

Code generation in Brian 2 From mathematical model descriptions to executable code

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Who is Brian?

- A neural simulator, originally started by Romain Brette and Dan Goodman in 2007 at ENS Paris
- Focuses on ease-of-use and flexibility, while being "reasonably fast"
- Written in Python, free-and-open-source
- Brian2: Rewrite of core parts of Brian, more consistent use of string-based model descriptions and code generation

Goodman DF and Brette R (2009). The Brian simulator. Frontiers in Neuroscience

String-based model descriptions



Mathematical description:

 $E_L = -70 \text{ mV}$ $v_{\text{th}} = -50 \text{ mV}$ $v_{\text{r}} = E_L$

$$\frac{\tau}{dt} = 20 \text{ ms}$$

$$\frac{dv}{dt} = -\frac{(v - E_L - I)}{\tau}$$

(v and I have units of volt)

When $v > v_{\text{th}}$: $v \leftarrow v_r$

Brian description:

- $E_L = -70*mV$ $v_r = E_L$ $v_th = -50*mV$ tau = 20*ms
- G = NeuronGroup(N,

1 1 1

```
dv/dt = -(v - E_L - I) / tau : volt
I : volt
''',
threshold='v>v_th',
reset='v=v_r')
```



Brian description:

E_L = -70*mV v_r = E_L v_th0 = -50*mV tau = 20*ms tau_th = 50*ms sigma = 4*mV

Modelling synaptic learning

Brian description:

Connecting neurons

String expressions referring to indices **i** and **j**:

- one-to-one connectivity: S.connect('i == j')
- local neighbourhood:
 S.connect('abs(i j) < 5')
- Random connectivity, avoiding self-connections: S.connect('i != j', p='1./(sigma*sqrt(2*pi))* ↔ exp(-(i - j)**2/(2*sigma**2)')
- Connecting cells depending on properties (stored as state variables): S.connect('sqrt((x_pre-x_post)**2) < 100*umetre')

Code generation

Code generation

- Event-based updates (impact of an incoming spike on a neuron, reset after a spike, ...) are formulated as "abstract code"
- Continuous updates (neuronal dynamics, synaptic dynamics, ...) are formulated as differential equations
 → need to be numerically integrated

 \rightarrow yields abstract code

 Syntax and consistency checks (e.g. units) happen on abstract code

Abstract code



Uses **sympy** for symbolic mathematics

Interface to code generation

Abstract code

```
_v = dt*(-v + w)/tau_v + v
_w = -dt*w/tau_w + w
v = _v
w = w
```

Code generation

Indices {'v': 'n_idx', 'w': 'n_idx'}

Executable code generation (C++)

for(int n idx=0; n idx< num neurons; n idx++)</pre> double v = ptr array neurongroup v[n idx]; double w = _ptr_array_neurongroup w[n idx]; **const** double w = -(dt) * w / tau w + w;**const double** v = dt * (-(v) + w) / tau v + v;w = w;v = v;ptr array neurongroup v[n idx] = v;ptr array neurongroup w[n idx] = w;

general template

Executable "Code object"

Abstract code

Abstract namespace

Indices

C++ code

```
for(int _n_idx=0; _n_idx<_num_neurons; _n_idx++)
{
    double v = _ptr_array_neurongroup_v[_n_idx];
    double w = _ptr_array_neurongroup_w[_n_idx];
    const double _w = -(dt) * w / tau_w + w;
    const double _v = dt * (-(v) + w) / tau_v + v
    w = _w;
    v = _v;
    _ptr_array_neurongroup_v[_n_idx] = v;
    _ptr_array_neurongroup_w[_n_idx] = w;
}</pre>
```

Namespace

Brian's "runtime" mode

Network.run:

Pseudocode, the actual run loop also takes care of # multiple clocks, etc.

```
while clock.t < run_time:
    # self.objects contains objects such as
    # NeuronGroup, Synapses, StateMonitor, ...
    for obj in self.objects:
        # The "code objects" are the objects doing
        # the actual computations, e.g. the numerical integration,
        # thresholding, resetting, synaptic propagation, etc.
        for code_object in object.code_objects:
            code_object.run()</pre>
```

clock.t += clock.dt

Brian's "standalone" mode

We are already generating code for the core computational parts – why not go all the way?

- Add memory allocation in target language
- Add simulation loop in target language
 → full simulation runs independent of Brian
- Also a possible starting point for simulations that connect to special hardware, etc.

Standalone mode: usage

from brian2 import *

E_L = -70*mV v_r = E_L v_th0 = -50*mV tau = 20*ms tau_th = 50*ms sigma = 4*mV N = 100

Standalone mode: usage

```
from brian2 import *
from brian2.devices.cpp_standalone import *
set device('cpp standalone')
```

```
E_L = -70*mV

v_r = E_L

v_th0 = -50*mV

tau = 20*ms

tau_th = 50*ms

sigma = 4*mV

N = 100
```

Standalone mode: generated files

output

- brianlib
 - clocks.h
 - common_math.h
 - dynamic_array.h
 - network.cpp
 - network.h
 - spikequeue.h
 - synapses.h

code_objects

- neurongroup_resetter_codeobject.cpp
- neurongroup_resetter_codeobject.h
- neurongroup_stateupdater_codeobject.cpp
- neurongroup_stateupdater_codeobject.h
- meurongroup_thresholder_codeobject.cpp
- neurongroup_thresholder_codeobject.h
- spikemonitor_codeobject.cpp
- spikemonitor_codeobject.h

— main

- main.cpp
- objects.cpp
- objects.h

— results

- _array_neurongroup_i
- ______array_neurongroup_I
- _array_neurongroup__spikespace
- _array_neurongroup_v
- _____array_neurongroup_v_th
- _____array_spikemonitor__count
- _dynamic_array_spikemonitor__i
- spikemonitor_codeobject_i
- spikemonitor_codeobject_t

static_arrays

__static_array_array_neurongroup_I

Summary: code generation

- Combine easy-to-use Python interface with efficient code
- Generating code doesn't mean to write a Python-to-X compiler: Model-specific code only uses a small subset of Python (assignments, arithmetic expressions)
- Allows to run scripts on devices where it couldn't run otherwise

Thanks for listening

Try it out (alpha version): https://pypi.python.org/pypi/Brian2

https://github.com/brian-team/brian2

Read the docs:

http://brian2.readthedocs.org

Discuss/ask/comment:

brian-development@googlegroups.com

General information about Brian

http://briansimulator.org

Stimberg, Goodman, Benichoux, and Brette. 'Equation-Oriented Specification of Neural Models for Simulations'. Frontiers in Neuroinformatics 8: (2014): 6

