

Xolotl: An Intuitive and Comprehensible Neuronal Simulator

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2 ABSTRACT

`xolotl` is an open-source neuronal simulator written in C++ with MATLAB wrappers. Complex models and networks can be designed efficiently using an intuitive language tightly coupled to the object-based architecture of the underlying C++ code. Models can be specified by adding conductances to compartment objects. The structure is modular, serialized, and searchable, permitting high-level programmatic control over nearly all features of the models. C++ templates are provided for developing new conductances, compartments, and integration schemata. It also includes a customizable graphical user interface ('puppeteer') for rapid prototyping and hand-tuning conductances in real-time. The modular structure and accessibility to all parameters, variables, and dynamics of the model network in MATLAB facilitate rapid construction and assessment of model networks. `xolotl` is freely available at <https://github.com/marderlab/xolotl>. This tool provides straightforward implementation and fast simulation of neuronal models while permitting full control over every aspect of the network and integration.

Keywords: simulator, MATLAB, C++, `xolotl`, conductance-based, computational, keyword, keyword

1 INTRODUCTION

`xolotl` (<https://github.com/sg-s/xolotl>) is a fast single-compartment and multi-compartment simulator in C++ with MATLAB wrappers (<https://www.mathworks.com/products/matlab.html>). Written with an emphasis on flexibility and speed, `xolotl` can simulate single-compartment conductance-based models, networks of these, and detailed multi-compartment models. `xolotl` exploits a novel automatic type system, `cpplab`, which binds MATLAB code to C++ header files, creating objects and classes *ad libitum* in MATLAB which reflect the underlying object-oriented code. `xolotl` implements `cpplab` to represent the nested structure of conductance-based models, and exploits the computational efficiency of the low-level programming language to quickly integrate models. For this reason, models can be implemented entirely in MATLAB with few lines of code.

Models are specified in MATLAB by a `xolotl` object which contains compartment objects which themselves contain conductances. Synapses belong to the `xolotl` object and connect compartments together. The high-level specification supports arbitrarily large network and multi-compartment morphologies.

`xolotl` provides parameter optimization capabilities through the algorithm-agnostic `procrustes` toolbox. Any network parameters accessible through the `xolotl` structure can be optimized using arbitrary algorithms and objective functions on multi-core computers and high-performance computing clusters.

The software has been implemented in MATLAB due to its ease-of-use and popularity among neuroscientists. `cpplab` provides a powerful backend for specifying and integrating models without relying on the significantly slower and limiting MATLAB `codegen`. Minimal experience with MATLAB is required to use `xolotl`, and all equations and integration methods are provided transparently to the end user. No string parsing of equations is required.

`xolotl` comes packaged with visualization functions, a graphical user interface for real-time manipulation of model parameters. Plotting of voltage, intracellular calcium, conductance gating functions, and time constants is provided by built-in `xolotl` methods. The `puppeteer` toolbox permits real-time tuning of any network parameters using numerical sliders in a graphical interface which displays the resultant membrane potential and intracellular calcium traces. The ease-of-use of these tools lends them to pedagogical applications and rapid exploration of toy models. This tool aims to simplify the investigation of dynamics of complex neural network models, facilitate collaborative modeling, and complement other tools being developed in the neuroinformatics community.

2 ARTICLE TYPES

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3 WORKED EXAMPLES

Using `xolotl` in MATLAB, users create a `xolotl` object and populate it with compartments, synapses, and controllers. Each field is a `cpplab` object constructed by a function call to `add`. The model is integrated with the `integrate` function where the membrane potential, intracellular calcium concentration, controller states, intrinsic currents, and synaptic currents can be outputs.

`xolotl` comes packaged with a library of pre-existing conductance and synapse objects which greatly simplify the task of constructing model neurons. These objects can be referenced by name and added directly to a compartment. Novel conductance dynamics can be easily written by modifying a template header file contained in the `xolotl` distribution.

3.1 Simulating a Hodgkin-Huxley Model

The seminal Hodgkin-Huxley model contains a fast inactivating sodium conductance which promotes spiking, a non-inactivating potassium delayed rectifier, and a passive leak current (1A). A compartment named HH with compartment properties of membrane capacitance $C_m = 10 \mu\text{F}/\text{mm}^2$ and surface area $A = 0.1 \text{ mm}^2$ can be specified by 1B. Compartment, conductance, synapse, and controller properties can be specified during the call to the `add` function, or after construction, using dot-notation in MATLAB (e.g.

62 `x.HH.Cm`). 1C shows the MATLAB command prompt after invoking the `xolotl` object `x`, displaying the
63 hierarchical structure inherent in conductance-based treatments of neurodynamics.

64 This model was constructed using conductances from Liu *et al* 1998. In the absence of applied positive
65 current, the model is quiescent and tonically spikes under 0.2 nA of applied current (1D). The `integrate`
66 function takes the applied current as an argument (e.g. `x.integrate(AppliedCurrent)`), so that the
67 `xolotl` object is agnostic to integration-specific perturbations. The `plot` function generates voltage and
68 intracellular calcium traces, where the voltage trace is colored by the dominant current. If the membrane
69 potential is increasing, the strongest instantaneous inward current colors the trace. Conversely, if the
70 membrane potential is decreasing, the strongest outward current colors the trace instead. 1F-I display the
71 results of the `show` function. Activation and inactivation steady-states and the voltage-dependent time
72 constants of these gating variables describe the conductance dynamics in absence of other channel types.

73 3.2 Performing a Voltage Clamp Experiment *in-silico*

74 `xolotl` can recapitulate the results of voltage clamp experiments. 2 displays the procedure to clamp the
75 membrane potential of a cell with a delayed rectifier potassium conductance. The second argument of the
76 `integrate` function determines the clamped voltage (e.g. `x.integrate([], VoltageClamp)`).
77 Isolated currents under voltage clamp approach the steady-state (2D-E) so that a current-voltage relation
78 at steady-state can be extracted (2F). The derivative of the IV curve is the steady-state conductance (2G).
79 Fitting a sigmoid to various powers yields a model for the current dynamics (2H-I). These figures describe
80 graphically the theoretical underpinnings of current analysis through voltage clamp and can serve as an
81 effective pedagogical tool for computational and quantitative neuroscience.

82 3.3 Simulating Network Models

83 Network models in `xolotl` consist of compartment objects connected by synapses. Synapses are
84 stored in a vector array as a field of the `xolotl` object in MATLAB. Presynaptic and postsynaptic labels
85 indicate the connectivity of the synapse. Figure 3 implements a model of the triphasic pyloric rhythm in
86 the stomatogastric ganglion of crustaceans. The pyloric model contains three compartments and seven
87 synapses (3A). This structure is reciprocated in the hierarchy of the `xolotl` object, where conductances
88 are contained within compartments (3B). Synapses are upper-level properties of the network which point
89 between two compartments (3C). This exploits vectorized operations in MATLAB and does not require
90 each synapse to possess a unique name. The `plot` function generates multiple subplots when called for a
91 network with multiple compartments (3D-F).

92 3.4 Simulating Integral Control

93 `xolotl` can implement homeostatic tuning rules as integral control. Figure 4 depicts generation of a
94 bursting neuron model from quiescent conditions. Calcium sensors supervene on maximal conductance
95 density (4). In `xolotl`, integral controllers are properties of conductances (4B-C). They modify properties
96 of the conductance in response to an error signal. In a demonstration adapted from O'Leary *et al.* 2013,
97 integral control changes maximal conductances to bring a neuron from quiescence into a bursting regime.
98 Maximal conductances increase from random initial conditions to a set which elicits the desired network
99 output by minimizing the error signal (4D-F).

4 MANUSCRIPT FORMATTING

4.1 Heading Levels

4.2 Level 2

4.2.1 Level 3

4.2.1.1 Level 4

4.2.1.1.1 Level 5

4.3 Equations

Equations should be inserted in editable format from the equation editor.

$$\sum x + y = Z \quad (1)$$

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CONFLICT OF INTEREST STATEMENT

The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

AUTHOR CONTRIBUTIONS

AH and SG-S wrote the manuscript and the code. EM provided funding and moral support.

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SUPPLEMENTAL DATA

- 131 Supplementary Material should be uploaded separately on submission, if there are Supplementary Figures,
132 please include the caption in the same file as the figure. LaTeX Supplementary Material templates can be
133 found in the Frontiers LaTeX folder.

DATA AVAILABILITY STATEMENT

- 134 The code to generate all figures can be found in the `xolotl` repository (<https://github.com/sg-s/xolotl>).
135

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FIGURE CAPTIONS

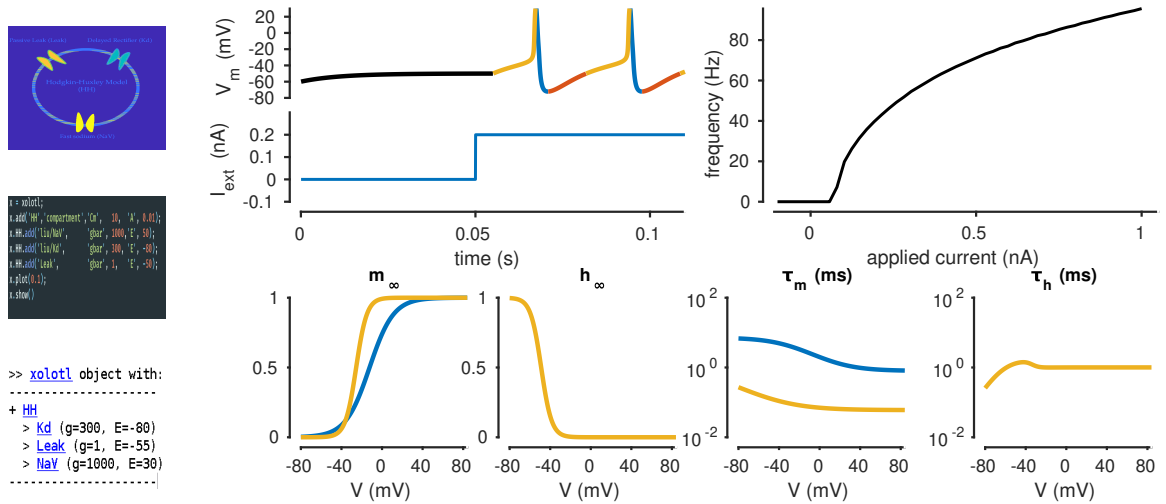


Figure 1. *xolotl* can quickly set up and simulate conductance-based models. (A) Cartoon of a Hodgkin-Huxley single-compartment neuron model with fast sodium, delayed rectifier, and leak currents. (B) Code snippet in MATLAB used to implement D, F-I. (C) *xolotl* schematic displayed in the MATLAB command prompt. (D) Simulated voltage trace with 0.1 nA applied current. Colors indicate the dominant current (gold is fast sodium, blue is delayed rectifier, red is leak). (E) Frequency-input relation displaying firing rate as a function of applied current. (F-G) Steady-state gating functions for activation (m) and inactivation (h) gating variables. Variables not plotted are unity for all voltage. (H-I) Voltage-dependence of time constants for activation (m) and inactivation (h) gating variables. Variables not plotted are unity for all voltage. Colors indicate conductance type (gold is fast sodium, blue is delayed rectifier, red is leak).

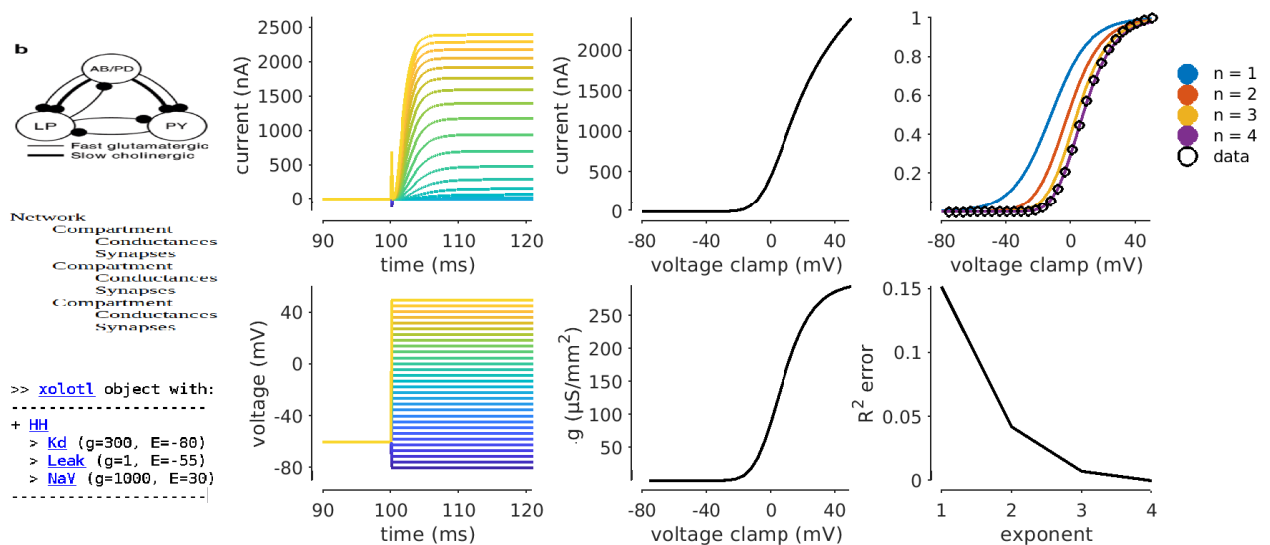


Figure 2. *xolotl* readily implements voltage clamp. (A) Cartoon of a cell with potassium conductance with experimentally-fixed voltage. (B) Structure of *xolotl* object in A. (C) Code snippet depicting integration under voltage clamp. (D-E) Current response to steps in voltage from a holding potential of $V_m = -60$ mV. (F) Current-voltage relation of the steady-state current ($t = 400$ ms) indicating a reversal potential of $E = -80$ mV and no inactivation. (G) Conductance-voltage relation at steady-state takes the form of a sigmoid. (H) Sigmoids m fit to the model as m^n data indicating that $n = 4$ is the best fit. (I) R^2 correlation of the sigmoid fits at various powers where $n = 4$ is an exact fit.

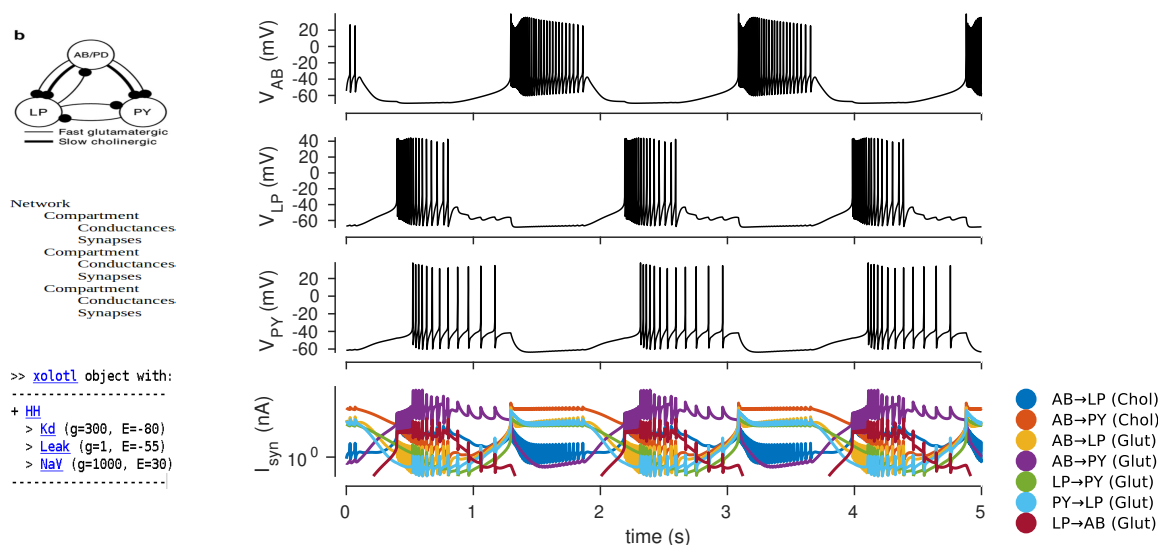


Figure 3. *xolotl* readily implements conductance-based network models. (A) Diagram of a network model of the pyloric rhythm in the crustacean stomatogastric ganglion (Prinz *et al.* 2004). (B) Hierarchical structure of a neuronal network considers compartments as components of the network and conductances and synapses as components of compartments. (C) *xolotl* implements conductances as fields of compartments and synapses as connections between compartments. (D-F) Simulated voltage trace of a model network for the three compartments. (G) Time series of synaptic currents in the simulated network with model cholinergic (chol) and glutamatergic (glut) synapses are outputs of the integration.

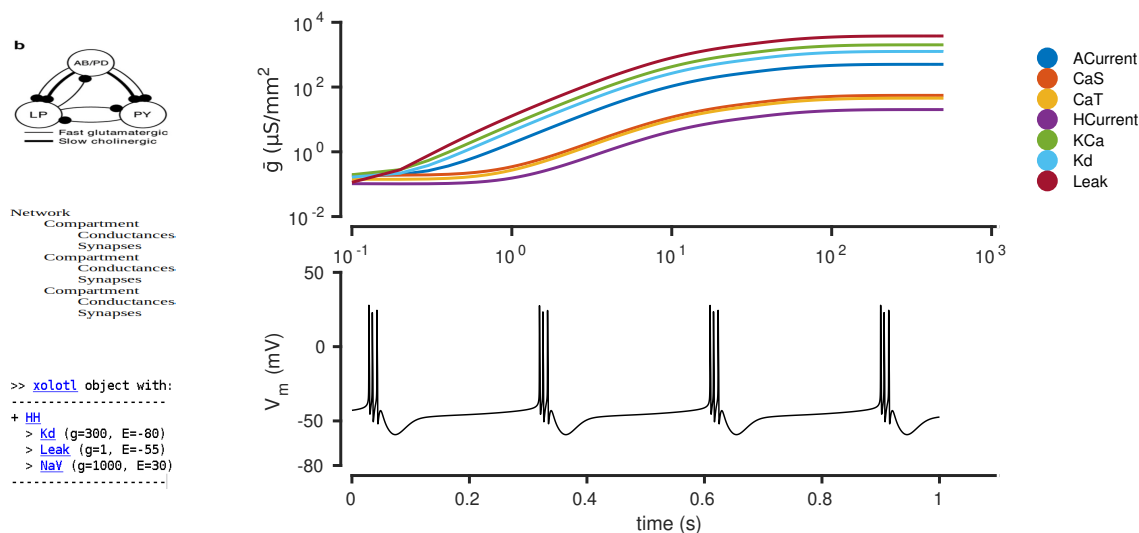


Figure 4. *xolotl* can implement homeostatic tuning rules as integral control. (A) Cartoon of a model neuron with integral control. (B) Hierarchical structure of a neuronal network considers controllers as components of compartments which act on conductances. (C) *xolotl* implements controllers XYZ. (D) Calcium sensors change maximal conductances to move a neuron from quiescence to a bursting state. (E) Voltage trace of the controlled neuron shows regular bursting activity.