

06__osc__transition

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(C) Gerold Baier, University College London, 2024

1 Transition to Oscillations in Two Coupled Variables

1.1 Mathematical Model

Single Variable, first-order differential equation

The equation:

$$\begin{aligned}\frac{dEx}{dt} &= h_{ex} - Ex + c_1 * \tanh(Ex) - c_2 * \tanh(In) \\ \frac{dIn}{dt} &= h_{in} - In + c_3 * \tanh(Ex) - c_4 * \tanh(In)\end{aligned}$$

where Ex and In are variables that changes with time t , h_{ex} and c_i are model parameters, \tanh is the tangens hyperbolicus.

For $c_2 = c_3 = 0$, the two variables are uncoupled and independent.

The model dynamics can be characterised by scans of parameter h_{ex} . Here, we use a slow continuous change of this parameter to simulate spontaneous qualitative transitions of dynamics in the human EEG.

1.2 Import Functions

```
[62]: from scipy.integrate import odeint
      from scipy.signal import find_peaks, butter, sosfilt

      from numpy import zeros, ones, tanh, mod, gradient, linspace, sign, log, \
      ↪ meshgrid
      from numpy import sqrt, fill_diagonal, ndarray, amax, amin, where
      from numpy import asarray, array, around, arange, corrcoef, flip, var
      from numpy.random import default_rng

      from matplotlib.pyplot import subplots, xticks, yticks, axes
```

1.3 Model and Functions

```
[34]: def sigmoid(u):  
    return tanh(u)  
  
def single_oscillator(y, t, h_ex, h_in, pars):  
    tau_ex, tau_in, c_1, c_2, c_3, c_4 = pars  
  
    dydt = (  
        (h_ex - y[0] + c_1*sigmoid(y[0]) - c_2*sigmoid(y[1]))*tau_ex,  
        (h_in - y[1] + c_3*sigmoid(y[0]) - c_4*sigmoid(y[1]))*tau_in,  
    )  
  
    return dydt  
  
def single_oscillator_plus_driving(y, t, h_ex, h_in, pars, sr, time_stop,  
    driving, driving_strength):  
    tau_ex, tau_in, c1, c2, c3, c4 = pars  
  
    index = int(t*sr)  
  
    if index >= time_stop*sr:  
        dydt = zeros(2)  
        return dydt  
  
    h_ex_driven = h_ex + driving_strength*driving[index]  
  
    dydt = (  
        (h_ex_driven - y[0] + c1*sigmoid(y[0]) - c2*sigmoid(y[1]))*tau_ex,  
        (h_in - y[1] + c3*sigmoid(y[0]) - c4*sigmoid(y[1]))*tau_in,  
    )  
  
    return dydt  
  
def single_oscillator_plus_driving_plus_noise(y, t, h_ex, h_in, pars, sr,  
    time_stop, driving, driving_strength, random_data, random_strength):  
    tau_ex, tau_in, c1_ex, c2_ex, c3_in, c4_in = pars  
  
    index = int(t*sr)  
  
    if index >= time_stop*sr:
```

```

dydt = zeros(2)

return dydt

h_ex_driven = h_ex + driving_strength*driving[index]

dydt = (
    (h_ex_driven - y[0] + c1_ex*sigmoid(y[0]) - c2_ex*sigmoid(y[1]) +
     random_strength*random_data[index, 0])*tau_ex,
    (h_in - y[1] + c3_in*sigmoid(y[0]) - c4_in*sigmoid(y[1]) +
     random_strength*random_data[index, 1])*tau_in
)

return dydt

```

```

[122]: def plot_series(time, data, time_begin, time_end, sr):

    N = data.shape[1]//2

    name_vars = ('Ex', 'In')

    no_vars = 2*N

    fig, ax = subplots(ncols=2*N, figsize=(6, 4))

    for ind in arange(no_vars):

        ax[ind].plot(time[time_begin*sr:time_end*sr], data[time_begin*sr:
↪time_end*sr, ind], linewidth=2, c='b')
        ax[ind].set_xticks(linspace(0, time_end-time_begin, 5));
        ax[ind].set_xticklabels(linspace(0, time_end-time_begin, 5));
        ax[ind].set_xlabel('Time', fontsize=12);
        ax[ind].set_ylabel(name_vars[ind], fontsize=12)
        y_min, y_max = ax[ind].get_ylim()
        ax[ind].set_yticks(linspace(y_min, y_max, 3));
        ax[ind].set_yticklabels(around(linspace(y_min, y_max, 3),1),
↪fontsize=14);

    fig.tight_layout()

    return fig, ax

def plot_series_statespace(time, data, time_begin, time_end, sr):

```

```

N = data.shape[1]//2

name_vars = ('Ex', 'In')

no_vars = 2*N

fig, ax = subplots(ncols=2*N, figsize=(6, 4))

ax[0].plot(time[time_begin*sr:time_end*sr], data[time_begin*sr:time_end*sr, 0], linewidth=2, c='b')
ax[0].set_xticks(linspace(0, time_end-time_begin, 5));
ax[0].set_xticklabels(linspace(0, time_end-time_begin, 5));
ax[0].set_xlabel('Time', fontsize=12);
y_min, y_max = ax[0].get_ylim()
ax[0].set_yticks(linspace(y_min, y_max, 3));
ax[0].set_yticklabels(around(linspace(y_min, y_max, 3),1), fontsize=14);
ax[0].set_ylabel(name_vars[0], fontsize=12);

ax[1].plot(data[time_begin*sr:time_end*sr, 1], data[time_begin*sr:
time_end*sr, 0], linewidth=2, c='b')
x_min, x_max = ax[1].get_xlim()
ax[1].set_xticks(linspace(y_min, y_max, 3));
ax[1].set_xticklabels(linspace(0, time_end-time_begin, 3));
ax[1].set_xlabel(name_vars[1], fontsize=12);
ax[1].set_ylabel(name_vars[0], fontsize=12)
y_min, y_max = ax[1].get_ylim()
ax[1].set_yticks(linspace(y_min, y_max, 3));
ax[1].set_yticklabels(around(linspace(y_min, y_max, 3),1), fontsize=14);
ax[1].set_ylabel(name_vars[0], fontsize=12);

fig.tight_layout()

return fig, ax

```

```

[4]: def plot_bifdiagram(results_min_f, results_max_f,
                        results_min_b, results_max_b,
                        par_set):

    N = len(results_min_f)

    fig, ax = subplots()

    for xe, ye in zip(par_set, results_max_f[0]):

        if not isinstance(ye, ndarray):
            ax.scatter(xe, ye, c='r', s=5)

```

```

        else:
            ax.scatter([xe] * len(ye), ye, c='m', s=50, marker='x')

    for xe, ye in zip(par_set, results_min_f[0]):

        if not isinstance(ye, ndarray):
            ax.scatter(xe, ye, c='r', s=5)
        else:
            ax.scatter([xe] * len(ye), ye, c='m', s=50, marker='x')

    for xe, ye in zip(flip(par_set), results_max_b[0]):

        if not isinstance(ye, ndarray):
            ax.scatter(xe, ye, c='r', s=5)
        else:
            ax.scatter([xe] * len(ye), ye, c='b', s=20, marker='P')

    for xe, ye in zip(flip(par_set), results_min_b[0]):

        if not isinstance(ye, ndarray):
            ax.scatter(xe, ye, c='r', s=5)
        else:
            ax.scatter([xe] * len(ye), ye, c='b', s=20, marker='P')

    ax.set_xticks(linspace(par_min, par_max, 5));
    ax.set_xticklabels(around(linspace(par_min, par_max, 5), 2), fontsize=16);
    ax.set_xlabel('Parameter', fontsize=16)

    ax.set_ylabel('Ex', fontsize=14)

    y_min, y_max = ax.get_ylim()

    ax.set_yticks(linspace(y_min, y_max, 3));
    ax.set_yticklabels(around(linspace(y_min, y_max, 3), 2), fontsize=14);

    fig.tight_layout()

    return fig, ax

```

1.4 Time Series

```

[135]: # Excitatory input parameter
h_ex_0    = -4.2
h_in_0    = -4

# Supercritical Hopf parameters

```

```

pars = (1, 1, 4, 6, 6, 0)
# Bistability parameters
# pars = (1, 1, 4, 1, 6, 0)

# Initial conditions
SEED = 123

rng = default_rng()

y_ini = rng.uniform(size=2)
# y_ini = y[-1, :]

# Time array
time_stop = 100
sr         = 1000
time       = linspace(start=0, stop=time_stop, num=time_stop*sr)

# Simulation
y = odeint(func=single_oscillator, y0=y_ini, t=time,
           args=(h_ex_0, h_in_0, pars),
           hmax=0.1)

# Show final values of all variables
print('End of run:', list(around(y[-1,:],3)))
print('')

```

End of run: [-2.076, -9.814]

```

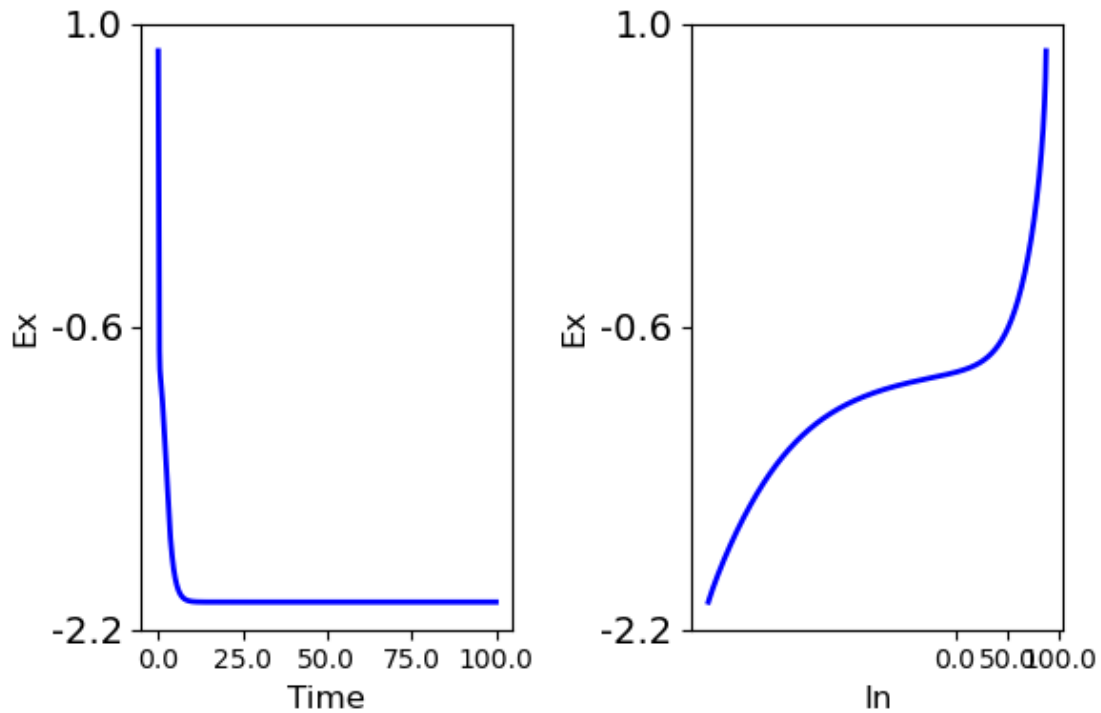
[130]: time_begin, time_end = 0, time_stop

fig, ax = plot_series_statespace(time, y, time_begin, time_end, sr)

title_chars = 'Figs/SNIC_Timeseries_h_ex' + str(h_ex_0) + '.png'
# fig.savefig(title_chars, format='png')
print(title_chars)

```

Figs/SNIC_Timeseries_h_ex-4.2.png



1.5 Bifurcation Diagram

```
[131]: # Initial conditions
y_ini = y[-1, :]

# Bifurcation parameter range
steps = 50

par_min, par_max = -5, 2

par_set = linspace(par_min, par_max, steps)

# Stop time
time_stop = 500
time = linspace(start=0, stop=time_stop, num=time_stop*sr)

results_max_f      = dict()
results_max_inds_f = dict()
results_min_f      = dict()
results_min_inds_f = dict()

rows = time.size
```

```

# Simulation "forward"
for par in par_set:

    h_ex = par

    y_f = odeint(func=single_oscillator, y0=y_ini, t=time,
                  args=(h_ex, h_in_0, pars),
                  hmax=0.1)

    for num, series in enumerate(y_f[rows//2:,-1:2].T):

        if var(series) < 0.00005:

            if num not in results_max_f:

                results_max_f[num] = [series[-1]]
                results_max_inds_f[num] = [0]
                results_min_f[num] = [series[-1]]
                results_min_inds_f[num] = [0]

            else:

                results_max_f[num].append(series[-1])
                results_max_inds_f[num].append(0)
                results_min_f[num].append(series[-1])
                results_min_inds_f[num].append(0)

        else:

            y_f_max_inds = find_peaks(series, distance=100)
            y_f_maxs = series[y_f_max_inds[0]]

            y_f_min_inds = find_peaks(-series, distance=100)
            y_f_mins = series[y_f_min_inds[0]]

            if num not in results_max_f:

                results_max_f[num] = [y_f_maxs]
                results_max_inds_f[num] = [y_f_max_inds]
                results_min_f[num] = [y_f_mins]
                results_min_inds_f[num] = [y_f_min_inds]

            else:

                results_max_f[num].append(y_f_maxs)
                results_max_inds_f[num].append(y_f_max_inds)

```



```

        results_min_f[num].append(y_f_mins)
        results_min_inds_f[num].append(y_f_min_inds)

    if par != par_set[-1]:

        y_ini = y_f[-1, :]

results_max_b      = dict()
results_max_inds_b = dict()
results_min_b      = dict()
results_min_inds_b = dict()

# Simulation "backward"
for par in flip(par_set):

    h_ex = par

    y_b = odeint(func=single_oscillator, y0=y_ini, t=time,
                  args=(h_ex, h_in_0, pars),
                  hmax=0.1)

    for num, series in enumerate(y_b[rows//2:,-1:2].T):

        if var(series) < 0.00005:

            if num not in results_max_b:

                results_max_b[num]      = [series[-1]]
                results_max_inds_b[num] = [0]
                results_min_b[num]      = [series[-1]]
                results_min_inds_b[num] = [0]

            else:

                results_max_b[num].append(series[-1])
                results_max_inds_b[num].append(0)
                results_min_b[num].append(series[-1])
                results_min_inds_b[num].append(0)

        else:

            y_b_max_inds = find_peaks(series, distance=100)
            y_b_maxs     = series[y_b_max_inds[0]]

            y_b_min_inds = find_peaks(-series, distance=100)
            y_b_mins     = series[y_b_min_inds[0]]

```

```

        if num not in results_max_b:

            results_max_b[num]      = [y_b_maxs]
            results_max_inds_b[num] = [y_b_max_inds]
            results_min_b[num]      = [y_b_mins]
            results_min_inds_b[num] = [y_b_min_inds]

        else:

            results_max_b[num].append(y_b_maxs)
            results_max_inds_b[num].append(y_b_max_inds)
            results_min_b[num].append(y_b_mins)
            results_min_inds_b[num].append(y_b_min_inds)

    y_ini = y_b[-1, :]

print('')
print('Scan complete!')
print('')

```

Scan complete!

```

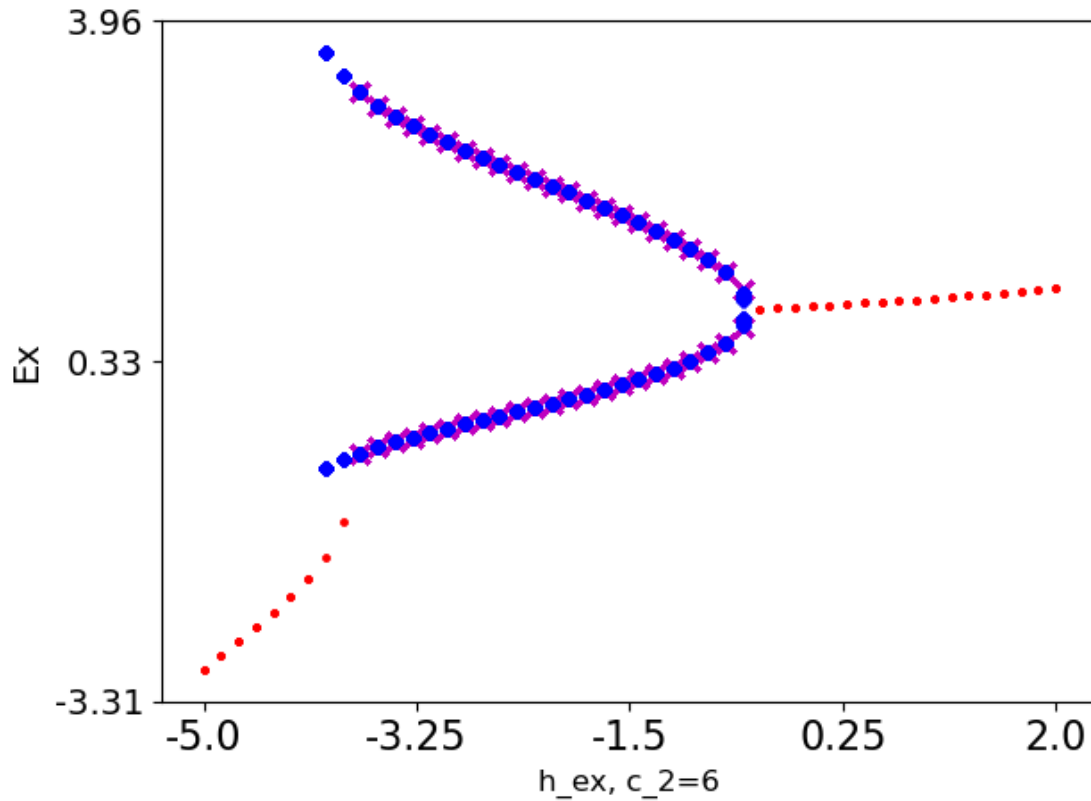
[132]: # Plot
fig, ax = plot_bifdiagram(results_min_f, results_max_f,
                          results_min_b, results_max_b, par_set)

ax.set_xlabel('h_ex, c_2=' + str(pars[3]), fontsize=12)

title_chars = '1osc_Bifs' + '_h_ex, c_2=' + str(pars[3]) + '.png'
# fig.savefig(title_chars, format='png')
title_chars

```

[132]: '1osc_Bifs_h_ex, c_2=6.png'



1.6 Transition to oscillation

```
[136]: # Set Initial conditions:
y_ini = y[-1, :]

# Time array
time_stop = 500
time      = linspace(start=0, stop=time_stop, num=time_stop*sr)

# Initial parameter value
h_ex_0    = -4.2

# Driving
driving = linspace(0, 1, time.size)
driving_strength = 0.5

y = odeint(func=single_oscillator_plus_driving, y0=y_ini, t=time,
           args=(h_ex_0, h_in_0, pars, sr, time_stop, driving,
↳driving_strength), hmax=0.1)
```

```
print('End of run:', y.shape)
print('')
```

End of run: (500000, 2)

```
[137]: time_begin, time_end = 0, time_stop

fig, ax = plot_series_statespace(time, y, time_begin, time_end, sr)

if pars[1] == 1:

    title_chars = 'Figs/N=1/Onset_SNIC_h_ex.png'

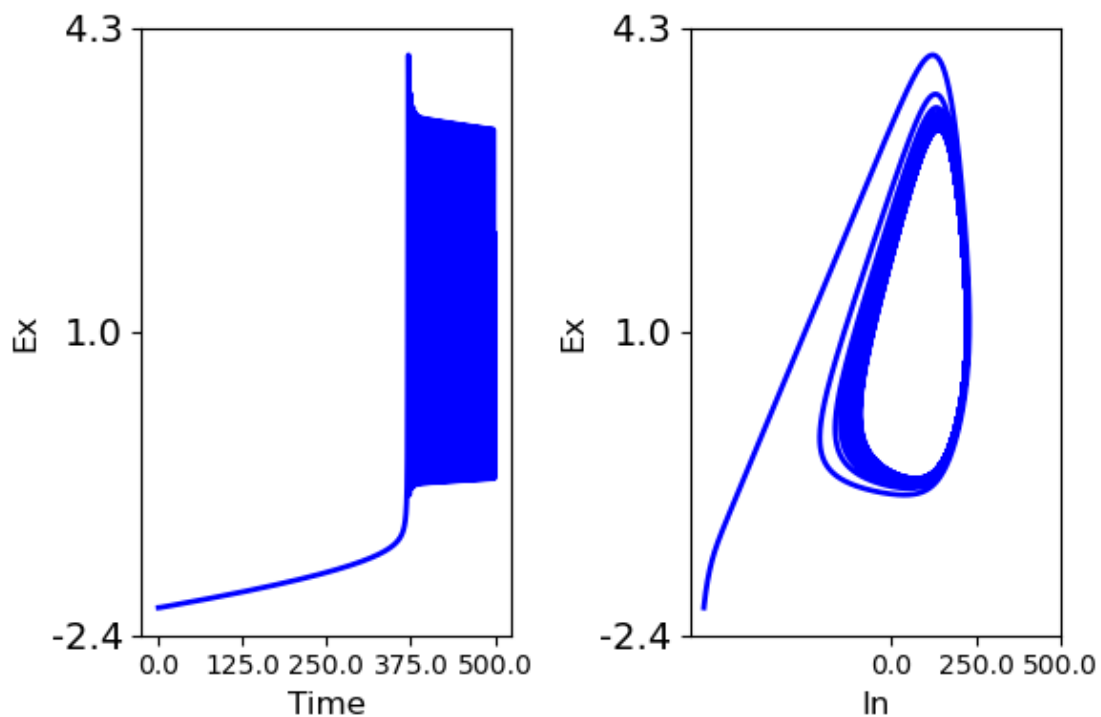
elif pars[1] == 2:

    title_chars = 'Figs/N=1/Onset_Homoclinic_h_ex.png'

# fig.savefig(title_chars, dpi=300, format='png', bbox_inches='tight')

print(title_chars)
```

Figs/N=1/Onset_SNIC_h_ex.png



1.7 With added noise

```
[138]: # Set Initial conditions:
# y_ini = y[-1, :]

# Time array
time_stop = 50
time       = linspace(start=0, stop=time_stop, num=time_stop*sr)

# Initial parameter value
h_ex_0     = -4.2

# Driving
driving = linspace(0, 1, time.size)
driving_strength = 1.0

# Noise
SEED = 123

rng = default_rng(SEED)

random_data = rng.normal(size=(time.size, y.shape[1]))

order, band_low, band_high = 5, 1, 10

sos = butter(order, (band_low, band_high), btype='bandpass', fs=sr,
             ↪output='sos')

random_data_filtered = zeros((time.size, y.shape[1]))

for index, column in enumerate(random_data.transpose()):
    forward    = sosfilt(sos, column)
    backwards  = sosfilt(sos, forward[-1::-1])
    random_data_filtered[:, index] = backwards[-1::-1]

random_strength = 10

# Simulation /Add noise arrays to params
y = odeint(func=single_oscillator_plus_driving_plus_noise, y0=y_ini, t=time,
           args=(h_ex_0, h_in_0, pars, sr, time_stop, driving, driving_strength,
                random_data_filtered, random_strength), hmax=0.1)

print('End of run:')
print('')
```

End of run:

```
[139]: time_begin, time_end = 0, time_stop

fig, ax = plot_series_statespace(time, y, time_begin, time_end, sr)

if pars[1] == 1:

    title_chars = 'Figs/N=1/Onset_SNIC_h_ex.png'

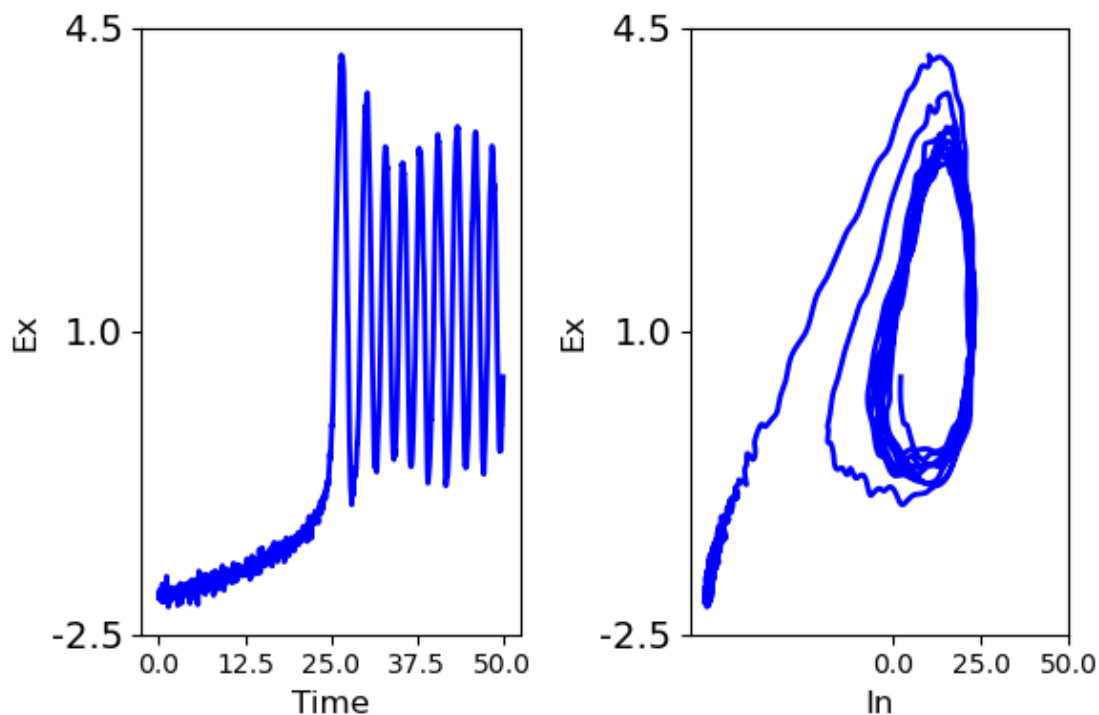
elif pars[1] == 2:

    title_chars = 'Figs/N=1/Onset_Homoclinic_h_ex.png'

# fig.savefig(title_chars, dpi=300, format='png', bbox_inches='tight')

print(title_chars)
```

Figs/N=1/Onset_SNIC_h_ex.png



```
[ ]:
```

2 Try it Yourself

Re-run the above code with these parameter settings:

`pars = (1.2, 0.1, 4, 6, 6, 0)`

`h_ex_0 = 0.2`

`driving_strength = -2`

to simulate an onset with small fast oscillations that grow in amplitude. This is another common type of seizure onset in humans.

3 Notes on the Reading

3.1 A taxonomy of seizure dynamotypes, Maria Luisa Saggio et al

In this paper, transitions to epileptic seizures as recorded in invasive EEG are classified according to a small number of bifurcations in two-variable dynamical systems. Seizure onset is considered as a transition from fixed point to oscillations and there are four types of such transitions as a function of changes in a single parameter. Saddle-node in invariant cycle (SNIC) lead to sudden onset of large amplitude slow frequency oscillations. Supercritical Hopf bifurcations results in small amplitude fast oscillations with increasing amplitude. Figure 1 is a schematic of all possible types of this low-dimensional approach. Appendix 1, page 29, gives clinical examples of seizure onsets and their interpretation according to dynamical systems theory.