

# NEMO

A PARALLEL DISCRETE EVENT NEUROMORPHIC HARDWARE  
SIMULATION MODEL

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# OVERVIEW

- Neuromorphic Computing & Hardware
- Demand and Potential
- Introducing NeMo
- Experimental Setup and Results
- Future Work

# NEUROMORPHIC COMPUTATION

## NEW COMPUTING MODELS

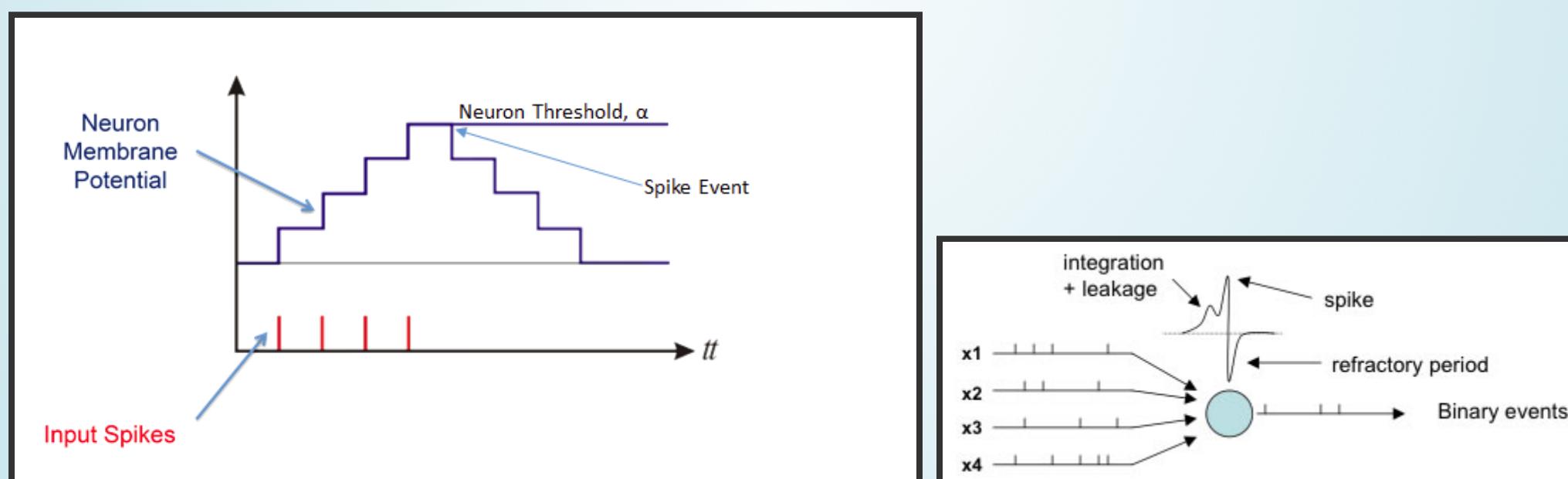
Neuromorphic computing is based on spiking neural networks

- Based on Artificial Neural Network (ANN) concept
- Third generation neuron simulations
- Designed to simulate biological functions
  - Initially not for general computation

# SPIKING NEURONS

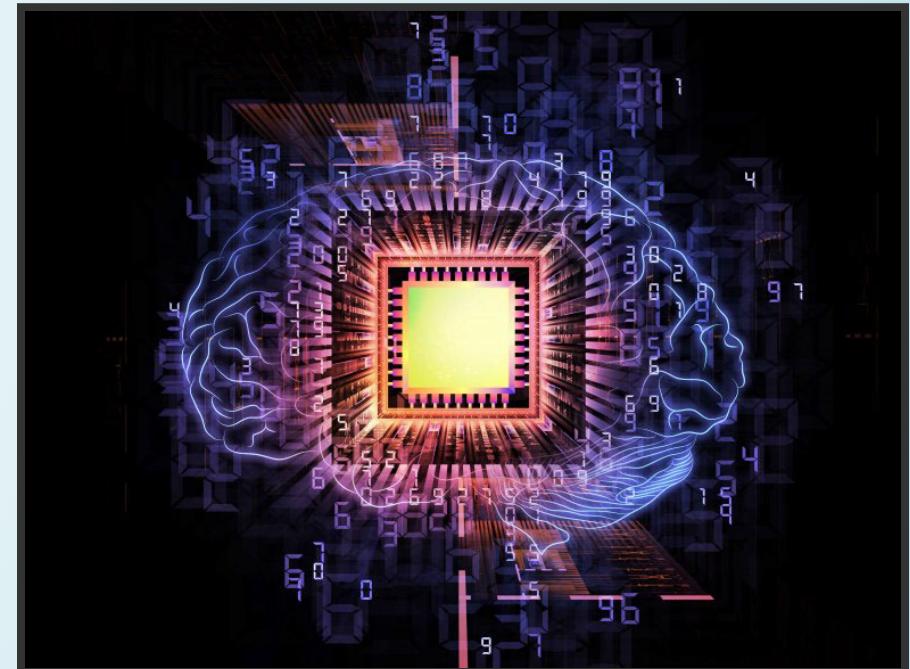
Unlike a traditional Neural Network Neuron:

- Include concept of time
  - State changes over a series of time-steps.
- Neurons do not need to fire at every time-step  $t$
- Activation level is increased with input spikes



# SPIKING NEURON MODELS

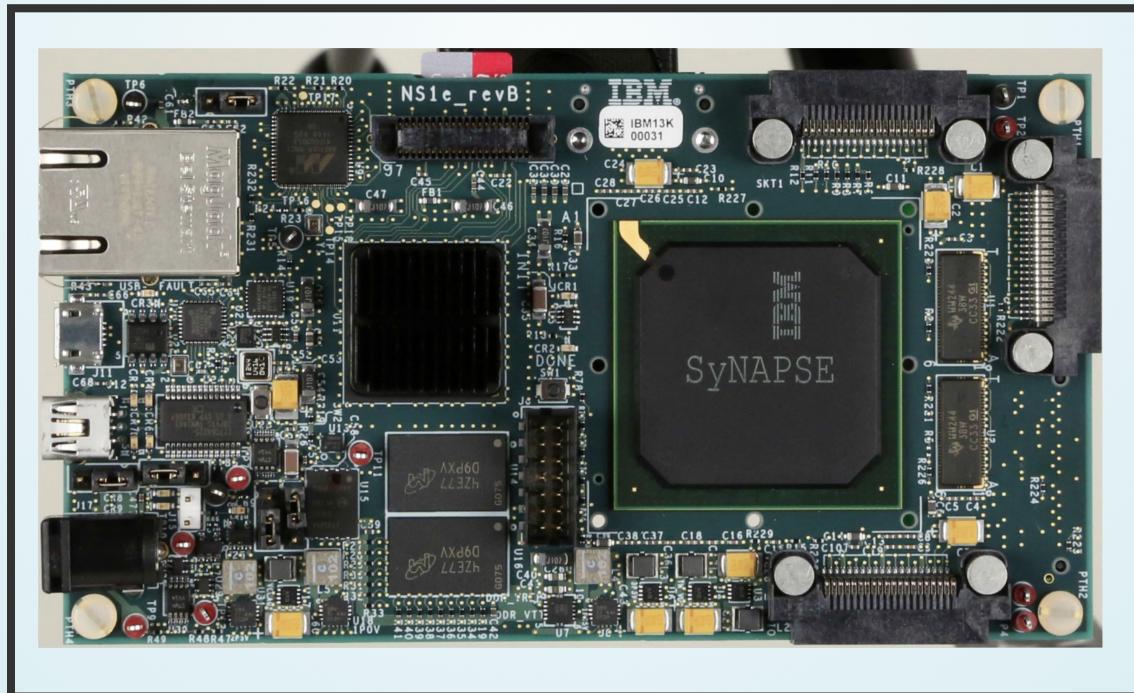
- First scientific model developed in 1952
- Commonly used models include:
  - Hodgkin and Huxley Model
  - Integrate and Fire
  - Leaky Integrate and Fire
  - Izhikevich's Simple Neuron Model
  - Many More



# NEUROMORPHIC HARDWARE

"Non von Neumann" computation

- Does not operate using traditional algorithms
- Tiny power requirements:  $\lesssim 60mW$
- Great at data classification



# DEMAND AND POTENTIAL

- Supercomputers are Transitioning to “Fat Nodes”
- Accelerator cards are becoming increasingly important
  - GPU, Intel PHI
- Neuromorphic hardware excels at pattern recognition
- Potential for managing power, predicting errors, and monitoring application performance.
- Need a model to simulate various neuromorphic hardware designs
- We wish to simulate neuromorphic hardware operating within a supercomputer simulation.

# NEMO IMPLEMENTATION

## NEMO FEATURES

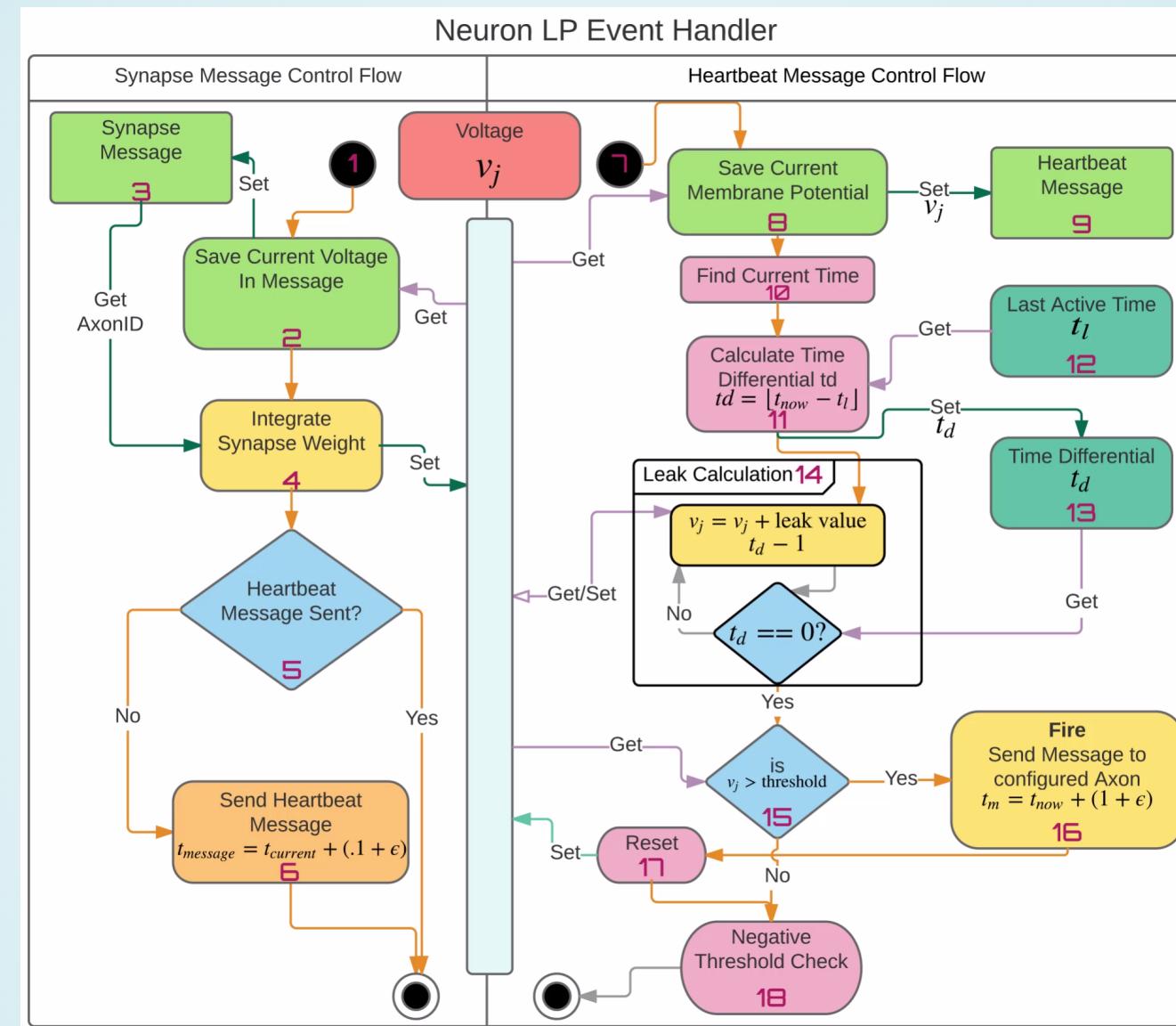
- Provides an open framework for simulation of new hardware designs:
  - One neuron per core to thousands of neurons per core
  - Weighted synapses
  - Different spiking neuron models
- Supports simulation of non-IBM Hardware
- "Heartbeat" messages that provide implicit synchronization
- Message fanout reduction through message routing
- Reverse computation for optimistic scheduling
- Supports very large simulation models

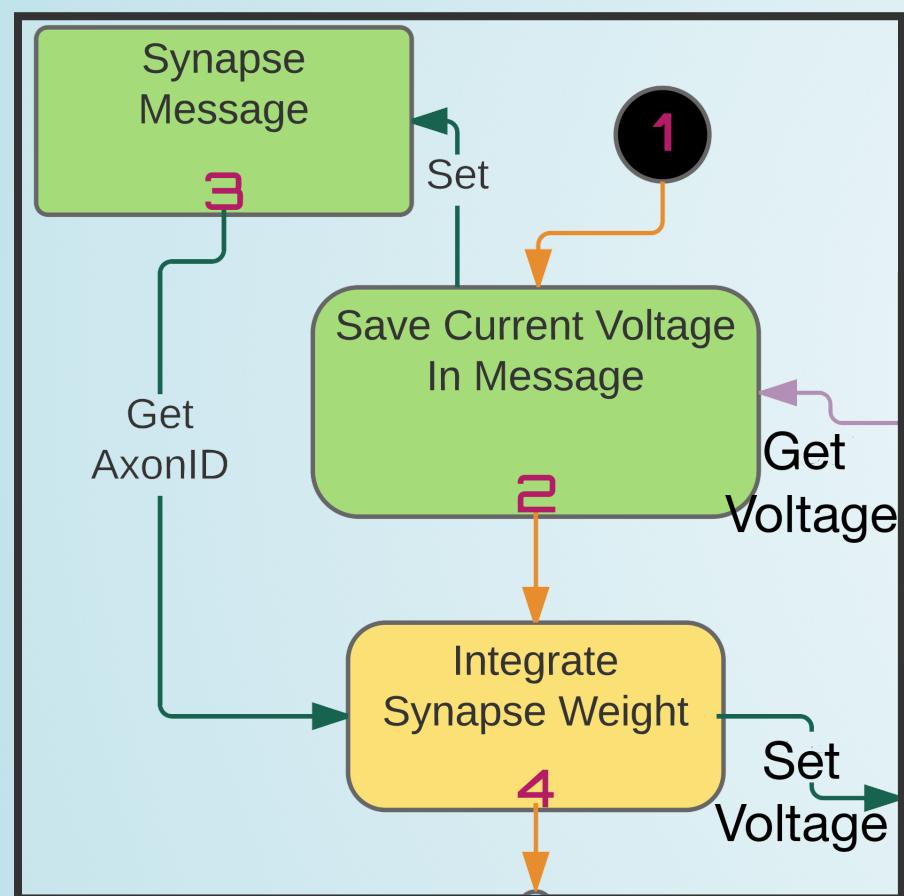
## NEMO FEATURES

- Implemented using ROSS (Rensselaer's Optimistic Simulation System)
- ROSS provides optimistic and conservative parallel discrete event simulation
- NeMo functions best in optimistic mode

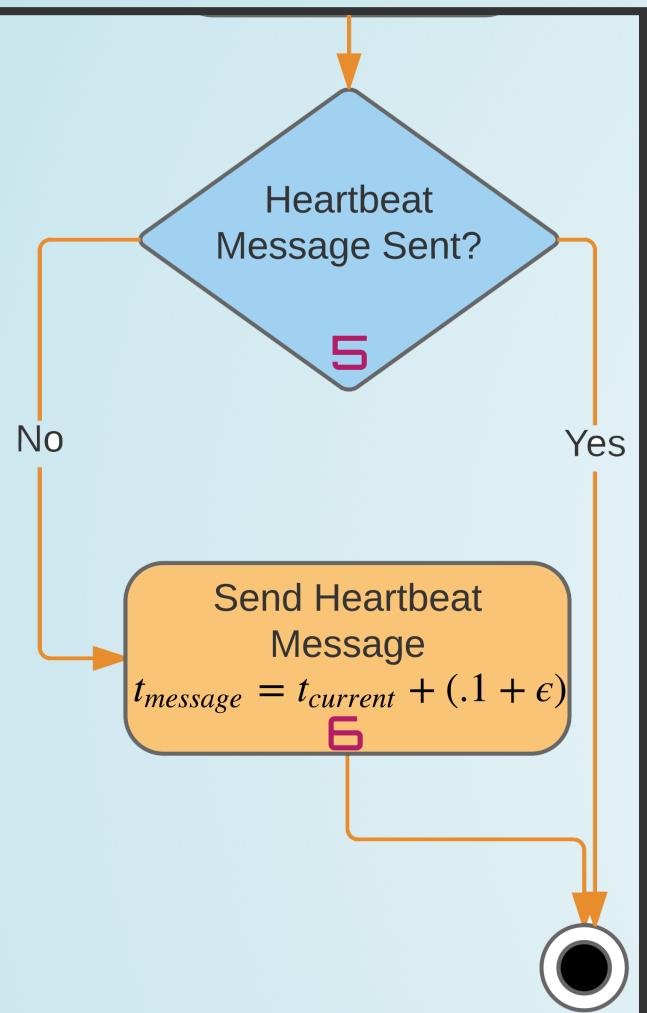
## COMPARED WITH IBM COMPASS

- NeMo is open source and freely available to use
- NeMo is able to simulate non-IBM hardware
- NeMo supports various hardware architectures - IBM Blue Gene as well as Intel



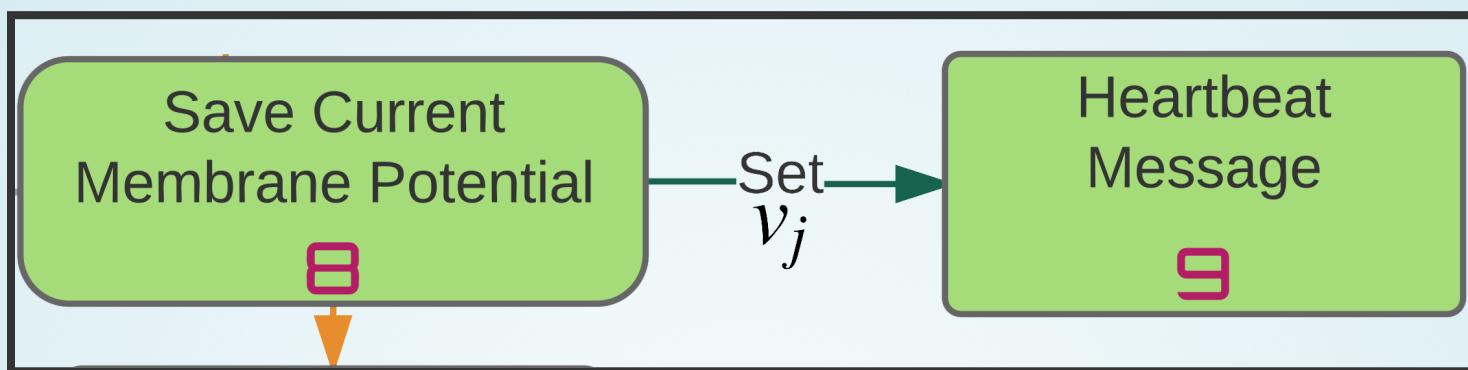


- Synapses send event messages to the neuron
- Neurons integrate the synapse weight at this point
- Previous voltage is saved in message



## Synapse Message Part 2

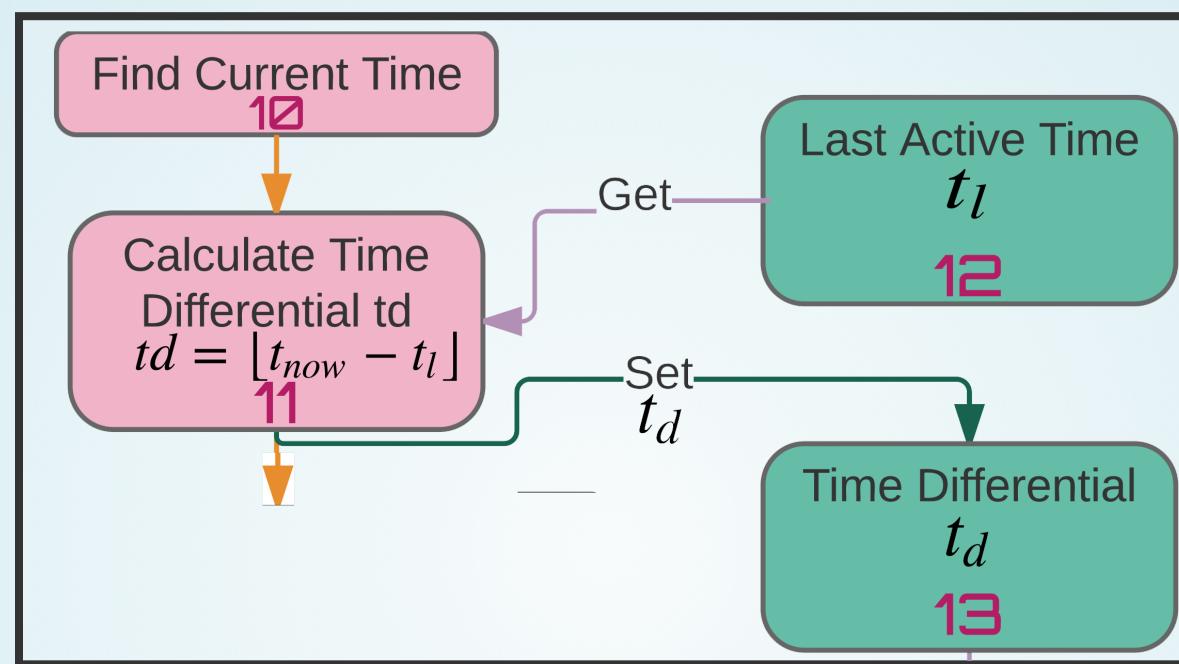
- Neurons check to see if a "heartbeat" message has been sent
- Send heartbeat message if one has not been sent



## HEARTBEAT MESSAGE RECEIVED

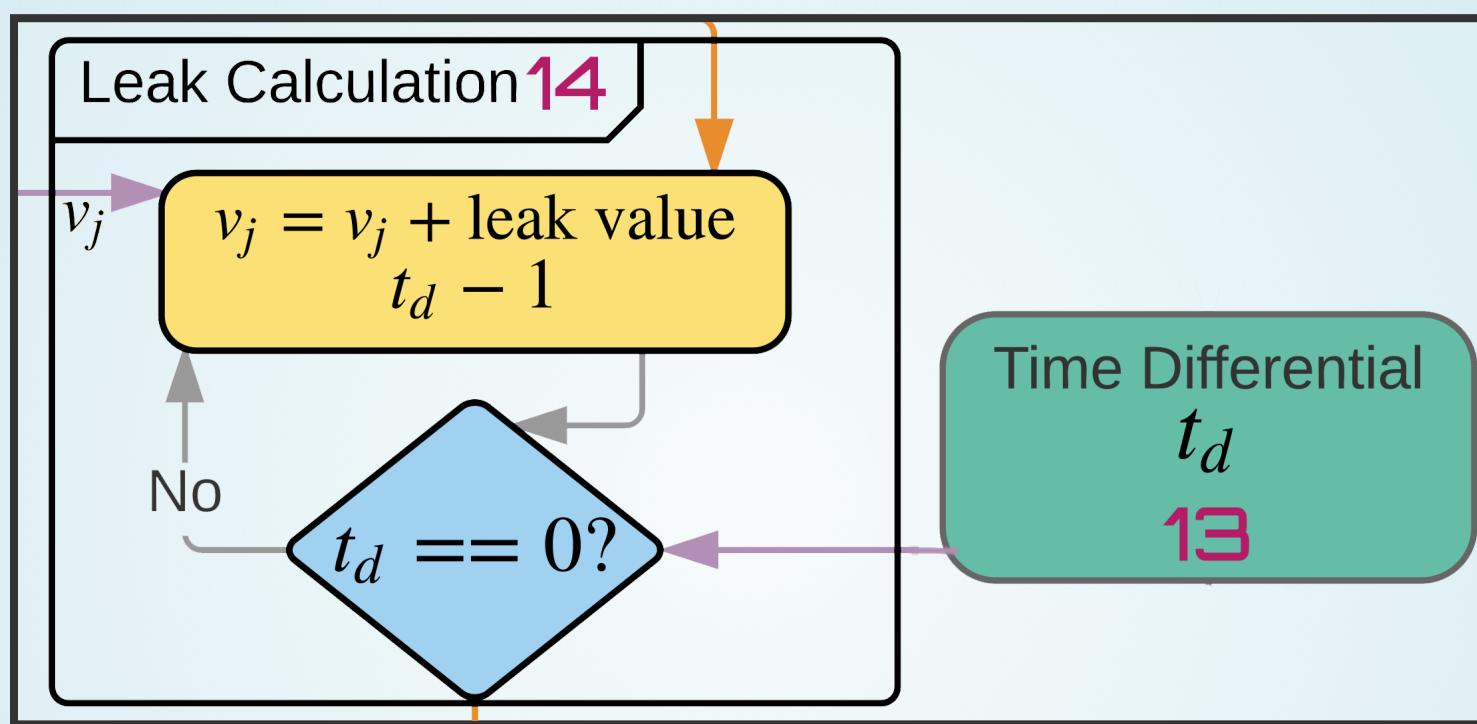
Neurons receiving a heartbeat message have either:

- Had synaptic activity this tick
- Have the potential to be self-firing
  - Positive leak
  - Negative leak with specific reset values
  - And More!
- Neurons save current membrane potential in the message



Neurons then:

- Retreive current time
- Calculate the number of ticks since last activating
- Store this as  $t_d$

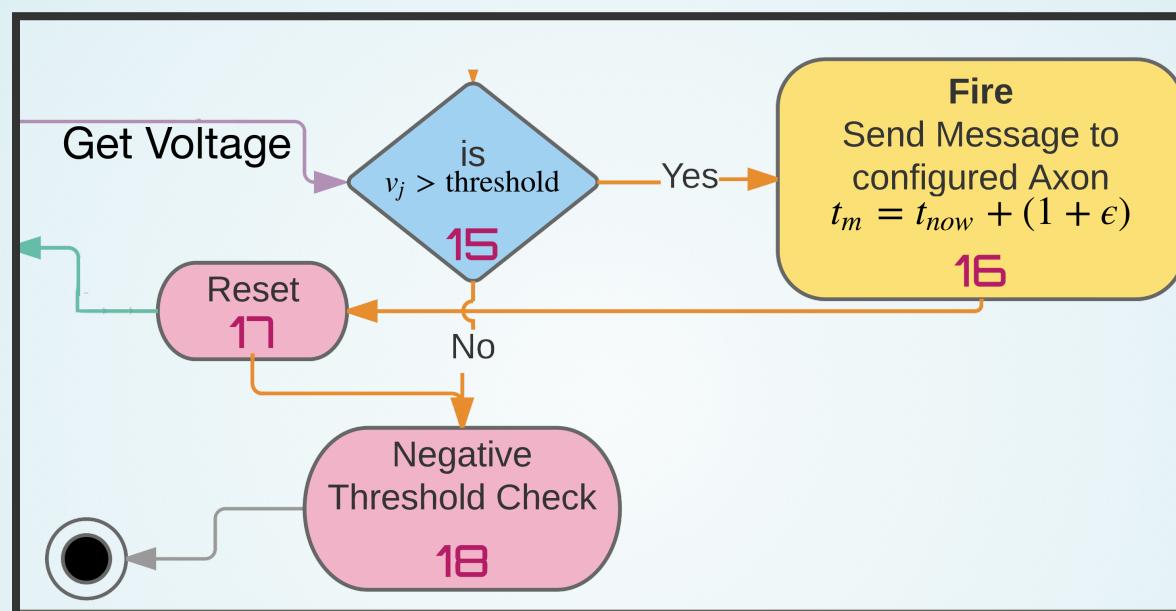


Leak calculation:

```

while (td > 0) {
    voltage = voltage - leakFunction();
    td--;
}
  
```

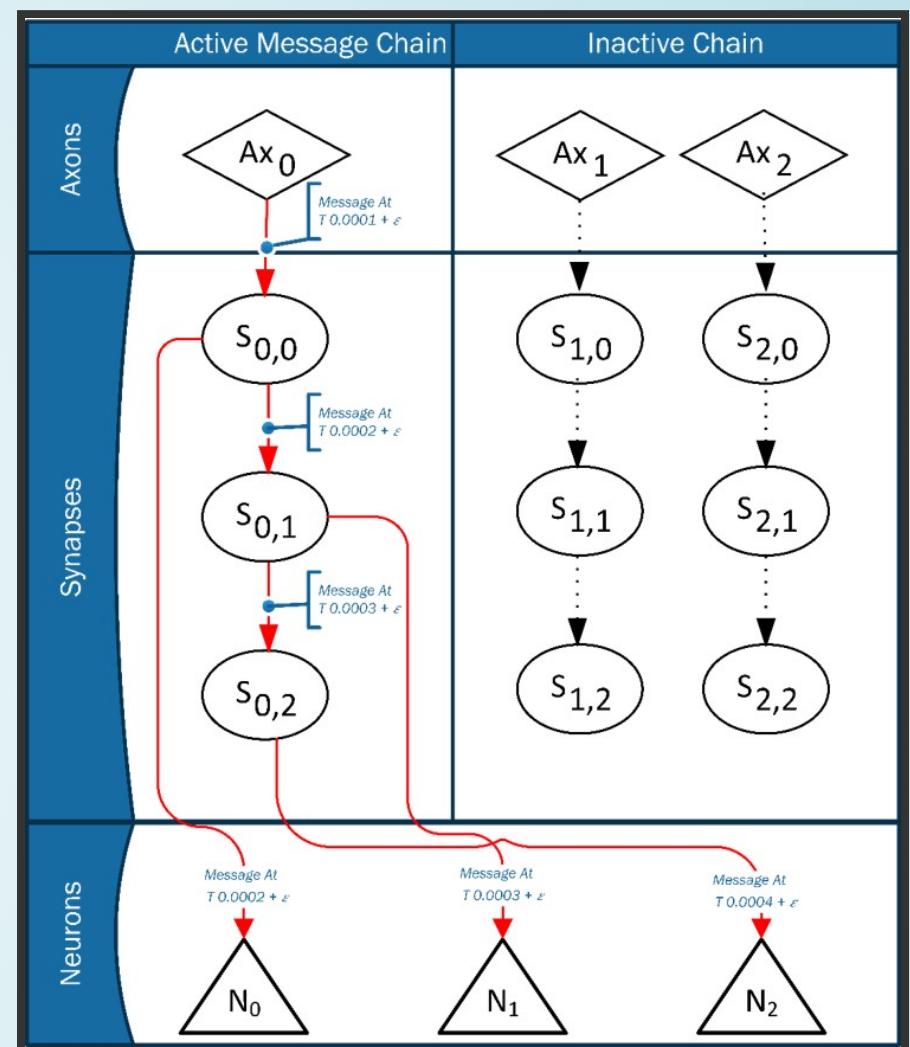
## THRESHOLD, RESET & FIRE



- Check if voltage ( $V_j$ ) is greater than the threshold ( $\alpha$ )  
 $V_j \geq \alpha$
- Send fire message
- Reset or check negative threshold

# NEMO EVENT FLOW

Given an event at  $Ax_0$ (axon 0):

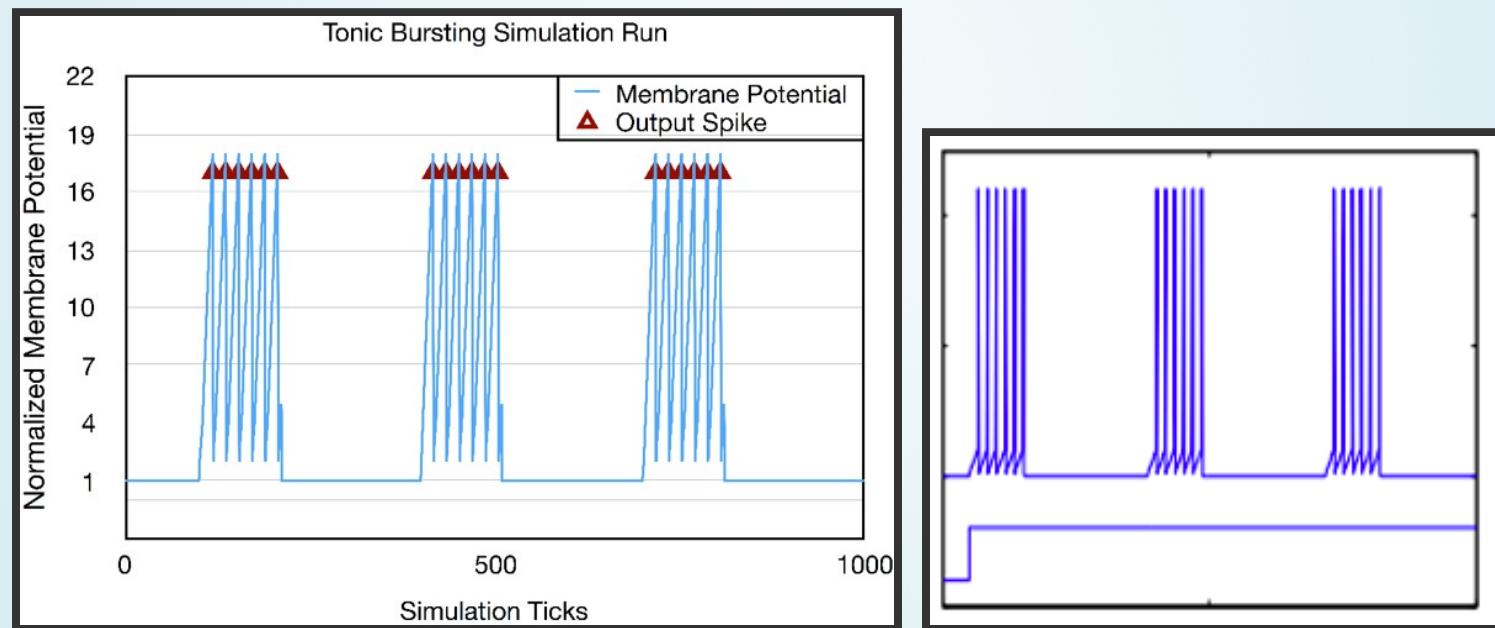


- Synapse  $S_{0,0}$  through  $S_{0,2}$  activate
- All neurons integrate
- Neurons that have weight 0 for an axon do not send heartbeat

# NEMO EXPERIMENTAL RESULTS

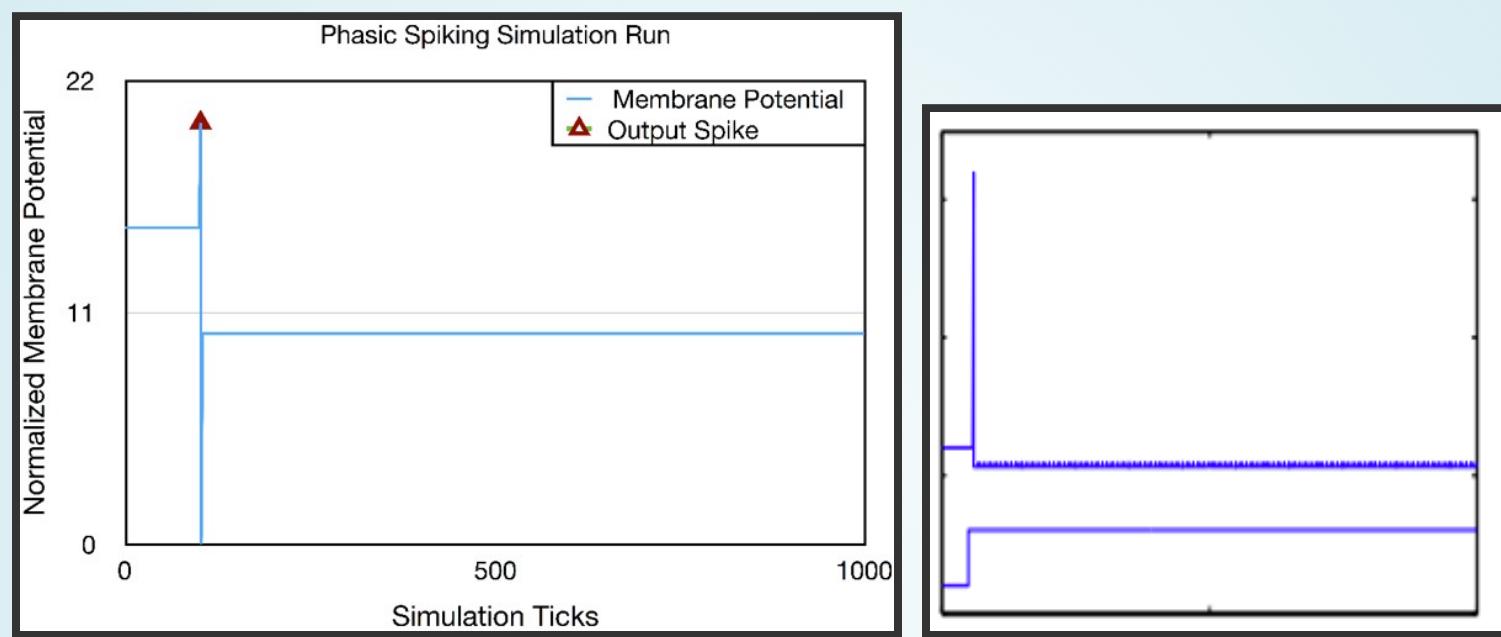
## MODEL VALIDATION

- The IBM Model is able to reproduce Izhikevich's Interesting Biological Neurons
- We recreated two of these models using NeMo and IBM's published parameters



### *Tonic Bursting Function Results*

These are two of Izhikevich's Neuron Models, reproduced by Cassidy et al.



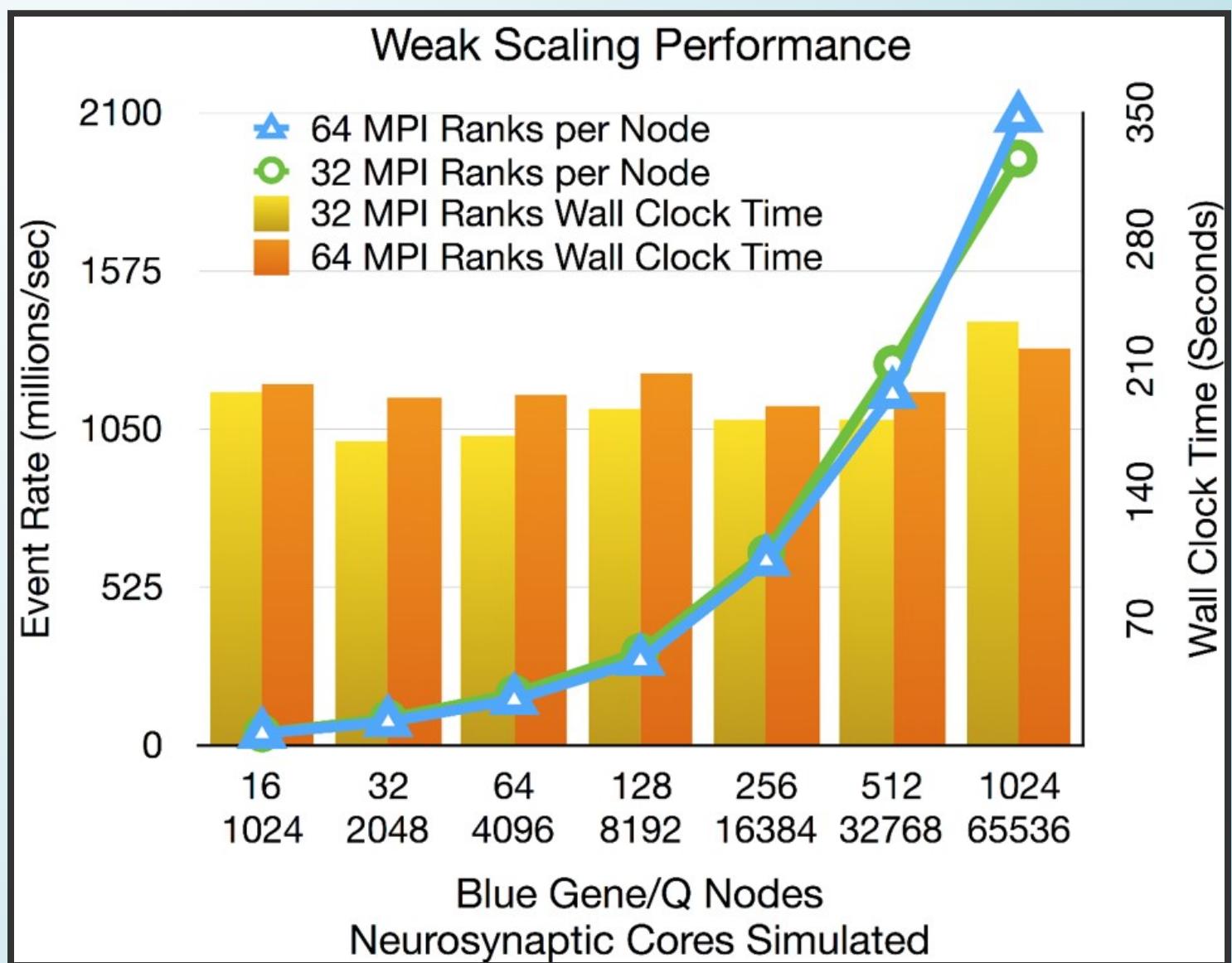
### *Phasic Spiking Results*

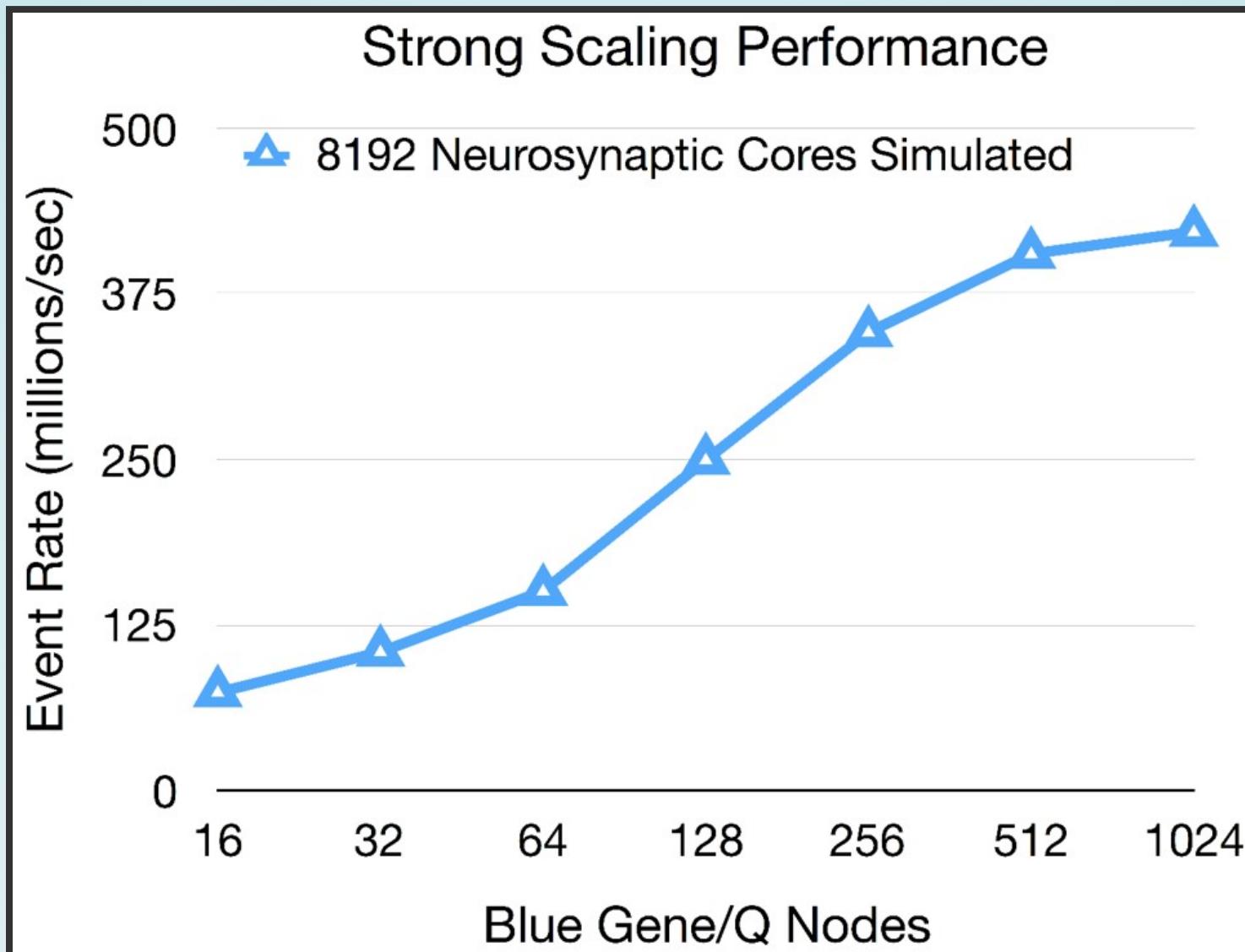
These are two of Izhikevich's Neuron Models, reproduced by Cassidy et al.

## PERFORMANCE EXPERIMENTS

- Evaluation System
  - Center for Computational Innovations (CCI) IBM Blue Gene/Q
    - 64 Hardware Threads / Node
    - 16 GB Memory per Node
- Ran a randomized network with 80% remote (off-core) probability
- Neurons were organized in an "identity matrix" of 1-1 activations:
  - When neuron  $n$  received a message from axon  $n$ , it would spike to a random axon.

- NeMo Weak Scaling Performance
- Running a randomly connected network
- 80% remote message probability
- Generally excellent results





- Strong scaling shows initially excellent results
- The simulation begins to lose parallelism
- Larger simulation runs should improve strong scaling

# FUTURE WORK

The next steps for NeMo include:

- "Super Synapse" - Reduced memory/message footprint
- Integration with CODES
- JSON neuromorphic core design support
- Pythonic I/O for integration with spiking neural network tools

Thank You!  
References