



Integrating NeuroML 2 with PyNN, Brian & CSA

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Overview

- Quick introduction to cells & synapses in LEMS & NeuroML 2
- Incorporating PyNN standard cells into this framework
- CSA & PyNN -> NeuroML 2
- NeuroML 2 -> Brian
- Future interaction of NineML & LEMS

Example: Fitzhugh-Nagumo cell model

$$\begin{aligned}\dot{V} &= V - V^3/3 - W + I \\ \dot{W} &= 0.08(V + 0.7 - 0.8W)\end{aligned}$$

Simplified version of 4
variable HH model
2 state variables, 2 ODEs

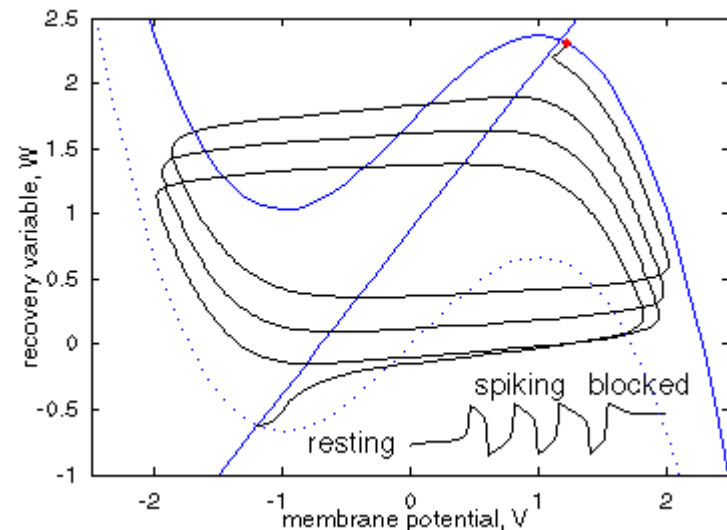


Image from Scholarpedia.org

Original model

$$\begin{aligned}\dot{V} &= V - V^3/3 - W + I \\ \dot{W} &= 0.08(V + 0.7 - 0.8W)\end{aligned}$$

Model expressed in LEMS (Low Entropy Model Specification language)

```
<ComponentType name="fitzHughNagumoCell" extends="abstractCell">
```

```
<Parameter name="I" dimension="none"/>
```

```
<Constant name="SEC" dimension="time" value="1s"/>
```

```
<Exposure name="V" dimension="none"/>
```

```
<Exposure name="W" dimension="none"/>
```

```
<Behavior>
```

```
<StateVariable name="V" dimension="none" exposure="V"/>
```

```
<StateVariable name="W" dimension="none" exposure="W"/>
```

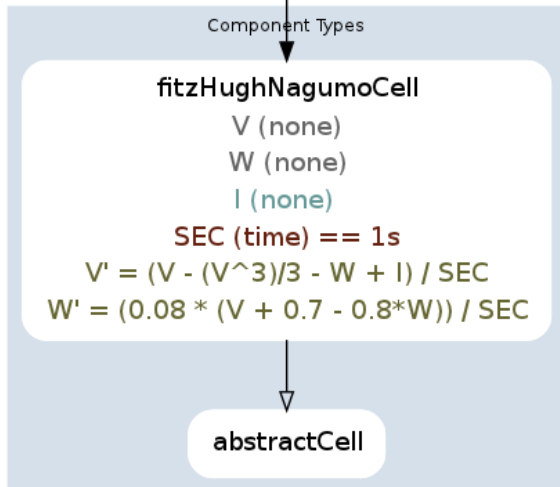
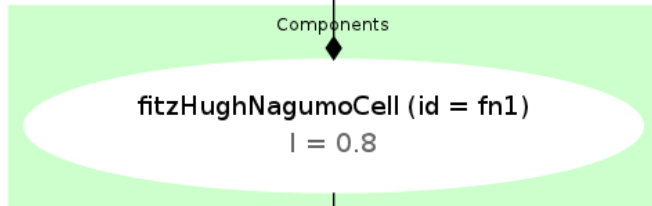
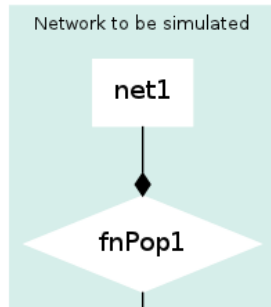
```
<TimeDerivative variable="V" value="(V - (V^3)/3 - W + I) / SEC"/>
```

```
<TimeDerivative variable="W" value="(0.08 * (V + 0.7 - 0.8*W)) / SEC"/>
```

```
</Behavior>
```

Component instance in NeuroML v2.0

A

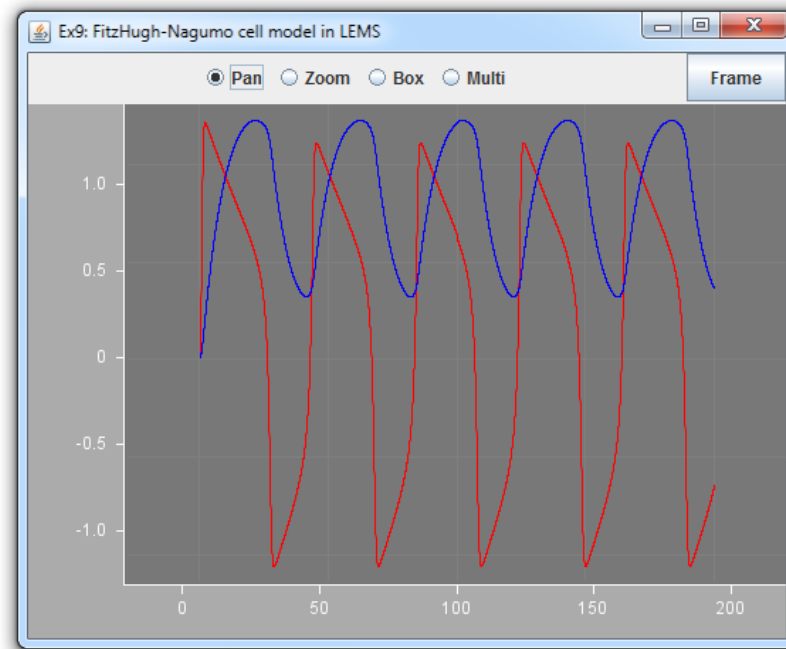


B

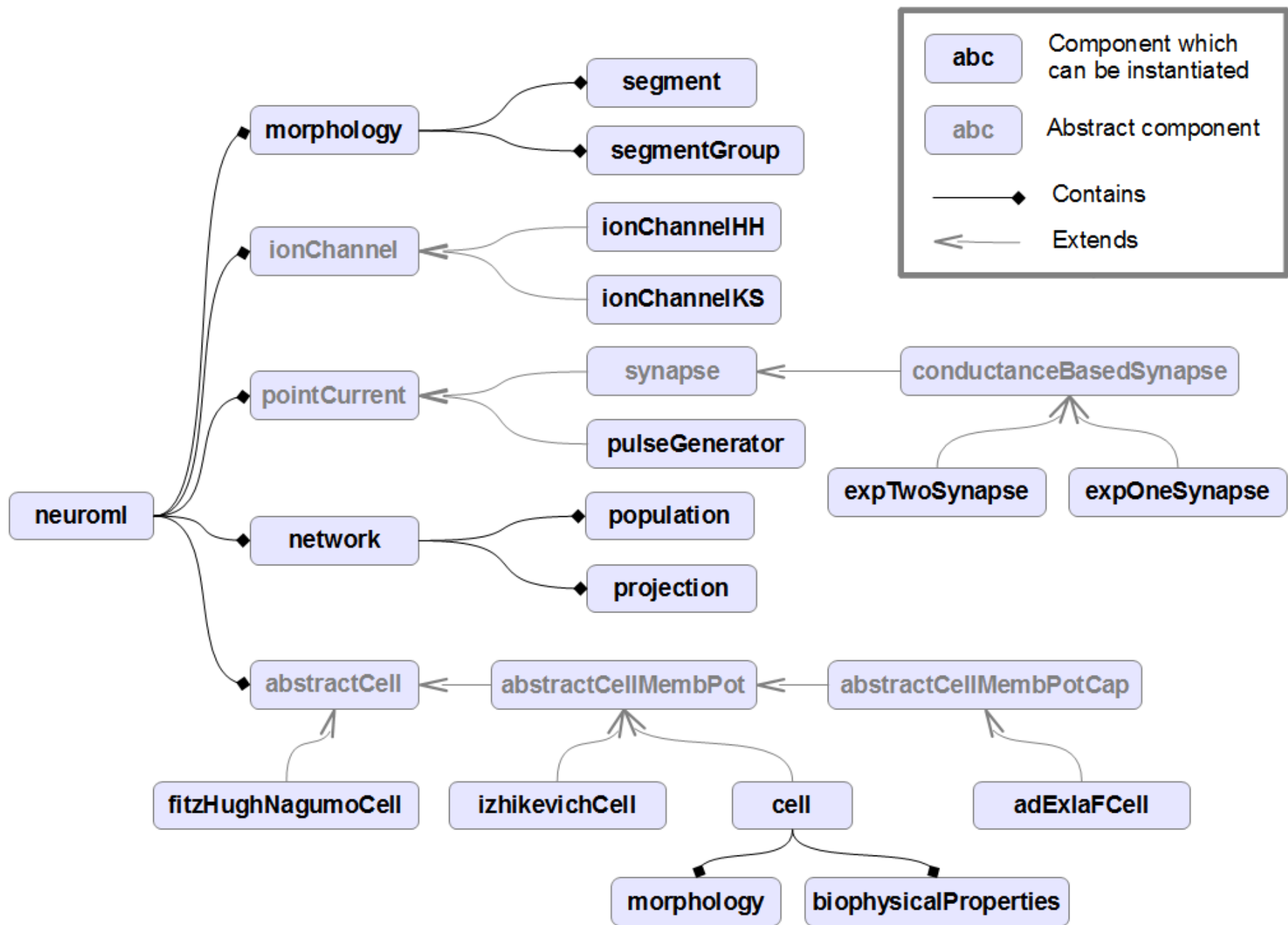
```
<fitzHughNagumoCell id="fn1" I="0.8"/>

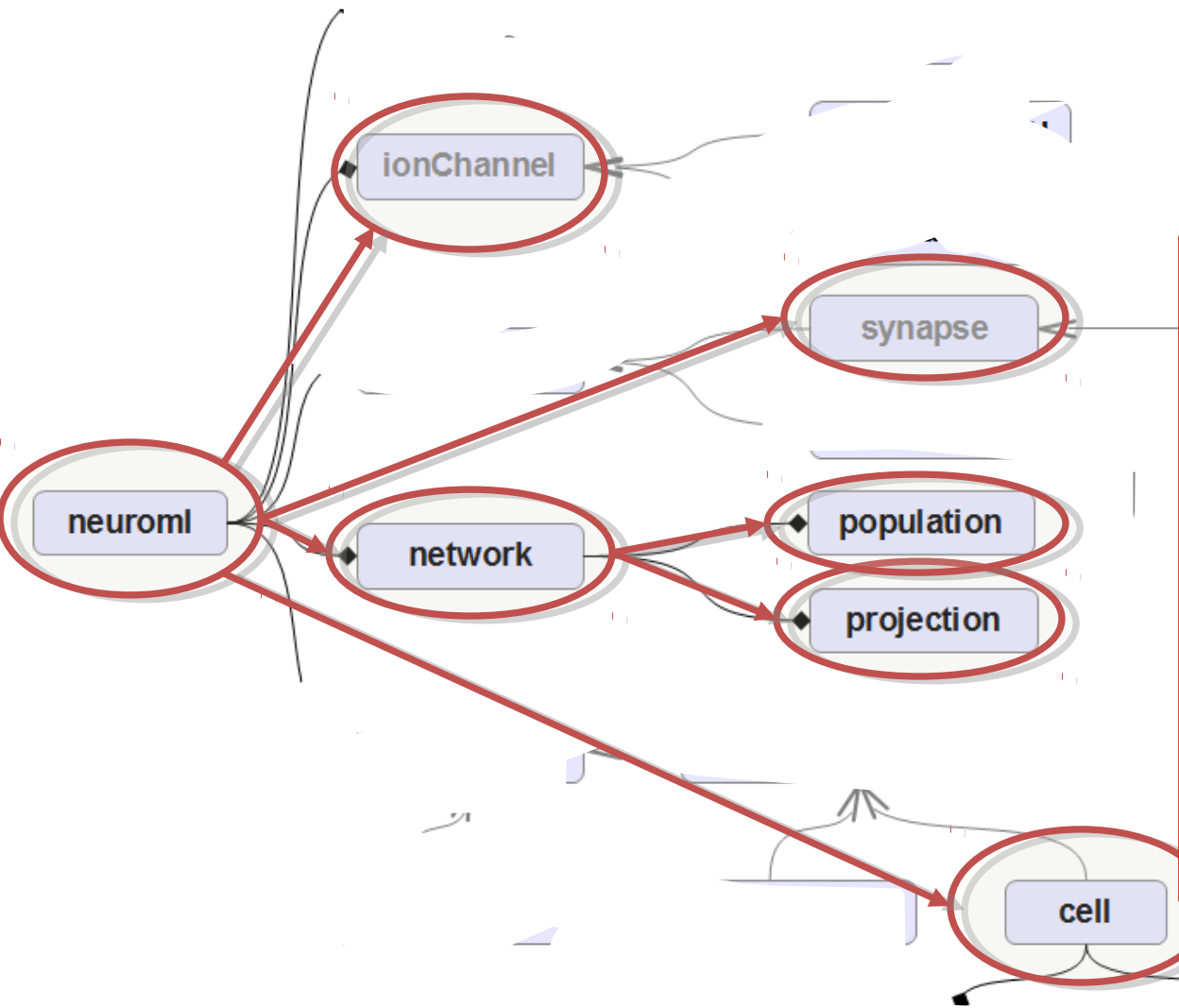
<network id="net1">
  <population id="fnPop1" component="fn1" size="1"/>
</network>
```

C



V
W





abc	Component which can be instantiated
abc	Abstract component
	Contains
	Extends

```

<neuroml>
  <ionChannel id="HH_Na" >
  </ionChannel>

  <synapse id="ExpSyn" >
  </synapse>

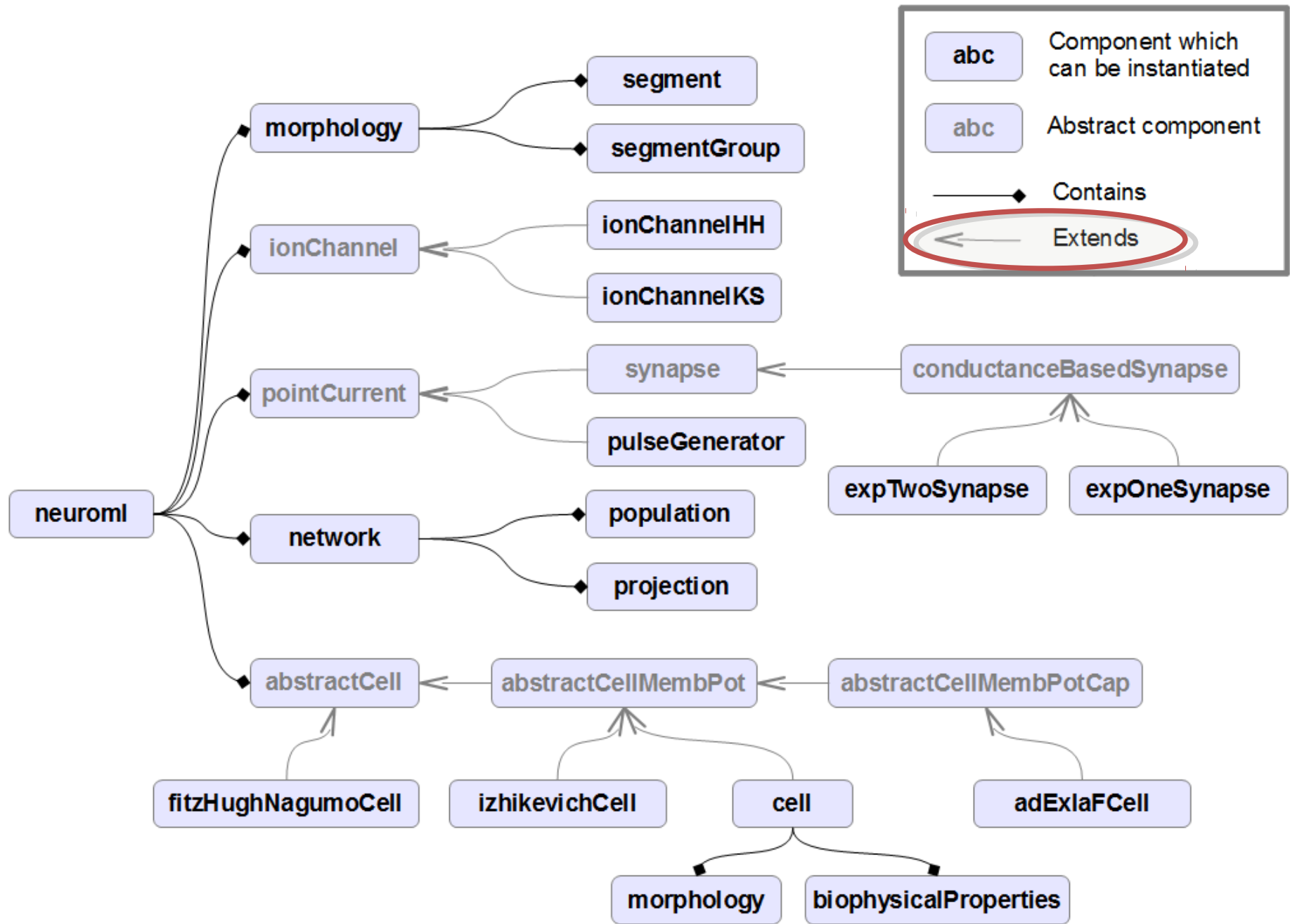
  <cell id="ExcCell" >
  </cell>

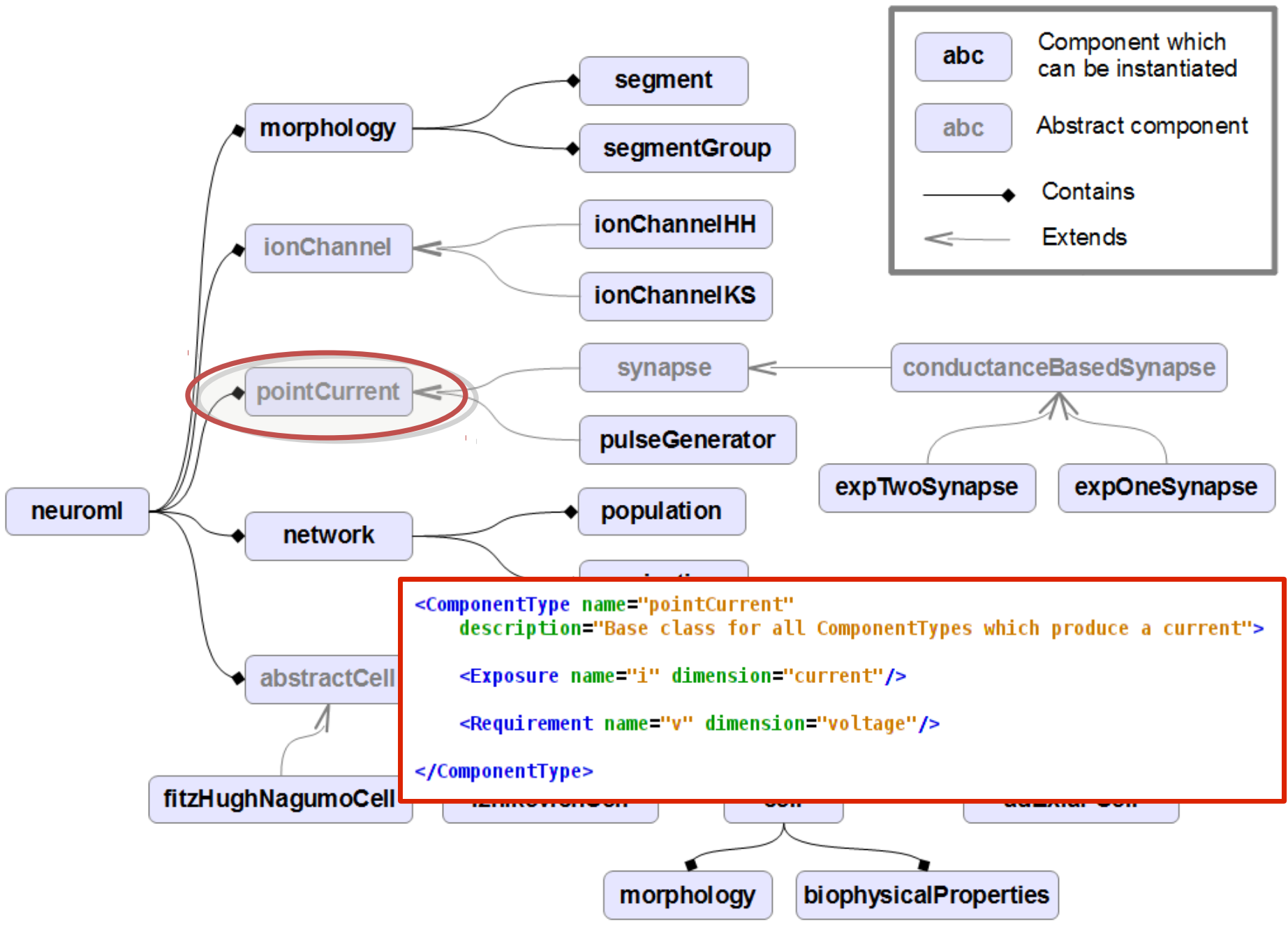
  <network id="PyrCellNet">

    <population id="Population1">
    </population>

    <projection id="Proj1">
    </projection>

  </network>
</neuroml>
  
```





abc

Component which can be instantiated

abc

Abstract component

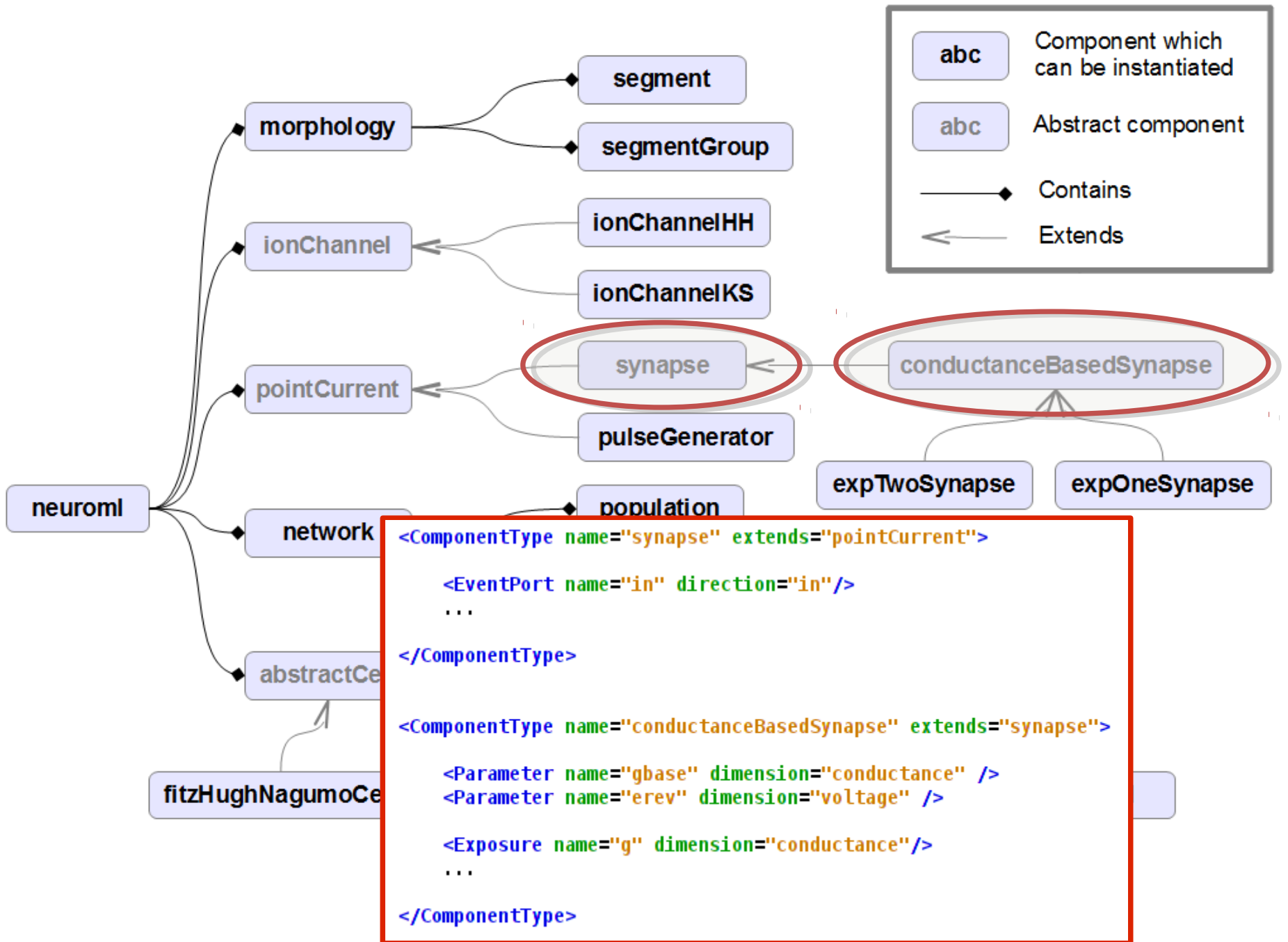
Contains

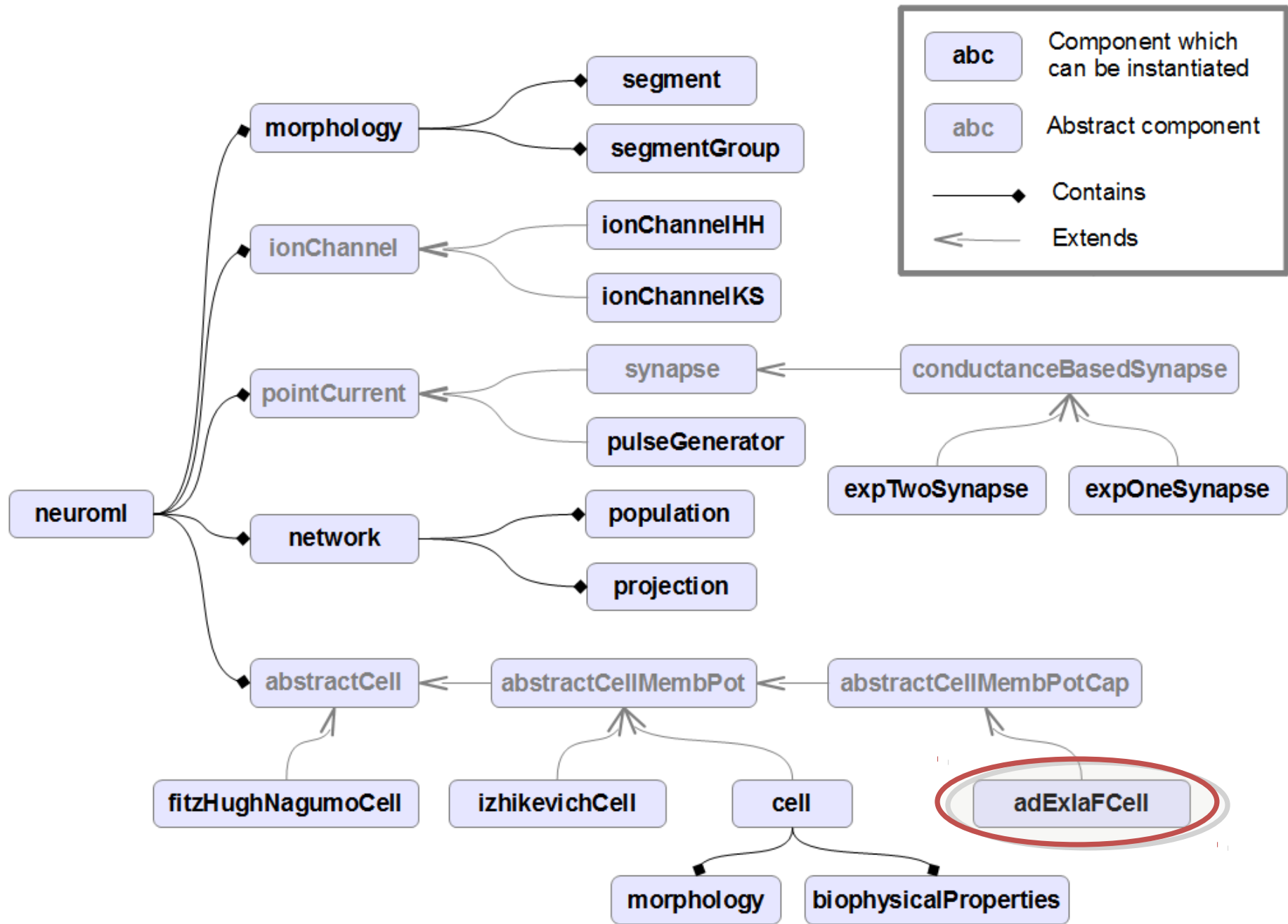
Extends

```

<ComponentType name="pointCurrent"
  description="Base class for all ComponentTypes which produce a current">
  <Exposure name="i" dimension="current"/>
  <Requirement name="v" dimension="voltage"/>
</ComponentType>

```



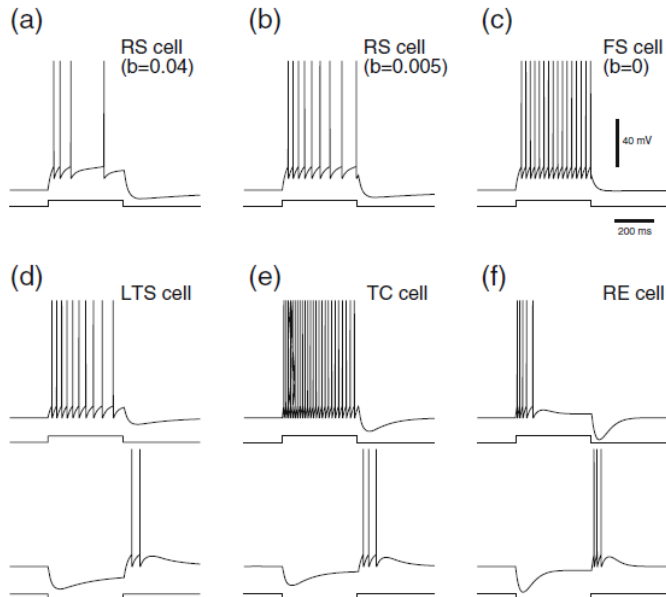


Adaptive Exponential Integrate & Fire cell

$$C \frac{dV}{dt} = -g_L(V - E_L) + g_L \Delta_T \exp\left(\frac{V - V_T}{\Delta_T}\right) - g_e(t)(V - E_e) - g_i(t)(V - E_i) - w$$

$$\tau_w \frac{dw}{dt} = a(V - E_L) - w$$

At spike time ($V > 20$ mV): $V \rightarrow E_L$
 $w \rightarrow w + b$



adExIaFCell

w (current)

I (current)

iSyn (current)

iMemb (current)

v (voltage)

g_L (conductance), E_L (voltage), V_T (voltage),
 thresh (voltage), reset (voltage), ΔT (voltage),
 tauw (time), lamp (current), I_{del} (time),
 I_{dur} (time), a (conductance), b (current)

$iSyn = \text{synapses}[*]/i$ (REDUCE: add)

$iMemb = -1 * g_L * (v - E_L) + g_L * \Delta T * \exp((v - V_T) / \Delta T) - w + I + iSyn$

$v' = iMemb / C$

$w' = (a * (v - E_L) - w) / \tau_w$

IF ($v > \text{thresh}$) THEN

($v = \text{reset}$) AND ($w = w + b$)

IF (($t > I_{del}$) AND ($t < (I_{del} + I_{dur})$)) THEN

($I = \text{lamp}$)

IF ($t > (I_{del} + I_{dur})$) THEN

($I = 0$)

abstractCellMembPotCap

iSyn (current)

iMemb (current)

v (voltage)

C (capacitance)

abstractCellMembPot

v (voltage)

abstractCell

adExIaFCell

w (current)

I (current)

iSyn (current)

iMemb (current)

v (voltage)

gL (conductance), EL (voltage), VT (voltage),
thresh (voltage), reset (voltage), delT (voltage),
tauw (time), Iamp (current), Idel (time),
Idur (time), a (conductance), b (current)

iSyn = synapses[*]/i (REDUCE: add)

iMemb = $-1 * gL * (v - EL) + gL * delT * \exp((v - VT) / delT) - w + I + iSyn$

$v' = iMemb / C$

$w' = (a * (v - EL) - w) / tauw$

IF (v > thresh) THEN

(v = reset) AND (w = w + b)

IF ((t > Idel) AND (t < (Idel + Idur))) THEN

(I = Iamp)

IF (t > (Idel + Idur)) THEN

(I = 0)

abstractCellMembPotCap

iSyn (current)

iMemb (current)

v (voltage)

C (capacitance)

abstractCellMembPot

v (voltage)

abstractCell

adExIaFCell

w (current)

I (current)

iSyn (current)

iMemb (current)

v (voltage)

gL (conductance), EL (voltage), VT (voltage),
thresh (voltage), reset (voltage), delT (voltage),

tauw (time), Iamp (current), Idel (time),

Idur (time), a (conductance), b (current)

$iSyn = \text{synapses[*]}/i$ (REDUCE: add)

$iMemb = -1 * gL * (v - EL) + gL * delT * \exp((v - VT) / delT) - w + I + iSyn$

$v' = iMemb / C$

$w' = (a * (v - EL) - w) / tauw$

IF (v > thresh) THEN

(v = reset) AND (w = w + b)

IF ((t > Idel) AND (t < (Idel + Idur))) THEN

(I = Iamp)

IF (t > (Idel + Idur)) THEN

(I = 0)

abstractCellMembPotCap

iSyn (current)

iMemb (current)

v (voltage)

C (capacitance)

abstractCellMembPot

v (voltage)

abstractCell

Incorporating PyNN into NeuroML 2

- PyNN is a Python package for simulator independent specification of neuronal network models
- Model code can be developed using the PyNN API and then run using NEURON, NEST, PCSIM, Brian or MOOSE, by replacing:

```
from pyNN.neuron import *
```

with

```
from pyNN.nest import *
```

Incorporating PyNN into NeuroML 2

- Initial implementation allowing export of the network structure to NeuroML 2 using:
from pyNN.neuroml2 import *
- Maps instances of PyNN standard cell models on to equivalent defined in LEMS
- Maps populations in PyNN to <population ...> in LEMS
- Maps connections generated by PyNN to <explicitConnection ...> in LEMS


```
...  
ifcell = create(IF_cond_exp, { 'i_offset' : 0.1, 'tau_refrac' : 3.0,  
                              'v_thresh' : -51.0, 'tau_syn_E' : 2.0,  
                              'tau_syn_I' : 5.0, 'v_reset' : -70.0,  
                              'e_rev_E' : 0., 'e_rev_I' : -80.})  
  
spike_sourceE = create(SpikeSourceArray, {'spike_times': [float(i) for i in range(5,105,10)]})  
spike_sourceI = create(SpikeSourceArray, {'spike_times': [float(i) for i in range(155,255,10)]})  
  
connE = connect(spike_sourceE, ifcell, weight=0.006, synapse_type='excitatory',delay=2.0)  
connI = connect(spike_sourceI, ifcell, weight=0.02, synapse_type='inhibitory',delay=4.0)  
  
...
```

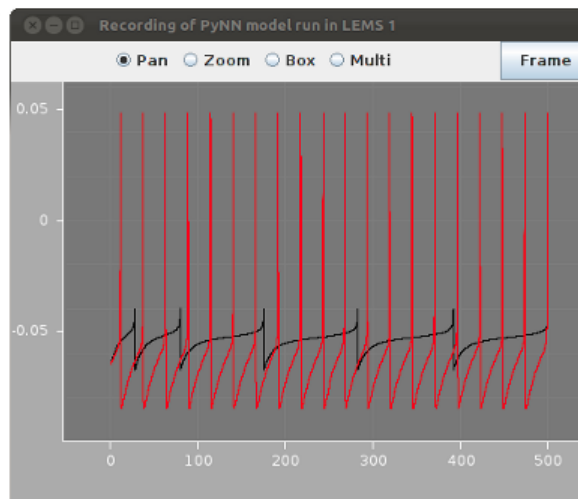
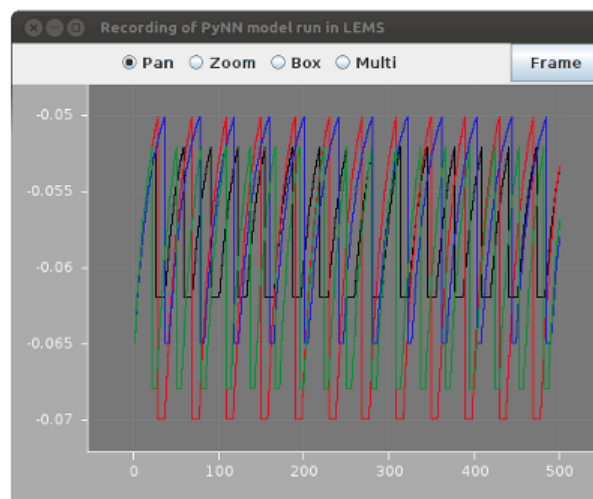
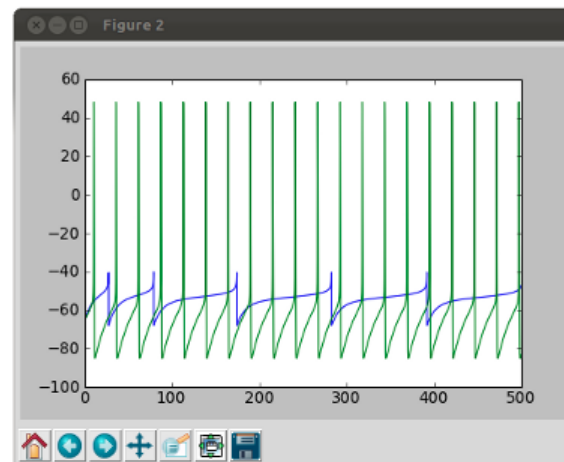
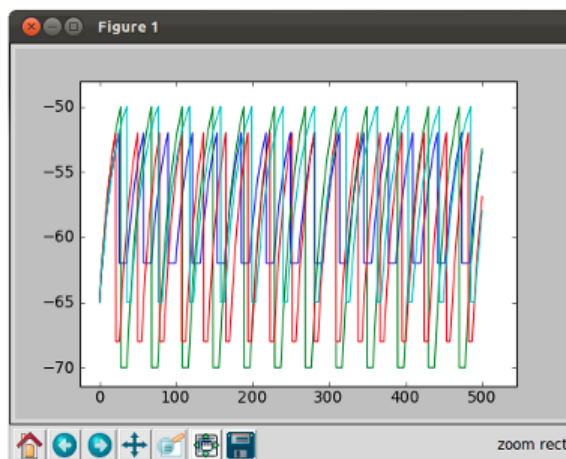


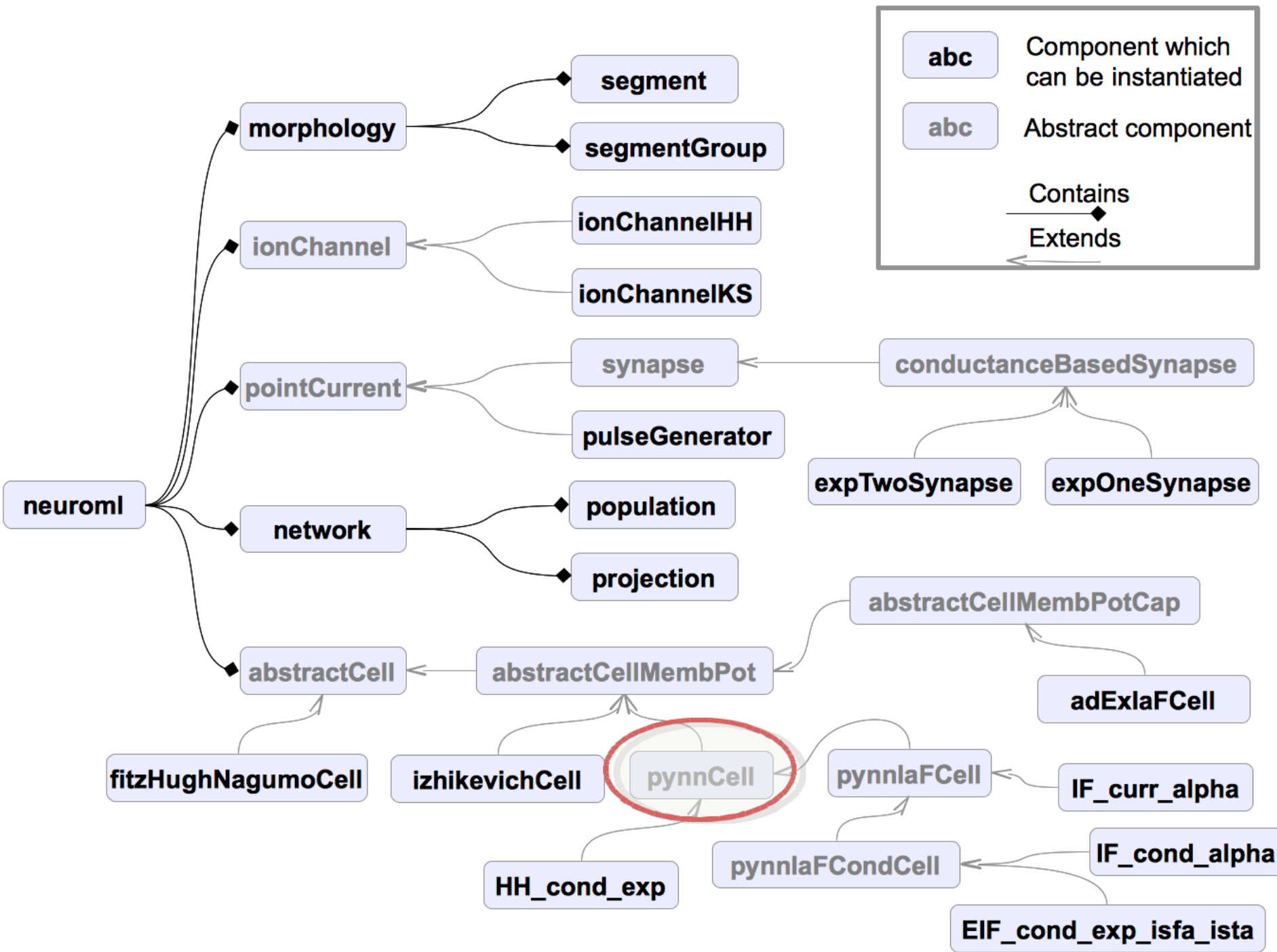
```
<?xml version="1.0" ?>
<neuroml id="PyNN2NeuroMLv2"
  xmlns="http://www.neuroml.org/schema/neuroml2"
  xmlns:xsi="http://www.w3.org/2001/XMLSchema-instance"
  xsi:schemaLocation="http://www.neuroml.org/schema/neuroml2
  http://neuroml.svn.sourceforge.net/viewvc/neuroml/NeuroML2/Schemas/NeuroML2/NeuroML_v2alpha.xsd">

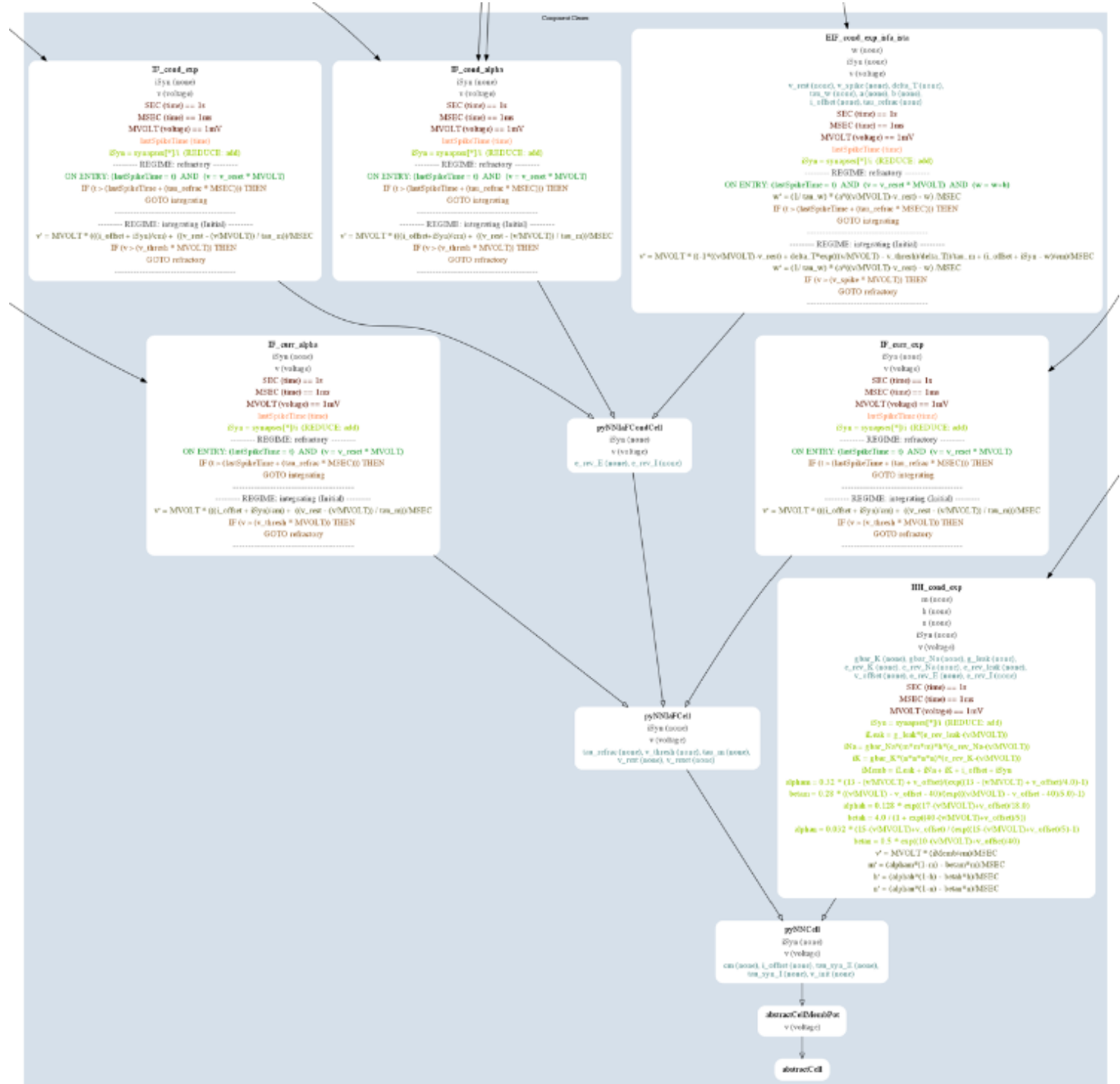
  <IF_cond_exp cm="1.0" e_rev_E="0.0" e_rev_I="-80.0" i_offset="0.1" id="cell_population0"
    tau_m="20.0" tau_refrac="3.0" tau_syn_E="2.0" tau_syn_I="5.0" v_init="-65"
    v_reset="-70.0" v_rest="-65.0" v_thresh="-51.0">
    <notes>
      Component for PyNN IF_cond_exp cell type
    </notes>
  </IF_cond_exp>
  <expCondSynapse e_rev="0.0" id="syn_e_cell_population0" tau_syn="2.0"/>
  <expCondSynapse e_rev="-80.0" id="syn_i_cell_population0" tau_syn="5.0"/>
  <spikeArray id="cell_population1">
    <spike time="5.000000ms" />
    <spike time="15.000000ms" />
    <spike time="25.000000ms" />
    <spike time="35.000000ms" />
    <spike time="45.000000ms" />
    <spike time="55.000000ms" />
    <spike time="65.000000ms" />
    <spike time="75.000000ms" />
    <spike time="85.000000ms" />
    <spike time="95.000000ms" />
  </spikeArray>
  <spikeArray id="cell_population2">
    <spike time="155.000000ms" />
  </spikeArray>
</neuroml>
```



```
<spikeArray id="cell_population1">
  <spike time="5.000000ms"/>
  <spike time="15.000000ms"/>
  <spike time="25.000000ms"/>
  <spike time="35.000000ms"/>
  <spike time="45.000000ms"/>
  <spike time="55.000000ms"/>
  <spike time="65.000000ms"/>
  <spike time="75.000000ms"/>
  <spike time="85.000000ms"/>
  <spike time="95.000000ms"/>
</spikeArray>
<spikeArray id="cell_population2">
  <spike time="155.000000ms"/>
  <spike time="165.000000ms"/>
  <spike time="175.000000ms"/>
  <spike time="185.000000ms"/>
  <spike time="195.000000ms"/>
  <spike time="205.000000ms"/>
  <spike time="215.000000ms"/>
  <spike time="225.000000ms"/>
  <spike time="235.000000ms"/>
  <spike time="245.000000ms"/>
</spikeArray>
<network id="network_PyNN2NeuroMLv2">
  <population component="cell_population0" id="population0" size="1"/>
  <population component="cell_population1" id="population1" size="1"/>
  <population component="cell_population2" id="population2" size="1"/>
  <synapticConnectionWD delay="2.0ms" from="population1[0]"
    synapse="syn_e_cell_population0" to="population0[0]" weight="0.006"/>
  <synapticConnectionWD delay="4.0ms" from="population2[0]"
    synapse="syn_i_cell_population0" to="population0[0]" weight="0.02"/>
</network>
</neuroml>
```







NeuroML & Connection Set Algebra

- PyNN & CSA already well integrated...

```
input_population = Population(10, SpikeSourceArray, {'spike_times': spike_times }, label="input")
output_population = Population(10, IF_curr_exp, cell_params, label="output")
```

```
g=csa.grid2d(3)
d=csa.euclidMetric2d(g,g)
connector = CSAConnector(csa.cset(csa.random(0.5), csa.gaussian(0.1,1.0)*d, 1.0))
```

```
projection = Projection(input_population, output_population, connector, rng=rng)
```

- Connections described with CSA can be used in PyNN scripts & exported to NeuroML 2

```
<spike time="351.090862ms"/>
<spike time="358.827891ms"/>
<spike time="362.685506ms"/>
<spike time="365.812919ms"/>
<spike time="375.407497ms"/>
<spike time="403.008045ms"/>
<spike time="412.715018ms"/>
</spikeArray>
<IF_curr_exp cm="1.0" i_offset="0.0" id="cell_output" tau_m="20.0" tau_refrac="2.0" tau_syn_E="2.0" tau_syn_I="2.0" v_init="-65"
v_reset="-65.0" v_rest="-65.0" v_thresh="-50.0">

</IF_curr_exp>
<expCurrSynapse id="syn_e_cell_output" tau_syn="2.0"/>
<expCurrSynapse id="syn_i_cell_output" tau_syn="2.0"/>
<network id="network_PyNN2NeuroMLv2">
  <population component="cell_input" id="input" size="10"/>
  <population component="cell_output" id="output" size="10"/>
  <synapticConnectionWD delay="1.0ms" from="input[3]" synapse="syn_e_cell_output" to="output[0]" weight="0.00386592013947"/>
  <synapticConnectionWD delay="1.0ms" from="input[7]" synapse="syn_e_cell_output" to="output[0]" weight="8.63504075338e-13"/>
  <synapticConnectionWD delay="1.0ms" from="input[8]" synapse="syn_e_cell_output" to="output[0]" weight="4.9891093928e-20"/>
  <synapticConnectionWD delay="1.0ms" from="input[9]" synapse="syn_e_cell_output" to="output[0]" weight="0.0"/>
  <synapticConnectionWD delay="1.0ms" from="input[0]" synapse="syn_e_cell_output" to="output[1]" weight="0.00386592013947"/>
  <synapticConnectionWD delay="1.0ms" from="input[1]" synapse="syn_e_cell_output" to="output[1]" weight="1.0"/>
  <synapticConnectionWD delay="1.0ms" from="input[2]" synapse="syn_e_cell_output" to="output[1]" weight="0.00386592013947"/>
  <synapticConnectionWD delay="1.0ms" from="input[4]" synapse="syn_e_cell_output" to="output[1]" weight="0.00386592013947"/>
  <synapticConnectionWD delay="1.0ms" from="input[8]" synapse="syn_e_cell_output" to="output[1]" weight="8.63504075338e-13"/>
  <synapticConnectionWD delay="1.0ms" from="input[9]" synapse="syn_e_cell_output" to="output[1]" weight="0.0"/>
  <synapticConnectionWD delay="1.0ms" from="input[1]" synapse="syn_e_cell_output" to="output[2]" weight="0.00386592013947"/>
  <synapticConnectionWD delay="1.0ms" from="input[2]" synapse="syn_e_cell_output" to="output[2]" weight="1.0"/>
  <synapticConnectionWD delay="1.0ms" from="input[6]" synapse="syn_e_cell_output" to="output[2]" weight="4.9891093928e-20"/>
```


NeuroML & Brian interoperability

- Brian is a pure Python simulator
- Easy to specify new neuron models

```
from brian import *
eqs = '''
dv/dt = (ge+gi-(v+49*mV))/(20*ms) : volt
dge/dt = -ge/(5*ms) : volt
dgi/dt = -gi/(10*ms) : volt
'''
P = NeuronGroup(4000, eqs, threshold=-50*mV, reset=-60*mV)
P.v = -60*mV
Pe = P.subgroup(3200)
Pi = P.subgroup(800)
Ce = Connection(Pe, P, 'ge', weight=1.62*mV, sparseness=0.02)
Ci = Connection(Pi, P, 'gi', weight=-9*mV, sparseness=0.02)
M = SpikeMonitor(P)
run(1*second)
raster_plot(M)
show()
```

Example

```
<neuroml xmlns="http://www.neuroml.org/schema/neuroml2"
  xmlns:xsi="http://www.w3.org/2001/XMLSchema-instance"
  xsi:schemaLocation="http://www.neuroml.org/schema/neuroml2 ../Schemas/NeuroML2/NeuroML_v2alpha.xsd"
  id="NML2_AbstractCells">

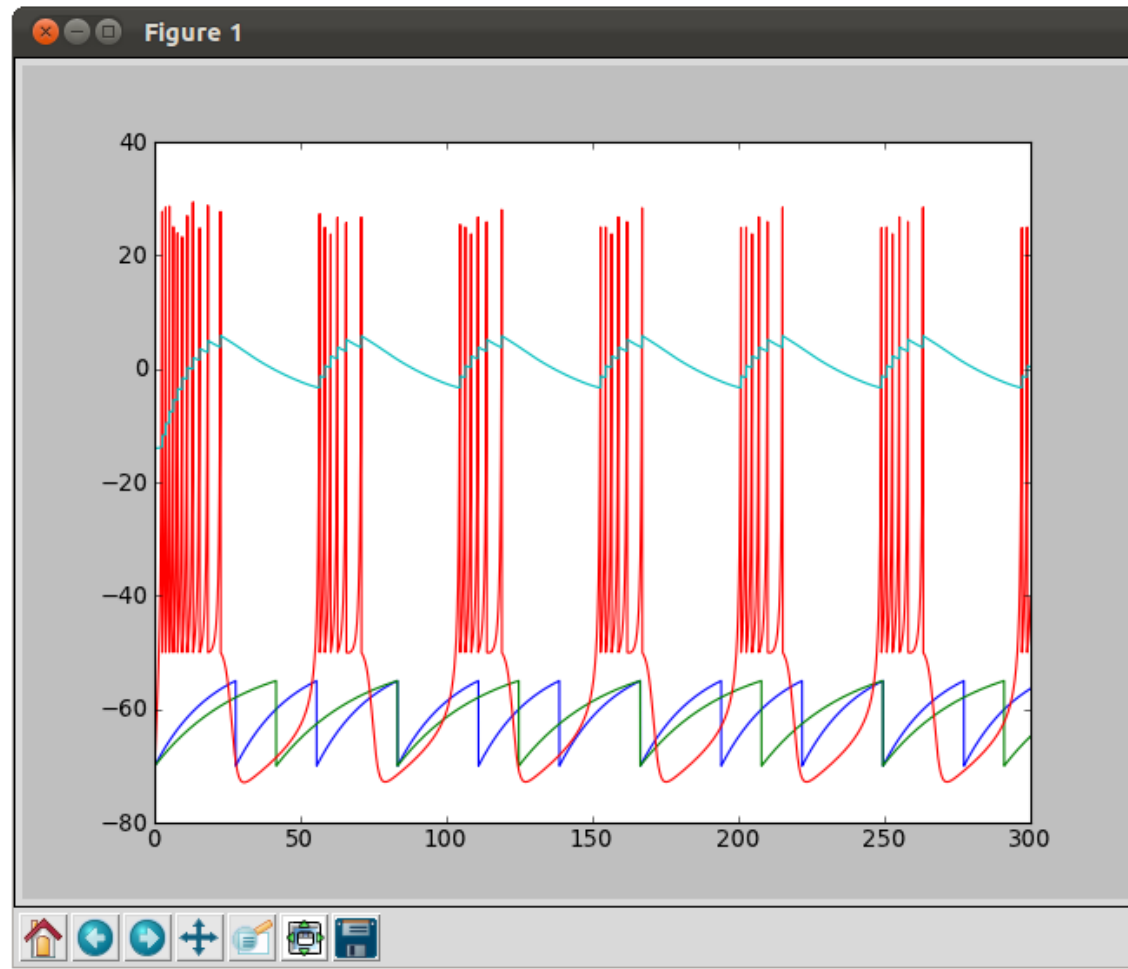
  <iafTauCell id="iafTau" leakReversal="-50mV" thresh="-55mV" reset="-70mV" tau="30ms"/>

  <iafCell id="iafCell" leakReversal="-50mV" thresh="-55mV" reset="-70mV" C="0.2nF" leakConductance="0.01uS"/>

  <izhikevichCell id="izBurst" v0 = "-70mV" thresh = "30mV" a = "0.02" b = "0.2" c = "-50.0" d = "2" Iamp="15" Idel="0ms" Idur="2000ms"/>

</neuroml>
```

```
eqs = '''
dv/dt = (0.04 * v**2 / MVOLT + 5 * v + (140.0 - U + I) * MVOLT)/MSEC : mvolt
dU/dt = a * (b * v / MVOLT - U) / MSEC : 1
I : 1
MSEC : msecond
MVOLT : mvolt
v0 : mvolt
a : 1
b : 1
c : 1
d : 1
thresh : mvolt
Iamp : 1
Idel : msecond
Idur : msecond
'''
```



LEMS/NeuroML & NineML

Much overlap between current NineML abstraction layer &
LEMS

Component - Component

ComponentType - ComponentClass

Parameter - Parameter

NineML missing key concepts like composition, extension

Proposed framework for interaction:

Have consistent way to “flatten” LEMS descriptions

Useful too for mappings to Brian, MATLAB, SBML...

Conclusions

- Declarative specifications of cell model behaviour useful for cross simulator interoperability
- Procedural specification of network structure affords great flexibility
- Combination of the two will be useful for moving forward & sharing models