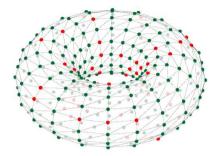


MANCHESIER Running PyNN Simulations on SpiNNaker

MANCHESTER Spiking Neural Networks



Andrew Rowley, Alan Stokes

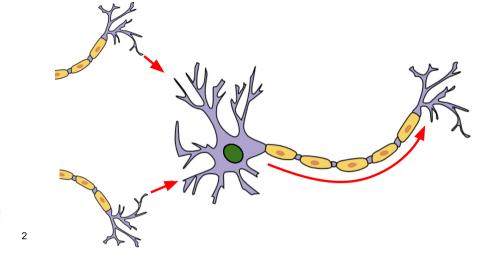
SpiNNaker Workshop, September 2016





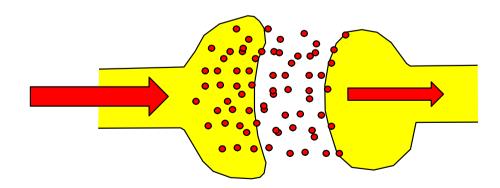






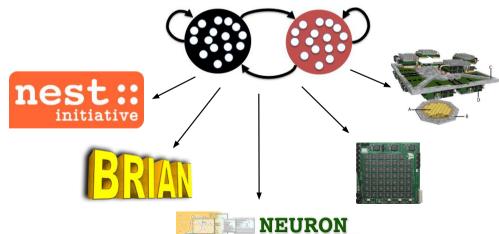


Spiking Neural Networks



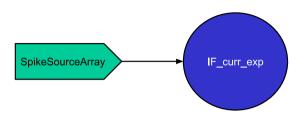


MANCHESTER What is PyNN?





MANCHESTER A Simple PyNN Network



MANCHESTER A Simple PyNN Network

import pyNN.spiNNaker as p

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MANCHESTER A Simple PyNN Network

import pyNN.spiNNaker as p p.setup(timestep=1.0)

MANCHESIER A Simple PyNN Network

import pyNN.spiNNaker as p p.setup(timestep=1.0) pop_1 = p.Population(1, p.IF_curr_exp, {}, label="pop_1")

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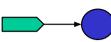
MANCHESTER A Simple PyNN Network

```
import pyNN.spiNNaker as p
p.setup(timestep=1.0)
pop 1 = p.Population(1, p.IF curr exp, {}, label="pop 1")
input = p.Population(1, p.SpikeSourceArray,
                     {'spike times': [0]}, label="input")
```



MANCHESTER A Simple PyNN Network

```
import pyNN.spiNNaker as p
p.setup(timestep=1.0)
pop 1 = p.Population(1, p.IF curr exp, {}, label="pop 1")
input = p.Population(1, p.SpikeSourceArray,
                     {'spike times': [0]}, label="input")
input proj = p.Projection(input, pop 1, p.OneToOneConnector(
    weights=5.0, delays=1))
```

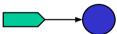


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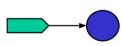
MANCHESIER A Simple PyNN Network

```
import pyNN.spiNNaker as p
p.setup(timestep=1.0)
pop 1 = p.Population(1, p.IF curr exp, {}, label="pop 1")
input = p.Population(1, p.SpikeSourceArray,
                     {'spike times': [0]}, label="input")
input_proj = p.Projection(input, pop_1, p.OneToOneConnector(
    weights=5.0, delays=1))
pop 1.record()
pop 1.record v()
```



MANCHESIER A Simple PyNN Network

```
import pyNN.spiNNaker as p
p.setup(timestep=1.0)
pop 1 = p.Population(1, p.IF curr exp, {}, label="pop 1")
input = p.Population(1, p.SpikeSourceArray,
                     {'spike times': [0]}, label="input")
input_proj = p.Projection(input, pop_1, p.OneToOneConnector(
    weights=5.0, delays=1))
pop 1.record()
pop 1.record v()
p.run(10)
```

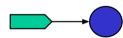


MANCHESTER Edit ~/.spynnaker.cfg

```
[Machine]
#----
# Information about the target SpiNNaker board or machine:
                  The name or IP address of the target board
# version:
                  Version of the Spinnaker Hardware Board (1-5)
# machineTimeStep: Internal time step in simulations in usecs.
# timeScaleFactor: Change this to slow down the simulation time
                      relative to real time.
machineName
                = None
version
                = None
#machineTimeStep = 1000
#timeScaleFactor = 1
```

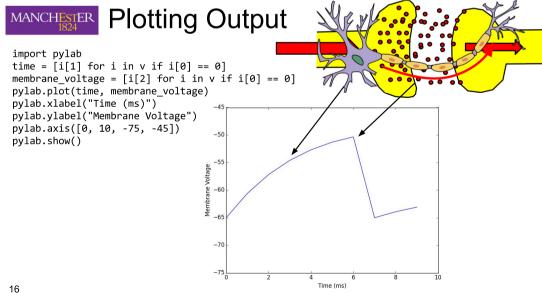
MANCHESTER A Simple PyNN Network

```
import pyNN.spiNNaker as p
p.setup(timestep=1.0)
pop 1 = p.Population(1, p.IF curr exp, {}, label="pop 1")
input = p.Population(1, p.SpikeSourceArray,
                     {'spike times': [0]}, label="input")
input_proj = p.Projection(input, pop_1, p.OneToOneConnector(
    weights=5.0, delays=1))
pop 1.record()
pop 1.record v()
p.run(10)
spikes = pop_1.getSpikes()
v = pop 1.get v()
```



MANCHESTER Edit ~/.spynnaker.cfg

```
[Machine]
#----
# Information about the target SpiNNaker board or machine:
                  The name or IP address of the target board
# version:
                  Version of the Spinnaker Hardware Board (1-5)
# machineTimeStep: Internal time step in simulations in usecs.
# timeScaleFactor: Change this to slow down the simulation time
                      relative to real time.
#----
machineName
               = 192.168.240.253
               = 3
version
#machineTimeStep = 1000
#timeScaleFactor = 1
```



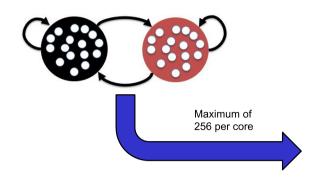
MANCHESTER 1824

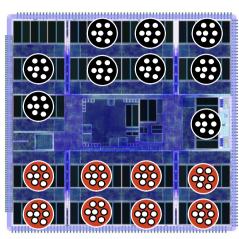
Plotting Output

```
import pylab
spike_time = [i[1] for i in spikes]
spike_id = [i[0] for i in spikes]
pylab.plot(spike_time, spike_id, ".")
pylab.xlabel("Time (ms)")
pylab.ylabel("Neuron ID")
pylab.axis([0, 10, -1, 1])
pylab.show()
-0.5
-0.5
-1.0
2 4 6 8 10
```



Limitations of PyNN on SpiNNaker: Neurons Per Core

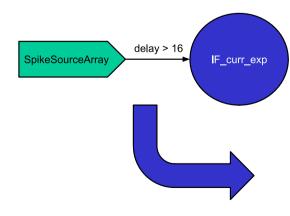


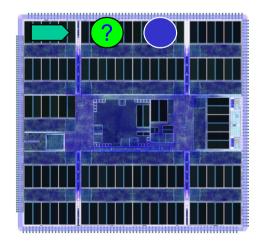


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MANCHESTER 1824

Limitations of PyNN on SpiNNaker: Number of cores available

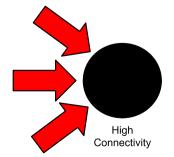




MANCHESTER 1824

SpiNNaker-Specific PyNN

import pyNN.spiNNaker as p
p.setup(timestep=1.0)
p.set_number_of_neurons_per_core(p.IF_curr_exp, 100)
pop_1 = p.Population(1, p.IF_curr_exp, {}, label="pop_1")



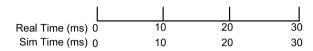


Plasticity

MANCHESTER 1824

Configuration with spynnaker.cfg

[Machine]



MANCHESTER 1824

Configuration with spynnaker.cfg

[Machine]

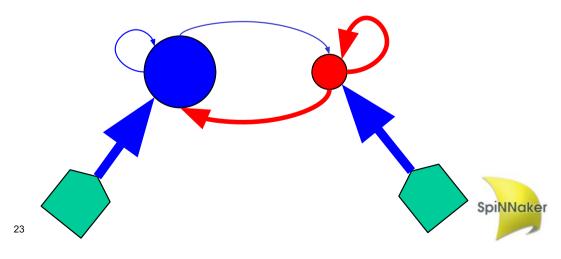






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Balanced Random Network



MANCHESTER

Balanced Random Network

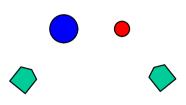
import pyNN.spiNNaker as p
import pylab
from pyNN.random import RandomDistribution

p.setup(timestep=0.1)
n_neurons = 1000
n_exc = int(round(n_neurons * 0.8))
n_inh = int(round(n_neurons * 0.2))



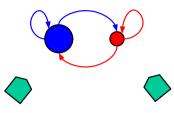
Balanced Random Network

```
pop_exc = p.Population(n_exc, p.IF_curr_exp, {},
                       label="Excitatory")
pop inh = p.Population(n inh, p.IF curr exp, {},
                       label="Inhibitory")
stim exc = p.Population(n exc, p.SpikeSourcePoisson,
                        {"rate": 10.0}, label="Stim Exc")
stim inh = p.Population(n inh, p.SpikeSourcePoisson,
                        {"rate": 10.0}, label="Stim Inh")
```



MANCHESTER Balanced Random Network

```
conn exc = p.FixedProbabilityConnector(0.1, weights=0.2,
                                       delays=2.0)
conn inh = p.FixedProbabilitvConnector(0.1, weights=-1.0.
                                       delays=2.0)
p.Projection(pop exc, pop exc, conn exc, target="excitatory")
p.Projection(pop exc, pop inh, conn exc, target="excitatory")
p.Projection(pop inh, pop inh, conn inh, target="inhibitory")
p.Projection(pop inh, pop exc, conn inh, target="inhibitory")
```



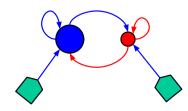




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Balanced Random Network

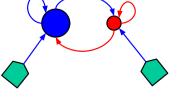
```
delays stim = RandomDistribution("uniform", [1.0, 1.6])
conn stim = p.OneToOneConnector(weights=2.0,
                                delays=delays stim)
p.Projection(stim exc, pop exc, conn stim, target="excitatory")
p.Projection(stim inh, pop inh, conn stim, target="excitatory")
```



Balanced Random Network

```
pop exc.initialize("v", RandomDistribution("uniform",
                                            [-65.0, -55.0]))
pop inh.initialize("v", RandomDistribution("uniform",
                                            [-65.0, -55.0]))
pop exc.record()
p.run(1000)
```



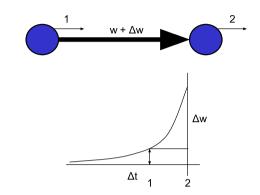




Balanced Random Network

```
spikes = pop_exc.getSpikes()
pylab.plot([i[1] for i in spikes], [i[0] for i in spikes], ".")
pylab.xlabel("Time (ms)")
pylab.ylabel("Neuron ID")
pylab.axis([0, 1000, -1, n exc + 1])
pylab.show()
```

Spike Time Dependent Plasticity: Potentiation

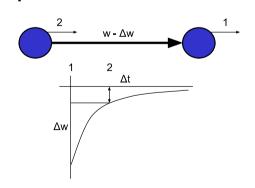




SpiNNaker

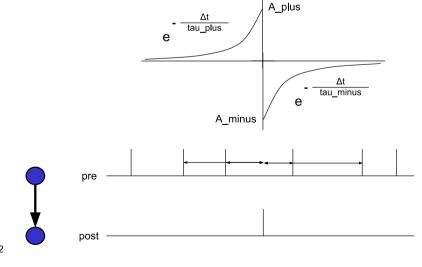
30

Spike Time Dependent Plasticity: Depression





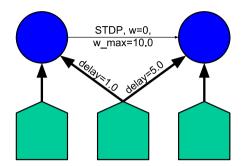
STDP Rules







MANCHESTER STDP in PyNN







Spi<mark>NNake</mark>r

1

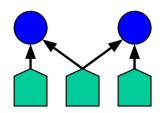
MANCHESTER 1824

STDP in PyNN

```
training = p.Population(
    n_neurons, p.SpikeSourcePoisson,
    {"rate": 10.0, "start": 2000.0, "duration": 1000.0},
    label="Training")

p.Projection(pre_noise, pre_pop, p.OneToOneConnector(weights=2.0))
p.Projection(post_noise, post_pop, p.OneToOneConnector(weights=2.0))

p.Projection(training, pre_pop, p.OneToOneConnector(weights=5.0, delays=1.0))
p.Projection(training, post_pop, p.OneToOneConnector(weights=5.0, delays=10.0))
```



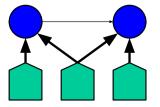


MANCHESTER STDP in PyNN

```
timing_rule = p.SpikePairRule(tau_plus=20.0, tau_minus=20.0)
weight_rule = p.AdditiveWeightDependence(
    w_max=5.0, w_min=0.0, A_plus=0.5, A_minus=0.5)

stdp_model = p.STDPMechanism(
    timing_dependence=timing_rule, weight_dependence=weight_rule)

stdp_projection = p.Projection(
    pre_pop, post_pop, p.OneToOneConnector(weights=0.0, delays=5.0),
    synapse_dynamics=p.SynapseDynamics(slow=stdp_model))
```







MANCHESTER STDP in PyNN

```
p.run(5000)
pre spikes = pre pop.getSpikes()
post_spikes = post_pop.getSpikes()
print stdp projection.getWeights()
p.end()
             [ 4.83886719 5.
                                    4.27246094 5.
                                                         3.80957031 4.87060547
               4.25048828 5.
                                               4.57763672 5.
                                                                              5.
                                                                    5.
                                                                              5.
               5.
                         3.76953125 5.
                                                                   5.
                                                                              5.
               5.
                         5.
                                    5.
                                              5.
                                                                   5.
                                                                              5.
               4.81591797 5.
                                    5.
                                              5.
                                                         5.
                                                                   5.
               2.73339844 5.
                                    5.
                                              5.
                                                         5.
                                                                   5.
                                                                              5.
               5.
                         5.
                                    5.
                                              5.
                                                         5.
                                                                   5.
                                                                              5.
               5.
                         5.
                                    5.
                                                         4.98046875 5.
                                              5.
                                                                              5.
                                                         4.23388672 5.
               5.
                         5.
                                    5.
                                              5.
                                                                              5.
               5.
                         5.
                                    5.
                                              5.
                                                         4.69433594 5.
                                                                              5. SpiNNaker
               5.
                         5.
                                                                   5.
               3.85400391 5.
                                                         4.07617188 5.
                                                                              5.
               5.
                         5.
```

MANCHESTER STDP in PyNN

```
pylab.figure()
pylab.xlim((0, 5000))
pylab.plot([i[1] for i in pre_spikes], [i[0] for i in pre_spikes], "r.")
pylab.plot([i[1] for i in post_spikes], [i[0] for i in post_spikes], "b.")
pylab.xlabel('Time/ms')
pylab.ylabel('spikes')
pylab.show()
```