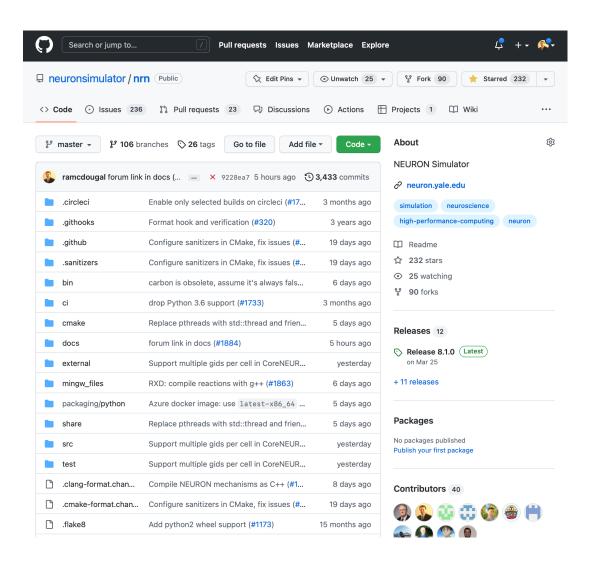
# Other resources

- Setup
  - git init
- Stage new/modified files for next commit:
  - git add FILENAME
- See what has changed
  - git diff
- Commit a version (so you can return to it later); you will be prompted to enter a commit message:
  - git commit
- Push to an external repository:
  - git push
- Checkout a version of a file from 2 commits ago:
  - git checkout HEAD~2 FILENAME





About -

Resources \*

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Register

**Portal** 

## DO YOU NEED...

- ✓ Free and open access to Supercomputers, High-Performance, High-Throughput & cloud computing for research?
- ✓ Access to popular computational neuroscience tools and data processing software?
- ✓ Web portal and programmatic access to supercomputing resources?
- ✓ A place to disseminate your research software and tools to the neuroscience community?



NSG is an easy to use interface for neuroscience researchers to access HPC, HTC and cloud resources for simulation and data processing.

Contact us at nsghelp@sdsc.edu

NSG facilitates access and use of High Performance Computing, High Throughput Computing, and academic cloud computing resources by neuroscientists for applications that include computational modeling, data processing, and development and dissemination of neuroscience software. Software currently available via NSG includes NEURON, NetPyNE, BMTK, NEST, BRIAN 2, BluePyOpt, MATLAB, EEGLAB, FREESURFER, PyTorch, TensorFlow, DEAP; this list is updated based on user needs.

http://www.nsgportal.org

LFPy

Module LFPy

Cell classes

Point processes

**Networks** 

Forward models

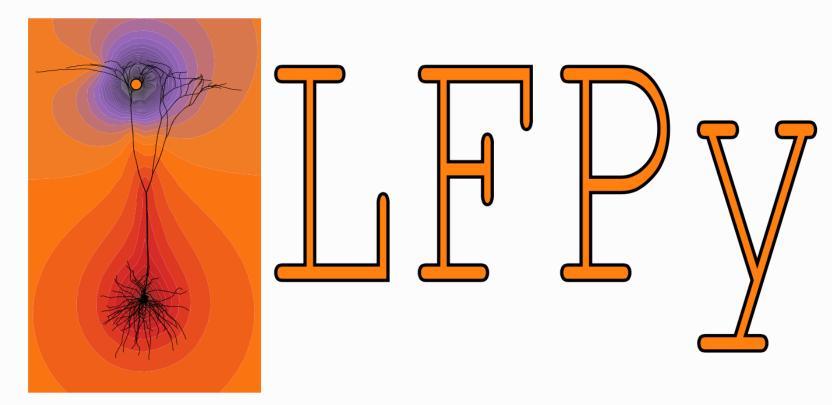
**Current Dipole Moment forward models** 

**Current Source Density (CSD)** 

Misc.



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## Welcome to LFPy's documentation!

(Looking for the old LFPy v1.\* documentation? Follow link)

## **Contents**





## BluePyOpt: Leveraging Open Source Software and Cloud Infrastructure to **Optimise Model Parameters in Neuroscience**

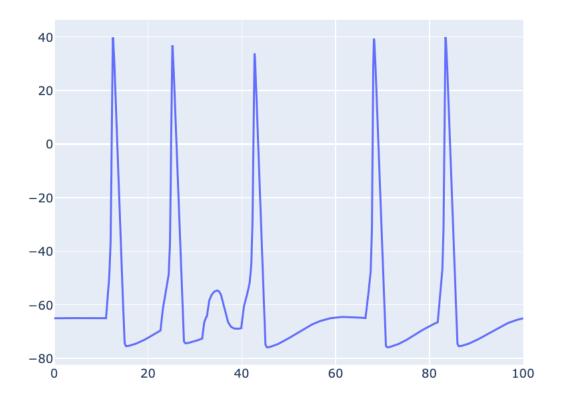
Werner Van Geit 1\*, Michael Gevaert 1, Giuseppe Chindemi 1, Christian Rössert 1, Jean-Denis Courcol<sup>1</sup>, Eilif B. Muller<sup>1</sup>, Felix Schürmann<sup>1</sup>, Idan Segev<sup>2,3</sup> and Henry Markram 1,4\*

<sup>1</sup> Blue Brain Project, École Polytechnique Fédérale de Lausanne, Geneva, Switzerland, <sup>2</sup> Department of Neurobiology, Alexander Silberman Institute of Life Sciences, The Hebrew University of Jerusalem, Jerusalem, Israel, 3 The Edmond and Lily Safra Centre for Brain Sciences, The Hebrew University of Jerusalem, Jerusalem, Israel, <sup>4</sup> Laboratory of Neural Microcircuitry, Brain Mind Institute, École Polytechnique Fédérale de Lausanne, Lausanne, Switzerland

**OPEN ACCESS** 

At many scales in neuroscience, appropriate mathematical models take the form of complex dynamical systems. Parameterizing such models to conform to the multitude of available experimental constraints is a global non-linear optimisation problem with a

# pip install bluepyopt



## eFEL:

**Electrophys Feature Extraction Library** 

## features:



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**Tutorials** 

Function Reference by Module Contributing to Elephant

Release Notes

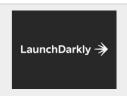
Acknowledgments

Authors and contributors

Citing Elephant

### Quick search

Go



**LaunchDarkly:** Change how you deliver software with the #1 platform for managing feature flags

# Elephant - Electrophysiology Analysis Toolkit

*Elephant* (Electrophysiology Analysis Toolkit) is an emerging open-source, community centered library for the analysis of electrophysiological data in the Python programming language.

The focus of Elephant is on generic analysis functions for spike train data and time series recordings from electrodes, such as the local field potentials (LFP) or intracellular voltages. In addition to providing a common platform for analysis codes from different laboratories, the Elephant project aims to provide a consistent and homogeneous analysis framework that is built on a modular foundation. Elephant is the direct successor to Neurotools and maintains ties to complementary projects such as ephyviewer and neurotic for raw data visualization.

The input-output data format is either <u>Neo</u>, <u>Quantity</u> or <u>Numpy</u> array. Quantity is a Numpy-wrapper package for handling physical quantities like seconds, milliseconds, Hz, volts, etc. Quantity is used in both Neo and Elephant.

### Visualization of Elephant analysis objects

<u>Viziphant</u> package is developed by Elephant team and provides a high-level API to easily generate plots and interactive visualizations of neuroscientific data and analysis results. The API uses and extends the same structure as in Elephant to ensure intuitive usage for scientists that are used to Elephant.

### Table of Contents

- Installation
- Tutorials
- Function Reference by Module
- Contributing to Elephant
- Release Notes
- Acknowledgments
- · Authors and contributors
- Citing Elephant

elephant.readthedocs.io



### NeuroML Documentation

Q Search this book...

#### **USER DOCUMENTATION**

Mission and Aims

How to use this documentation

Getting started with NeuroML

Finding and sharing NeuroML models

Creating NeuroML models

Validating NeuroML Models

Visualising NeuroML Models

Simulating NeuroML Models

Optimising/fitting NeuroML Models

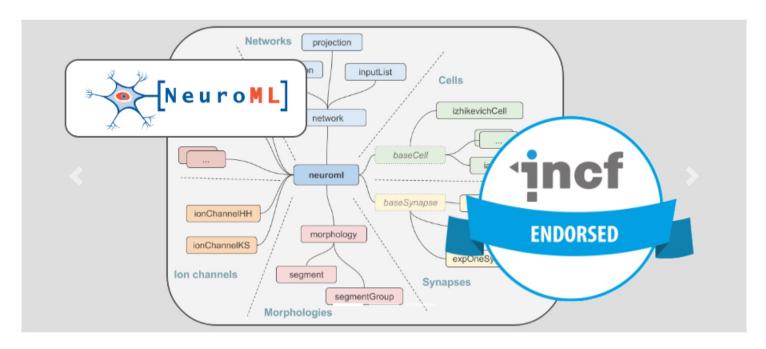
Schema/Specification





## NeuroML

A model description language for computational neuroscience.



NeuroML is an international, collaborative initiative to develop a language for describing detailed models of neural systems, which will serve as a standard data format for defining and exchanging descriptions of neuronal cell and network models. NeuroML is:

To convert a NeuroML model to NEURON, use e.g. pynml model nml -neuron

V

docs.neuroml.org



About

**Neuron Types** 

**Electrophysiology Properties** 

**Articles** 

**FAQs** 

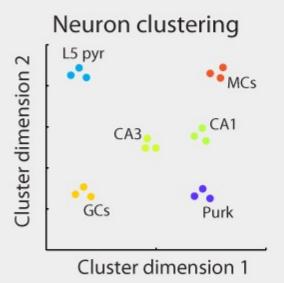
Data/API

Get Involved

### NeuroElectro: organizing information on cellular neurophysiology.

The goal of the NeuroElectro Project is to extract information about the electrophysiological properties (e.g. resting membrane potentials and membrane time constants) of diverse neuron types from the existing literature and place it into a centralized database.

#### Physiology database Published literature Olfactory Bulb Mitral Cell Novel subcellular distribution pattern of A-type K+ Input resistance 200 MΩ channels on neuronal surface. Unique clustering of A-type potassium channels on Extracted from Literature -65 mV $V_rest$ different cell types of the main olfactory bulb. Spike width 1 ms oretical and functional studies predicted a highly non-uniform distribution of voltage-gated ion channels in the neuronal surface. This was confirmed by recent immunolocalization experiments for Na+, Ca2+, perpolarization activated mixed cation and K+ channels. These experiments also indicated that some K+ nannels were clustered in synaptic or non-synaptic membrane specializations. Here we analysed the subcellular distribution of Kr4.2 and Kv4.3 subunits in the rat main olfactory bulb at high resolution to address whether clustering characterizes their distribution, and whether they are concentrated in synaptic or CA1 Pyramidal Cell non-synaptic junctions. The cell surface distribution of the Kv4.2 and Kv4.3 subunits is highly non-uniform. Strong Kv4.2 subunit-immunopositive clusters were detected in intercellular junctions made by mitral, external tufted and granule cells (GCs). We also found Kv4.3 subunit-immunopositive clusters in periglomerular (PGC), nput resistance 400 MΩ leep short-axon and GCs. In the juxtaglomerular region some calretinin-immunopositive glial cells envirage neighboring PGC somata in a cap-like manner. Kv4.3 subunit clusters are present in the cap membrane that directly contacts the PGC, but not the one that faces the neuropil. In membrane specializations established by -70 mV V rest members of the same cell type. K+ channels are enriched in both membranes, whereas specializations between different cell types contain a high density of channels asymmetrically. None of the K+ channel-rich junctions showed any of the ultrastructural features of known chemical synapses. Our study provides evidence Spike width .5 ms for highly non-uniform subcellular distributions of A-type K+ channels and predicts their involvements in novel



Our goal is to facilitate the discovery of neuron-to-neuron relationships and better understand the role of functional diversity across neuron types.

## Pandas:

File storage, data frames, and databases

Saving as CSV with pandas:

```
import pandas as pd
pd.DataFrame({
    "t": t, "v": v}
).to_csv("data.csv", index=False)
```

Loading from CSV with pandas:

```
import pandas as pd
data = pd.read_csv("data.csv")
t = h.Vector(data["t"])
v = h.Vector(data["v"])
```

Saving data to a database:

```
import pandas as pd
import sqlite3
import time
# do all the following inside a loop
start = time.perf counter()
# do the simulation here
data = pd.DataFrame(
   "dx": [dx],
   "L": L,
    "diam": diam,
    "alpha": alpha,
    "temperature": h.celsius,
    "spike_half_width": spike_half_width,
    "calculated_value": calculated_value,
    "runtime": time.perf counter() - start,
with sqlite3.connect(DB FILENAME) as conn:
  data.to_sql("data", conn,
              if exists="append", index=False)
```



A python package to facilitate the development, parallel simulation, optimization and analysis of biological neuronal networks using the NEURON simulator.



Funded by:

