

Report 4

Computational Neuroscience

Computer Assignment 4

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```
In [1]: %matplotlib notebook
import torch
import numpy as np
```

```
In [2]: from cnsproject.network.neural_populations import LIFPopulation
from cnsproject.plotting.plotting import plotting
from cnsproject.utils import noise_function, step_noise_function, incremental_step_noise_function
from cnsproject.network.monitors import Monitor
from cnsproject.network.connections import Connection, randomNormalConnect, fullyConnect, randomUniformConnect
```

Global Variables

time parameter shows how often (seconds*scale/dt) we want to run our neuron. *dt* means with what resolution (*scale*) we want our seconds move forward. There will be 3 population which two of them are excitatory and one of them is inhibitory. The number of neurons in each population considered 100 neurons.

```
In [3]: time = 1000
scale = 100
dt = 1
neuron_size = 100
shape_ep1 = (int(neuron_size),)
shape_ep2 = (int(neuron_size),)
shape_ip1 = (int(neuron_size),)
```

Walkthrough

The default implemented model is a normal random connection and heterogeneous population.

First we need to specify the input current.

```
In [4]: I_ep1 = incremental_step_noise_function(
        time = time, I_value = 20, scale = scale, neuron_size = shape_ep1[0], gap = 5
    )
I_ep2 = incremental_step_noise_function(
        time = time, I_value = 20, scale = scale, neuron_size = shape_ep2[0], gap = 3
    )
I_ip1 = step_noise_function(time = time, I_value = 20, scale = scale, neuron_size = shape_ip1[0])
```

Then the 3 population that mentioned before, are created by LIF Population model.

```
In [5]: ep1 = LIFPopulation(
        shape = shape_ep1, spike_trace = True, additive_spike_trace = True,
        tau_s = 10, trace_scale = 1., is_inhibitory = False,
        learning = False, R = 1, C = 20, threshold = -40, dt = dt
    )
    ep2 = LIFPopulation(
        shape = shape_ep2, spike_trace = True, additive_spike_trace = True,
        tau_s = 10, trace_scale = 1., is_inhibitory = False,
        learning = False, R = 1, C = 20, threshold = -40, dt = dt
    )
    ip1 = LIFPopulation(
        shape = shape_ip1, spike_trace = True, additive_spike_trace = True,
        tau_s = 10, trace_scale = 1., is_inhibitory = True,
        learning = False, R = 1, C = 20, threshold = -40, dt = dt
    )
```

Here we creat connection between the populations. We have a connection from each excitatory population to the inhibitory population and vice versa. And each excitatory population has a connection with itself.

```
In [6]: con_ep1_ip1 = Connection(
        pre = ep1, post = ip1, lr = None, weight_decay = 0.0,
        J = 1, tau_s = 10, trace_scale = 1., dt = dt, connectivity= randomNormalConnect,
        wmean=20., wstd=5.
    )
    con_ep2_ip1 = Connection(
        pre = ep2, post = ip1, lr = None, weight_decay = 0.0,
        J = 1, tau_s = 10, trace_scale = 1., dt = dt, connectivity= randomNormalConnect,
        wmean=20., wstd=5.
    )
    con_ip1_ep1 = Connection(
        pre = ip1, post = ep1, lr = None, weight_decay = 0.0,
        J = 1, tau_s = 10, trace_scale = 1., dt = dt, connectivity= randomNormalConnect,
        wmean=20., wstd=5.
    )
    con_ip1_ep2 = Connection(
        pre = ip1, post = ep2, lr = None, weight_decay = 0.0,
        J = 1, tau_s = 10, trace_scale = 1., dt = dt, connectivity= randomNormalConnect,
        wmean=20., wstd=5.
    )
    con_ep1_ep1 = Connection(
        pre = ep1, post = ep1, lr = None, weight_decay = 0.0,
        J = 1, tau_s = 10, trace_scale = 1., dt = dt, connectivity= randomNormalConnect,
        wmean=20., wstd=5.
    )
    con_ep2_ep2 = Connection(
        pre = ep2, post = ep2, lr = None, weight_decay = 0.0,
        J = 1, tau_s = 10, trace_scale = 1., dt = dt, connectivity= randomNormalConnect,
        wmean=20., wstd=5.
    )
```

For each population we need to configure a monitor.

```
In [7]: monitor_ep1 = Monitor(ep1, state_variables=["s", "u"])
        monitor_ep1.set_time_steps(time, dt)
        monitor_ep1.reset_state_variables()

        monitor_ep2 = Monitor(ep2, state_variables=["s", "u"])
        monitor_ep2.set_time_steps(time, dt)
        monitor_ep2.reset_state_variables()

        monitor_ip1 = Monitor(ip1, state_variables=["s", "u"])
        monitor_ip1.set_time_steps(time, dt)
        monitor_ip1.reset_state_variables()
```

At a time, each population will forward its current and then compute its effect in each connection.

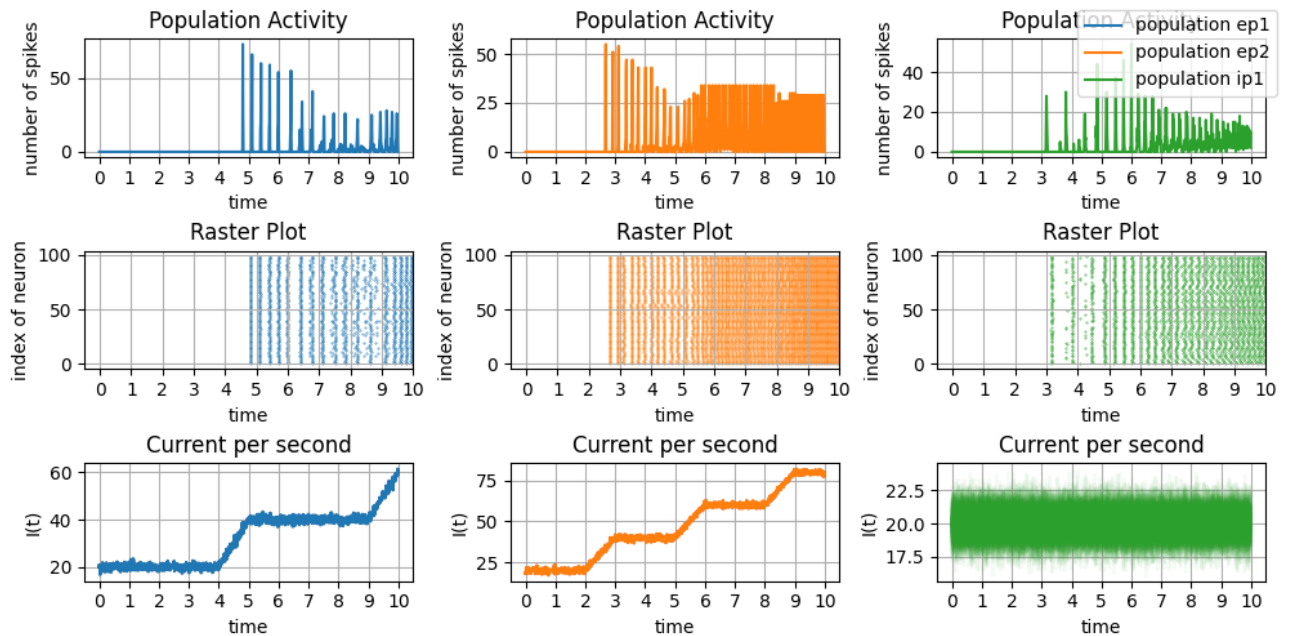
```
In [8]: out_ep1_ip1 = 0
        out_ep2_ip1 = 0
        out_ip1_ep1 = 0
        out_ip1_ep2 = 0
        out_ep1_ep1 = 0
        out_ep2_ep2 = 0
        for i in range(time):
            ep1.forward(I_ep1[i] - out_ip1_ep1 + out_ep1_ep1)
            ep2.forward(I_ep2[i] - out_ip1_ep2 + out_ep2_ep2)
            ip1.forward(I_ip1[i] + out_ep1_ip1 + out_ep2_ip1)
            out_ep1_ip1 = con_ep1_ip1.compute()
            out_ep2_ip1 = con_ep2_ip1.compute()
            out_ip1_ep1 = con_ip1_ep1.compute()
            out_ip1_ep2 = con_ip1_ep2.compute()
            out_ep1_ep1 = con_ep1_ep1.compute()
            out_ep2_ep2 = con_ep2_ep2.compute()
            monitor_ep1.record()
            monitor_ep2.record()
            monitor_ip1.record()

        s_ep1 = torch.transpose(monitor_ep1.get("s")*1, 0, 1)
        s_ep2 = torch.transpose(monitor_ep2.get("s")*1, 0, 1)
        s_ip1 = torch.transpose(monitor_ip1.get("s")*1, 0, 1)
```

In the end lets plot the output.

```
In [9]: plot = plotting()

plot.plot_three_population_activity_init(time/scale)
plot.plot_three_population_activity_update(s_ep1, I_ep1, s_ep2, I_ep2, s_ip1, I_ip1, n1=
"ep1", n2="ep2", n3="ip1")
plot.show()
```



Population Behavior

Zero Current to Inhibitory Population

If we change the input of the inhibitory population to zero, we can see the effect of the other excitatory populations on it and they cause it to spike.

```
In [10]: time = 1000
neuron_size = 100
shape_ep1 = (int(neuron_size),)
shape_ep2 = (int(neuron_size),)
shape_ip1 = (int(neuron_size),)

I_ep1 = incremental_step_noise_function(time = time, I_value = 20, scale = scale, neuron_size = shape_ep1[0], gap = 5)
I_ep2 = incremental_step_noise_function(time = time, I_value = 20, scale = scale, neuron_size = shape_ep2[0], gap = 3)
I_ip1 = torch.zeros(time, neuron_size)

ep1 = LIFPopulation(
    shape = shape_ep1, spike_trace = True, additive_spike_trace = True, tau_s = 10,
    trace_scale = 1.,
    is_inhibitory = False, learning = False, R = 1, C = 20, threshold = -40, dt = dt
)
ep2 = LIFPopulation(
    shape = shape_ep2, spike_trace = True, additive_spike_trace = True, tau_s = 10,
    trace_scale = 1.,
```

```

        is_inhibitory = False, learning = False, R = 1, C = 20, threshold = -40, dt = dt
    )
    ip1 = LIFPopulation(
        shape = shape_ip1, spike_trace = True, additive_spike_trace = True, tau_s = 10,
        trace_scale = 1.,
        is_inhibitory = True, learning = False, R = 1, C = 20, threshold = -40, dt = dt
    )

    con_ep1_ip1 = Connection(
        pre = ep1, post = ip1, lr = None, weight_decay = 0.0,
        J = 1, tau_s = 10, trace_scale = 1., dt = dt, connectivity= randomNormalConnect,
        wmean=20., wstd=5.
    )
    con_ep2_ip1 = Connection(
        pre = ep2, post = ip1, lr = None, weight_decay = 0.0,
        J = 1, tau_s = 10, trace_scale = 1., dt = dt, connectivity= randomNormalConnect,
        wmean=20., wstd=5.
    )
    con_ip1_ep1 = Connection(
        pre = ip1, post = ep1, lr = None, weight_decay = 0.0,
        J = 1, tau_s = 10, trace_scale = 1., dt = dt, connectivity= randomNormalConnect,
        wmean=20., wstd=5.
    )
    con_ip1_ep2 = Connection(
        pre = ip1, post = ep2, lr = None, weight_decay = 0.0,
        J = 1, tau_s = 10, trace_scale = 1., dt = dt, connectivity= randomNormalConnect,
        wmean=20., wstd=5.
    )
    con_ep1_ep1 = Connection(
        pre = ep1, post = ep1, lr = None, weight_decay = 0.0,
        J = 1, tau_s = 10, trace_scale = 1., dt = dt, connectivity= randomNormalConnect,
        wmean=20., wstd=5.
    )
    con_ep2_ep2 = Connection(
        pre = ep2, post = ep2, lr = None, weight_decay = 0.0,
        J = 1, tau_s = 10, trace_scale = 1., dt = dt, connectivity= randomNormalConnect,
        wmean=20., wstd=5.
    )

    monitor_ep1 = Monitor(ep1, state_variables=["s", "u"])
    monitor_ep1.set_time_steps(time, dt)
    monitor_ep1.reset_state_variables()

    monitor_ep2 = Monitor(ep2, state_variables=["s", "u"])
    monitor_ep2.set_time_steps(time, dt)
    monitor_ep2.reset_state_variables()

    monitor_ip1 = Monitor(ip1, state_variables=["s", "u"])
    monitor_ip1.set_time_steps(time, dt)
    monitor_ip1.reset_state_variables()

    out_ep1_ip1 = 0
    out_ep2_ip1 = 0
    out_ip1_ep1 = 0
    out_ip1_ep2 = 0
    out_ep1_ep1 = 0
    out_ep2_ep2 = 0
    for i in range(time):

```

```

ep1.forward(I_ep1[i] - out_ip1_ep1 + out_ep1_ep1)
ep2.forward(I_ep2[i] - out_ip1_ep2 + out_ep2_ep2)
ip1.forward(I_ip1[i] + out_ep1_ip1 + out_ep2_ip1)
out_ep1_ip1 = con_ep1_ip1.compute()
out_ep2_ip1 = con_ep2_ip1.compute()
out_ip1_ep1 = con_ip1_ep1.compute()
out_ip1_ep2 = con_ip1_ep2.compute()
out_ep1_ep1 = con_ep1_ep1.compute()
out_ep2_ep2 = con_ep2_ep2.compute()
monitor_ep1.record()
monitor_ep2.record()
monitor_ip1.record()

```

```

s_ep1 = torch.transpose(monitor_ep1.get("s")*1, 0, 1)
s_ep2 = torch.transpose(monitor_ep2.get("s")*1, 0, 1)
s_ip1 = torch.transpose(monitor_ip1.get("s")*1, 0, 1)

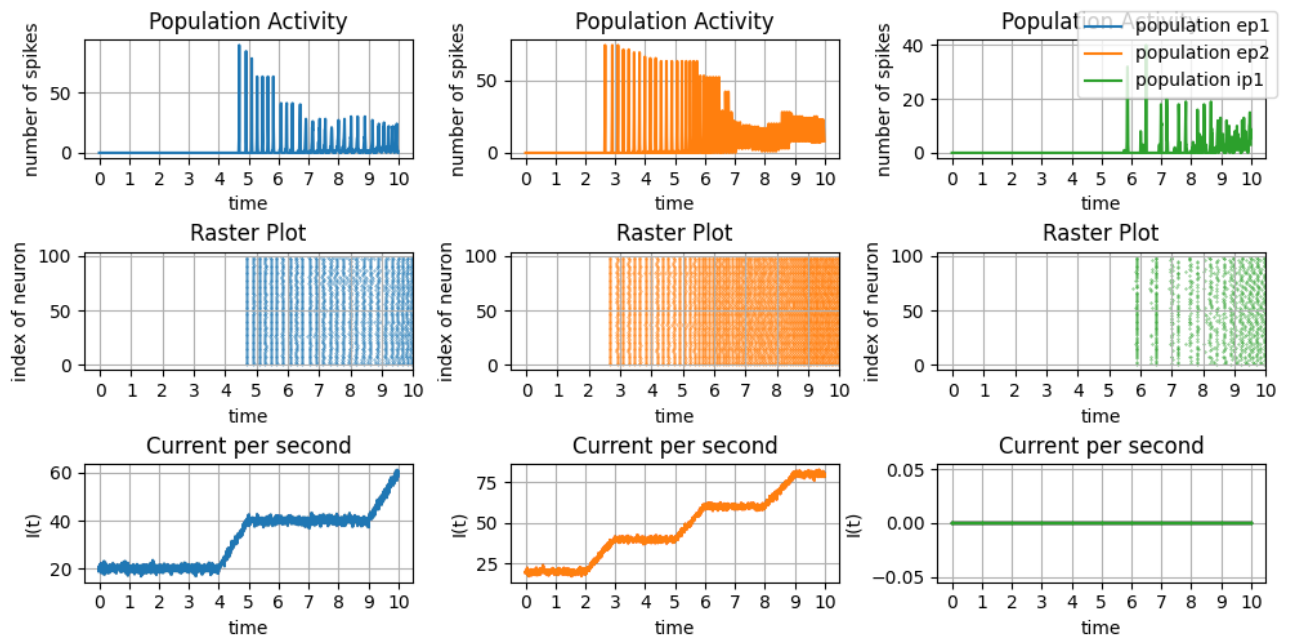
```

```
plot = plotting()
```

```

plot.plot_three_population_activity_init(time/scale)
plot.plot_three_population_activity_update(s_ep1, I_ep1, s_ep2, I_ep2, s_ip1, I_ip1, n1=
"ep1", n2="ep2", n3="ip1")
plot.show()

```



Random Noise as Input

Lets change the input current from a step random current to a random current. It seems that for a period of time we decide based on the first population and for another time we can decide based on the second population.

```
In [12]: time = 1000
         neuron_size = 100
         shape_ep1 = (int(neuron_size),)
         shape_ep2 = (int(neuron_size),)
         shape_ip1 = (int(neuron_size),)
```



```

I_ep1 = noise_function(time = time, neuron_size = shape_ep1[0])
I_ep2 = noise_function(time = time, neuron_size = shape_ep2[0])
I_ip1 = torch.zeros(time, neuron_size)

ep1 = LIFPopulation(
    shape = shape_ep1, spike_trace = True, additive_spike_trace = True, tau_s = 10,
    trace_scale = 1.,
    is_inhibitory = False, learning = False, R = 1, C = 20, threshold = -40, dt = dt
)
ep2 = LIFPopulation(
    shape = shape_ep2, spike_trace = True, additive_spike_trace = True, tau_s = 10,
    trace_scale = 1.,
    is_inhibitory = False, learning = False, R = 1, C = 20, threshold = -40, dt = dt
)
ip1 = LIFPopulation(
    shape = shape_ip1, spike_trace = True, additive_spike_trace = True, tau_s = 10,
    trace_scale = 1.,
    is_inhibitory = True, learning = False, R = 1, C = 20, threshold = -40, dt = dt
)

con_ep1_ip1 = Connection(
    pre = ep1, post = ip1, lr = None, weight_decay = 0.0,
    J = 1, tau_s = 10, trace_scale = 1., dt = dt, connectivity= randomNormalConnect,
    wmean=20., wstd=5.
)
con_ep2_ip1 = Connection(
    pre = ep2, post = ip1, lr = None, weight_decay = 0.0,
    J = 1, tau_s = 10, trace_scale = 1., dt = dt, connectivity= randomNormalConnect,
    wmean=20., wstd=5.
)
con_ip1_ep1 = Connection(
    pre = ip1, post = ep1, lr = None, weight_decay = 0.0,
    J = 1, tau_s = 10, trace_scale = 1., dt = dt, connectivity= randomNormalConnect,
    wmean=20., wstd=5.
)
con_ip1_ep2 = Connection(
    pre = ip1, post = ep2, lr = None, weight_decay = 0.0,
    J = 1, tau_s = 10, trace_scale = 1., dt = dt, connectivity= randomNormalConnect,
    wmean=20., wstd=5.
)
con_ep1_ep1 = Connection(
    pre = ep1, post = ep1, lr = None, weight_decay = 0.0,
    J = 1, tau_s = 10, trace_scale = 1., dt = dt, connectivity= randomNormalConnect,
    wmean=20., wstd=5.
)
con_ep2_ep2 = Connection(
    pre = ep2, post = ep2, lr = None, weight_decay = 0.0,
    J = 1, tau_s = 10, trace_scale = 1., dt = dt, connectivity= randomNormalConnect,
    wmean=20., wstd=5.
)

monitor_ep1 = Monitor(ep1, state_variables=["s", "u"])
monitor_ep1.set_time_steps(time, dt)
monitor_ep1.reset_state_variables()

monitor_ep2 = Monitor(ep2, state_variables=["s", "u"])
monitor_ep2.set_time_steps(time, dt)
monitor_ep2.reset_state_variables()

```

```

monitor_ip1 = Monitor(ip1, state_variables=["s", "u"])
monitor_ip1.set_time_steps(time, dt)
monitor_ip1.reset_state_variables()

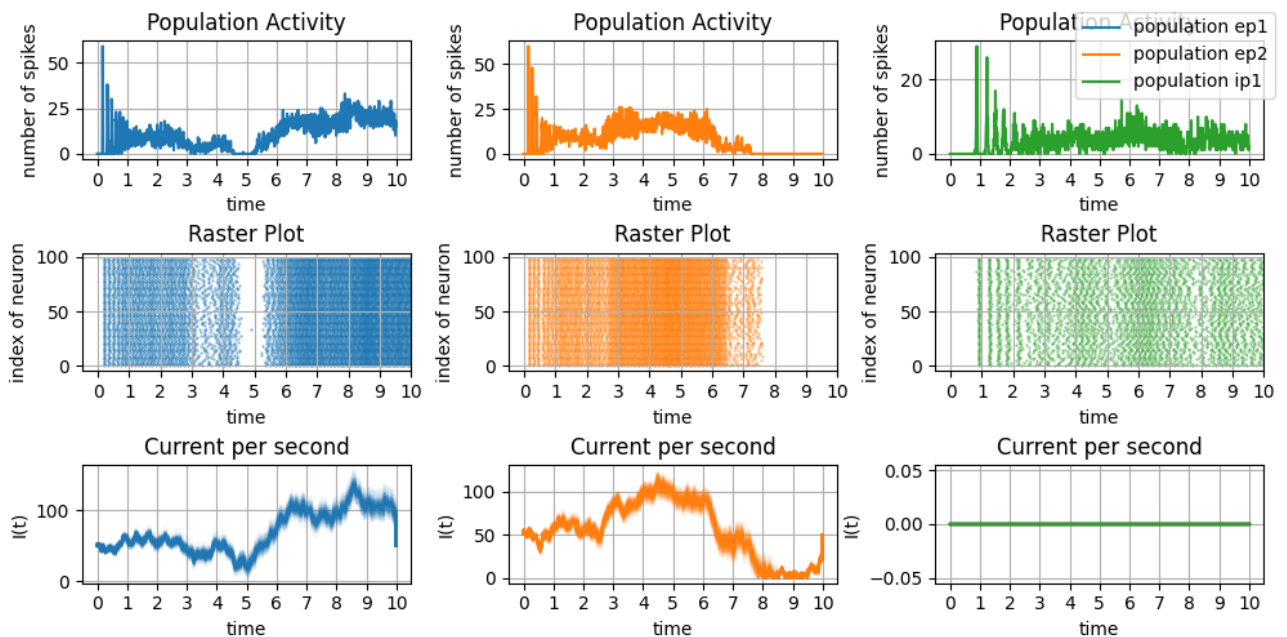
out_ep1_ip1 = 0
out_ep2_ip1 = 0
out_ip1_ep1 = 0
out_ip1_ep2 = 0
out_ep1_ep1 = 0
out_ep2_ep2 = 0
for i in range(time):
    ep1.forward(I_ep1[i] - out_ip1_ep1 + out_ep1_ep1)
    ep2.forward(I_ep2[i] - out_ip1_ep2 + out_ep2_ep2)
    ip1.forward(I_ip1[i] + out_ep1_ip1 + out_ep2_ip1)
    out_ep1_ip1 = con_ep1_ip1.compute()
    out_ep2_ip1 = con_ep2_ip1.compute()
    out_ip1_ep1 = con_ip1_ep1.compute()
    out_ip1_ep2 = con_ip1_ep2.compute()
    out_ep1_ep1 = con_ep1_ep1.compute()
    out_ep2_ep2 = con_ep2_ep2.compute()
    monitor_ep1.record()
    monitor_ep2.record()
    monitor_ip1.record()

s_ep1 = torch.transpose(monitor_ep1.get("s")*1, 0, 1)
s_ep2 = torch.transpose(monitor_ep2.get("s")*1, 0, 1)
s_ip1 = torch.transpose(monitor_ip1.get("s")*1, 0, 1)

plot = plotting()

plot.plot_three_population_activity_init(time/scale)
plot.plot_three_population_activity_update(s_ep1, I_ep1, s_ep2, I_ep2, s_ip1, I_ip1, n1=
"ep1", n2="ep2", n3="ip1")
plot.show()

```



Uniform Random Connection

The next simulation has random step noise but the difference is instead of normal random connection, I used uniform random connection. It cause a more uniform activity for each population and cause for neurons spikes toghether.

```

In [17]: time = 1000
neuron_size = 100
shape_ep1 = (int(neuron_size),)
shape_ep2 = (int(neuron_size),)
shape_ip1 = (int(neuron_size),)

I_ep1 = incremental_step_noise_function(time = time, I_value = 20, scale = scale, neuron_size = shape_ep1[0], gap = 5)
I_ep2 = incremental_step_noise_function(time = time, I_value = 20, scale = scale, neuron_size = shape_ep2[0], gap = 3)
I_ip1 = torch.zeros(time, neuron_size)

ep1 = LIFPopulation(
    shape = shape_ep1, spike_trace = True, additive_spike_trace = True, tau_s = 10,
    trace_scale = 1.,
    is_inhibitory = False, learning = False, R = 1, C = 20, threshold = -40, dt = dt
)
ep2 = LIFPopulation(
    shape = shape_ep2, spike_trace = True, additive_spike_trace = True, tau_s = 10,
    trace_scale = 1.,
    is_inhibitory = False, learning = False, R = 1, C = 20, threshold = -40, dt = dt
)
ip1 = LIFPopulation(
    shape = shape_ip1, spike_trace = True, additive_spike_trace = True, tau_s = 10,
    trace_scale = 1.,
    is_inhibitory = True, learning = False, R = 1, C = 20, threshold = -40, dt = dt
)

con_ep1_ip1 = Connection(
    pre = ep1, post = ip1, lr = None, weight_decay = 0.0,
    J = 1, tau_s = 10, trace_scale = 1., dt = dt, connectivity= randomUniformConnect
, wmax=20., wmin=5.
)
con_ep2_ip1 = Connection(
    pre = ep2, post = ip1, lr = None, weight_decay = 0.0,
    J = 1, tau_s = 10, trace_scale = 1., dt = dt, connectivity= randomUniformConnect
, wmax=20., wmin=5.
)
con_ip1_ep1 = Connection(
    pre = ip1, post = ep1, lr = None, weight_decay = 0.0,
    J = 1, tau_s = 10, trace_scale = 1., dt = dt, connectivity= randomUniformConnect
, wmax=20., wmin=5.
)
con_ip1_ep2 = Connection(
    pre = ip1, post = ep2, lr = None, weight_decay = 0.0,
    J = 1, tau_s = 10, trace_scale = 1., dt = dt, connectivity= randomUniformConnect
, wmax=20., wmin=5.
)
con_ep1_ep1 = Connection(
    pre = ep1, post = ep1, lr = None, weight_decay = 0.0,
    J = 1, tau_s = 10, trace_scale = 1., dt = dt, connectivity= randomUniformConnect
, wmax=20., wmin=5.
)

```

```

con_ep2_ep2 = Connection(
    pre = ep2, post = ep2, lr = None, weight_decay = 0.0,
    J = 1, tau_s = 10, trace_scale = 1., dt = dt, connectivity= randomUniformConnect
, wmax=20., wmin=5.
)

monitor_ep1 = Monitor(ep1, state_variables=["s", "u"])
monitor_ep1.set_time_steps(time, dt)
monitor_ep1.reset_state_variables()

monitor_ep2 = Monitor(ep2, state_variables=["s", "u"])
monitor_ep2.set_time_steps(time, dt)
monitor_ep2.reset_state_variables()

monitor_ip1 = Monitor(ip1, state_variables=["s", "u"])
monitor_ip1.set_time_steps(time, dt)
monitor_ip1.reset_state_variables()

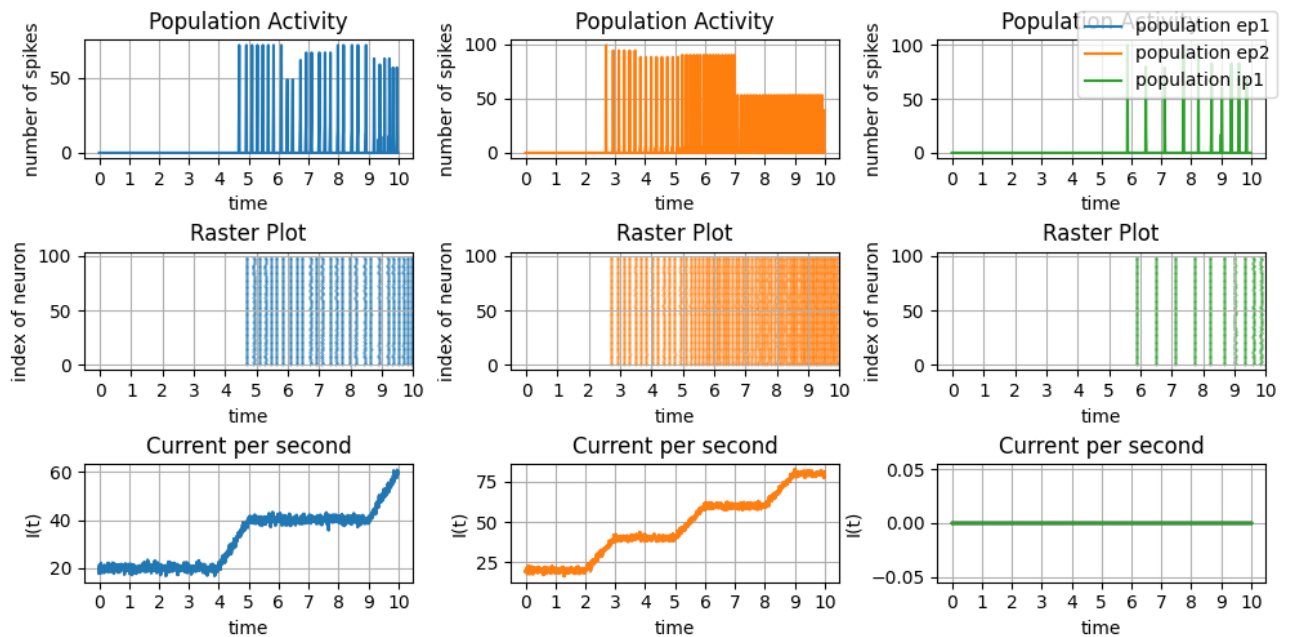
out_ep1_ip1 = 0
out_ep2_ip1 = 0
out_ip1_ep1 = 0
out_ip1_ep2 = 0
out_ep1_ep1 = 0
out_ep2_ep2 = 0
for i in range(time):
    ep1.forward(I_ep1[i] - out_ip1_ep1 + out_ep1_ep1)
    ep2.forward(I_ep2[i] - out_ip1_ep2 + out_ep2_ep2)
    ip1.forward(I_ip1[i] + out_ep1_ip1 + out_ep2_ip1)
    out_ep1_ip1 = con_ep1_ip1.compute()
    out_ep2_ip1 = con_ep2_ip1.compute()
    out_ip1_ep1 = con_ip1_ep1.compute()
    out_ip1_ep2 = con_ip1_ep2.compute()
    out_ep1_ep1 = con_ep1_ep1.compute()
    out_ep2_ep2 = con_ep2_ep2.compute()
    monitor_ep1.record()
    monitor_ep2.record()
    monitor_ip1.record()

s_ep1 = torch.transpose(monitor_ep1.get("s")*1, 0, 1)
s_ep2 = torch.transpose(monitor_ep2.get("s")*1, 0, 1)
s_ip1 = torch.transpose(monitor_ip1.get("s")*1, 0, 1)

plot = plotting()

plot.plot_three_population_activity_init(time/scale)
plot.plot_three_population_activity_update(s_ep1, I_ep1, s_ep2, I_ep2, s_ip1, I_ip1, n1=
"ep1", n2="ep2", n3="ip1")
plot.show()

```



Fully Connected model

The next simulation has random step noise but the difference is instead of normal random connection, I used fully connection. It has less effect on the other population because its output will distribute among all the other population's neuron so it will cause less or no spikes for a inhibitory population without input current.

```
In [15]: time = 1000
neuron_size = 100
shape_ep1 = (int(neuron_size),)
shape_ep2 = (int(neuron_size),)
shape_ip1 = (int(neuron_size),)

I_ep1 = incremental_step_noise_function(time = time, I_value = 20, scale = scale, neuron_size = shape_ep1[0], gap = 5)
I_ep2 = incremental_step_noise_function(time = time, I_value = 20, scale = scale, neuron_size = shape_ep2[0], gap = 3)
I_ip1 = torch.zeros(time, neuron_size)

ep1 = LIFPopulation(
    shape = shape_ep1, spike_trace = True, additive_spike_trace = True, tau_s = 10,
    trace_scale = 1.,
    is_inhibitory = False, learning = False, R = 1, C = 20, threshold = -40, dt = dt
)
ep2 = LIFPopulation(
    shape = shape_ep2, spike_trace = True, additive_spike_trace = True, tau_s = 10,
    trace_scale = 1.,
    is_inhibitory = False, learning = False, R = 1, C = 20, threshold = -40, dt = dt
)
ip1 = LIFPopulation(
    shape = shape_ip1, spike_trace = True, additive_spike_trace = True, tau_s = 10,
    trace_scale = 1.,
    is_inhibitory = True, learning = False, R = 1, C = 20, threshold = -40, dt = dt
)

con_ep1_ip1 = Connection(
    pre = ep1, post = ip1, lr = None, weight_decay = 0.0,
```

```

        J = 1, tau_s = 10, trace_scale = 1., dt = dt, connectivity= fullyConnect
    )
    con_ep2_ip1 = Connection(
        pre = ep2, post = ip1, lr = None, weight_decay = 0.0,
        J = 1, tau_s = 10, trace_scale = 1., dt = dt, connectivity= fullyConnect
    )
    con_ip1_ep1 = Connection(
        pre = ip1, post = ep1, lr = None, weight_decay = 0.0,
        J = 1, tau_s = 10, trace_scale = 1., dt = dt, connectivity= fullyConnect
    )
    con_ip1_ep2 = Connection(
        pre = ip1, post = ep2, lr = None, weight_decay = 0.0,
        J = 1, tau_s = 10, trace_scale = 1., dt = dt, connectivity= fullyConnect
    )
    con_ep1_ep1 = Connection(
        pre = ep1, post = ep1, lr = None, weight_decay = 0.0,
        J = 1, tau_s = 10, trace_scale = 1., dt = dt, connectivity= fullyConnect
    )
    con_ep2_ep2 = Connection(
        pre = ep2, post = ep2, lr = None, weight_decay = 0.0,
        J = 1, tau_s = 10, trace_scale = 1., dt = dt, connectivity= fullyConnect
    )

    monitor_ep1 = Monitor(ep1, state_variables=["s", "u"])
    monitor_ep1.set_time_steps(time, dt)
    monitor_ep1.reset_state_variables()

    monitor_ep2 = Monitor(ep2, state_variables=["s", "u"])
    monitor_ep2.set_time_steps(time, dt)
    monitor_ep2.reset_state_variables()

    monitor_ip1 = Monitor(ip1, state_variables=["s", "u"])
    monitor_ip1.set_time_steps(time, dt)
    monitor_ip1.reset_state_variables()

    out_ep1_ip1 = 0
    out_ep2_ip1 = 0
    out_ip1_ep1 = 0
    out_ip1_ep2 = 0
    out_ep1_ep1 = 0
    out_ep2_ep2 = 0
    for i in range(time):
        ep1.forward(I_ep1[i] - out_ip1_ep1 + out_ep1_ep1)
        ep2.forward(I_ep2[i] - out_ip1_ep2 + out_ep2_ep2)
        ip1.forward(I_ip1[i] + out_ep1_ip1 + out_ep2_ip1)
        out_ep1_ip1 = con_ep1_ip1.compute()
        out_ep2_ip1 = con_ep2_ip1.compute()
        out_ip1_ep1 = con_ip1_ep1.compute()
        out_ip1_ep2 = con_ip1_ep2.compute()
        out_ep1_ep1 = con_ep1_ep1.compute()
        out_ep2_ep2 = con_ep2_ep2.compute()
        monitor_ep1.record()
        monitor_ep2.record()
        monitor_ip1.record()

    s_ep1 = torch.transpose(monitor_ep1.get("s")*1, 0, 1)
    s_ep2 = torch.transpose(monitor_ep2.get("s")*1, 0, 1)
    s_ip1 = torch.transpose(monitor_ip1.get("s")*1, 0, 1)

```

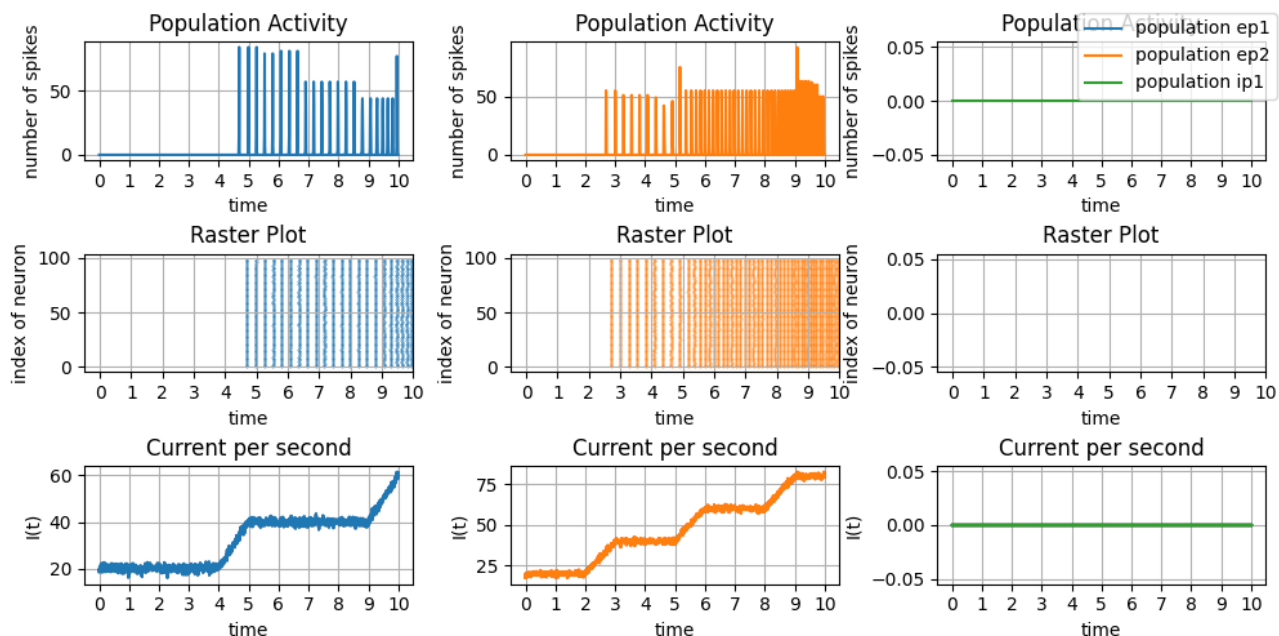


```
plot = plotting()
```

```
plot.plot_three_population_activity_init(time/scale)
```

```
plot.plot_three_population_activity_update(s_ep1, I_ep1, s_ep2, I_ep2, s_ip1, I_ip1, n1="ep1", n2="ep2", n3="ip1")
```

```
plot.show()
```



Same Current as Input

If we get a same current to all the populations, the result is as follows:

```
In [23]: time = 1000
neuron_size = 100
shape_ep1 = (int(neuron_size),)
shape_ep2 = (int(neuron_size),)
shape_ip1 = (int(neuron_size),)

I_ep1 = incremental_step_noise_function(time = time, I_value = 20, scale = scale, neuron
_size = shape_ep1[0], gap = 6)
I_ep2 = torch.clone(I_ep1)
I_ip1 = torch.clone(I_ep1)

ep1 = LIFPopulation(
    shape = shape_ep1, spike_trace = True, additive_spike_trace = True, tau_s = 10,
    trace_scale = 1.,
    is_inhibitory = False, learning = False, R = 1, C = 20, threshold = -40, dt = dt
)
ep2 = LIFPopulation(
    shape = shape_ep2, spike_trace = True, additive_spike_trace = True, tau_s = 10,
    trace_scale = 1.,
    is_inhibitory = False, learning = False, R = 1, C = 20, threshold = -40, dt = dt
)
ip1 = LIFPopulation(
    shape = shape_ip1, spike_trace = True, additive_spike_trace = True, tau_s = 10,
    trace_scale = 1.,
```

```

        is_inhibitory = True, learning = False, R = 1, C = 20, threshold = -40, dt = dt
    )

    con_ep1_ip1 = Connection(
        pre = ep1, post = ip1, lr = None, weight_decay = 0.0,
        J = 1, tau_s = 10, trace_scale = 1., dt = dt, connectivity= randomNormalConnect,
        wmean=20., wstd=5.
    )
    con_ep2_ip1 = Connection(
        pre = ep2, post = ip1, lr = None, weight_decay = 0.0,
        J = 1, tau_s = 10, trace_scale = 1., dt = dt, connectivity= randomNormalConnect,
        wmean=20., wstd=5.
    )
    con_ip1_ep1 = Connection(
        pre = ip1, post = ep1, lr = None, weight_decay = 0.0,
        J = 1, tau_s = 10, trace_scale = 1., dt = dt, connectivity= randomNormalConnect,
        wmean=20., wstd=5.
    )
    con_ip1_ep2 = Connection(
        pre = ip1, post = ep2, lr = None, weight_decay = 0.0,
        J = 1, tau_s = 10, trace_scale = 1., dt = dt, connectivity= randomNormalConnect,
        wmean=20., wstd=5.
    )
    con_ep1_ep1 = Connection(
        pre = ep1, post = ep1, lr = None, weight_decay = 0.0,
        J = 1, tau_s = 10, trace_scale = 1., dt = dt, connectivity= randomNormalConnect,
        wmean=20., wstd=5.
    )
    con_ep2_ep2 = Connection(
        pre = ep2, post = ep2, lr = None, weight_decay = 0.0,
        J = 1, tau_s = 10, trace_scale = 1., dt = dt, connectivity= randomNormalConnect,
        wmean=20., wstd=5.
    )

    monitor_ep1 = Monitor(ep1, state_variables=["s", "u"])
    monitor_ep1.set_time_steps(time, dt)
    monitor_ep1.reset_state_variables()

    monitor_ep2 = Monitor(ep2, state_variables=["s", "u"])
    monitor_ep2.set_time_steps(time, dt)
    monitor_ep2.reset_state_variables()

    monitor_ip1 = Monitor(ip1, state_variables=["s", "u"])
    monitor_ip1.set_time_steps(time, dt)
    monitor_ip1.reset_state_variables()

    out_ep1_ip1 = 0
    out_ep2_ip1 = 0
    out_ip1_ep1 = 0
    out_ip1_ep2 = 0
    out_ep1_ep1 = 0
    out_ep2_ep2 = 0
    for i in range(time):
        ep1.forward(I_ep1[i] - out_ip1_ep1 + out_ep1_ep1)
        ep2.forward(I_ep2[i] - out_ip1_ep2 + out_ep2_ep2)
        ip1.forward(I_ip1[i] + out_ep1_ip1 + out_ep2_ip1)
        out_ep1_ip1 = con_ep1_ip1.compute()
        out_ep2_ip1 = con_ep2_ip1.compute()

```

```

out_ip1_ep1 = con_ip1_ep1.compute()
out_ip1_ep2 = con_ip1_ep2.compute()
out_ep1_ep1 = con_ep1_ep1.compute()
out_ep2_ep2 = con_ep2_ep2.compute()
monitor_ep1.record()
monitor_ep2.record()
monitor_ip1.record()

```

```

s_ep1 = torch.transpose(monitor_ep1.get("s")*1, 0, 1)
s_ep2 = torch.transpose(monitor_ep2.get("s")*1, 0, 1)
s_ip1 = torch.transpose(monitor_ip1.get("s")*1, 0, 1)

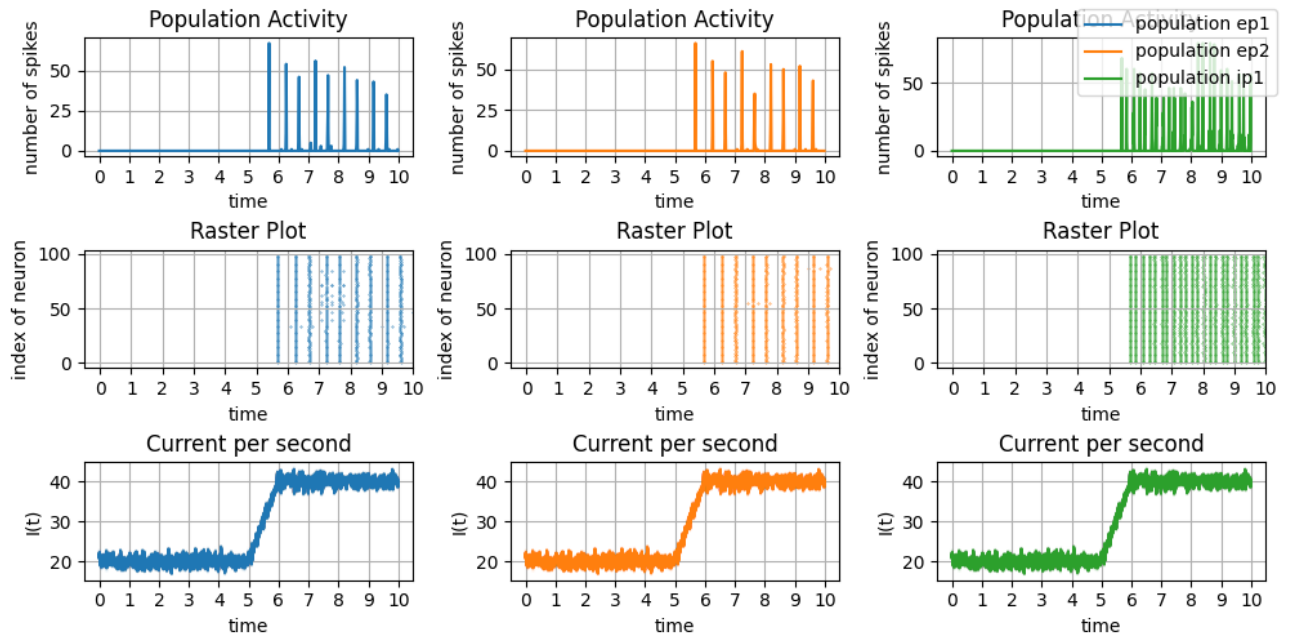
```

```
plot = plotting()
```

```

plot.plot_three_population_activity_init(time/scale)
plot.plot_three_population_activity_update(s_ep1, I_ep1, s_ep2, I_ep2, s_ip1, I_ip1, n1=
"ep1", n2="ep2", n3="ip1")
plot.show()

```



One Strong Current

As we can see below there may be a time when the inhibitory population won't allow another population spike although in the previous parts it spikes.

```
In [24]: time = 1000
         neuron_size = 100
         shape_ep1 = (int(neuron_size),)
         shape_ep2 = (int(neuron_size),)
         shape_ip1 = (int(neuron_size),)

         I_ep1 = incremental_step_noise_function(time = time, I_value = 20, scale = scale, neuron
         _size = shape_ep1[0], gap = 6)
         I_ep2 = torch.clone(I_ep1)
         I_ep2[5*scale:,:] = I_ep2[5*scale:,:] + 20
         I_ip1 = I_ep1
```

```

ep1 = LIFPopulation(
    shape = shape_ep1, spike_trace = True, additive_spike_trace = True, tau_s = 10,
    trace_scale = 1.,
    is_inhibitory = False, learning = False, R = 1, C = 20, threshold = -40, dt = dt
)
ep2 = LIFPopulation(
    shape = shape_ep2, spike_trace = True, additive_spike_trace = True, tau_s = 10,
    trace_scale = 1.,
    is_inhibitory = False, learning = False, R = 1, C = 20, threshold = -40, dt = dt
)
ip1 = LIFPopulation(
    shape = shape_ip1, spike_trace = True, additive_spike_trace = True, tau_s = 10,
    trace_scale = 1.,
    is_inhibitory = True, learning = False, R = 1, C = 20, threshold = -40, dt = dt
)

con_ep1_ip1 = Connection(
    pre = ep1, post = ip1, lr = None, weight_decay = 0.0,
    J = 1, tau_s = 10, trace_scale = 1., dt = dt, connectivity= randomNormalConnect,
    wmean=20., wstd=5.
)
con_ep2_ip1 = Connection(
    pre = ep2, post = ip1, lr = None, weight_decay = 0.0,
    J = 1, tau_s = 10, trace_scale = 1., dt = dt, connectivity= randomNormalConnect,
    wmean=20., wstd=5.
)
con_ip1_ep1 = Connection(
    pre = ip1, post = ep1, lr = None, weight_decay = 0.0,
    J = 1, tau_s = 10, trace_scale = 1., dt = dt, connectivity= randomNormalConnect,
    wmean=20., wstd=5.
)
con_ip1_ep2 = Connection(
    pre = ip1, post = ep2, lr = None, weight_decay = 0.0,
    J = 1, tau_s = 10, trace_scale = 1., dt = dt, connectivity= randomNormalConnect,
    wmean=20., wstd=5.
)
con_ep1_ep1 = Connection(
    pre = ep1, post = ep1, lr = None, weight_decay = 0.0,
    J = 1, tau_s = 10, trace_scale = 1., dt = dt, connectivity= randomNormalConnect,
    wmean=20., wstd=5.
)
con_ep2_ep2 = Connection(
    pre = ep2, post = ep2, lr = None, weight_decay = 0.0,
    J = 1, tau_s = 10, trace_scale = 1., dt = dt, connectivity= randomNormalConnect,
    wmean=20., wstd=5.
)

monitor_ep1 = Monitor(ep1, state_variables=["s", "u"])
monitor_ep1.set_time_steps(time, dt)
monitor_ep1.reset_state_variables()

monitor_ep2 = Monitor(ep2, state_variables=["s", "u"])
monitor_ep2.set_time_steps(time, dt)
monitor_ep2.reset_state_variables()

monitor_ip1 = Monitor(ip1, state_variables=["s", "u"])
monitor_ip1.set_time_steps(time, dt)

```

```

monitor_ip1.reset_state_variables()

out_ep1_ip1 = 0
out_ep2_ip1 = 0
out_ip1_ep1 = 0
out_ip1_ep2 = 0
out_ep1_ep1 = 0
out_ep2_ep2 = 0
for i in range(time):
    ep1.forward(I_ep1[i] - out_ip1_ep1 + out_ep1_ep1)
    ep2.forward(I_ep2[i] - out_ip1_ep2 + out_ep2_ep2)
    ip1.forward(I_ip1[i] + out_ep1_ip1 + out_ep2_ip1)
    out_ep1_ip1 = con_ep1_ip1.compute()
    out_ep2_ip1 = con_ep2_ip1.compute()
    out_ip1_ep1 = con_ip1_ep1.compute()
    out_ip1_ep2 = con_ip1_ep2.compute()
    out_ep1_ep1 = con_ep1_ep1.compute()
    out_ep2_ep2 = con_ep2_ep2.compute()
    monitor_ep1.record()
    monitor_ep2.record()
    monitor_ip1.record()

s_ep1 = torch.transpose(monitor_ep1.get("s")*1, 0, 1)
s_ep2 = torch.transpose(monitor_ep2.get("s")*1, 0, 1)
s_ip1 = torch.transpose(monitor_ip1.get("s")*1, 0, 1)

plot = plotting()

plot.plot_three_population_activity_init(time/scale)
plot.plot_three_population_activity_update(s_ep1, I_ep1, s_ep2, I_ep2, s_ip1, I_ip1, n1=
"ep1", n2="ep2", n3="ip1")
plot.show()

```

