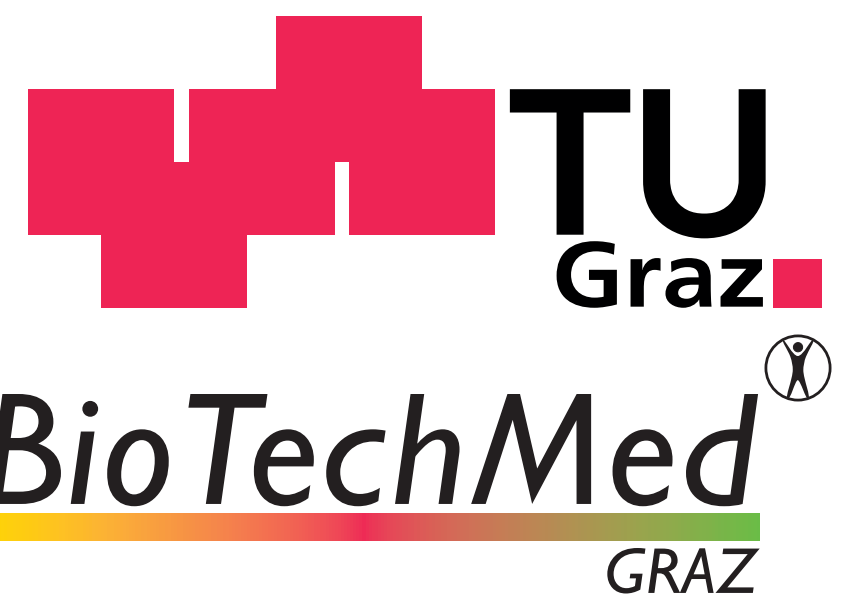


REAL-TIME INTERACTION WITH AN ELECTROPHYSIOLOGICAL CANCER CELL MODEL

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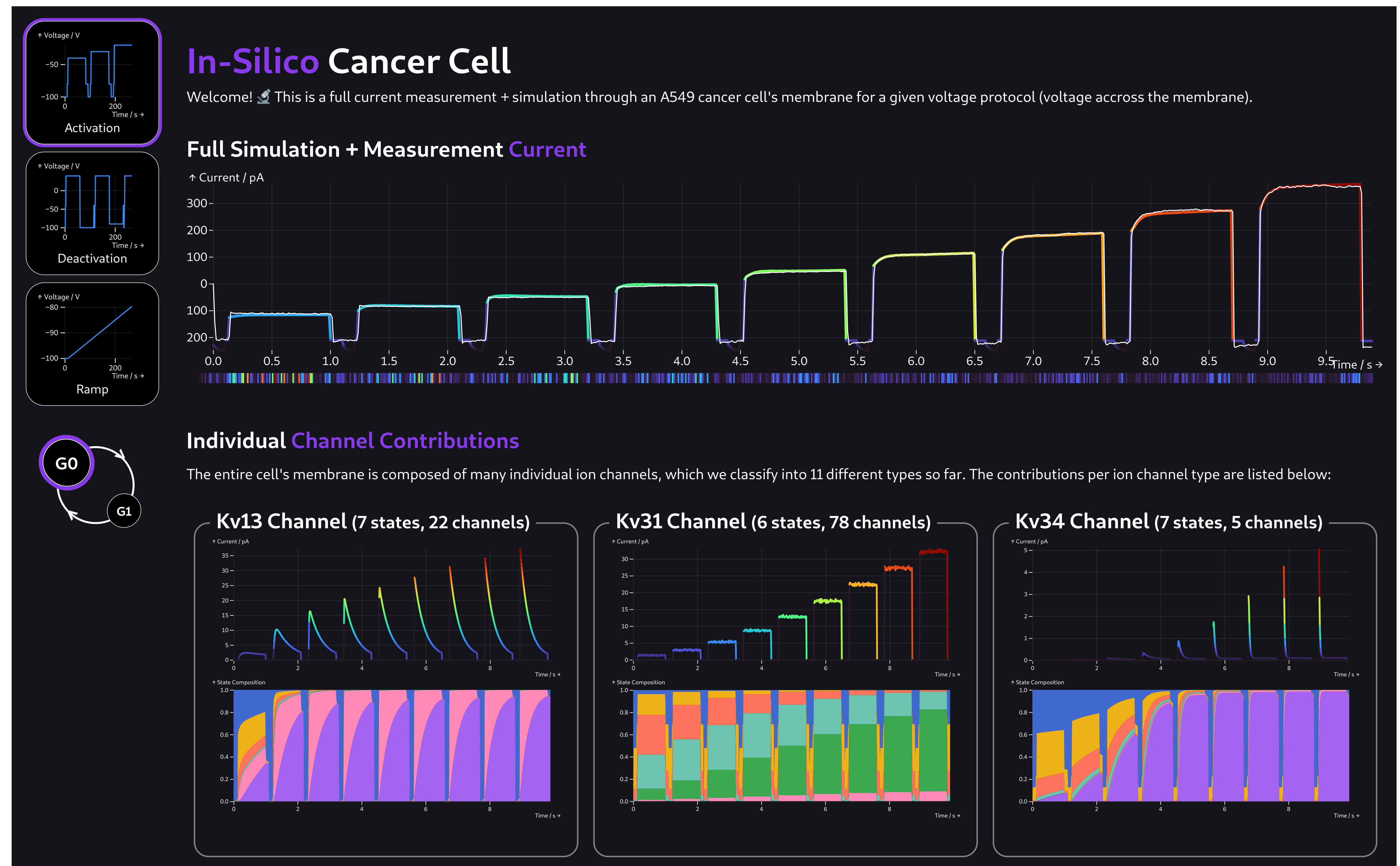
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Introduction

We improve on the A549 electrophysiological cancer cell model introduced in [1, 2], combining numerical methods with an efficient implementation to reduce simulation time to a level where it is feasible for live interaction. More specifically, we were able to accelerate the simulation with adaptive timestepping and a highly efficient implementation in the Rust programming language, while we also managed to approach the corresponding inverse problem using a quadratic program, solving it within milliseconds. We introduce a visualisation approach of the entire model in the form of a live simulation dashboard available online, running directly in the browser. The entire source code is freely available on GitHub and reusable through three different channels: the simulation interface (powered by compilation to WebAssembly), the Rust linkable library implementation and a Python package.

Live, In-Browser Cell Simulation Interface



Available live on in-silico.hce.tugraz.at.

Model

The whole cell current $I : T \rightarrow \mathbb{R}$ over time $t \in T \subset \mathbb{R}^+$ is the sum of all individual channel contributions $I_k, k \in \{1, \dots, M\}$ over $M \in \mathbb{N}$ channel types

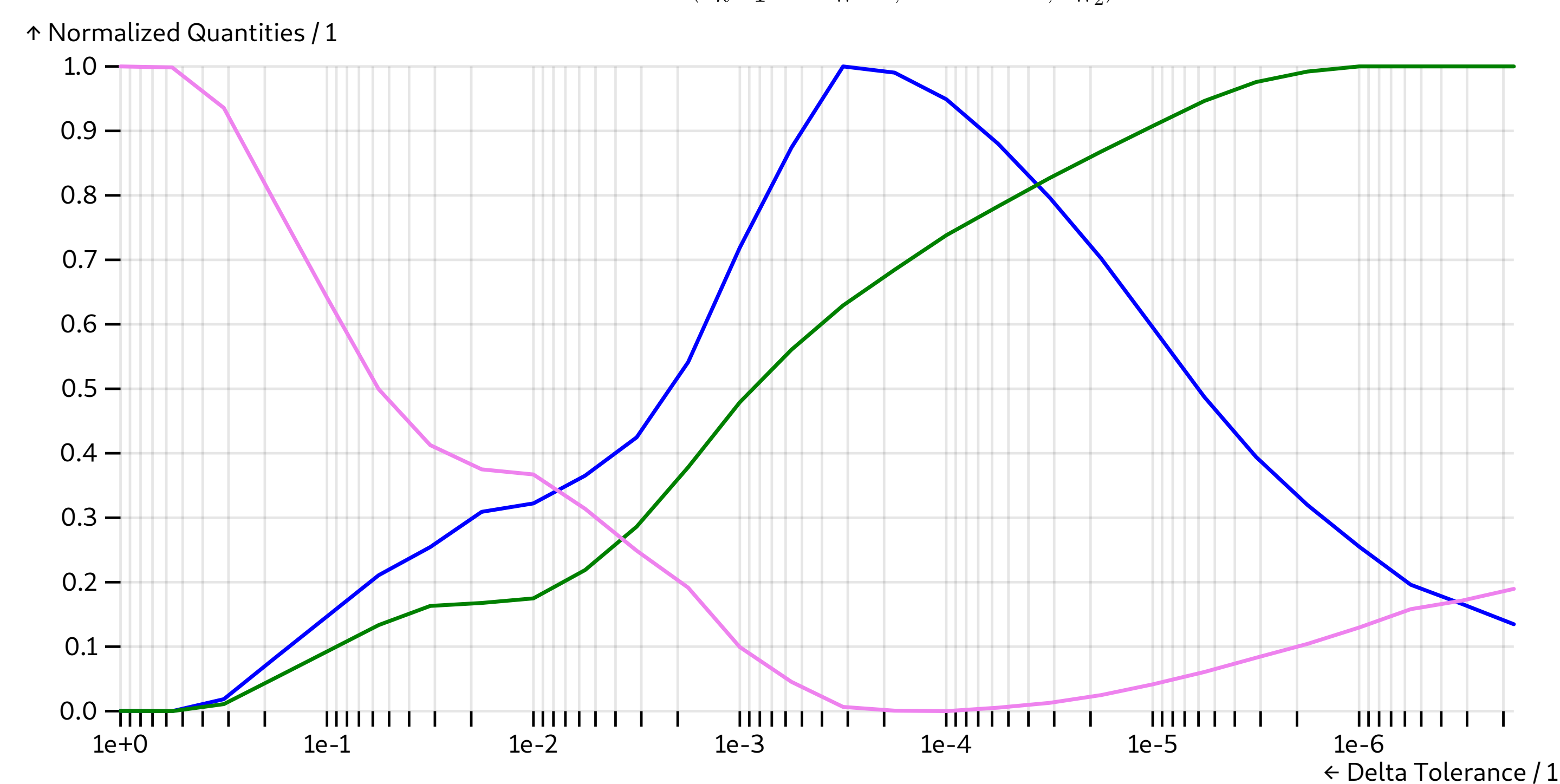
$$I(t) := \sum_{k=1}^M N_k I_k(t) = \sum_{k=1}^M N_k g_{k,p_o,k} (V(t) - E_k), \quad (1)$$

At each time step,

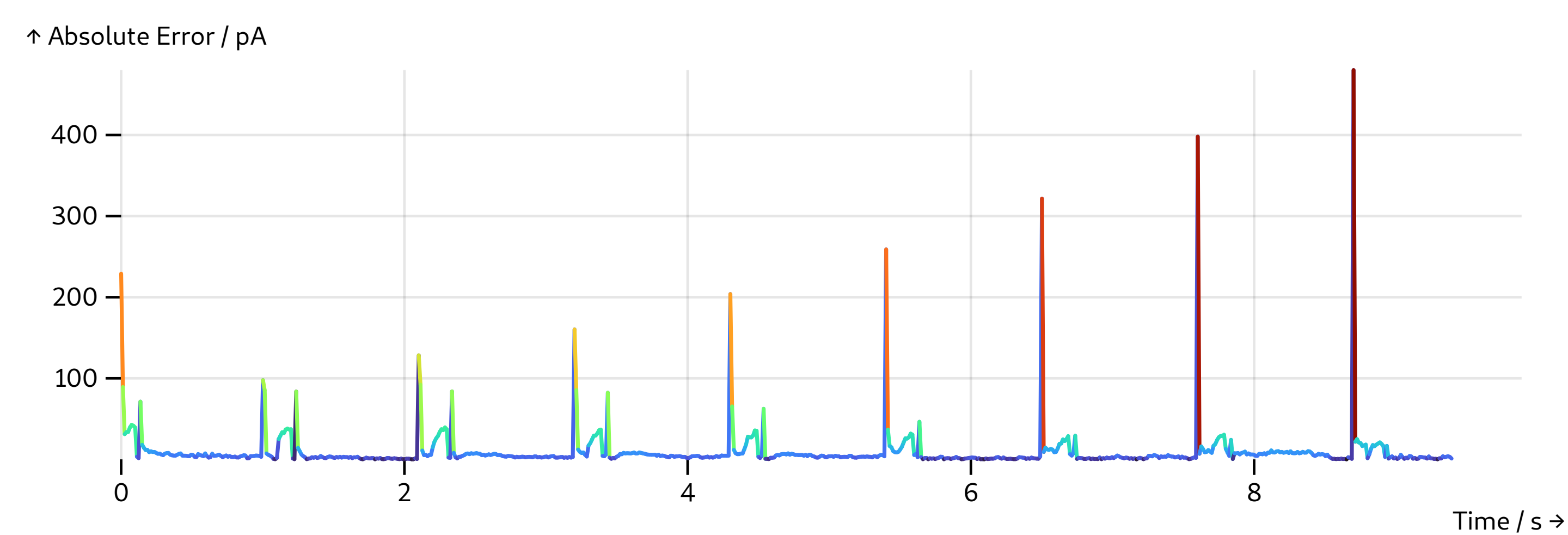
$$\mathbf{s}_{k,n+1} = H_k(V(t_n), \mathbf{C}(t_n), t_n) \mathbf{s}_{k,n}. \quad (2)$$

Adaptive Timestepping

$$(\Delta t)_{n+1} = (\Delta t)_n \left(\frac{\Delta^{\text{tol}}}{\sum_{k=1}^M N_k \|\mathbf{s}_{k,n+1} - \mathbf{s}_{k,n}\|_2} \right)^{1/2}, \quad (3)$$



Pointwise Error between Simulation and Measurements



Conclusion

simply run: `pip install in-silico-cancer-cell`

References

Formulation as a Quadratic Program

We want to find

$$N_{\text{opt}} = \arg \min_{N \in \mathbb{N}_0^M} \frac{1}{2} \|RN - \mathbf{I}_{\text{meas}}\|_2^2, \quad (4)$$

with $\mathbf{I}_{\text{meas}} \in \mathbb{R}^{N_t}$ the experimentally measured current and $R \in \mathbb{R}^{N_t \times M}$ the matrix of all currents I_k per channel type. Letting $\mathbf{d} := \mathbf{I}_{\text{meas}}$ for brevity,

$$N_{\text{opt}} \approx \arg \min_{\mathbf{x} \in \mathbb{R}_+^M} f(\mathbf{x}) = \arg \min_{\mathbf{x} \in \mathbb{R}_+^M} \frac{1}{2} \|R\mathbf{x} - \mathbf{d}\|_2^2,$$

with cost function $f : \mathbb{R}^M \rightarrow \mathbb{R}^+$, which we manipulate to

$$\begin{aligned} f(\mathbf{x}) &= \frac{1}{2} (R\mathbf{x} - \mathbf{d})^T (R\mathbf{x} - \mathbf{d}) \\ &= \frac{1}{2} (\mathbf{x}^T R^T R \mathbf{x} - \mathbf{x}^T R^T \mathbf{d} - \mathbf{d}^T R \mathbf{x} + \mathbf{d}^T \mathbf{d}) \\ &= \frac{1}{2} (\mathbf{x}^T P \mathbf{x} + \mathbf{x}^T \mathbf{q} + \mathbf{q}^T \mathbf{x}) + \mathcal{O}(1) \\ &= \frac{1}{2} \mathbf{x}^T P \mathbf{x} + \mathbf{q}^T \mathbf{x} + \mathcal{O}(1) \end{aligned}$$

where we let $P := R^T R \in \mathbb{R}^{M \times M}$ and $\mathbf{q} := -R^T \mathbf{d} \in \mathbb{R}^M$ and leave out the constant $\mathbf{d}^T \mathbf{d}$ as $\mathcal{O}(1)$. We can express the nonnegativity constraint $\mathbf{x} \geq 0$ as an equality constraint using a slack variable $\mathbf{s} \in \mathbb{R}_+^M$,

$$-\mathbf{x} + \mathbf{s} = \mathbf{0} \quad \Leftrightarrow \quad A\mathbf{x} + \mathbf{s} = \mathbf{b},$$

where we set $A := -\mathbb{1} \in \mathbb{R}^{M \times M}$ and $\mathbf{b} := \mathbf{0} \in \mathbb{R}^M$. This leaves us with a constrained quadratic program,

$$\min_{\mathbf{x} \in \mathbb{R}_+^M} \frac{1}{2} \mathbf{x}^T P \mathbf{x} + \mathbf{q}^T \mathbf{x}, \quad (5)$$

$$s.t. \quad A\mathbf{x} + \mathbf{s} = \mathbf{b}, \quad \mathbf{s} \in \mathbb{R}_+^M. \quad (6)$$

The integer solution can then be obtained from rounding,

$$N_{\text{opt}} = \lfloor \mathbf{x} \rfloor \in \mathbb{N}_0^M.$$

Comparison of Optimization Methods

Algorithm	Abbreviation	Runtime / ms	RMSE / pA
Particle Swarm Optimization	PSO	22571	27.69
Gradient Descent + More Thuente	GD	18924	32.34
Limited-Memory BFGS + Hager Zhang	LBFGS	4845	32.20
Non-Negative Least Squares [3]	NNLS	318	28.00
Quadratic Program	QP	18	28.13