

Finite element simulation of ionic electrodiffusion in cellular geometries

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simula



The emerging EMI framework use a geometrically explicit representation of the cellular domains

Find the intracellular and extracellular potentials

$\phi_i = \phi_i(\mathbf{x}, t)$ and $\phi_e = \phi_e(\mathbf{x}, t)$, and the transmembrane current $I_M = I_M(\mathbf{x}, t)$ s.t.:

$$-\nabla \cdot (\sigma_i \nabla \phi_i) = 0 \quad \text{in } \Omega_i, \quad (1)$$

$$-\nabla \cdot (\sigma_e \nabla \phi_e) = 0 \quad \text{in } \Omega_e, \quad (2)$$

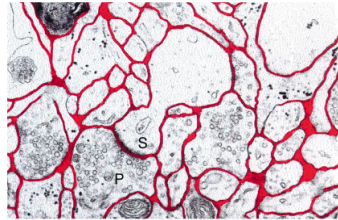
$$\phi_M = \phi_i - \phi_e \quad \text{at } \Gamma, \quad (3)$$

$$\sigma_e \nabla \phi_e \cdot \mathbf{n}_e = -\sigma_i \nabla \phi_i \cdot \mathbf{n}_i = I_M \quad \text{at } \Gamma, \quad (4)$$

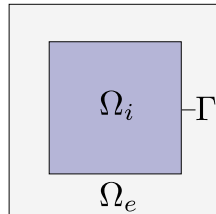
$$\frac{\partial \phi_M}{\partial t} = \frac{1}{C_M} (I_M - I_{\text{ion}}) \quad \text{at } \Gamma. \quad (5)$$

- Ion concentrations are assumed to be constant in space and time – often an accurate approximation, but not always ...

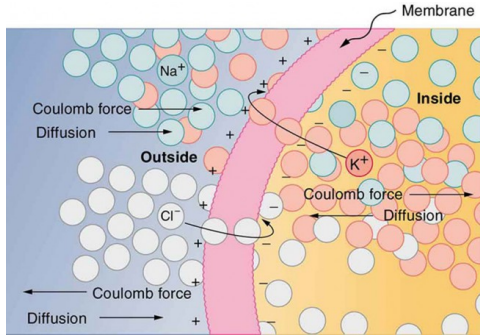
[Krassowska & Neu 1994], [Ying & Henriquez 2007],
[Tveito et al. 2017]



Rat cortex with ECS in red [Nicholson, 1998]

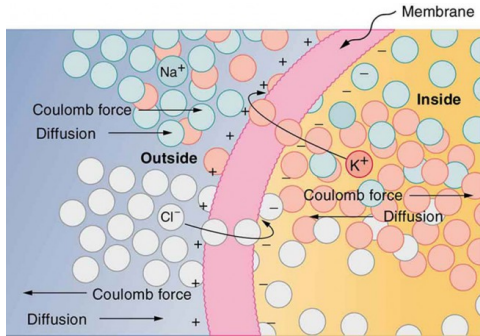


Movement of ions is fundamental in brain signalling and cellular swelling

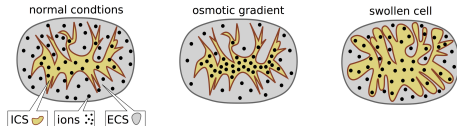


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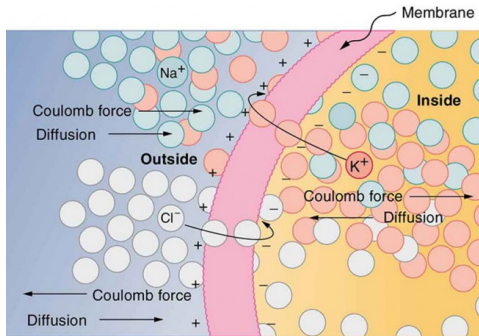


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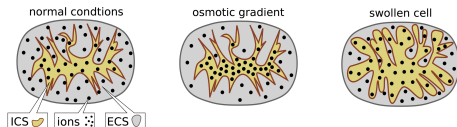


[Hubel, 2016]

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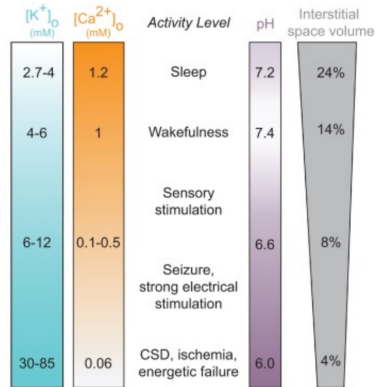


[courses.lumenlearning.com]



[Hubel, 2016]

The extracellular ion composition changes with local neuronal activity and across brain states



[Rasmussen, 2021]

A computational framework for ionic electrodiffusion in brain tissue with geometrical explicit representation of the cells

KNP-EMI

In a (tissue) domain $\Omega = \Omega_i \cup \Omega_e \subset \mathbb{R}^d$, where Ω_i (with boundary Γ) and Ω_e represent respectively intracellular and extracellular regions, with (ion) species $k \in K$ (e.g. Na^+ , K^+ , Cl^-).

For each compartment $r \in \{i, e\}$ and species k , $x \in \Omega_r$, $t > 0$, find the:

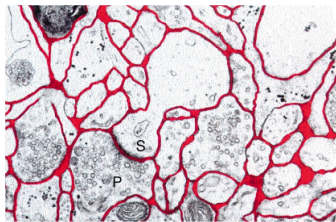
- concentrations $[k]_r(x, t)$,
- electrical potentials $\phi_r(x, t)$,

and at the interface, $x \in \Gamma$, $t > 0$, find the:

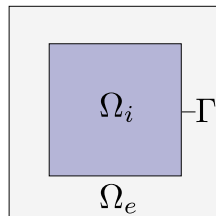
- transmembrane current $I_M(x, t)$.

[Mori & Peskin, 2009]

[Ellingsrud et al., 2020]



Rat cortex with ECS in red [Nicholson, 1998]



A computational framework for ionic electrodiffusion in brain tissue with explicit representation of the cells (KNP-EMI)

Conservation of ions for the bulk of each region:

$$\frac{\partial [k]_r}{\partial t} + \nabla \cdot J_r^k = 0, \quad \text{in } \Omega_r.$$

Ion flux densities are given by:

$$J_r^k = -D^k \nabla [k]_r - \frac{D^k z^k}{\psi} [k] \nabla \phi_r. \quad \text{in } \Omega_r,$$

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"KNP assumption" of bulk electroneutrality:

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[Pods, 2017]

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Interface conditions:

$$\phi_i - \phi_e = \phi_M, \quad \text{on } \Gamma,$$

$$I_M \equiv F \sum_k z^k J_i^k \cdot n_i = -F \sum_k z^k J_e^k \cdot n_e, \quad \text{on } \Gamma,$$

$$J_i^k \cdot n_i = \frac{I_{\text{ion}}^k + \alpha_i^k I_{\text{cap}}}{F z^k}, \quad \text{on } \Gamma,$$

$$-J_e^k \cdot n_e = \frac{I_{\text{ion}}^k + \alpha_e^k I_{\text{cap}}}{F z^k}, \quad \text{on } \Gamma.$$

[Pods, 2017]

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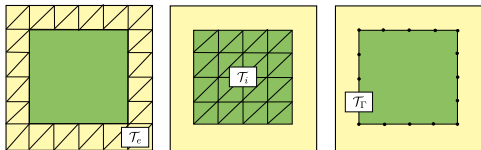
The strongly coupled and non-linear KNP-EMI equations are numerically and computationally challenging to solve

Numerical strategy:

- Split PDEs from ODEs (two-step first order)
- Finite difference ODE and PDE time discretizations (explicit handling of non-linear terms)
- Mortar finite element scheme for PDEs

$$[k]_{r,h}(t) \in V_{r,h}, \quad \phi_{r,h}(t) \in T_{r,h}, \quad l_{M,h}(t) \in S_h,$$

where $V_{r,h}$, $T_{r,h}$, S_h are constructed using continuous piecewise linear polynomials.



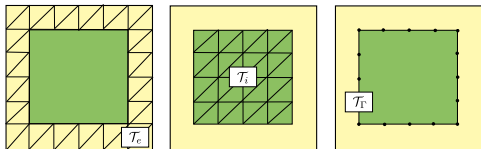
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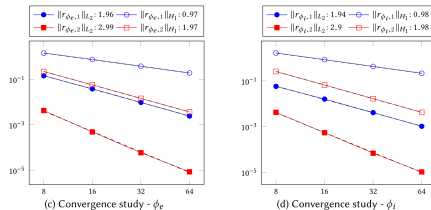
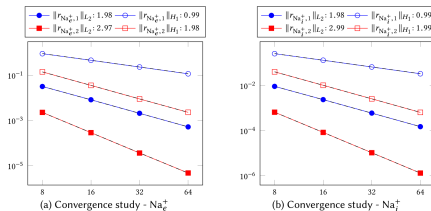
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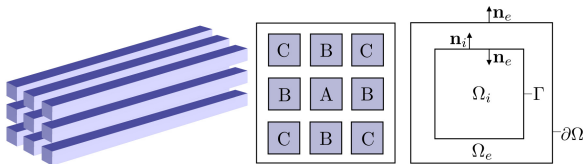
For a problem with a **smooth manufactured solution** we observe expected convergence rates:



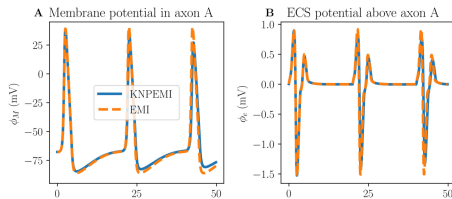
Comparing KNP-EMI and EMI: Do diffusive currents affect ephaptic coupling through the ECS in unmyelinated axon bundles?

In an **idealized axon bundle** with cell gaps of $0.1\mu\text{m}$, action potentials are induced (via a synaptic current) every 20 seconds in either:

- Axon A (1 active neighbour), or
- Axons B and C (8 active neighbour).



[Ellingsrud et al., 2020]



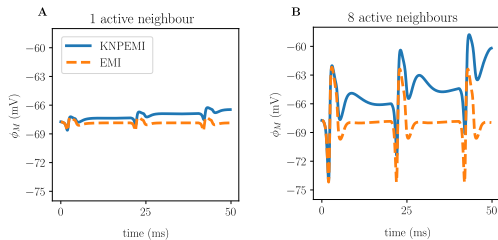
Diffusive currents contribute to ECS potential shifts in the KNP-EMI framework:

$$\text{EMI} \quad \nabla \cdot (\sigma_e \nabla \phi_e) = 0, \text{ in } \Omega_e,$$

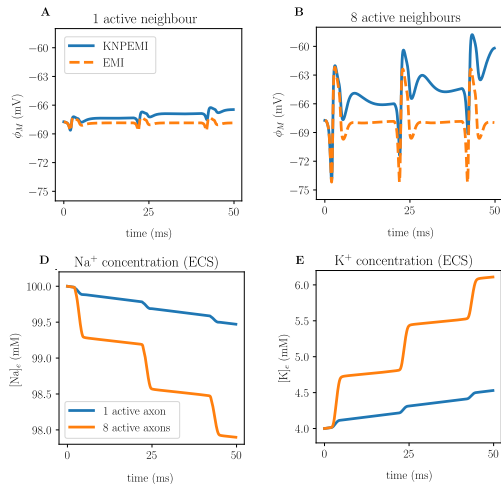
$$\text{KNP-EMI} \quad \nabla \cdot (\sigma_e \nabla \phi_e + \nabla b_e) = 0, \text{ in } \Omega_e,$$

where $b_e = F \sum_k z^k D_e^k [k]_e$.

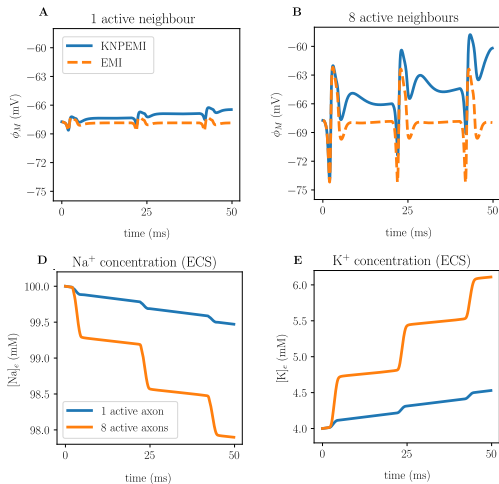
Diffusive currents do not strengthen the *electrical* ephaptic coupling (via the extracellular potential), however we see *diffusive* ephaptic coupling



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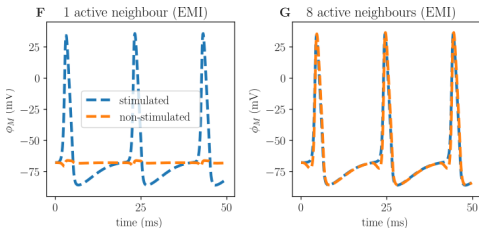


Ephaptic coupling is **inversely proportional** to the **extracellular conductivity**:

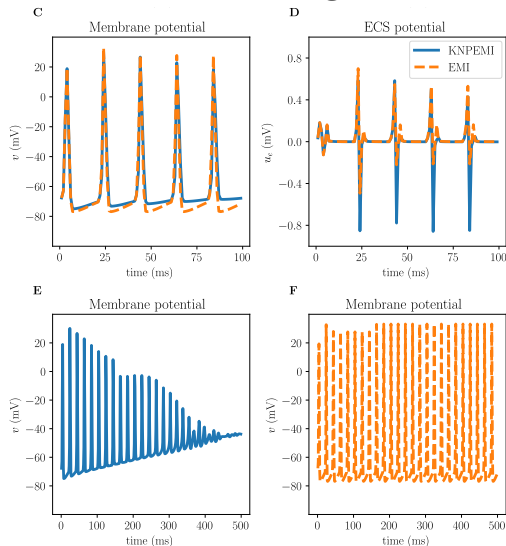
$$\sigma_i = \frac{F}{\psi} \sum_k D_i^k[k]_i (z^k)^2 = 2.01 \quad \sigma_i = 1.0, \text{ (S/m)}$$

$$\sigma_e = \frac{F}{\psi} \sum_k D_e^k[k]_e (z^k)^2 = 1.31 \quad \sigma_e = 0.1, \text{ (S/m)}$$

[Bokil et al., 2001]



During hyperactivity, the KNP-EMI and EMI models differ due to shifts in the ion concentration gradients

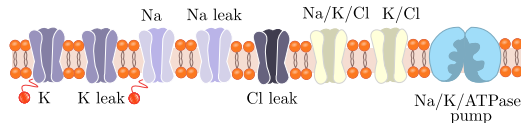


Standard Hodgkin-Huxley model and homeostasis mechanisms:

$$I_{ion}^{Na} = I_{leak}^{Na} + I^{Na} + 3I_{ATP} + I_{NKCC1}$$

$$I_{ion}^{K} = I_{leak}^{K} + I^K - 2I_{ATP} + I_{NKCC1} + I_{KCC2}$$

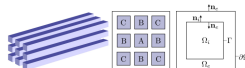
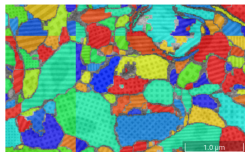
$$I_{ion}^{Cl} = I_{leak}^{Cl} - 2I_{NKCC1} - I_{KCC2},$$



We consider two firing regimes:

- Normal activity (1 Hz, not shown)
- Hyperactivity (50 Hz)

Exciting time: Extreme modelling of excitable tissue



Ambition

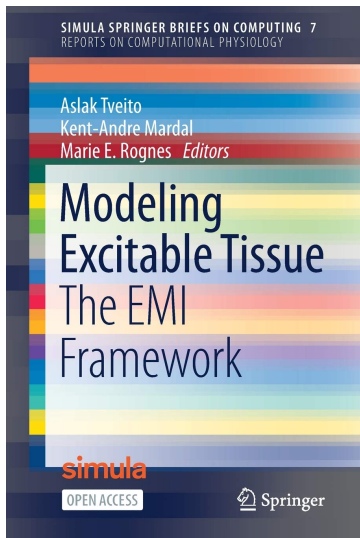
To establish mathematical and technological foundations for modelling and simulation of **electrical, chemical and mechanical interplay between brain cells at unprecedented detail**, allowing for pioneering in-silico studies of brain signalling, volume balance and clearance.

Topics and expected outcomes

- ◇ Well-posed general mathematical and numerical framework allowing for geometrically-explicit representations of moving excitable cells;
- ◇ New computational geometries and models, highly scalable algorithms, and solution software for high-resolution high-realism simulations of excitable cell ensembles – all distributed as open source;
- ◇ New physiological insight into inter-neuronal and astrocyte membrane mechanisms and their role in brain homeostasis and learning.

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1	Derivation of a Cell-Based Mathematical Model of Excitable Cells . . .	1
	Karoline Horgmo Jæger and Aslak Tveito	
2	A Cell-Based Model for Ionic Electrodiffusion in Excitable Tissue . .	14
	Ada J. Ellingsrud, Cécile Daversin-Catty and Marie E. Rognes	
3	Modeling Cardiac Mechanics on a Sub-Cellular Scale	28
	Åshild Telle, Samuel T. Wall and Joakim Sundnes	
4	Operator Splitting and Finite Difference Schemes for Solving the EMI Model	44
	Karoline Horgmo Jæger, Kristian Gregorius Hustad, Xing Cai and Aslak Tveito	
5	Solving the EMI Equations using Finite Element Methods	56
	Miroslav Kuchta, Kent-André Mardal and Marie E. Rognes	
6	Iterative Solvers for EMI Models	70
	Miroslav Kuchta and Kent-André Mardal	
7	Improving Neural Simulations with the EMI Model	87
	Alessio Paolo Buccino, Miroslav Kuchta, Jakob Schreiner, and Kent-André Mardal	



Marie E. Rognes



Geir Halmes



Gaute Einevoll



Andreas Solbrå



Finite Element Simulation of Ionic Electrodiffusion in Cellular Geometries

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