PROGRAMMING NEUROGRID I THE PYTHON WAY BEN VARKEY BENJAMIN PEIRAN GAO **KWABENA BOAHEN**

Practical Stuff

- One Neurogrid system will be available for class with at least 16 chips--a million neurons!
- Access to Neurogrid will be time-shared through a sign up website after next Wednesday
- Available time slots depending on lab members' schedule;
- Both hardware and software are under constant state of development:
 - Expect frustrating bugs and crashes
 - TA and other lab members will be there to help
- Goal is to make the experience mutually beneficial for you and the lab

Topics for Today:

- Synapse, Arbor, Soma and Neuron Models
- Hierarchical network constructs
- Making vertical (i.e., topographic) connections
- Mapping network to hardware
- Saving data from hardware (basic)

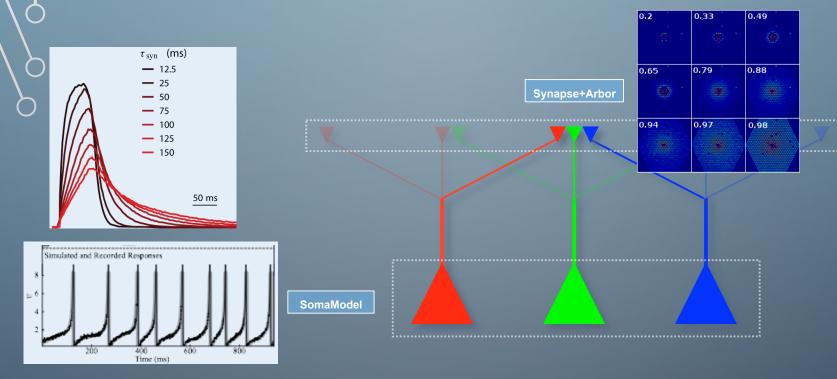
Topics for Next Time:

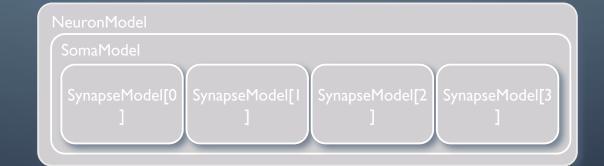
- Making horizontal connections
- Sending stimulus
- Running experiments
- Possibly more....

From NEST to Neurogrid

	NEST	Neurogrid
Models	Identical	Heterogeneous
Network	Flat, small	Hierarchical, large
Speed	Size-dependent	Real-time

Hierarchical NeuronModel



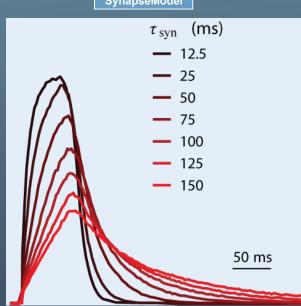


Building Synapse and Arbor

>>> syn = Synapse("cond_syn", {"erev": .1, "g_max": 40, "tau_syn": 2e-3, "t_xmt": 5e-3, "lambda": .2)

Default parameter (shown) used if omitted

SynapseModel



Inhibitory: 0<erev<.5 Shunting: .

5<erev<1.5

Excitatory: 1.5<erev

Default: erev=0.1

	O y napour a son
	Model+
0.2	0.33

0.2	0.33	0.49
400	(8)	0
0.65	0.79	0.88
0.94	0.97	0.98

$$I_{syn} = g_{syn}(t)(E_{rev} - V_m)$$

$$\tau_{syn}\dot{g}(t) = -g(t) + \sum_{i} pulse(t-t_i)$$

$$w(r) = \frac{1}{Z} \frac{\lambda'}{\sqrt{r}}, \quad 0 < \lambda < 1$$

Normalized such that total conductance equals g_{syn}(t)

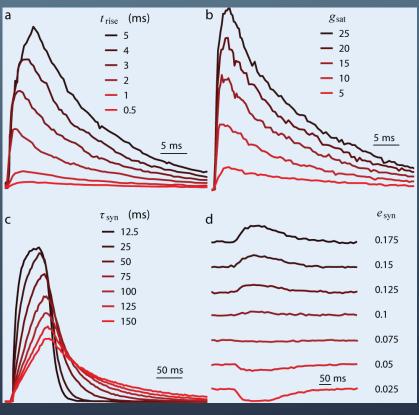
SYNAPSE DYNAMICS

$$\frac{\delta(t-t_k)}{g_{int}p_{rise}(t-t_k)} = \frac{1}{g_{int}-t_{rise}}$$

$$g_{int}(t)$$

- The cleft sets g_{syn}(t)'s rise time, t_{rise}, by producing a unit pulse, p_{rise}(t), with width t_{rise}.
- The receptor sets $g_{syn}(t)$'s steady-state value, g_{sat} , by scaling $p_{rise}(t)$, and its time constant, τ_{syn} .





Ben's data

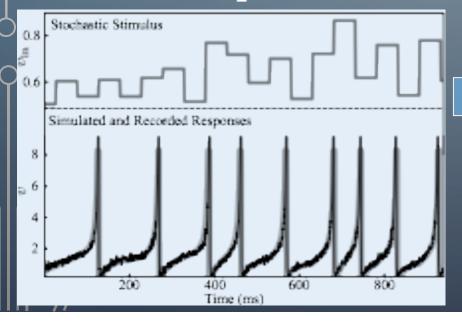
Building Soma and Neuron

Default parameters (shown) used if omitted

```
>>> som_quad = Soma("quadratic", {"x0": .9, "tau": 10e-3, "tau_ref": 2e-3})
>>> nrn_quad = Neuron("quad_neuron", som_quad)
>>> som_cub = Soma("cubic", {"x0": .385, "tau": 10e-3, "tau_ref": 2e-3})
>>> nrn_cub = Neuron("cub_neuron", som_cub)
>>> som_quad_ad = Soma("quadratic_adaptive", {"x0": .5, "tau": 10e-3, "tau_ref": 2e-3, "g_inf": 0.02})
>>> som_cub_ad = Soma("cubic_adaptive", {"x0": .385, "tau": 10e-3, "tau_ref": 2e-3, "g_inf": 0.02})
```

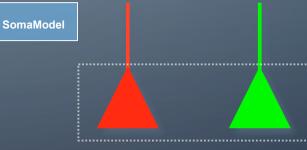
Quadratic Model:

$$\tau \dot{x} = -x + \frac{1}{2}x^2 + x_0, t_{ref}$$



Cubic Model:

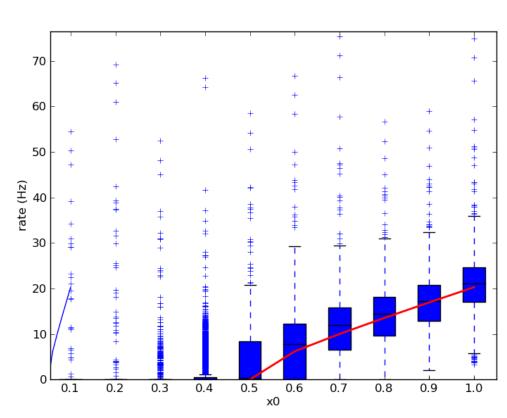
$$\tau \dot{x} = -x + \frac{1}{3}x^3 + x_0, t_{ref}$$



Bring the Neuron Model Together

```
>>> # Making inhibitory synapse and arbor
>>> syn_i = Synapse("cond_syn", {"erev": .5, "tau_syn": 10e-3, "lambda": .8})
>>>
>>> # Making excitatory synapse and arbor
>>> syn_e = Synapse("cond_syn", {"erev":2.5, "tau_syn": 2e-3, "lambda": .8})
>>>
>>> # Making quadratic SomaModel
>>> som quad = Soma("quadratic", {"x0": .9, "tau": 10e-3})
>>>
>>> # Add arbors to the soma
>>> som quad.AddASynapse(arb e)
>>> som_quad.AddSynapse(arb_i)
>>>
>>> # Putting together the Neuron model
>>> nrn_quad = Neuron("quad_neuron", som_quad)
```

Driving neuron with input current

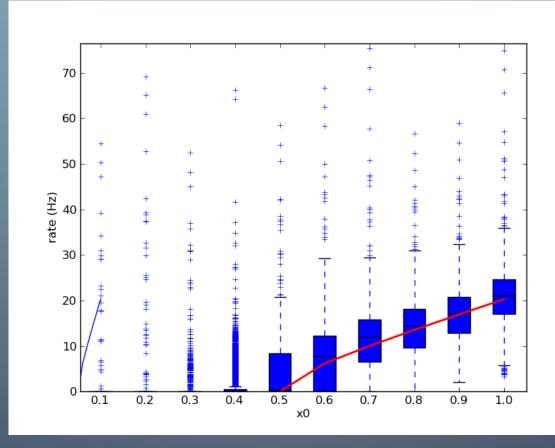


Neurons are neterogeneous

- Peiran's data
- Mean rate matches theory (red)
- Rate increases sublinearly (square root for quadratic)

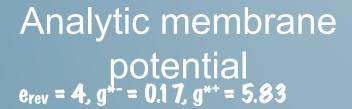
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Show firing rate distribution and biological comparison



- Starts spiking at x0>0.5
- Rate increases sublinearly (square root for quadratic)
- Neurons are heterogeneous

DRIVING NEURON WITH SYNAPSE



10

15

Gsyn

5.82

5.75

3.00

0.25

0.18

20

Quadratic positivefeedback can't
overcome
conductances larger
than a maximum value (g^{*+})

Such large shunts hold the membrane potential close to e_{rev} , preventing spiking even when $e_{rev} \gg 1$

Next week Neurogrid III lecture on building networks