modified form
\[ p_k(\omega) = \frac{z_k(\omega)}{\sum_m z_m(\omega) + \gamma_k(\omega)}. \]  

Thus we arrive at the generalisation of the “sum-over-trips” formalism that covers arbitrary resonant dendritic trees. In contrast to the original “sum-over-trips” formulation these arguments are developed in Laplace space and it remains to transform back to the temporal domain. To do this we first write \( V_i(X, \omega) \) in the unscaled spatial coordinates as
\[ \sum_j \int_0^\infty dY H_j(\gamma_i(\omega)X, \gamma_j(\omega)Y, \omega) \gamma_j(\omega)A_j(\gamma_j(\omega)Y, \omega). \]  

Introducing \( G_{ij}(X, Y, \omega) = H_{ij}(\gamma_i(\omega)X, \gamma_j(\omega)Y, \omega) / (D_j \gamma_j(\omega)) \) we have that
\[ V_i(X, \omega) = \sum_j \int_0^\infty dY G_{ij}(X, Y, \omega)[U_j(Y, \omega) + I_j(Y, \omega)], \]  

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where \( I_i(X, \omega) = I_{inj,i}(X, \omega)/C_i \) and
\[
U_i(X, \omega) = V_i(X, t = 0) - \frac{1}{C_i} \sum_k L_{k,i} I_{k,i}(X, t = 0) r_{k,i} + \omega L_{k,i}. \tag{45}
\]

After taking the inverse Laplace transform of (44) we obtain
\[
V_i(X, t) = \sum_j \left[ \int_0^\mathcal{L}_j dY G_{ij}(X, Y, t) U_j(Y, t = 0) + t \int_0^\mathcal{L}_j ds \int_0^s dY G_{ij}(X, Y, t - s) I_j(Y, s) \right]. \tag{46}
\]

Hence we identify the inverse Laplace transform of \( G_{ij}(X, Y, \omega) \), namely \( G_{ij}(X, Y, t) \), as the Green’s function of the resonant dendritic tree. Note that we can build this function using a combination of the “sum-over-trips” approach and a rescaling according to the resonant properties of each segment, before finally taking an inverse Laplace transform. Moreover, this naturally generalises the original approach of Abbott et al. (1991) not only to resonant dendritic trees, but to those with different passive membrane properties on each segment (i.e. \( \tau_i \) and \( D_i \) differ across segments).

For the case of \( \chi_k(\omega) = \gamma_m(\omega) \) then \( p_k(\omega) \) is independent of \( \omega \). In this case for the choice \( p_k = 1/2 \) a node acts no differently from a point on a single cable and we see that \( r_k^{-1} = \sum_{m \neq k} r_m^{-1} \). Since the axial resistance \( r_k \) scales as \( d_k^{-2} \sqrt{D_k} \), where \( d_k \) is the radius of the \( k \)th branch, then we recover Rall’s equivalent cylinder condition, \( d_k^{3/2} = \sum_{m \neq k} d_m^{3/2} \), as first noted in (Abbott et al. 1991). Note that the rules for computing \( G_{ij}(X, Y, \omega) \) require that trips start at \( X \) and end at \( Y \). To compute the function \( G_{ji}(Y, X, \omega) \) the reverse trips from \( Y \) to \( X \) have to be used. If the trips are generated in reverse order the probability factors associated with crossing the nodes will be different. However, a simple calculation shows that the result of summing over reversed as opposed to original trips differs only by a constant factor \( D_i r_i/(D_j r_j) \). This leads to the simple reciprocity identity
\[
G_{ij}(X, Y, t) = \frac{D_i r_i}{D_j r_j} G_{ji}(Y, X, t). \tag{47}
\]

In summary the Green’s function for an arbitrary branched resonant dendritic tree is given by the Bromwich integral (inverse Laplace transform)
\[
\frac{1}{2\pi i} \int_{c-i\infty}^{c+i\infty} e^{-s t} \frac{H_{ij}(\gamma_i(\omega) X, \gamma_j(\omega) Y, \omega)}{G_{ij}(\omega)} \, ds. \tag{48}
\]

Here, \( \gamma_i(\omega) \) is given by (28) and \( H_{ij}(X, Y) \) is given by (40). This last equation expresses \( H_{ij} \) in terms of an infinite sum of terms involving the fundamental response function \( H_\infty(x) = e^{-1|x|/2} \) and the length of a frequency dependent ‘trip’. The trip coefficients \( A_{\text{trip}}(\omega) \) are generated according to the rules in Abbott et al. (1991), under the replacement of \( p_k \) by \( p_k(\omega) \) as given by (41) and (42). For any practical computational implementation the number of terms in the sum for \( H_{ij} \) must be truncated. Since contributions of the individual terms for long trips (fixed \( \omega \)) are small, this truncation can be done by introducing a length cutoff. Numerical inversion of the Laplace transform can be efficiently performed using fast Fourier transforms. We have checked the validity of this approach by comparing code written in both C++ and Matlab with brute force simulations performed in NEURON (Carnevale and Hines 2006) for a wide range of realistic neuronal geometries. Moreover, in many cases a good approximation is reached by including only the four shortest trips. In the next section we present some examples of this approach.

4 Resonances associated with \( I_h \)

Many neurons exhibit resonances whereby subthreshold oscillatory behaviour is amplified for inputs at preferential frequencies (Hutcheon and Yarom 2000). A nice example is that of the subthreshold frequency preference seen in neurons of rat sensorimotor cortex (Hutcheon et al. 1996). In response to suprathreshold inputs, this frequency preference leads to an increased likelihood of firing for stimulation near the resonant frequency. It is known that the nonlinear ionic current \( I_h \) is partly responsible for this resonance. Indeed \( I_h \) plays a variety of important roles in many neuronal and non-neuronal cell types (Pape 1996), and it is believed that the presence of \( I_h \) in dendrites could have a significant impact on the integration of subthreshold synaptic activity (Magee 1998). As such it is interesting to apply the framework we have developed above to the specific case of the \( I_h \) current. Moreover, to gain insight into the resonant effects of \( I_h \) it is useful to explore both idealised geometries and more realistic reconstructed cells.

Here we focus on the \( I_h \) model of Magee (1998), given by
\[
I_h = g_h(V - V_h) f, \tag{49}
\]
where \( f \) is a single gating variable. The details of this model are given in Appendix B. In Fig. 2 we plot the membrane voltage of an unbranched semi-infinite resonant dendrite, with a uniformly distributed nonlinear \( I_h \) current, in response to a constant current injection of finite duration. Also shown is the plot of the voltage from the linearised model. The closeness of the two curves emphasises the usefulness of the quasi-active membrane description in approximating a fully nonlinear model of \( I_h \).
For the membrane with the ‘LRC’ circuit shown in Fig. 1 (with a single inductance) the natural frequency is given by

\[ \omega_* = \frac{\sqrt{CL - Cr}}{CL}. \]  

(50)

For an unbranched infinite resonant dendrite with a homogeneous distribution of electrical properties along its length the function \( G_\infty(X, \omega) \) given by (21) will have its maximum at the frequency \( \omega_* \) for any \( X \). In contrast a branched system will not exhibit such a stimulus location independent property.

To study how branches with differing natural frequencies interact we first consider the case of two semi-infinite branches (with common passive parameters \( \tau, D \) and \( C \)), with natural frequencies \( \omega_{*)1} = \left(\sqrt{CL_i - Cr_i}\right)/\left(CL_i\right) \) connected together at \( X = 0 \), such that \( \omega_{*)2} > \omega_{*)1} \). The response functions \( G_{11}(X, Y, \omega) \) and \( G_{12}(X, Y, \omega) \) are easily constructed for an applied stimulus at location \( Y \) on branch 1 and branch 2 respectively. The frequencies at which these functions reach their maximum define the resonant frequency of the (admittedly simple) dendritic tree, and thus satisfy \( \partial G_{ij}(X, Y, \omega) / \partial \omega = 0 \). Note that the function \( G_{12}(X, Y, \omega) \) consists of only one term \( 2p_2(\omega)H_\infty(\gamma_1(\omega)X + \gamma_2(\omega)Y) / (D_2\gamma_2'(\omega)) \), and the resonant frequency \( \Omega \) of the tree satisfies

\[ (\gamma_1'(X + \gamma_2'(Y)(r_1\gamma_2 + r_2\gamma_1) + r_1\gamma_2' + r_2\gamma_1') = 0, \]

where \( \gamma_1 = \gamma_1(\Omega) \) and \( \gamma_1' = d\gamma_1(\Omega)/d\omega \). The resonant frequency of the tree when the stimulus and response are taken on the same branch is obtained in a similar fashion (though the expression for \( G_{11}(X, Y, \omega) \) consists of two terms). In Fig. 3 we plot the resonant frequency of the tree, \( \Omega \), with varying stimulus point \( Y \) and fixed response point \( X \). This figure nicely demonstrates that if the natural frequency of the branches differ, then the frequency with which the whole tree responds depends on the location of the stimulus. Loosely speaking the response on branch 1 (far from branch 2) has a frequency more like that of \( \omega_{*)1} \) when the stimulus is nearby, and a frequency more like that of \( \omega_{*)2} \) when the stimulus is on the other branch and far away. Figure 4 shows examples of voltage profiles and Green’s function on branch 1 in response to the injected current on branch 2.

4.2 Idealised geometries II

Here we consider the case of three semi-infinite dendrites emanating from a central node, as depicted in Fig. 5. As before, the stimulus is applied to either branch 1 or branch 2. The resonant frequency of the tree may be constructed in a similar fashion to that above. For example, the resonant frequency of the function \( G_{12}(X, Y, \omega) \) is found by solving

\[ (\gamma_1'(X + \gamma_2'(Y)(r_1r_3\gamma_2 + r_2r_3\gamma_1 + r_1r_2\gamma_3) + r_1r_3\gamma_2' + r_2r_3\gamma_1' + r_1r_2\gamma_3') = 0. \]

(52)

In Fig. 5 we plot the resonant frequency of the tree as recorded on branch 1 in response to a stimulus (on branches 1 and 2) as the natural frequency on branch 3 is varied. Fig. 5a shows an example where \( \omega_{*)1} = \omega_{*)2} \). If \( \omega_{*)2} \) is below some critical value,
then we find that $\Omega = \omega_1^* = \omega_2^*$. Above this critical value then $\Omega$ increases toward $\omega_3^*$ (and more so when the stimulus point is close to branch 3). Figure 5b shows an example when the natural frequencies of branches 1 and 2 differ. If branch 3 is passive or only weakly resonant ($\omega_3^* \approx 0$), then the system behaves much like that of the previous example (i.e. as a system of just two interacting resonant branches). As $\omega_3^*$ is increased from zero the resonant frequency, $\Omega$, of the tree (as measured on branch 1) increases toward $\omega_3^*$.

4.3 Idealised geometries III

Till now we have considered uniform distributions of conductances along a dendritic segment. However, it is now well known that dendritic trees can have nonuniform membrane conductances [see for example (London et al. 1999)]. One way to treat spatial dependencies is to break a single segment into many pieces, each with a distinct yet uniform parameter set. A piecewise constant approximation to any spatially varying conductance can then be naturally implemented on this segmented cable. Using such an approach we now briefly turn to an observation that the conductance of $I_h$ increases with the distance from the soma in pyramidal neurons (Kole et al. 2006; Li and Ascoli 2006).

A spatially varying conductance $g_h = g_h(X)$ leads to a spatially varying steady-state and from Eqs. (7) and (8) to spatially varying forms for the resistance and inductance. Direct numerical simulations of a long cable with the full nonlinear $I_h$ model (49) show that the steady state voltage change is less than 1 mV and the steady state change in the gating variable is less than 0.01 for a ten-fold magnification of $g_h$. Hence, it is reasonable to use the approximation $r_1(X) = Kg_h(X)$ and $L(X) = \tau r(X)$ (with the constant $K$ determined from (7) assuming a space-independent steady-state), for a piecewise constant function $g_h(X)$. Using a chain of 100 segments (for a dendrite of total length 200 $\mu$m) we took a piecewise constant approximation to the function $g(1 + 3X/100)$ [suggested by data from (Migliore et al. 2005)], for fixed $\bar{g} = 0.05$ mmho cm$^{-2}$, and considered the response at the soma ($X = 0$) to stimuli at increasing separation from the soma. In contrast to a spatially uniform conductance the time-to-peak of the signal at the soma remained approximately constant, whereas in the former case it increased with distance from the soma. In some sense we can view this as another example of how to achieve dendritic democracy (Häusser 2001), albeit this time for resonant dendrites.

In this example and the next the function $H_{ij}$ was constructed using four primary trips as well as their extensions generated by adding trips that start and end at the same point (as discussed in Sect. 3 for the case of two connected segments). Longer trips generated from the primary trips were compared with the imposed length cutoff and ignored if $L_{trip}(i, j, x, y, \omega) > 6$ for fixed $\omega$. The convergence of the solution constructed according to the above procedure was validated by introducing increasing length cutoffs as well as by direct numerical simulations.

4.4 A reconstructed cell

Here we apply our formalism to a real neuronal geometry. We consider a rat CA1 hippocampal pyramidal cell, visualised with differential interference contrast optics using infrared illumination. The geometry of this cell was reconstructed (indicating the presence of 396 branches along with the soma) and saved in .hoc file format (Carnevale and Hines 2006). A diagram of the reconstructed cell is shown in Fig. 6a. Dual
Fig. 5 An idealised dendritic tree with three semi-infinite resonant branches. Here we plot the resonant frequency $\Omega$ of the dendritic tree as measured on branch 1 (at $X = 200 \mu m$), whilst varying both the natural frequency on branch 3 and the point of stimulus (on either branch 1 or branch 2). Passive membrane parameters for all three branches as in Fig. 3.

a The resistances and inductances of branches 1 and 2 are the same and equal $r = 27,000 \Omega \text{cm}^2$ and $L = 2,300 H \text{cm}^2$, respectively.
b The resistance and inductance of branch 1 are $r_1 = 27,000 \Omega \text{cm}^2$ and $L_1 = 2,300 H \text{cm}^2$ respectively. The resistance and inductance of branch 2 are $r_2 = 13,500 \Omega \text{cm}^2$ and $L_2 = 1,150 H \text{cm}^2$ respectively. The resistance and inductance of branch 3 vary (between 67,500 and 5,400 $\Omega \text{cm}^2$ for $r_3$ and between 5,750 and 460 $H \text{cm}^2$ for $L_3$) for both a and b.

Simultaneous whole-cell patch-clamp recordings were made in this cell and data were acquired at 5 KHz. Before sampling the data were filtered at 1 KHz with an eight pole filter. A pulse current of amplitude $-300 pA$ and duration 400 ms (beginning at 10 ms) was injected at the dendritic trunk. Dendritic and somatic recordings in response to this current injection are shown in Fig. 6b, where the voltages are plotted with respect to rest (at about $-70 mV$). The observed oscillatory voltage overshoot is believed to be associated with an $I_h$ current. In the corresponding quasi-active model with a single inductive current pathway (built using the reconstructed cell data) we take a uniform distribution of $I_h$ channel conductances. In this case the resistance $r$ and inductance $L$ of the resonant membrane are the same for all branches. Dendritic and somatic voltage responses in the model are shown by the solid lines in Fig. 6b. The parameters $r$ and $L$ were fitted to the data, rather than assuming a specific model of $I_h$ and fixing them according to Eqs. (7) and (8). As can be seen from the close agreement between theory and experiment in Fig. 6 the “sum-over-trips” machinery generates quasi-active filters for realistic tree structures that can be used to organise and understand the structure of real dual electrode recording data.
Fig. 6  a Reconstructed rat CA1 hippocampal pyramidal cell.  b An example of dendritic (red) and somatic (blue) dual simultaneous recording (with respect to rest) in response to the current injection at the dendrite trunk (average from 20 sweeps). A pulse current with amplitude −300 pA is applied for a duration of 400 ms starting from 10 ms. The other two curves in b are dendritic and somatic voltage responses calculated from the model of the branched cell with resonant membrane. The model cell was stimulated at the dendrite (as shown in a) with the same current used in experimental recordings. Parameters across the tree: the specific cytoplasmic resistivity \( \rho_c \) = 100 \( \Omega \) cm, \( C = 1 \) \( \mu F/cm^2 \) and the resistance across a unit area of passive membrane \( R_h = 20,000 \) \( \Omega \) cm\(^2\). Thus \( \tau = 20 \) ms and diffusion coefficients \( D_h \) vary from branch to branch. The conductance of \( I_h \) is assumed to be uniform across the cell with the resistance and inductance of resonant membrane \( r = 24,000 \) \( \Omega \) cm\(^2\) and \( L = 2,700 \) H cm\(^2\), respectively. Note that the voltage drop across the electrode’s resistance is not fully compensated and the bridge balance is not corrected on the dendritic recording so that we shift our model results by an appropriate value of 10 mV.

5 Discussion

We have developed a “sum-over-trips” formalism for constructing the Green’s function for a branched dendritic tree and soma with quasi-active membrane. This generalises the original work of Abbott et al. (1991) for passive dendrites. Although in this paper we have focused on the ionic current \( I_h \), we stress here that our approach can handle any channel kinetics written in the standard language of dynamic gating variables. Such a framework for studying (linear) dendritic neuron models obviates the need for the numerical solution of an underlying set of PDEs. One advantage of using the Green’s function in comparison to a more standard compartmental approach is that this function has to be computed only once for a given dendritic structure. Thus changing the stimulation protocol does not require a whole new numerical simulation, rather just a convolution of the Green’s function with the new input. In Laplace space at fixed frequency, the computational time will depend on the geometry of the tree, and issues of algorithmic efficiency reduce to those discussed in detail by Cao and Abbott for passive dendrites (Cao and Abbott 1993). Although their approach can not treat resonant membrane our generalisation naturally does so with the introduction of a frequency parameter. The remaining computational overhead is then to perform a single inverse Laplace transform which can be done efficiently with standard numerical techniques. Moreover, our framework allows for a more analytical exploration of the way in which dendritic segments with differing natural frequencies contribute to an overall resonance at the level of the whole tree. It has also proved practical in use with real neuronal geometries. Indeed there is now a growing body of reconstructed cell data that can be uploaded from databases such as ModelDB, and used within the framework we have developed here. Of course these geometries must be supplemented with data governing the distribution of active ionic conductances. Alternatively, with access to dual potential recording data, one may recover the quasi-active properties of dendritic neurons using the theory and algorithms developed by Cox and Griffith (2001).

Two natural extensions of the work in this paper suggest themselves; (i) to cover tapered dendrites, and (ii) to cover active dendrites. If in the former case the underlying PDE model of the tapered model is linear [see (Scott 2002) for a recent discussion of tapering] then there may be some hope to extend the “sum-over-trips” formalism. How to recover quasi-active properties from tapered dendrites is already known (Cox and Raol 2004). The problem of treating truly active dendrites (Johnson et al. 1996) would seem at face value to be a substantially harder challenge. However, recent work on the spike-diffuse-spike model has shown how a system of point hot-spots embedded throughout a passive tree can provide a reasonable caricature of a tree with active conductances (Timofeeva et al., 2006). The solution of this model is expressed in part using the Green’s function of the tree without hot-spots. Obviously this can be obtained with the “sum-over-trips” techniques we have described in this paper. Both of these extensions are topics of current research and will be reported upon elsewhere.

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\(^{1}\) http://senselab.med.yale.edu/senselab/modeldb.
and then reflecting back along branch $k$ of the reflection from the node these trips pick up a factor to trips that start out from the node along branch $k$ that these are longer by the length $x$.

Using an extension of the argument in (Abbott 1992) we prove here that the the rules for generating the trip coefficients are consistent with the boundary conditions describing an arbitrary branched resonant dendritic tree.

Let $x$ denote the distance away from the node along the branch $k$ (see Fig. 7). The location of the stimulus $y$, the segment number $j$ and the variable $\omega$ are all considered to be arbitrary. Suppose that we sum all the trips starting from the node itself and ending at point $y$ on branch $j$. We denote the result of summing over all trips that initially leave the node along segment $k$ by $Q_{kj}(0, y, \omega)$. The result of all trips that leave the node along other branches $m$ is denoted by $Q_{mj}(0, y, \omega)$.

Trips that start out from $x$ and move away from the node are identical to trips that start out from the node itself along segment $k$. The only difference is that the trips in the first case are shorter by the length $x$. We denote the sum of such shortened trips by $Q_{kj}(-x, y, \omega)$. The argument $-x$ means that a distance $x$ has to be subtracted from the length of each trip summed to compute $Q$ (not that the trips start at the point $-x$).

Trips that start out from $x$ by moving toward the node and then reflecting back along branch $k$ are also identical to trips that start out from the node along branch $k$ except that these are longer by the length $x$. In addition, because of the reflection from the node these trips pick up a factor $2p_k(\omega) - 1$ according to the “sum-over-trips” rules. Therefore, the contribution to the solution $H_{kj}(x, y, \omega)$ from these trips is $(2p_k(\omega) - 1)Q_{kj}(x, y, \omega)$.

Finally, we have to take into account trips that start from $x$, move toward the node along branch $k$ and then leave the node by moving out along any of the radiating branches $m$, $m \neq k$. Crossing the node introduces a factor $2p_m(\omega)$ and the sum of such trips is given by $2p_m(\omega)Q_{mj}(x, y, \omega)$.

The full solution $H_{kj}(x, y, \omega)$ includes the contributions from all different types of trips we have been discussing.

Thus,

$$H_{kj}(x, y, \omega) = Q_{kj}(-x, y, \omega) + (2p_k(\omega) - 1)Q_{kj}(x, y, \omega) + \sum_{m \neq k} 2p_m(\omega)Q_{mj}(x, y, \omega).$$

The functions $Q$ in this formula consist of infinite sums over trips, but we do not need to know what they are to show that the solution $H_{kj}(x, y, \omega)$ satisfies the node boundary conditions. At a node point we have

$$H_{kj}(0, y, \omega) = \sum_m 2p_m(\omega)Q_{mj}(0, y, \omega).$$

The sum in the last formula is over all segments radiating from the node including branch $k$ and, thus, it shows that the solution at the point $x = 0$ is independent of $k$. Therefore $H_{kj}(x, y, \omega)$ obeys the boundary condition (32).

To prove the boundary condition (33) we use Eq. (53) to find that

$$\left. \frac{\partial H_{kj}(x, y, \omega)}{\partial x} \right|_{x=0} = \sum_m 2p_m(\omega) \left. \frac{\partial Q_{mj}(x, y, \omega)}{\partial x} \right|_{x=0} - 2 \left. \frac{\partial Q_{kj}(x, y, \omega)}{\partial x} \right|_{x=0}.$$

Now we multiply this result by $p_k(\omega)$ and sum over $k$ to get

$$\sum_k p_k(\omega) \left. \frac{\partial H_{kj}(x, y, \omega)}{\partial x} \right|_{x=0} =$$

$$\sum_m 2p_m(\omega) \left. \frac{\partial Q_{mj}(x, y, \omega)}{\partial x} \right|_{x=0} \left( \sum_k p_k(\omega) - 1 \right).$$

Using the property that $\sum_k p_k(\omega) = 1$ we have

$$\sum_k p_k(\omega) \left. \frac{\partial H_{kj}(x, y, \omega)}{\partial x} \right|_{x=0} = 0.$$

Since $p_k(\omega)$ is proportional to $z_k(\omega)$ the solution $H_{kj}(x, y, \omega)$ satisfies the boundary condition (33).

A similar derivation can show that the “sum-over-trips” rules used at the terminals are also correct. If we consider a terminal point instead of a node point, $p_m(\omega) = 0$ for $m \neq k$ in Eq. (53) and $p_k(\omega) = 0$ for an open end or $p_k(\omega) = 1$ for a closed end boundary condition at the terminal. Then Eqs. (54) and (57) indicate that (34) and (35) are obeyed at all open and closed terminal nodes.

In the presence of the soma we have to check that the solution $H_{kj}(x, y, \omega)$ satisfies the corresponding boundary conditions (32) and (38). The proof that the solution satisfies the boundary condition (32) is identical to that for a node point. To show that the boundary condition (38) is also satisfied we use the following properties for $Q_{kj}(x, y, \omega)$, namely

Fig. 7 Main trips at a node

Appendix A

like to acknowledge support from the DTI in the form of a UK-Texas Biosciences Initiative award.
Here we complete the details of the Appendix B. The potential \( V_h = -16 \) mV and the conductance \( g_{th} = 0.09 \) mmho cm\(^{-2}\). The functions that appear in the gating dynamics are \( \alpha_f(V) = w_{\infty}(V) \), \( \beta_f(V) = (1 - w_{\infty}(V))/\tau_f(V) \). Here (for temperature 27°C)

\[
\tau_f(V) = \frac{\exp[0.03326(V + 80)]}{0.00446(1 + \exp[0.08316(V + 80)])},
\]

\[
w_{\infty}(V) = \frac{1}{1 + \exp[(V + 92)/8]}.
\]

These properties are a direct consequence of the form of \( H_\infty(x) = e^{-|x|}/2 \). We then have that

\[
\frac{\partial H_{kj}(x, y, \omega)}{\partial x} \bigg|_{x=0} = -2p_k(\omega)Q_{mj}(0, y, \omega) + \sum_m 2\gamma_k(0, y, \omega).
\]

Note that, as compared to a simple node point, \( p_k(\omega) \) for a branch radiating from the soma takes the modified form given by (42). By multiplying Eq. (60) by \( z_k(\omega) \) and summing over \( k \) we obtain

\[
\sum_k z_k(\omega) \frac{\partial H_{kj}(x, y, \omega)}{\partial x} \bigg|_{x=0} = \sum_m 2z_m(\omega)Q_{mj}(0, y, \omega)\left(1 - \frac{\sum_m z_m(\omega) + \gamma_j(\omega)}{\sum_m z_m(\omega)}\right).
\]

This gives us

\[
\sum_k z_k(\omega) \frac{\partial H_{kj}(x, y, \omega)}{\partial x} \bigg|_{x=0} = \gamma_j(\omega) \sum_m \frac{2z_m(\omega)}{\sum_m z_m(\omega) + \gamma_j(\omega)} Q_{mj}(0, y, \omega).
\]

Using the equality (54) we recover the boundary condition (38).

### Appendix B

Here we complete the details of the \( I_h \) model given by (49) as described in (Magee 1998). The potential \( V_h = -16 \) mV and the conductance \( g_{th} = 0.09 \) mmho cm\(^{-2}\). The functions that appear in the gating dynamics are \( f_\infty(V) \equiv w_{\infty}(V) \), \( \alpha_f(V) = w_{\infty}(V)/\tau_f(V) \) and \( \beta_f(V) = (1 - w_{\infty}(V))/\tau_f(V) \). Here (for temperature 27°C)

\[
\tau_f(V) = \frac{\exp[0.03326(V + 80)]}{0.00446(1 + \exp[0.08316(V + 80)])},
\]

\[
w_{\infty}(V) = \frac{1}{1 + \exp[(V + 92)/8]}.
\]

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