



FACETS

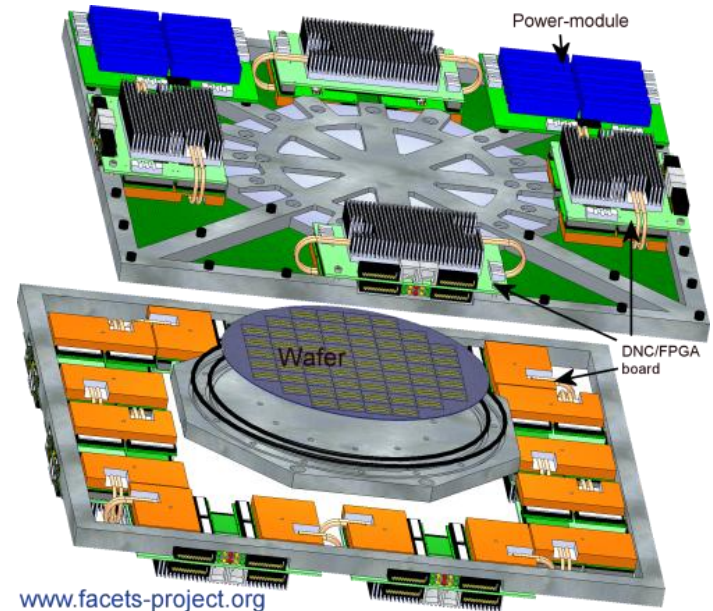
<http://www.facets-project.org>

Fast Analog Computing with Emergent Transient States

The FACETS Demonstrator

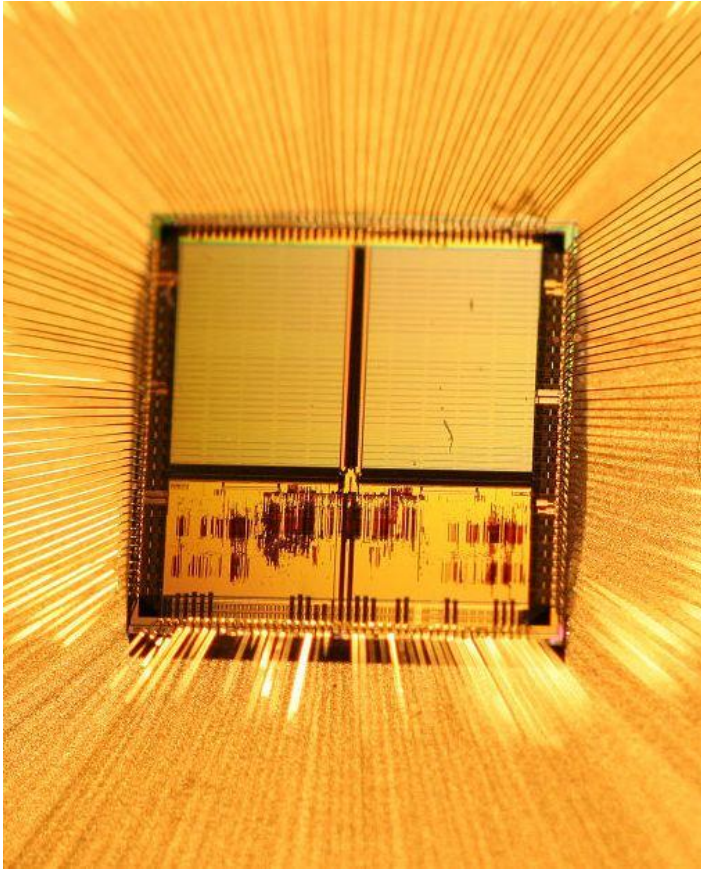
M. A. Petrovici
University of Heidelberg

Code Jam 3
Freiburg, 09.10.2009



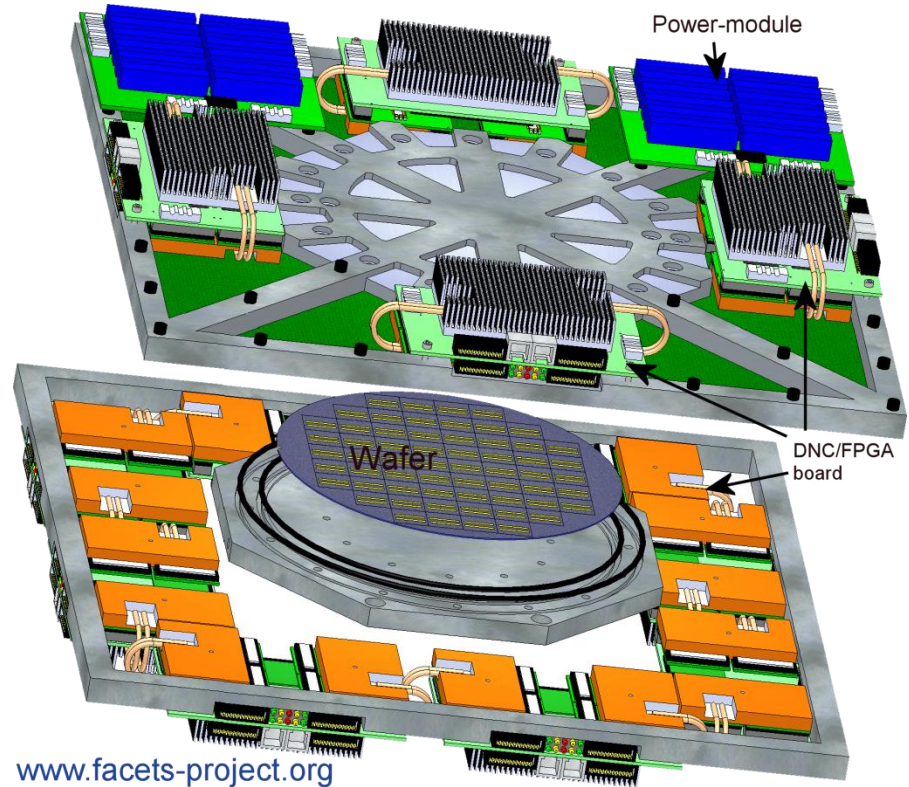
The FACETS Hardware

Stage 1: The Spikey Chip



384 neurons
 10^5 synapses

Stage 2: Waferscale integration



www.facets-project.org

$16 \cdot 10^4$ neurons
 $4 \cdot 10^7$ synapses

The FACETS Demonstrator is...

... a collection of cortical **neural network models** emulated with the **FACETS wafer-scale hardware** *

* as long as not yet available: dedicated executable system specification

...exhibiting functionality that can be demonstrated

...written in PyNN

...computable with established software simulators
(for verification, performance evaluation etc.)

Why bother ?

Reviewers & community are interested in:

Ability of the wafer-scale hardware system to serve as a flexible research tool in modeling neuroscience:

- despite hardware constraints, relevant experiments possible
- crucial interplay between hardware and software
- massive complexity of hardware configuration space can be handled by intelligent software
- operability by non-hardware experts

Ability of FACETS to fruitfully cooperate:

- Demonstrator is actively developed by UHEI, TUD, KTH, UNIC, ALUF, INCM
- hardware neuron and synapse models: result of discussions. FACETS has participated in their original development, e.g. EPFL.

Why not wait for the hardware ?

Testing and evaluation of all involved software layers

Virtual hardware allows to

- test software before hardware is available
- test without possible hardware-specific problems
- provide a preliminary PyNN module for off-line testing of experiments

Verification of possible hardware changes

e.g. optionally insert detailed HICANN model

⇒ **Important test-bench for preparation work**

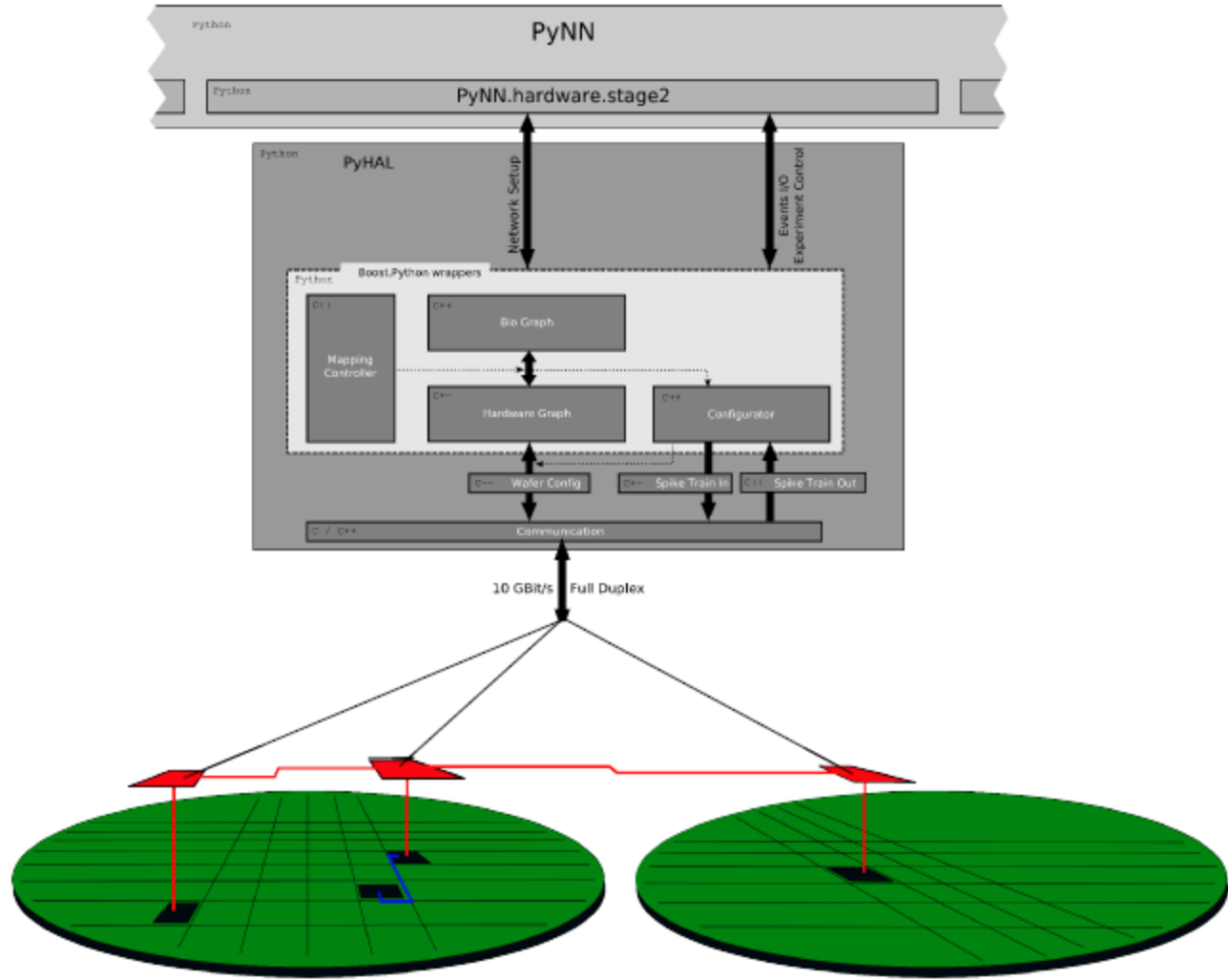
Simulating the emulator

Operating Interface

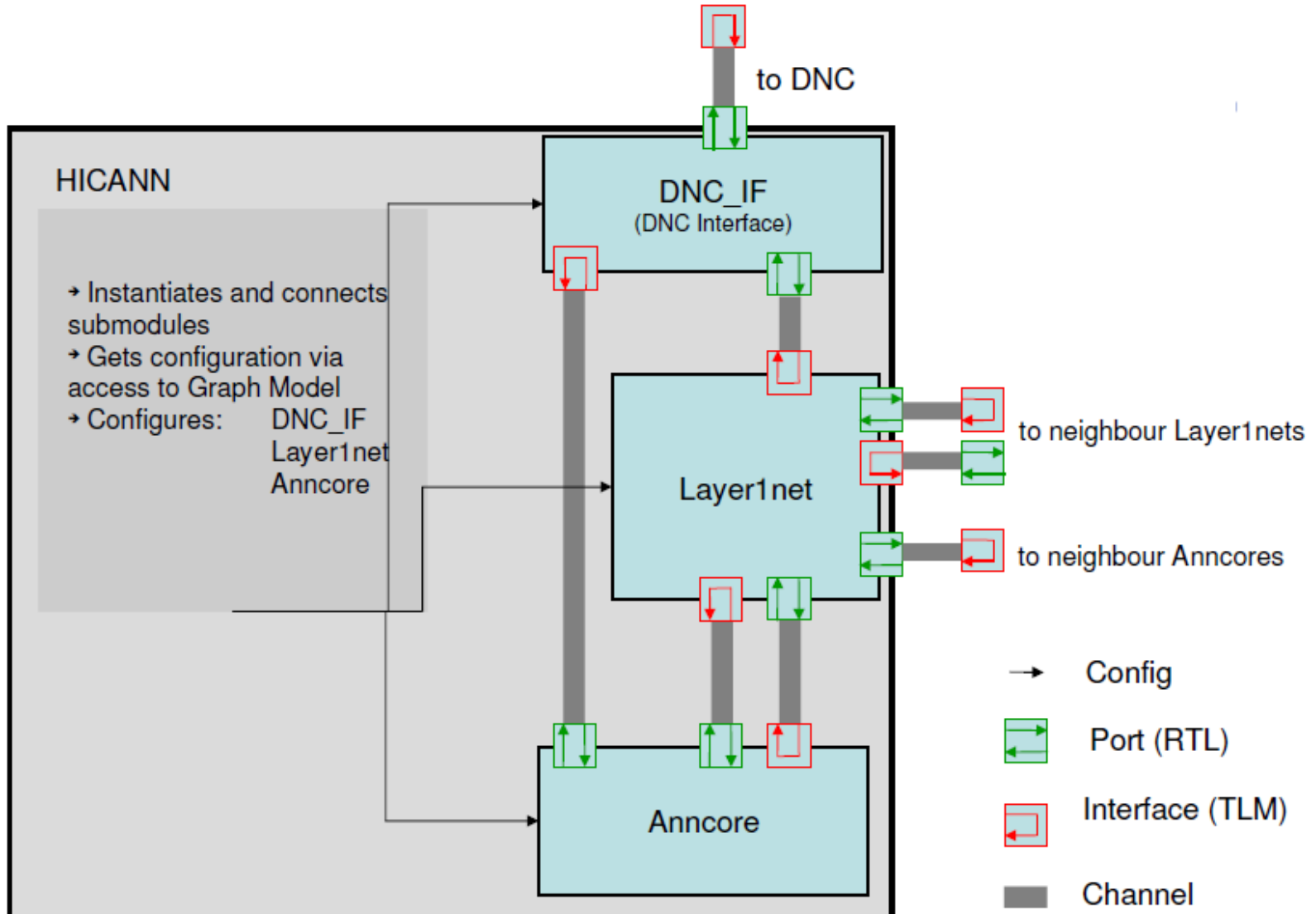
Mapping Software

Digital (Layer2)

Analog (Layer1)



Simulating the emulator



Experiments and demonstrations (to-do list...)

Evaluation of the involved software layers

- show full support of low-level and object-oriented PyNN API
- quantify performance of the mapping and routing algorithms
 - Execution time, memory consumption etc as function of network size, parameter heterogeneity, applied optimization strategies etc.
 - Neuron and synapse loss as function of ...
 - Histogram of layer1 hops as function of ...

Demonstrate correct functionality of the virtual hardware

- Unit tests

Experiments and demonstrations (to-do list...)

Model performance: Influence of hardware-specific imperfections or inhomogeneities

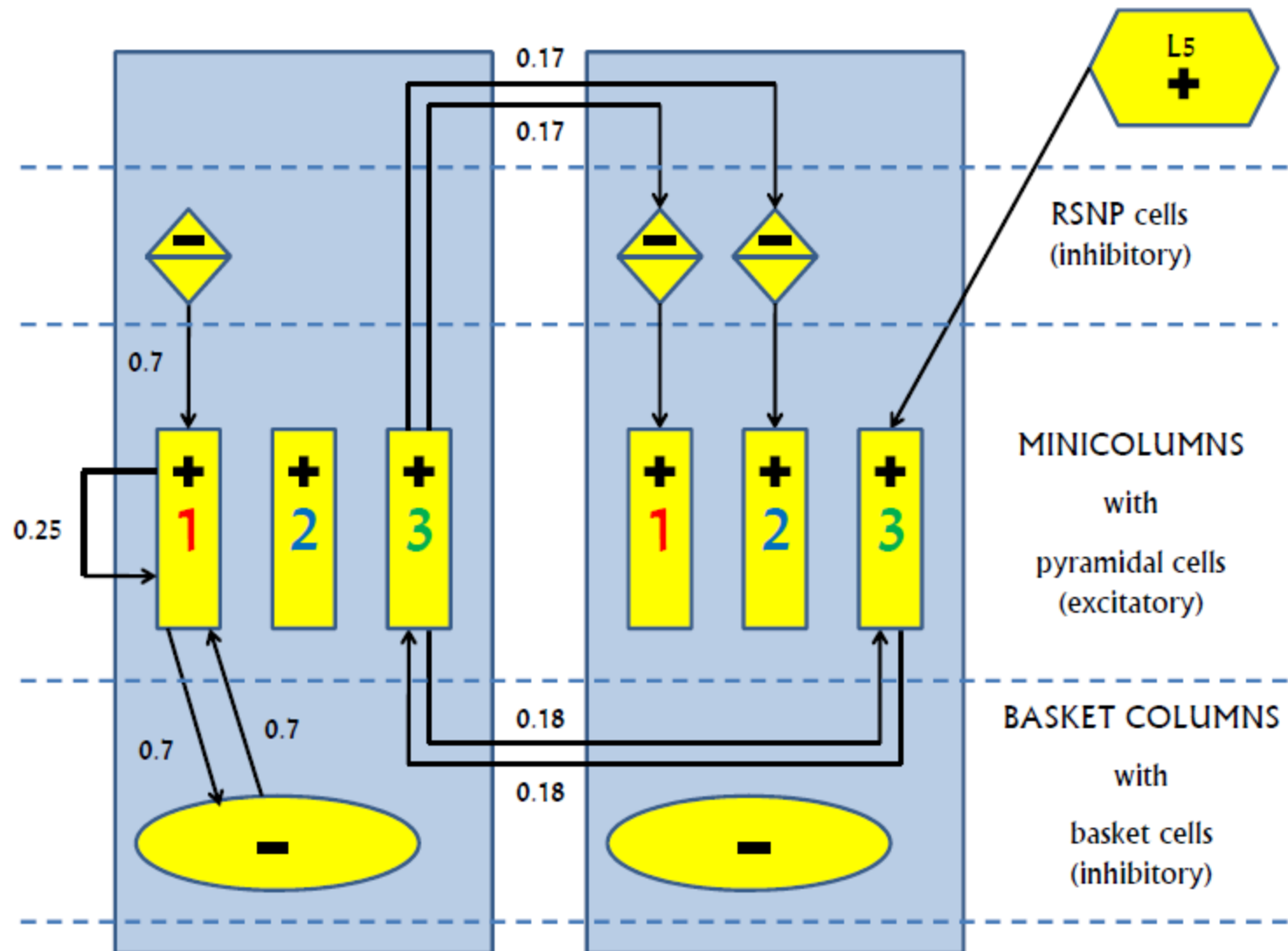
- neuron and synapse parameters (e.g. 4-bit weights, limited precision), signal transmission delays, ...
 - comparison software vs. (virtual) hardware
 - incorporate hardware-specific distortions into software model
- automatically extracted: “post-mapping” PyNN scripts that reflect mapping distortions

The Demonstrator models (so far)

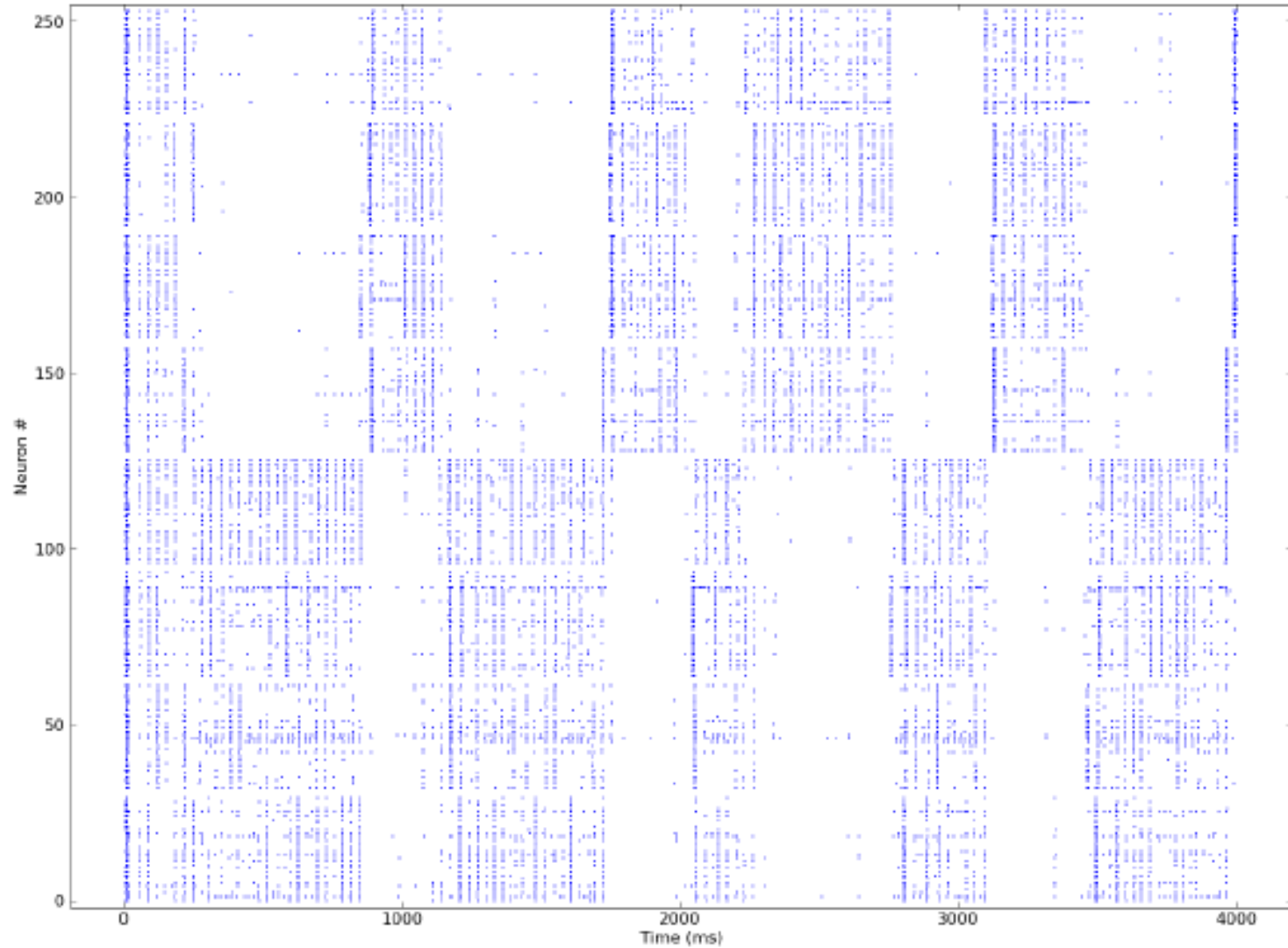
- **A layer 2/3 attractor memory**
(by KTH, Krishnamurty / Lansner)
- **A synfire chain model**
(by INCM and ALUF, Kremkow / Aertsen / Masson)
- **A model of self-sustaining cortical AI states**
(by UNIC, Davison / Destexhe)
- **Upcoming: Two-layer model by UNIC**

All written in PyNN, all scalable, basic versions can be mapped to hardware without synapse loss

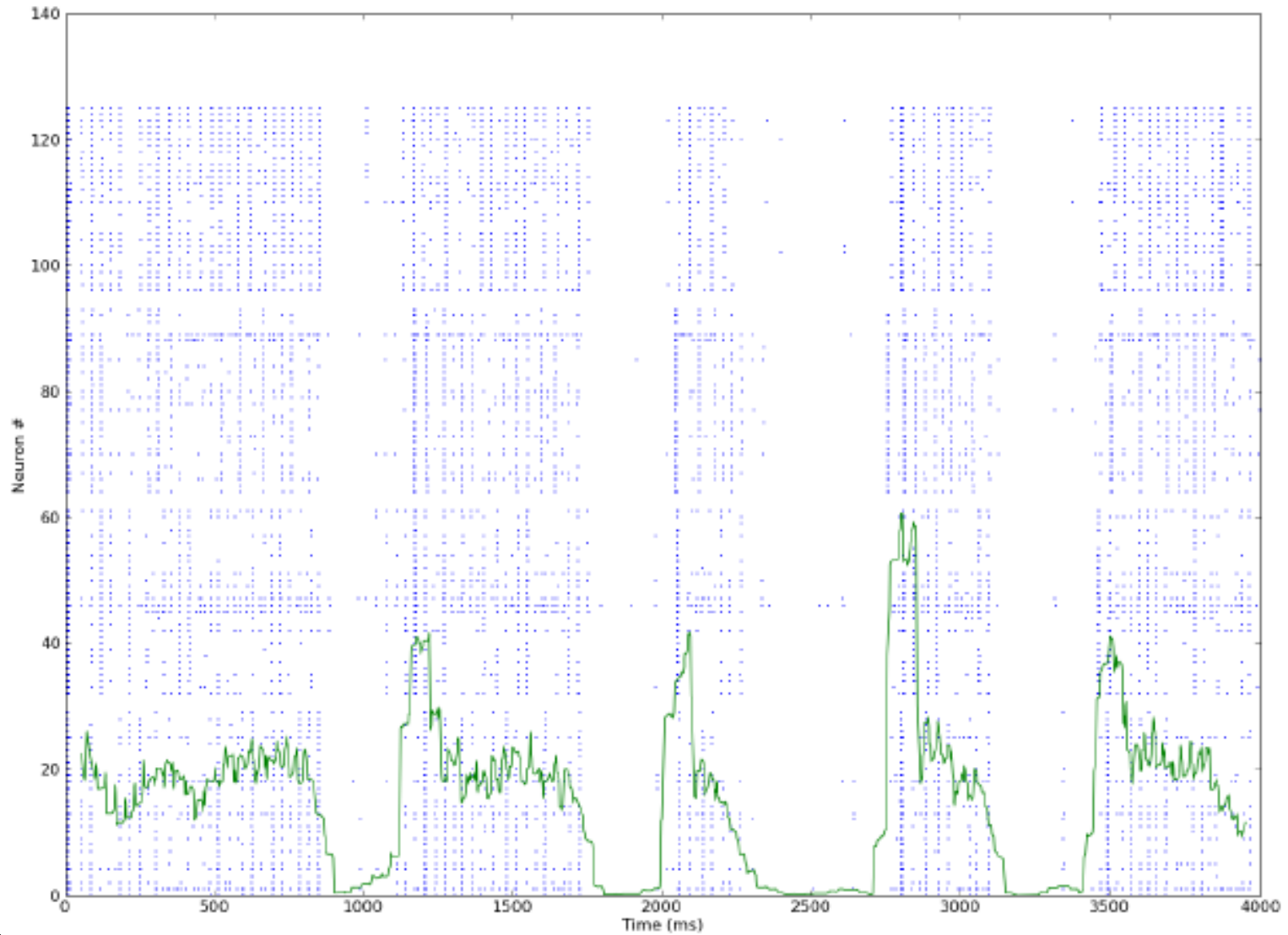
Attractor memory schematic



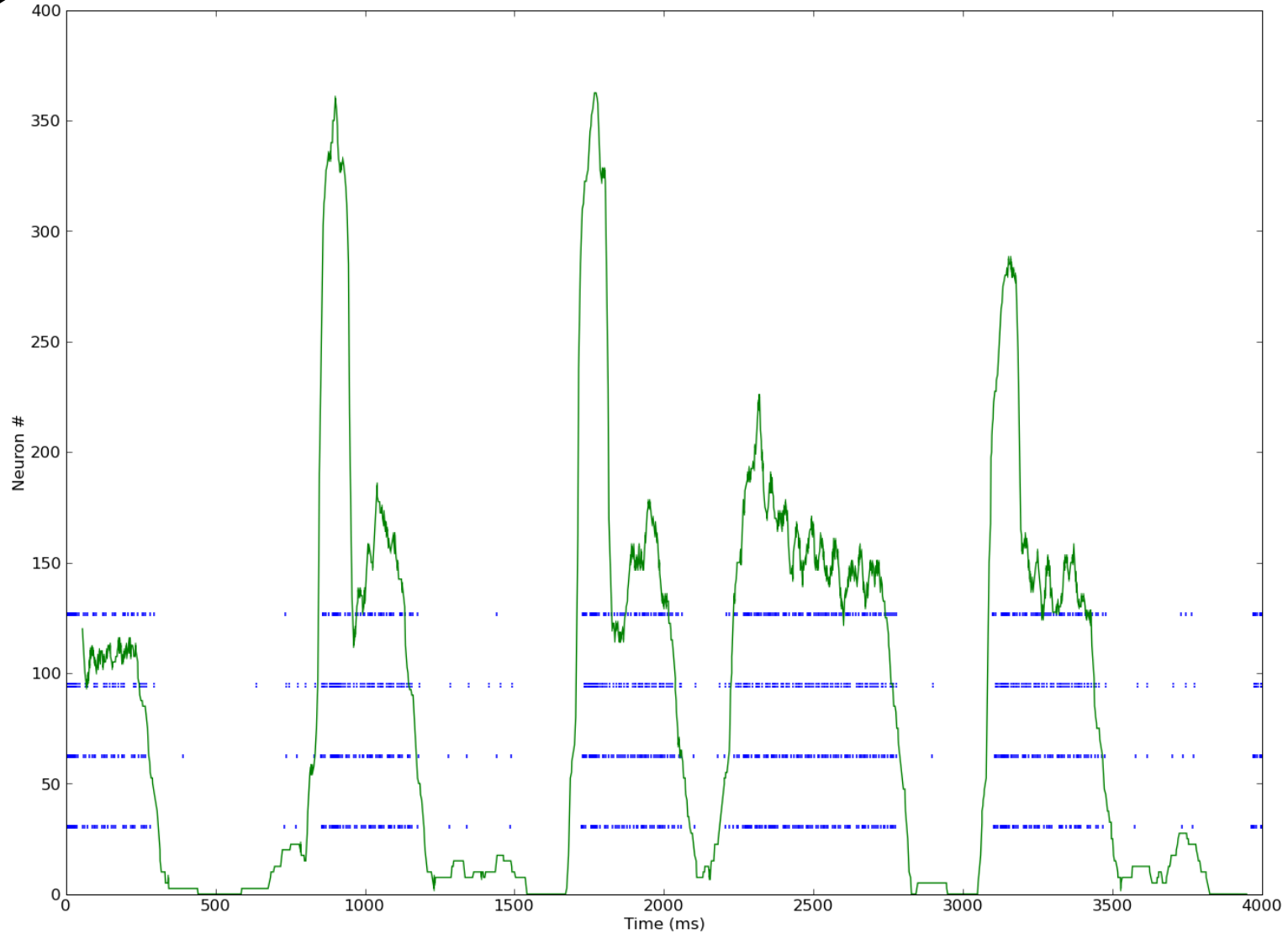
Attractor memory (NEST): 2 attractors



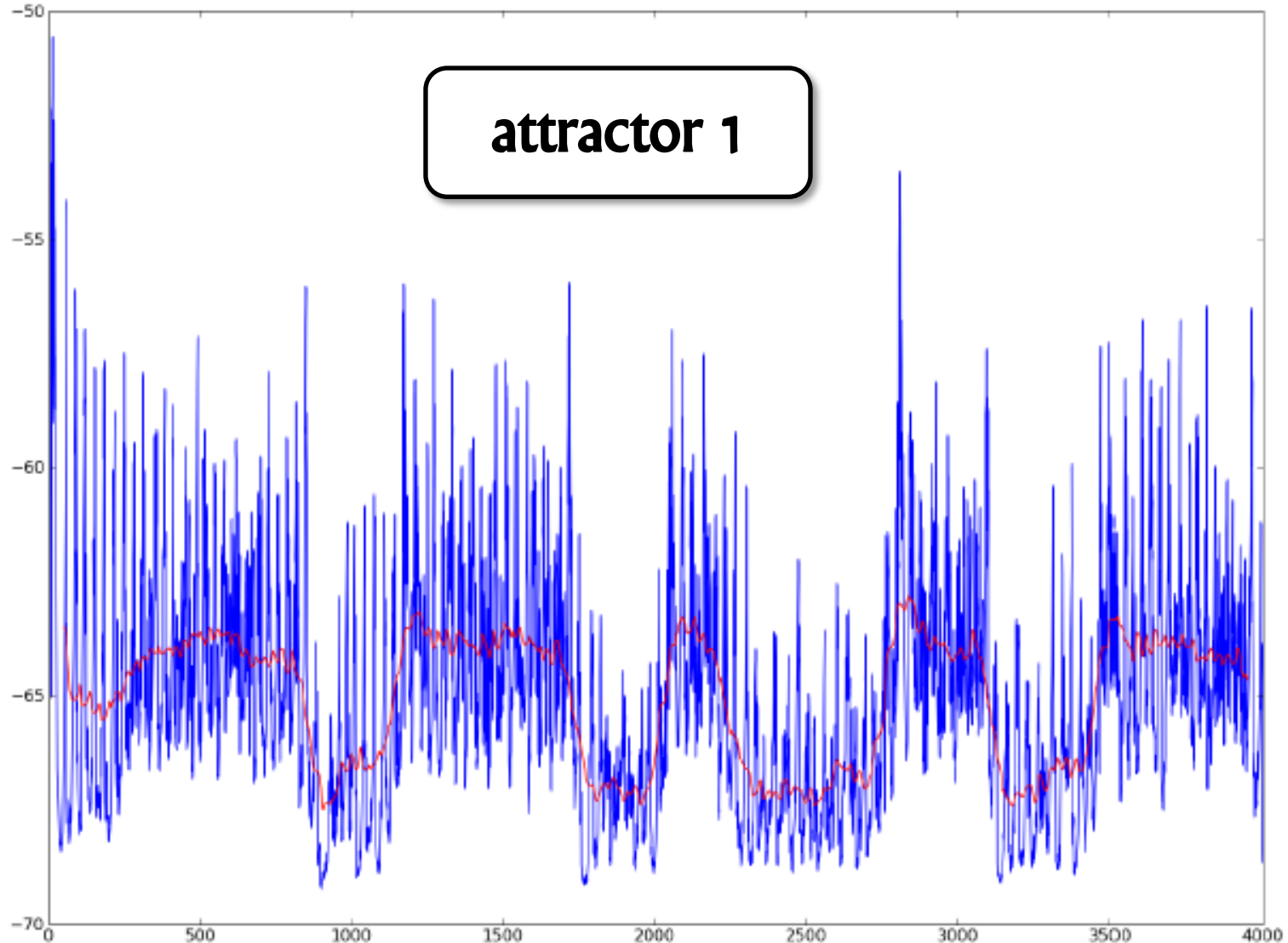
Attractor memory (NEST): pyramidal spike rate



Attractor memory (NEST): RSNP spike rate

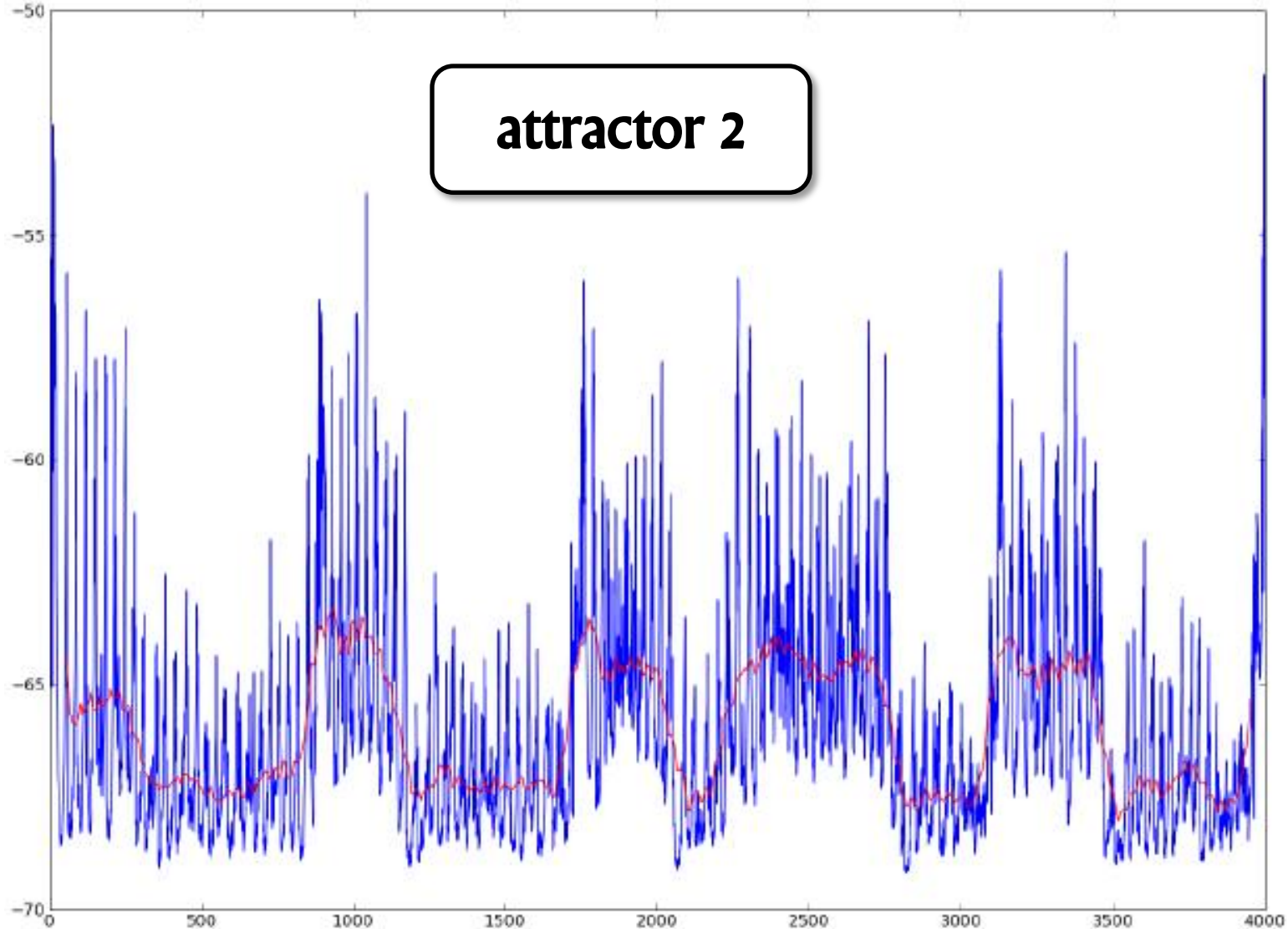


Attractor memory (NEST): average membrane potential

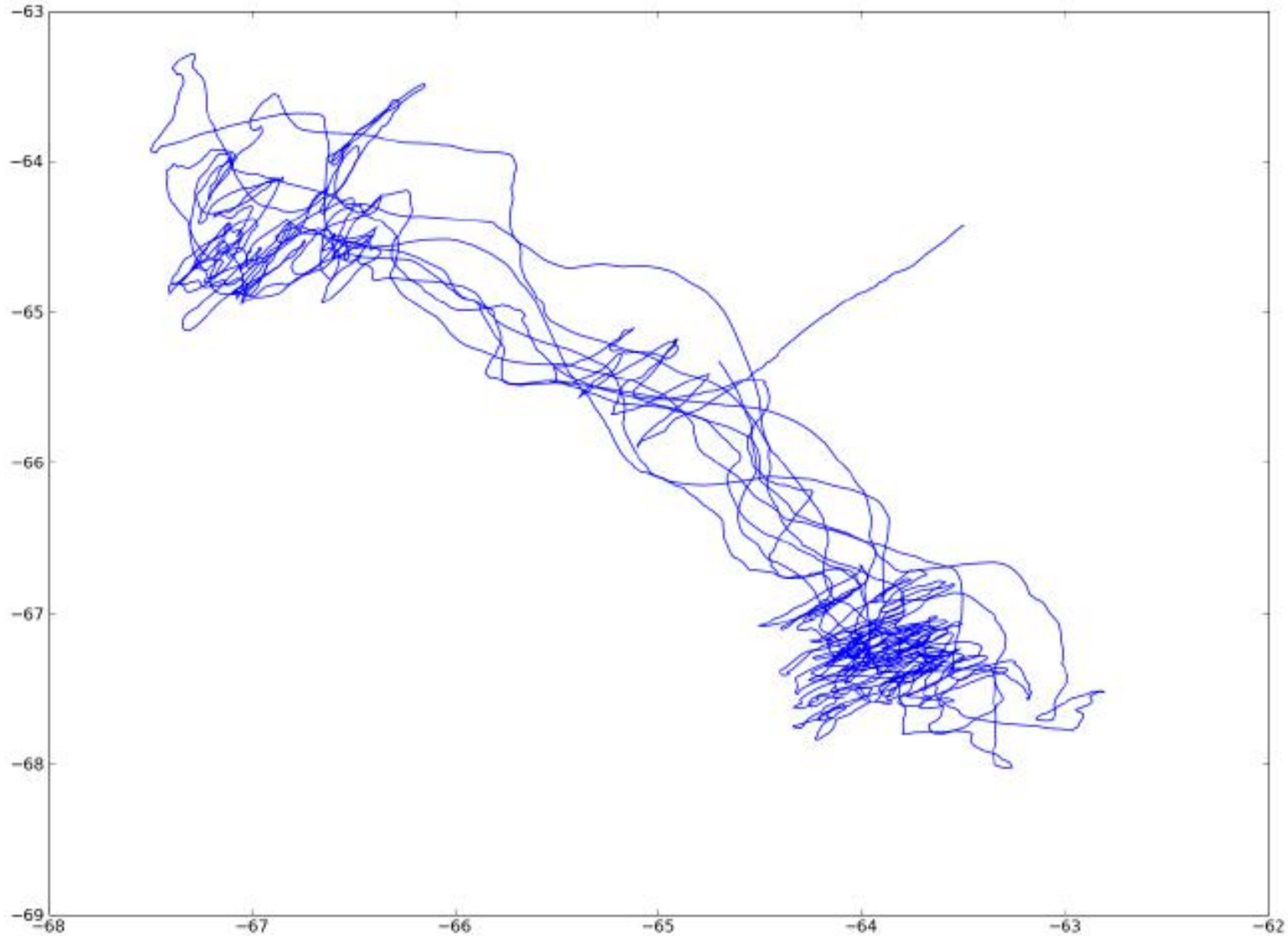


Attractor memory (NEST): average membrane potential

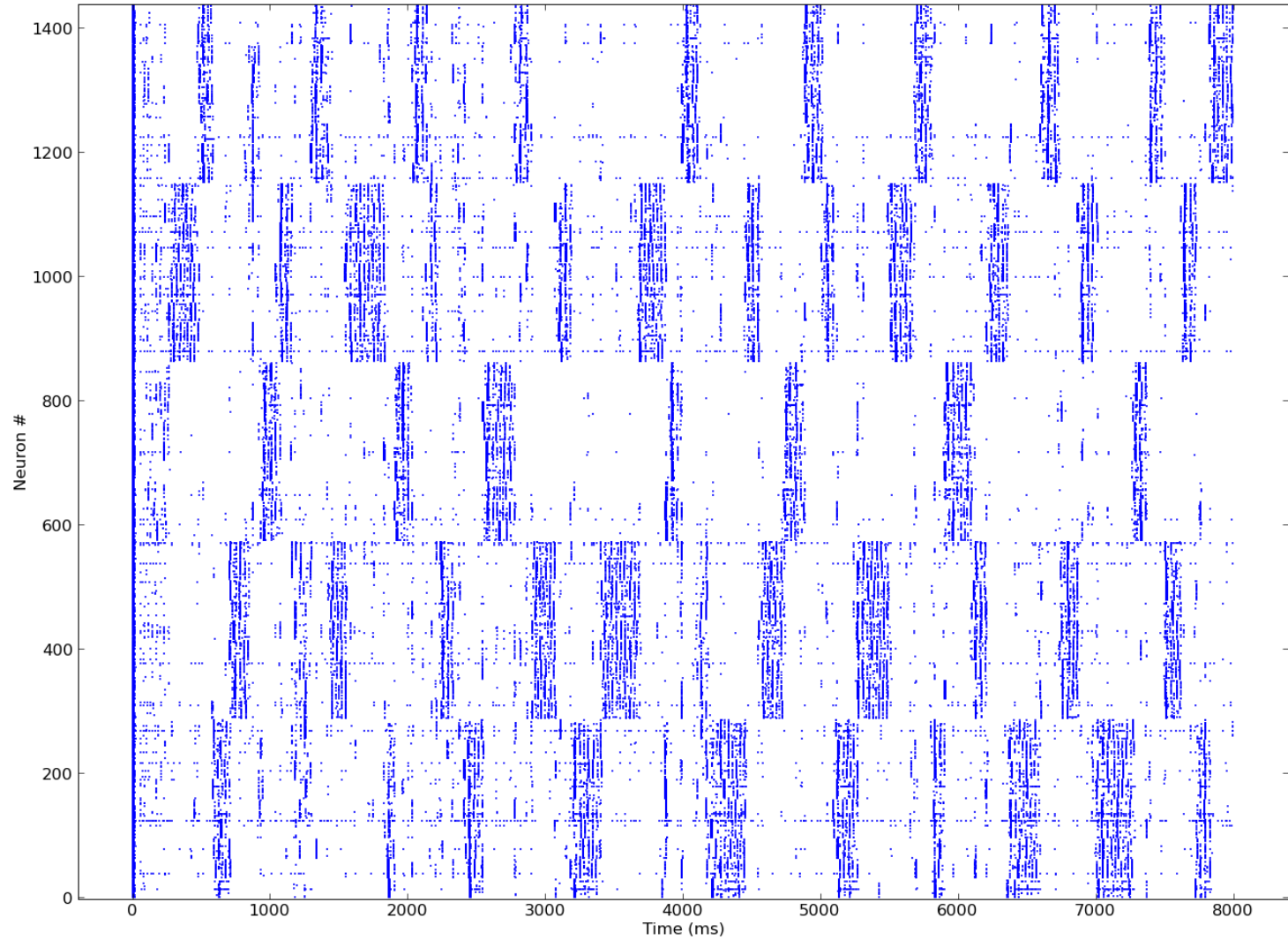
attractor 2



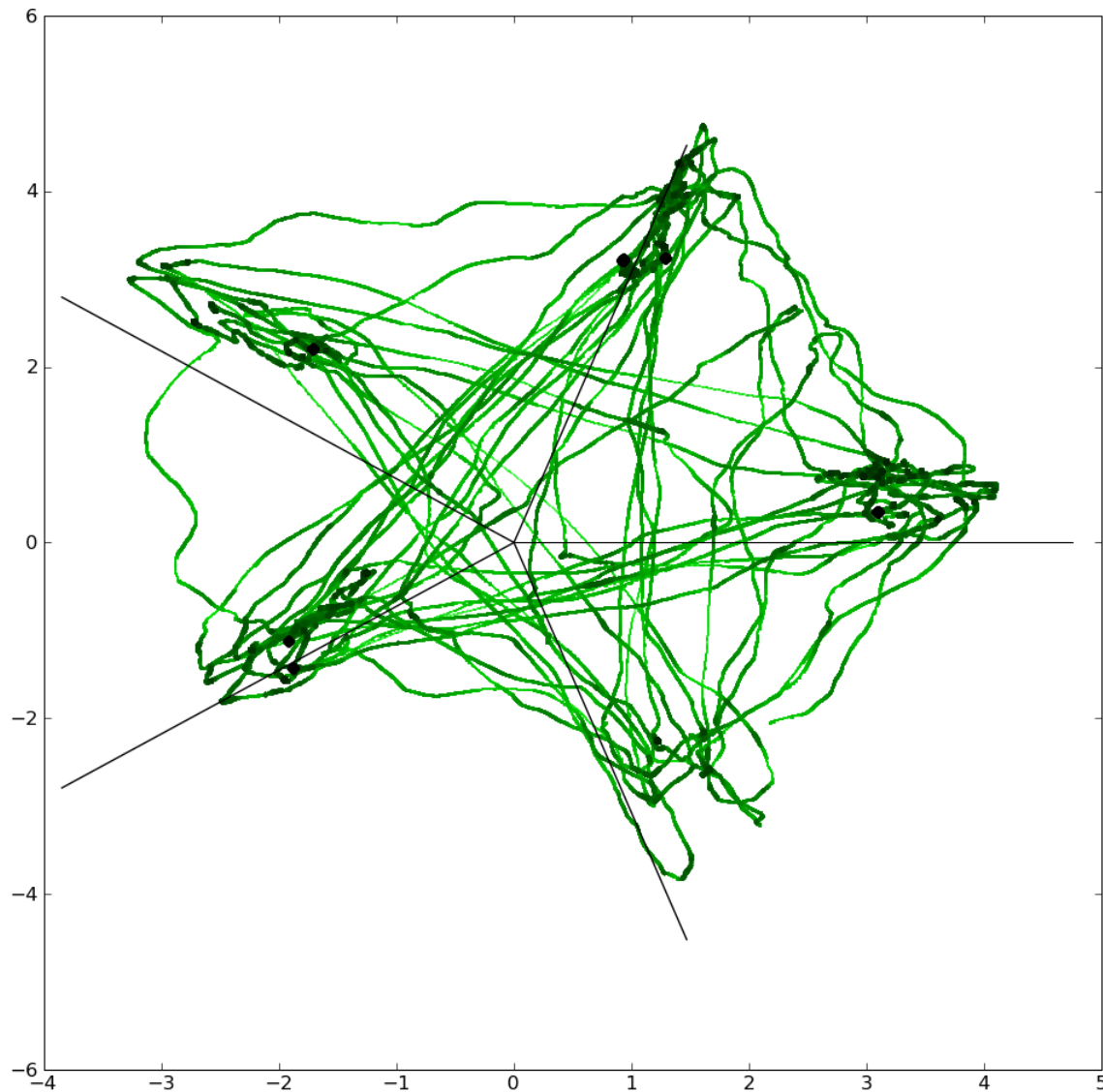
Attractor memory (NEST): phase diagram



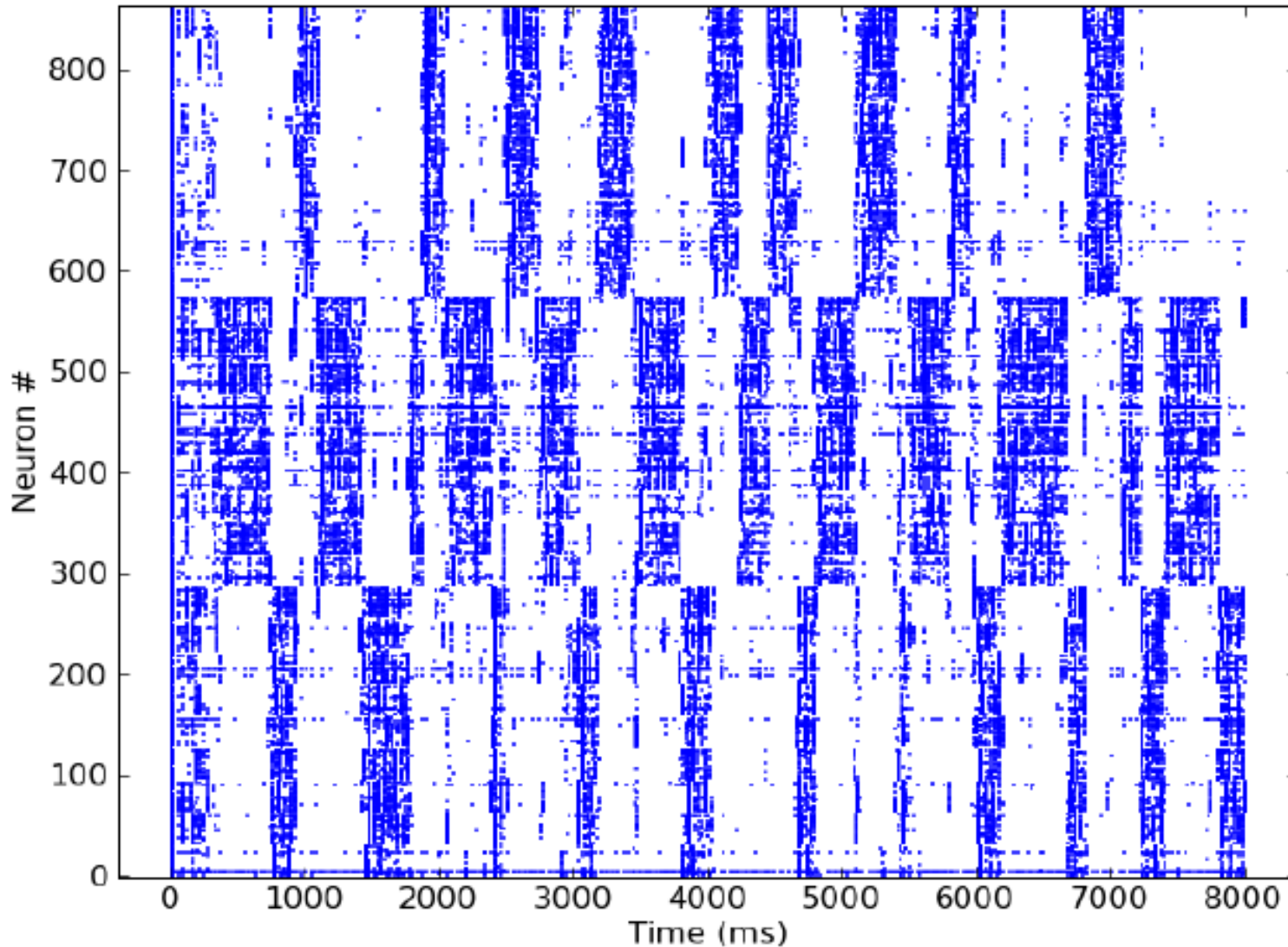
Attractor memory (NEST): 5 attractors



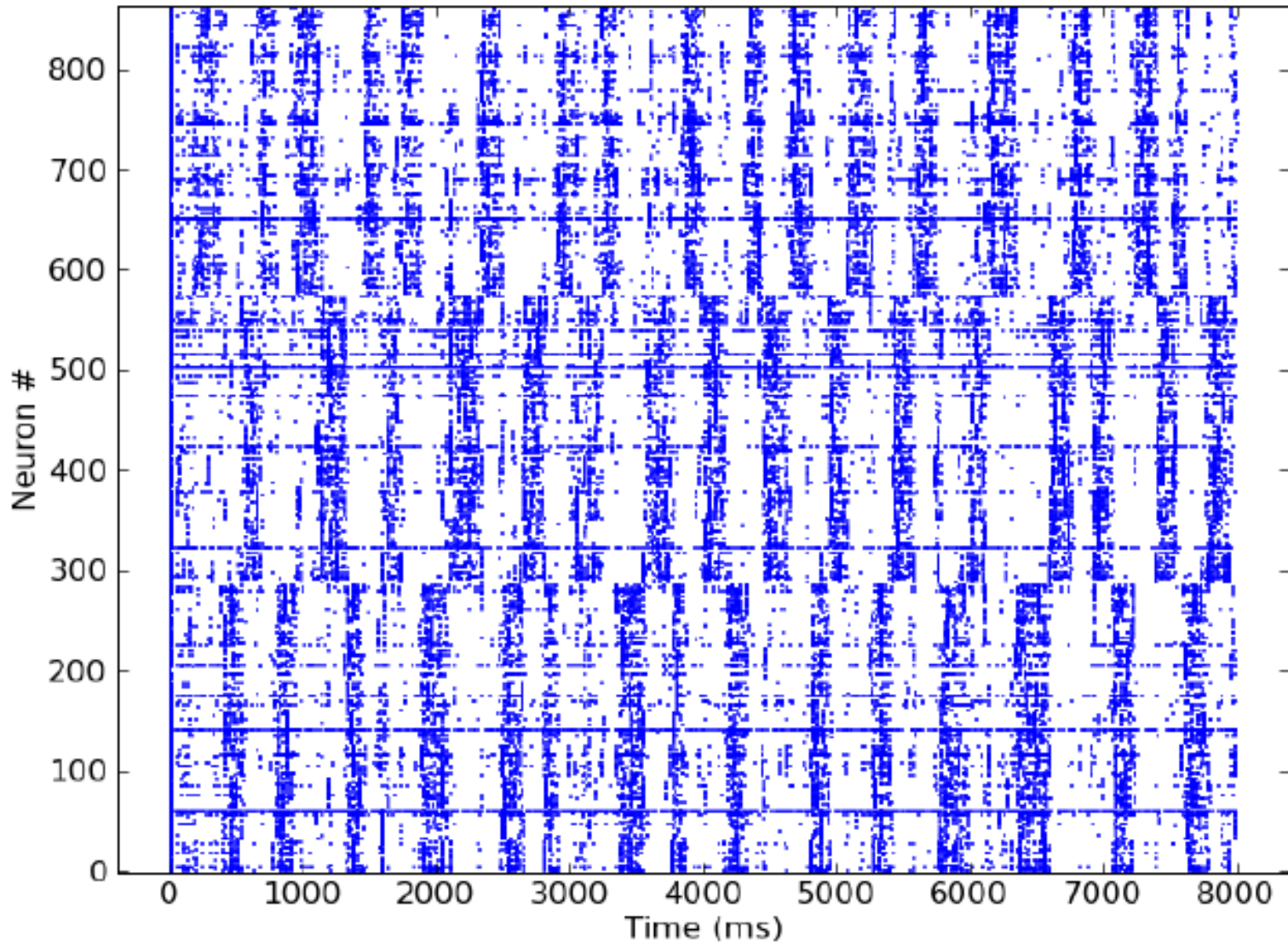
Attractor memory (NEST): “advanced” phase diagram



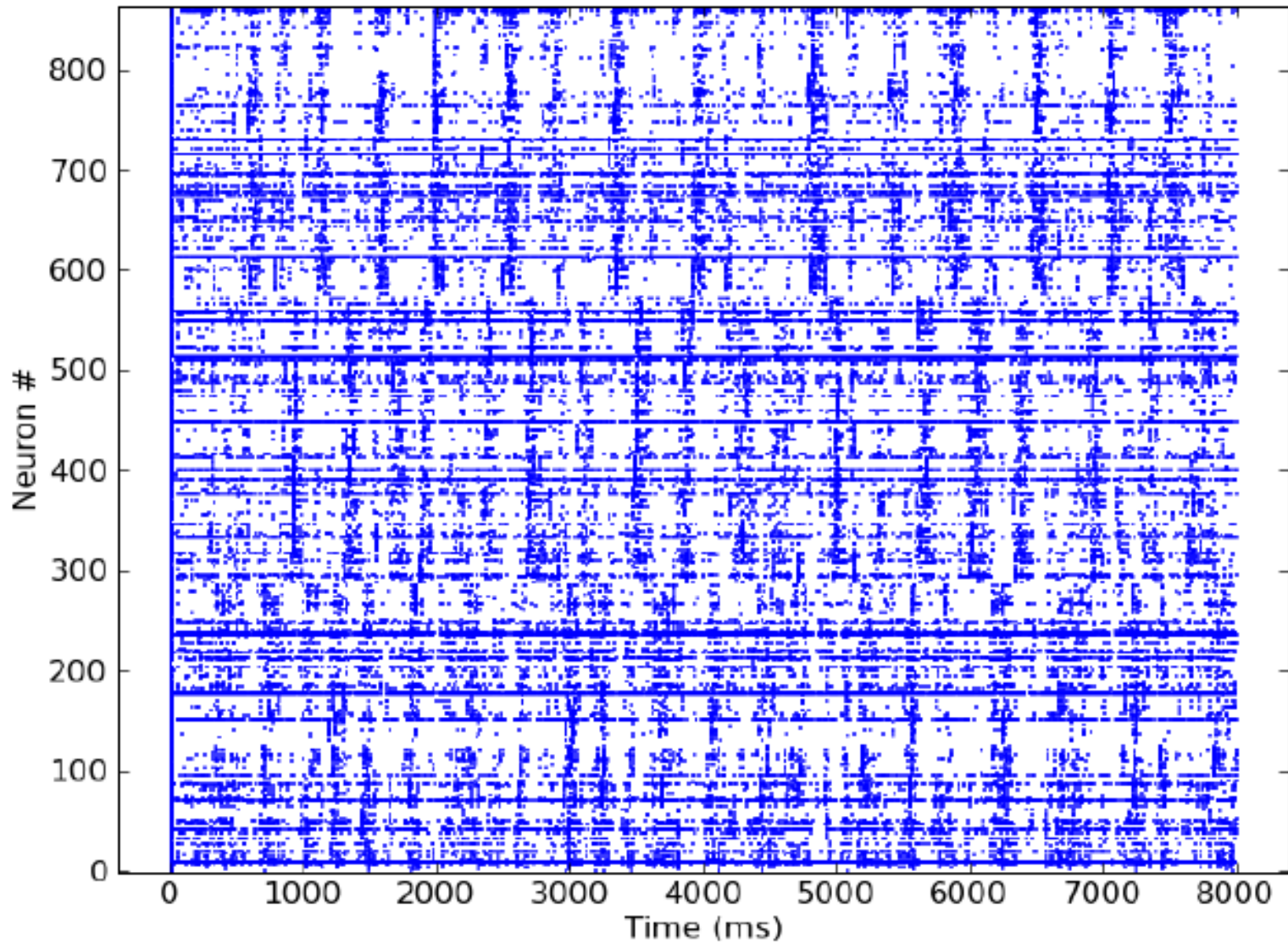
Attractor memory (NEST): 0% synaptic loss



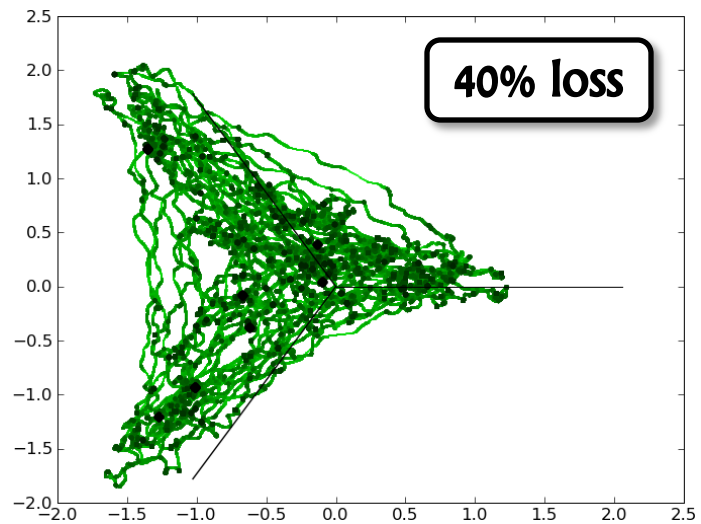
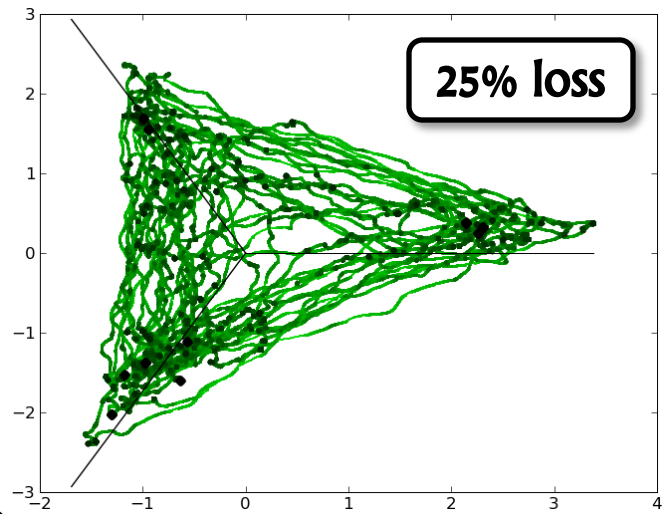
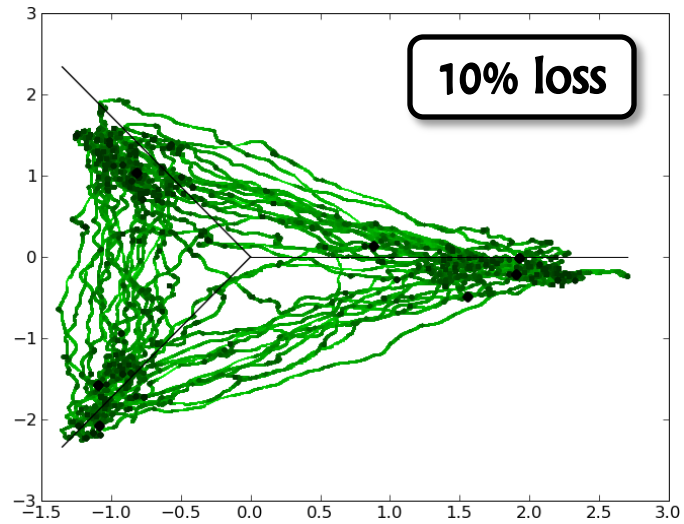
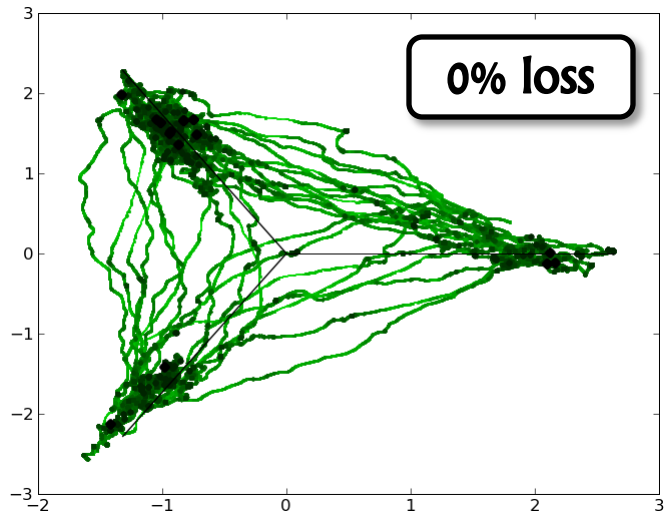
Attractor memory (NEST): 20% synaptic loss



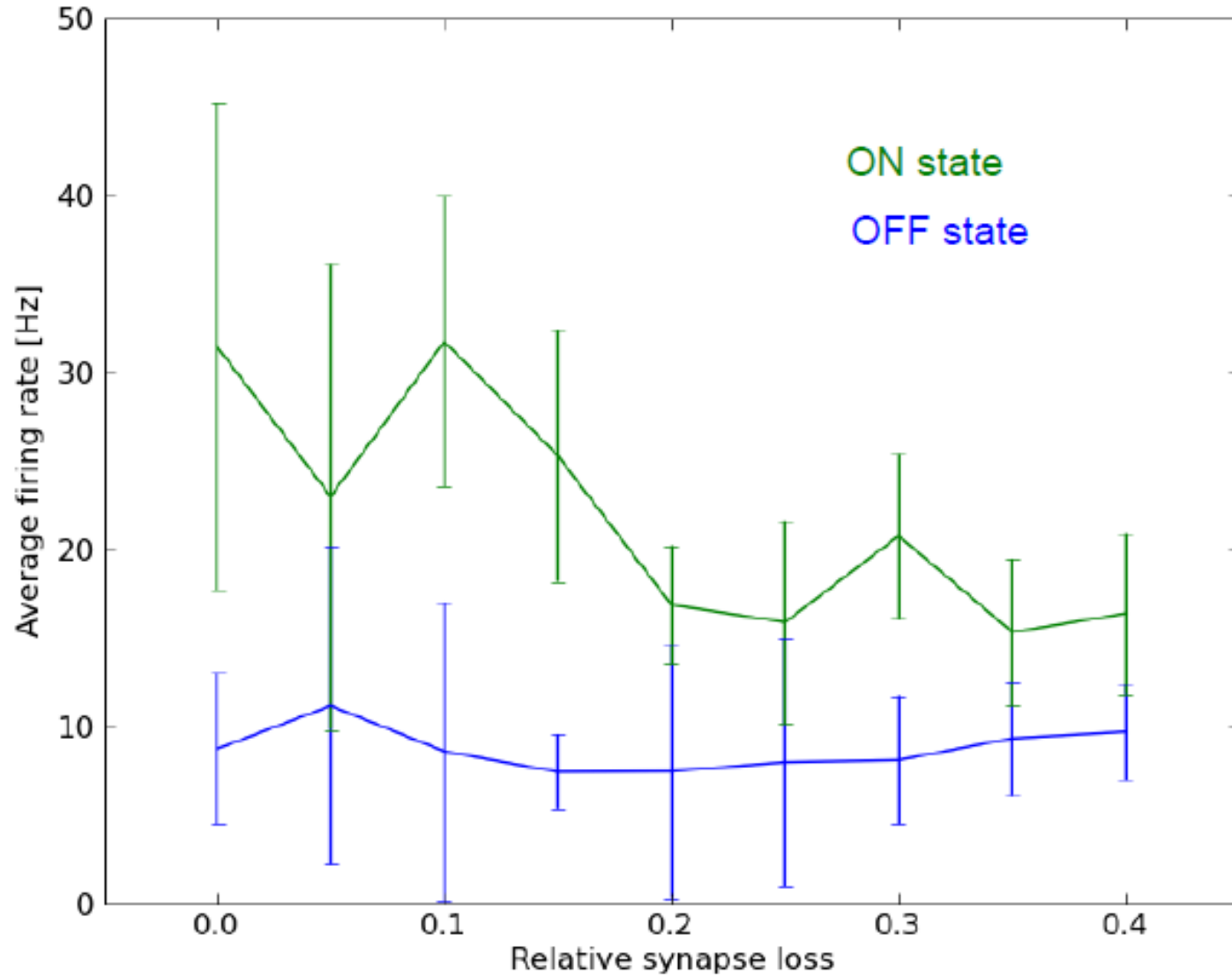
Attractor memory (NEST): 40% synaptic loss



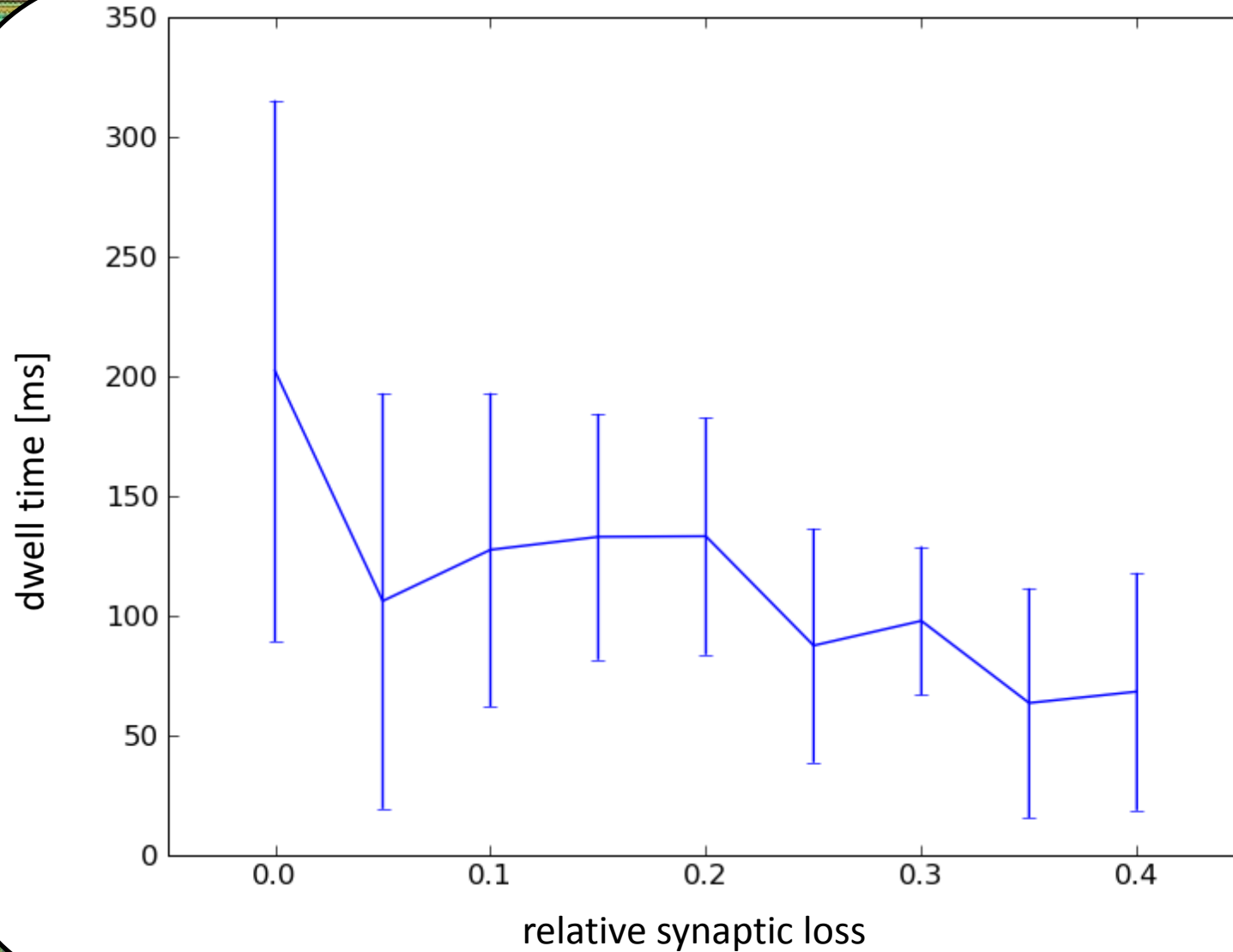
Attractor memory (NEST): phase diagrams



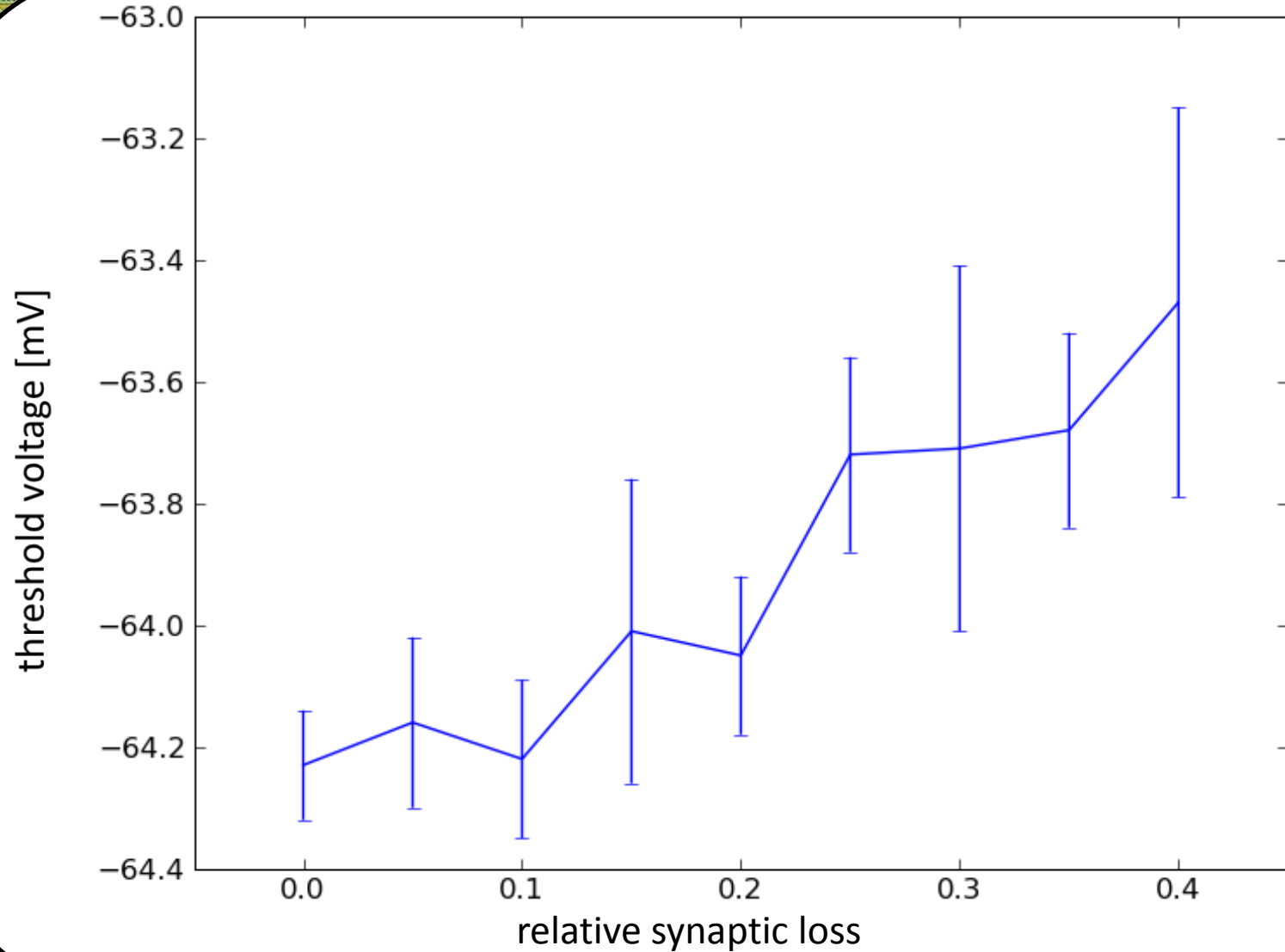
Attractor memory (NEST): firing rates



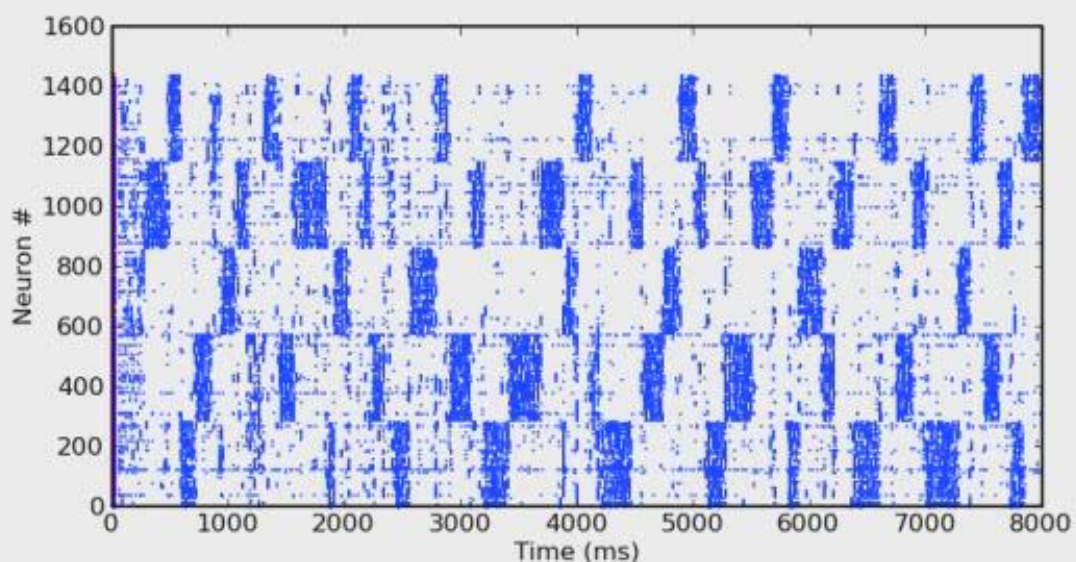
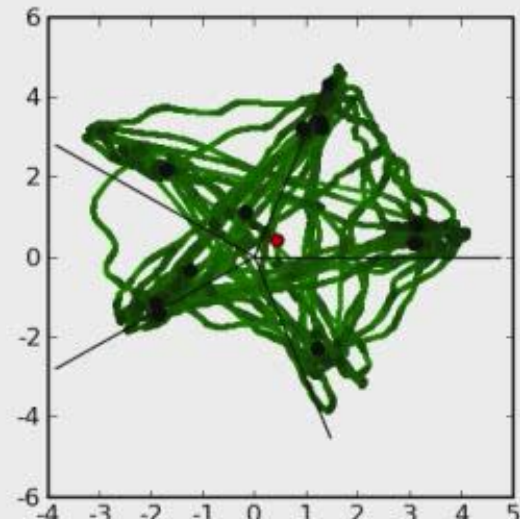
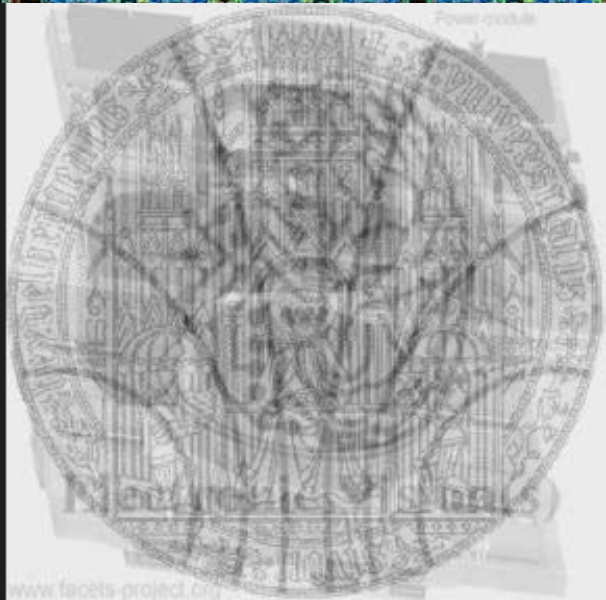
Attractor memory (NEST): attractor dwell times



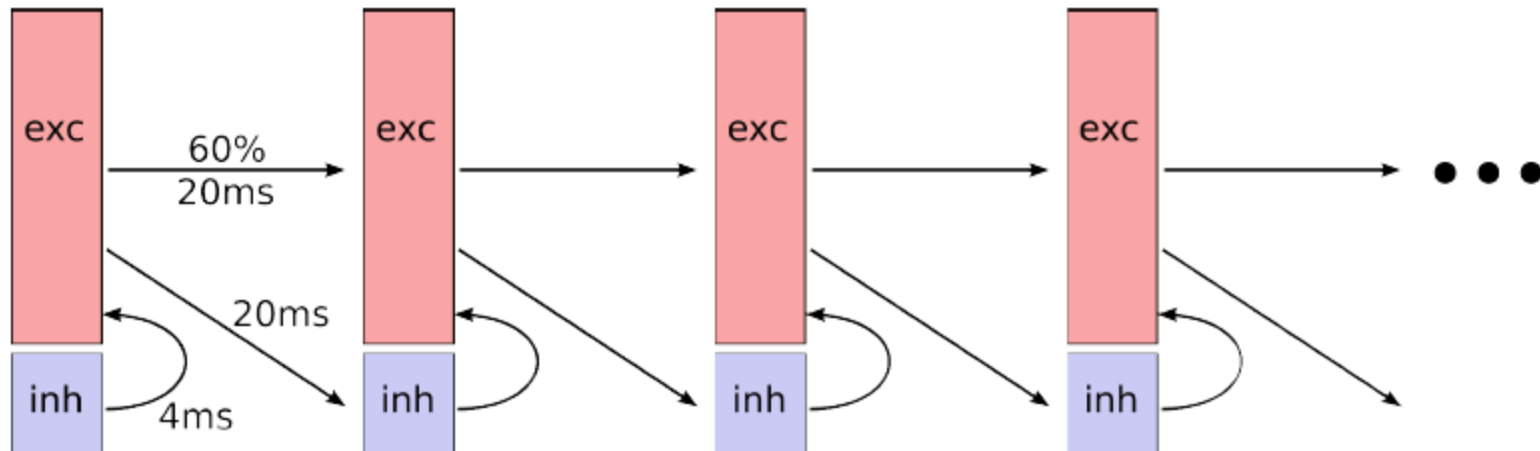
Attractor memory (NEST): threshold voltage



Attractor memory (NEST): pattern completion

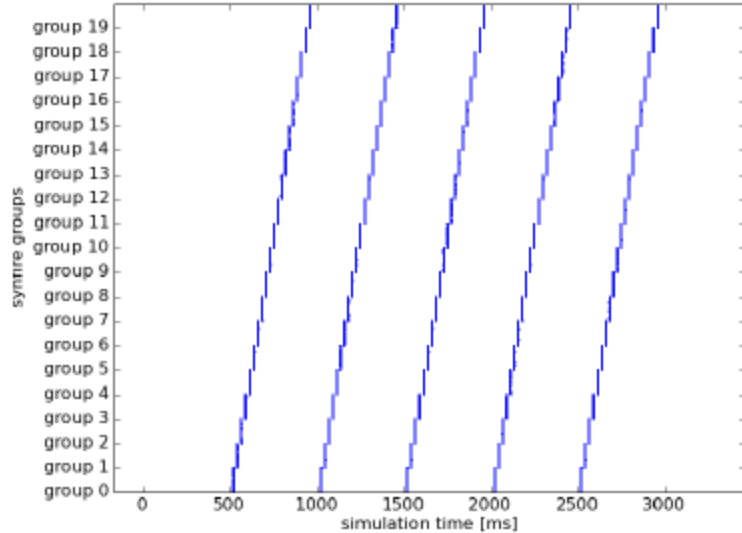


Synfire chain schematic

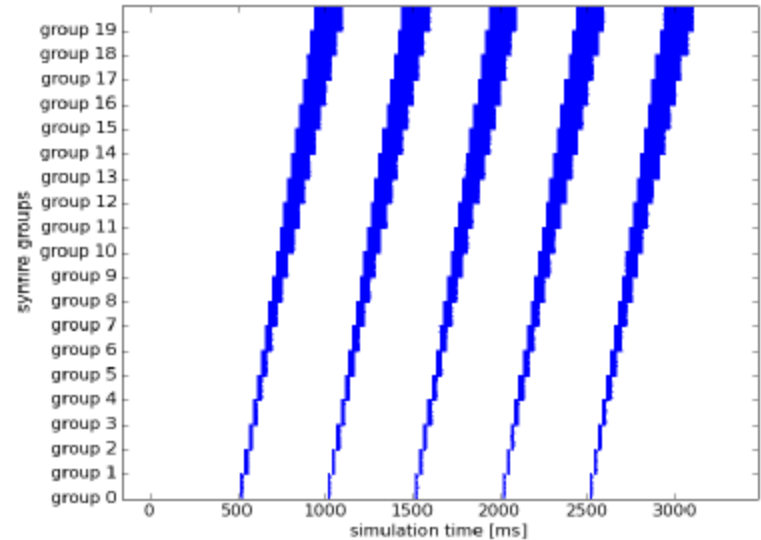


Synfire chain simulations (NEST)

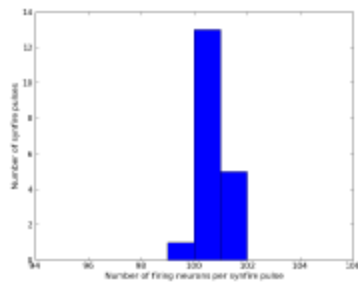
Synfire chain model (acc. to INCM-CNRS) with feed-forward inhibition



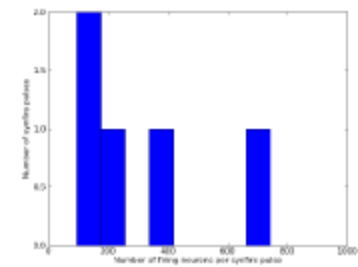
Synfire chain model (acc. to INCM-CNRS) without feed-forward inhibition



With feed-forward inhibition



Without



Technical infrastructure and visibility

- Wiki documentation (internal)
- Full code available in SVN repositories
- Experimental results in BSCW repository
- Tele-conferences
- Publications