Reproducible Models and Replicable Implementations Current Trends in Computational Neuroscience Hans Ekkehard Plesser, UMB Sharon Crook, ASU Andrew Davison, CNRS

> Presented at <u>SIAM Computational Science & Engineering</u> Reno, NV, 4 March 2011







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Computational Neuroscience



What is Computational Neuroscience?



The goal of neural modeling is to relate, in nervous systems, function to structure on the basis of operation. MacGregor & Lewis, 1977



Keeping it simple: point neurons



Then explore network dynamics!

Random network





Structured network







Hill & Tononi (2005)







Simulations are exciting — but reliable?

- Computational neuroscience
 - no conservation laws
 - no clear-cut separation of scales
 - no general agreement on which aspects of network activity are essential (spike rate vs spike time debate)
 - highly abstract models difficult to compare to experimental data quantitatively
- Highly dependent on *reliable* simulations
- Let's look at a real-life case





Case 2: The clipped Gaussian

- Well-known paper on plasticity
- Parameters chosen from Gaussian distribution, according to paper
- Results could not be reproduced independently
- Analysis of original C-code provided by authors:
 Parameters were chosen from *clipped* Gaussian



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The (sad) state of the art

- Few published results can be reproduced independently
- Authors often struggle to replicate their own results
- Systematic comparison and evaluation of models are rare
- Authors rarely discuss why and how models ended up as they are
- Models are seldom re-used





Reproduction vs Replication

Chris Drummond Replicability is not Reproducibility: Nor is it Good Science ICML Montreal 2009



Replication: necessary, difficult, & insufficient

- Replication
 - Re-create identical results
 - Essentially book-keeping
 - Requires tools & discipline
 - No new insights: tests implementation, not ideas
- Internal replication

Joe recreates results on original machine

• External replication

Jane recreates Joe's results on her machine using Joe's code

• Cross replication

Alice recreates Joe's results using a different simulator, based on a simulator-independent model description





How much detail does replication require?

- Very simple point-neuron model 10^{-12} $\dot{V} = -rac{V}{ au} + rac{I_E}{C}$ 10^{-13} Error [m] • Exact updating rule ($a = I_E \tau/C$) $V_{k+1} = V_k e^{-h/\tau} + a(1 - e^{-h/\tau})$ 10^{-15} 10^{-16} 10⁰ 1∩² 10^{1} • Two different implementations 10^{4} Number of steps V[k+1] = V[k] * exp(-h/tau) - a * expm1(-h/tau)V[k+1] = V[k] - (a - V[k]) * expm1(-h/tau)
- Different numerical properties
 Replication requires that we specify implementation!



But do such tiny differences matter?



Norwegian University of Life Sciences



Reproduction

- Independent
- Test hypotheses and models
- Validates concepts
- Requires reflection





From oil drops to first-passage times

Millikan's oil drop experiment



Erwin Schrödinger *Zur Theorie der Fall- und Steigversuche an Teilchen mit Brownscher Bewegung* Physikalische Zeitschrift **16**:289 (1915)

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Reproduction needs replication

- Reproducible models describe scientific ideas
- Independent reproduction will generally fail to replicate original results precisely
- Requires learned judgement of discrepancies
- Requires means to understand failure
- Replicable implementation





Improving Scientific Practice in Computational Neuroscience



Good model description practice

- Systematic approach to model description in papers
- Standardize tables/checklists
- Standards for graphic representation of models
- Nordlie, Gewaltig, & Plesser
 PLoS Comp Biol 5:e1000456 (2009)

Α	Model Summary
Populations	Three: excitatory, inhibitory, external input
Topology	-
Connectivity	Random convergent connections
Neuron model	Leaky integrate-and-fire, fixed voltage threshold, fixed absolute refractory time (voltage clamp)
Channel models	-
Synapse model	α-shaped current inputs
Plasticity	-
Input	Independent fixed-rate Poisson spike trains to all neurons (during initial stimulation period)
Measurements	Spike activity

В		Populations
Name	Elements	Size
E	laf neuron	$N_{E} = 4N_{I}$
I	laf neuron	NI
E _{ext}	Poisson generator	$C_E(N_{E}+N_{I})$

С			Connectivity
Name	Source	Target	Pattern
EE	E	E	Random convergent $C_{E} \rightarrow 1$, weight <i>J</i> , delay <i>D</i>
IE	E	I	Random convergent $C_{E} \to 1$, weight <i>J</i> , delay <i>D</i>
EI	I	E	Random convergent $C_{I} \rightarrow 1$, weight $-gJ$, delay D
II	I	I	Random convergent $C_{I} \rightarrow 1$, weight $-gJ$, delay D
Ext	E _{ext}	E∪I	Non-overlapping $C_{E} \to 1$, weight <i>J</i> , delay <i>D</i>

D	Neuron and Synapse Model								
Name	laf neuron								
Туре	Leaky integrate-and-fire, α -current input								
Subthreshold dynamics	$\begin{split} \tau \dot{V}(t) &= -V(t) + RI(t) \text{if} \qquad t > t^* + \tau_{\text{rp}} \\ V(t) &= V_{\text{r}} \qquad \text{else} \\ I(t) &= \ \frac{\tau}{R} \sum_{\tilde{t}} w \alpha (t - (\tilde{t} + \Delta)) \Theta (t - (\tilde{t} + \Delta)) \end{split}$								
Spiking	If $V(t-) < \theta \land V(t+) \ge \theta$ 1. set $t^* = t$ 2. emit spike with time-stamp t^*								
E	E Input								

E	input
Туре	Description
Poisson generators	Fixed rate $\nu_{\text{ext}},$ C_{E} generators per neuron, each generator projects to one neuron; active only during initial stimulation period

Spike activity as raster plots for subset of excitatory neurons



Professional, shared software

- Widely used packages replace homemade *ad hoc* code
- Currently: Neuron, NEST, Genesis, Moose, Brian, PCSim
- "Social" developments
 - Simulator review (Brette et al, 2007)
 - Teaching software at summer schools
 - Large-scale *scientific* projects (eg EU FACETS)
 - Neuroinformatics journals
 - Raising awareness among reviewers and editors
- "Technical" developments
 - Version control
 - Test suites

Software technology used in NEST development





ModelDB: Sharing models

- Curated database of computational neuroscience models
- Only published models
- Open to any software
- Nearly 700 models
- <u>http://</u> senselab.med.yale.edu/ modeldb/



Sparsely connected networks of spiking neurons (Brunel 2000)

Accession: 42020

The dynamics of networks of sparsely connected excitatory and inhibitory integrate-and-fire neurons are studied analytically (and with simulations). The analysis reveals a rich repertoire of states, including synchronous states in which neurons fire regularly; asynchronous states with stationary global activity and very irregular individual cell activity; and states in which the global activity oscillates but individual cells fire irregularly, typically at rates lower than the global oscillation frequency. See paper for more and details.

Reference: Brunel N (2000) Dynamics of sparsely connected networks of excitatory and inhibitory spiking neurons. <u>J Comput Neurosci</u> 8:183-208 [PubMed]

Citations Citation Browser

Model Information (Click of	on a link to find other models with that property)
Model Type:	Connectionist Network;
Brain Region(s)/Organism:	
Cell Type(s):	
Channel(s):	
Gap Junctions:	
Receptor(s):	
Gene(s):	
Transmitter(s):	
Simulation Environment:	NEST (formerly BLISS/SYNOD);
Model Concept(s):	Activity Patterns; Oscillations; Spatio-temporal Activity Patterns; Simplified Models;
Implementer(s):	Gewaltig, Marc-Oliver ;

brunel	Readme.txt for an implementation of the model associated with the paper:							
D <u>readme.txt</u> D <u>brunel.sli</u>	Brunel N (2000) Dynamics of sparsely connected networks of excitatory and inhibitory spiking neurons. J Comput Neurosci 8:183-208							
	The brunel.sli file was supplied by Marc-Oliver Gewaltig and runs under NEST. Please contact Marc-Oliver Gewaltig marc-oliver.gewaltig@honda-ri.de for more information.							





PyNN: A Meta-Simulator

- Python-based wrapper for many simulators
- Write model and simulation code once, run on all
- Facilitates model sharing and cross-validation
- Developed by Andrew Davison for FACETS project
- <u>http://neuralensemble.org/PyNN</u>



NeuroML: A model specification language

- XML-based language for model specification
- Multiple layers: channels, neuron morphologies, networks
- Code-generation for several simulators, including PyNN
- Facilitates model sharing and re-use
- <u>http://www.neuroml.org</u>

Provenance tracking: Sumatra

- Python package to enable systematic capture of the environment of numerical simulations/analyses
- Tracks simulation code, dependencies, platform information, results
- Developed by Andrew Davison as part of FACETS project
- http://neuralensemble.org/sumatra

000	Sumatra: TestProject: List of records												
▲ ► ▲ + ♦ http://127.0.0.1:8002/													
TestProject: List of records													
Delete	Label	Label Reason Outcome Duration Processes Simulator		Script	Script			Time	Tags				
include data 🚍						Name	Version	Repository	Main file	Version			
	<u>20100709-</u> <u>154255</u>		'Eureka! Nobel prize here we come.'	0.59 s		Python	2.5.2	/Users/andrew/tmp/SumatraTest	main.py	396c2020ca50	09/07/2010	15:42:55	
Θ	<u>20100709-</u> <u>154309</u>			0.59 s		Python	2.5.2	/Users/andrew/tmp/SumatraTest	main.py	396c2020ca50	09/07/2010	15:43:09	
	haggling	'determine whether the gourd is worth 3 or 4 shekels'	'apparently, it is worth NaN shekels.'	0.59 s		Python	2.5.2	/Users/andrew/tmp/SumatraTest	main.py	396c2020ca50	09/07/2010	15:43:20	<u>foobar</u>

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NineML: A model description standard

- Aims for community standard for declarative model descriptions
- Inspired by SBML and CellML
- Focus on networks of point neurons
- Under development by INCF Multi-scale Modeling Task Force

Perspectives

- Community increasingly aware of need for reproducibility and replicability
- Large-scale projects have led to development of valuable tools
- Summer schools educate PhD-students and post-docs in use of established modeling tools
- Neuroinformatics journals allow publication of domainspecific solutions
- International Neuroinformatics Coordinating Facility (<u>INCF</u>) stimulates debate and development
- <u>NEST Initiative</u> is devoted to furthering reliable simulations
- ➡ We have a long way to go, but we are (finally) moving!

Collaborators

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nest:: initiative

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Research Council of Norway (eVita program)

