

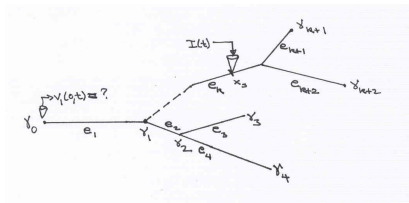
Inverse Problems for Neuronal Cables on Graphs

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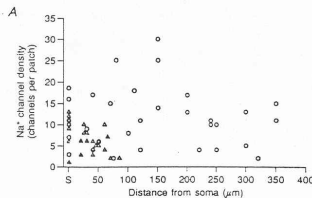
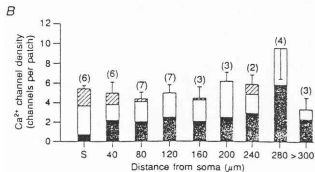
UMBC and University of Alaska Fairbanks

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Motivation



Can we use boundary data (voltage & current measurements) to recover various distributed parameters in a dendritic tree model (quantum tree graph)?



Single Branch, Single Conductance

1. JB, G Cracian, 2005: numerical method on

- single unknown ion density: $(1 + q(x))u_t + q(x)u = u_{xx}$,
 $0 < x < L$, $u_x(0, t) = f(t)$, $u_x(L, t) = 0$
- unknown spine density in (nonlinear) Baer-Rinzel model:

$$\begin{aligned}u_t + u &= u_{xx} + \rho n(x)(v - u) \\v_t + i_{HH}(v, m, n, h) &= \rho(u - v)\end{aligned}$$

2. D Wang, PhD thesis, UMBC, 2008: PDE-constrained
optimization methods
unknown spine density in Baer-Rinzel model

3. S Avdonin, JB, 2013, 2015 Boundary Control Theory on
single unknown conductance $u_t + g(x)u = u_{xx}$

Dendritic Trees

R.C. Cannon et al. / *Journal of Neuroscience Methods* 84 (1998) 49–54

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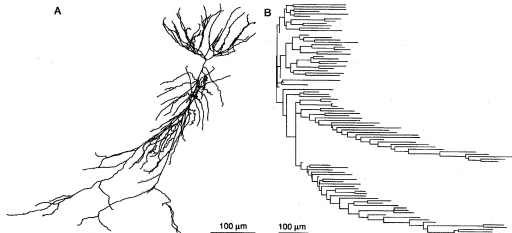
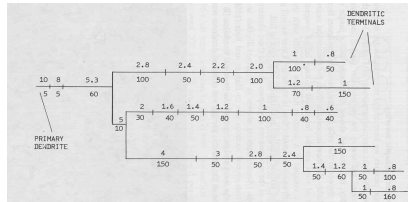


Fig. 3. A sample of representations available with the morphology editor as applied to the CA1 cell n184. A: a simple skeleton projection. B: the appropriate lengths and connections of all branches without direction information, in the form of a dendrogram.

motoneuron.jpg



Metric Graphs

Let $\Gamma = \{E, V\}$ be a finite compact metric graph.

$E = \{e_j\}_{j=1}^N$ is a set of edges and $V = \{\nu_j\}_{j=1}^M$ is a set of vertices. A graph is called a **metric graph** if every edge $e_j \in E$ is identified with an interval (a_{2j-1}, a_{2j}) of the real line with a positive length $l_j = |a_{2j-1} - a_{2j}|$, and a graph is a **tree** if it has no cycles. The edges are connected at the vertices ν_j which can be considered as equivalence classes of the edge end points $\{a_j\}$. Let $\{\gamma_1, \dots, \gamma_m\} = \partial\Gamma \subset V$ be the boundary vertices.

In this talk, **graph = (finite compact) metric tree graph**.

A Simple Tree Example

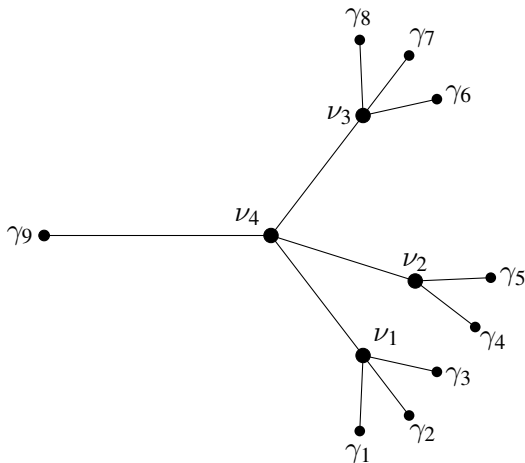


Fig. 1: A metric tree with $m = 9$ and $N = 12$

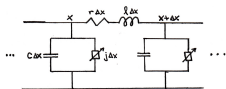
Quantum graph $\{\Gamma, H\}$: **differential operator** H on metric graph Γ , **coupled by specific vertex matching conditions**.

Applications include

- Oscillations of the flexible structures made of strings, beams, cables.
- hierarchical materials like ceramic or metallic foams, percolation networks and carbon and graphene nano-tubes.
- Our motivation comes from **neurobiology**; specifically dendritic trees of CNS neurons.

Control and inverse theories for PDEs on graphs constitute an important part of the rapidly developing area of applied mathematics — analysis on graphs.

Neuronal Cable Equation



Single dendritic branch \Rightarrow Circuit model \Rightarrow Cable model
 Non-dimensionalized neuronal cable equation has the form

$$v_t + \sum_{j=1}^k g_j(x)(v - E_j) = v_{xx}$$

Pick any $1 \leq i \leq k$, and let $u = v - E_i$, $E_{ji} = E_j - E_i$, and substitute into the equation:

$$u_t - u_{xx} + q(x)u = h(x)$$

where $q(x) = \sum_{j=1}^k g_j(x)$, and $h(x) = \sum_{j \neq i} E_{ji} g_j(x)$. The E_j s are assumed known, but the g_j s, hence $q(x)$ and $h(x)$, are assumed unknown a priori.

IBVP on a Graph

Consider a system described by neuronal cable theory on a graph (adding $p(t)$, with $p(0) \neq 0$):

$$\partial_t u_j - \partial_x^2 u_j + q_j(x)u_j = p(t)h_j(x) \quad \text{on } e_j \times (0, T) \quad \forall e_j \in E \quad (1)$$

or shortly, $\partial_t u - \partial_x^2 u + q(x)u = p(t)h(x) \quad \text{on } E \times (0, T)$

$$\begin{aligned} \mathbf{KN}: \quad & \sum_{e_j \sim \nu} \partial u_j(\nu, t) = 0 \text{ at each vertex } \nu \in V \setminus \partial\Gamma, \\ & u(\cdot, t) \text{ is continuous at each vertex, for } t \in [0, T] \end{aligned} \quad (2)$$

$$\partial u = f \quad \text{on } \partial\Gamma \times [0, T], \quad u|_{t=0} = 0 \quad \text{on } E. \quad (3)$$

In (2) (and below) $\partial u_j(\nu)$ denotes the derivative of u at the vertex ν taken along the edge e_j in the direction outwards the vertex. Also, $e_j \sim \nu$ means edge e_j is incident to vertex ν , and the sum is taken over all edges incident to ν .

IBVP and Inverse Problem

Let $\mathcal{H} = L^2(\Gamma)$ and $\mathcal{F}^T = L^2([0, T]; \mathbb{R}^m)$.

Theorem 1. If $f, p \in \mathcal{F}^T$, $h \in \mathcal{H}$, then for any $t \in [0, T]$, $u^f(\cdot, t) \in \mathcal{H}$ and $u^f \in C([0, T]; \mathcal{H})$.

For the inverse problem purposes we assume that $p \in H^1(0, T)$, $p(0) \neq 0$.

The **response operator**, $R^T : \mathcal{F}^T \rightarrow \mathcal{F}^T$, is defined as

$$(R^T f)(t) = u^f(\cdot, t)|_{\partial\Gamma}, \quad t \in [0, T].$$

Theorem 2. Operator R^T known for any $T > 0$ uniquely determines the graph topology, the lengths of the edges, the potentials q_j and sources h_j , $j = 1, \dots, N$.

Separating Problem into Two Auxiliary Problems

Solution of the problem (1)–(3) can be presented in the form

$u = y + z$:

$$\partial_t y - \partial_x^2 y + q(x)y = 0 \quad \text{on } E \times (0, T) \quad (4)$$

$$\partial y = f \quad \text{on } \partial\Gamma \times [0, T] \quad (5)$$

$$\partial_t z - \partial_x^2 z + q(x)z = p(t)h(x) \quad \text{on } E \times (0, T) \quad (6)$$

$$\partial z = 0 \quad \text{on } \partial\Gamma \times [0, T] \quad (7)$$

with the KN matching conditions and zero initial conditions.

We see that $z = u^0$ so, $y^f|_{\partial\Gamma} = u^f|_{\partial\Gamma} - u^0|_{\partial\Gamma} = R^T f - R^T 0$.

Solving Problem 1: Main Ideas

$$\mathcal{L}\phi \doteq -\phi'' + q(x)\phi = \lambda\phi \text{ on } \Gamma \setminus V, \text{ KN at } V \setminus \partial\Gamma, \phi'(\gamma_j, \lambda) = f_j, \\ f = \text{col}(f_1, \dots, f_m), \rightarrow \exists! \phi = \phi^f$$

1. BC method allows spectral data $\text{SD} = \{\lambda_n, \phi_n^f|_{\partial\Gamma}\}_{n \in \mathbb{N}}$ to be recovered from R^T

2. SD determines TW matrix function $M: \phi_{|\partial\Gamma}^f = M(\lambda)f$

3. TW matrix function M used in reduction method

Essence: recalculate M from original graph to smaller graph by 'pruning' the boundary edges. By iterative process we reduce original IP (Prob. 1) to IP for **companion wave problem** on single edge.

$$w_{tt} - w_{xx} + q(x)w = 0 \text{ on } e = (0, \ell), \text{ zero i.c.s,} \\ w_x(0, t) = f(t), w_x(\ell, t) = 0, \\ f \in \mathcal{H}^T \doteq \{f \in L^2(0, \ell) \mid \text{supp } f \subset [0, T]\}$$

Problem 1 continued

$$F(t) = \int_0^t f(s) ds \rightarrow$$

$$w^f(x, t) = -F(t - x) + \int_x^t \kappa(x, s) F(t - s) ds, \quad x < t, \quad w^f(x, t) = 0, \quad x \geq t$$

$$\kappa(x, t) = \text{sol. to a certain Goursat prob. on } 0 < x < t < T \leq \ell, \\ r(t) = \kappa(0, t)$$

$$\psi'' - q(x)\psi = 0, \quad \psi(0) = 1, \quad \psi'(0) = 0, \quad q \in L^1(0, \ell)$$

$$\exists f = f^T \in L^2(0, T) \text{ s.t. } w_t^{f^T}(x, T) = \psi(x), \quad x \leq T, \quad = 0, \quad x > T$$

$$f^T(t) - \frac{1}{2} \int_0^T \{r(|t - s|) + r(2T - t - s)\} f(s) ds = -1 \quad t \in [0, T]$$

$$w_t^{f^T}(T-, T) = -f^T(0+) = \psi(T) \in C^2 \Rightarrow q(T) = \frac{\psi''(T)}{\psi(T)}$$

See **Avdonin and Bell, J. Inv. Prob. and Imaging, 9(3)(2015)**

Solving Problem 2

$$z(x, t) = \sum_{n=1}^{\infty} \varphi_n(x) h_n \int_0^t p(t - \tau) e^{-\lambda_n \tau} d\tau,$$

where $h_n = \langle h, \varphi_n \rangle_{\mathcal{H}}$. Recalling $p(0) \neq 0$, $\mu(t) \doteq u^0(\cdot, t)|_{\partial\Gamma}$

$$\mu(t) = \int_0^t p(t - \tau) W(\tau) d\tau, \quad W(\tau) \doteq \sum_{n=1}^{\infty} h_n \varphi_n(0) e^{-\lambda_n \tau},$$

$$\mu'(t) = p(0) W(t) + \int_0^t p'(t - \tau) W(\tau) d\tau.$$

The last equation allows finding $W(t)$ as the solution to the VIE.

Problem 2 continued

Theorem 3 (controllability): implies that the family $\{\varphi_n(0)e^{-\lambda_n t}\}$ is minimal in $\mathcal{F}^T = L^2([0, T]; \mathbb{R}^m)$ (for any $T > 0$) with biorthogonal family $\{\theta_n(t)\}$.

Thus, with $W(\cdot)$ and SD determined, we then have the h_n s and h :

$$h_n = \langle W, \theta_n \rangle_{\mathcal{F}^T}, \quad h(x) = \sum_{n=1}^{\infty} h_n \varphi_n(x).$$

Recovery of original distributed parameters requires solving Problem 1 once, solving Problem 2 $k-1$ times.

See **Avdonin, Bell, and Nurtazina, Math. Meth. Appl. Sci. 40(2017)**

Scaled (linear) models:

- One unknown ion species: $(1 + q(x))u_t + q(x)u = u_{xx}$
 $(C_m = C_0 + C_1N(x), g = g_1N(x))$

- Unknown radius, known conductance:

$$u_t + q(x)u = \frac{1}{a(x)}(a(x)^2u_x)_x$$

First on single interval, then on a graph.

Next: Graphs with Cycles

cyclomatic number (first Betti number) = $|E| - |V| + 1 (> 0)$

Much more challenging extending control theory than for tree graphs

network.jpg

