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### 1 Complex Oscillations

Oscillators that interact can produce oscillations that are more complex. To interact, oscillators need to be connected with at least one coupling term. Here we study the symmetric interaction between two oscillators as a principal case of self-organised complex oscillations that are found in nonlinear dynamical systems.

#### 1.1 Two Coupled Oscillators with Excitatory Coupling

If we couple two oscillators, we obtain a system of four interacting differential equations: two variables for the first and two for the second oscillator.

The equations:

$$\begin{split} \frac{dEx_1}{dt} &= (h_{ex1} - Ex_1 + c_1 * tanh(Ex_1) - c_2 * tanh(In_1) + coup * Ex_2) * \tau_{ex} \\ \\ \frac{dIn_1}{dt} &= (h_{in1} - In_1 + c_3 * tanh(Ex_1) - c_4 * tanh(In_1)) * \tau_{in} \\ \\ \frac{dEx_2}{dt} &= (h_{ex2} - Ex_2 + c_1 * tanh(Ex_2) - c_2 * tanh(In_2) + coup * Ex_1) * \tau_{ex} \\ \\ \frac{dIn_2}{dt} &= (h_{in2} - In_2 + c_3 * tanh(Ex_2) - c_4 * tanh(In_2)) * \tau_{in} \end{split}$$

where  $Ex_1$ ,  $Ex_2$ ,  $In_1$  and  $In_2$  are the variables.

 $h_{ex1},\,h_{ex2},\,h_{in1},\,h_{in2},\,c_1,\,c_2,\,c_3,\,\mathrm{and}~c_4$  are model parameters.

tanh is the tangens hyperbolicus.

 $\tau_{ex}$ , and  $\tau_{in}$  are the time contants (assumed here to be identical for each oscillator).

coup is the coupling parameter. For coup = 0, there is no coupling and the dynamics is identical to the uncoupled oscillator.

To analyse, we assume that the two oscillators represent different spatial locations. The interaction is based on neurophysiological evidence. For interpretation look out for waveform, frequency composition, temporal and spatial asymmetry (symmetry breaking).

#### 1.2 Import Functions

```
[5]: from scipy.integrate import odeint from scipy.signal import find_peaks

from numpy import zeros, ones, tanh, mod, gradient, linspace, sign, log 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, show

from itertools import product
```

#### 1.3 Model and Functions

```
[14]: def sigmoid(u):
         return tanh(u)
     def N_oscillators(y, t, N, h_ex_rand, h_in_rand, coupling_matrix_E, pars):
         tau_ex, tau_in, c1, c2, c3, c4 = pars
         # Separate Variables
         y_ex = y[:-1:2]
         y_{in} = y[1::2]
         dy_ex, dy_in = zeros(N), zeros(N)
         dydt = zeros(2*N)
         for osc in arange(N):
             coup_E = sum(coupling_matrix_E[:, osc] * y_ex)
             dy_{ex}[osc] = (h_{ex}_{a}[osc] - y_{ex}[osc] + c1*sigmoid(y_{ex}[osc]) - u
       ⇒c2*sigmoid(y_in[osc]) + coup_E)*tau_ex
              dy_in[osc] = (h_in_rand[osc] - y_in[osc] + c3*sigmoid(y_ex[osc]) -__
       )*tau_in
      # Combine Variables
         dydt[:-1:2] = dy_ex
         dydt[1::2] = dy_in
         return dydt
```

```
[7]: def plot_series(time, data, time_begin, time_end, sr):
         N = data.shape[1]//2
         name_vars = ('Ex_', 'In_')
         no_vars = len(name_vars)
         if N == 1:
             fig, ax = subplots(ncols=len(name_vars), figsize=(8, 6))
             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)*sr, 5));
                 ax[ind].set_xticklabels([]);
                 ax[ind].set_xlabel('Time', fontsize=14);
                 ax[ind].set_ylabel(name_vars[ind], fontsize=14)
                 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);
         else:
             y_{max1} = amax(data[:, arange(0, 2*N, 2)])
             if y \max 1 > 0:
                 y_max1_ax = 1.1*y_max1
             else:
                 y_max1_ax = 0.9*y_max1
             y_{min1} = amin(data[:, arange(0, 2*N, 2)])
             if y_min1 > 0:
                 y_{min1}ax = 0.9*y_{min1}
             else:
                 y_{min1}ax = 1.1*y_{min1}
```

```
y_{max2} = amax(data[:, arange(1, 2*N+1, 2)])
      if y_max2 > 0:
           y_{max2_ax} = 1.1*y_{max2}
       else:
           y_max2_ax = 0.9*y_max2
      y_{min2} = amin(data[:, arange(1, 2*N+1, 2)])
      if y_min2 > 0:
           y_{min2}ax = 0.9*y_{min2}
       else:
           y_{min2} = 1.1*y_{min2}
       fig, ax = subplots(ncols=len(name_vars), nrows=N, figsize=(10, 6))
      for osc in arange(N):
           for ind in arange(2):
               if ind == 0:
                   ax[osc, ind].plot(time[time_begin*sr:time_end*sr],_
⇔data[time_begin*sr:time_end*sr, 2*osc+ind], linewidth=1, c='b')
                   ax[osc, ind].set_xticklabels([]);
                   ax[osc, ind].set_ylim(y_min1_ax, y_max1_ax)
                   ax[osc, ind].set_yticks(linspace(y_min1_ax, y_max1_ax, 3));
                   ax[osc, ind].set_yticklabels(around(linspace(y_min1_ax,_
\rightarrowy_max1_ax, 3),1), fontsize=14);
               elif ind == 1:
                   ax[osc, ind].plot(time[time_begin*sr:time_end*sr],_

data[time_begin*sr:time_end*sr, 2*osc+ind], linewidth=1, c='r')

                   ax[osc, ind].set_xticklabels([]);
                   ax[osc, ind].set_ylim(y_min2_ax, y_max2_ax)
                   ax[osc, ind].set_yticks(linspace(y_min2_ax, y_max2_ax, 3));
                   ax[osc, ind].set_yticklabels(around(linspace(y_min2_ax,_
\rightarrowy_max2_ax, 3),1), fontsize=14);
```

```
if osc == N-1:
    ax[N-1, ind].set_xlabel('Time', fontsize=14);
    # print(time[time_begin*sr:time_end*sr].size)
    # print((time_end-time_begin)*sr)
    ax[N-1, ind].set_xticks(linspace(time_begin, time_end, 5));
    ax[N-1, ind].set_xticklabels(linspace(time_begin, time_end, 5));
    ax[N-1, ind].set_xticklabels(linspace(time_begin, time_end, 5));
    label_text = name_vars[ind] + str(osc+1)
    ax[osc, ind].set_ylabel(label_text, fontsize=14)

fig.tight_layout()
    return fig, ax
```

```
[186]: 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(ncols=1, nrows=N, figsize=(5, 4))
           for osc in arange(N):
               # Forward
               for xe, ye in zip(par_set, results_max_f[osc]):
                   if not isinstance(ye, ndarray):
                       ax[osc].scatter(xe, ye, c='k', s=2)
                       ax[osc].scatter([xe] * len(ye), ye, s=2, marker='X', __

¬facecolors='r')
               for xe, ye in zip(par_set, results_min_f[osc]):
                   if not isinstance(ye, ndarray):
                       ax[osc].scatter(xe, ye, c='k', s=2)
                   else:
                       ax[osc].scatter([xe] * len(ye), ye, s=2, marker='X', u

¬facecolors='m')
               ax[osc].set_xticks(linspace(par_min, par_max, 6))
               ax[osc].set_xticklabels([])
```

```
y_min, y_max = ax[osc].get_ylim()
      ax[osc].set_yticks(linspace(y_min, y_max, 3));
      ax[osc].set_yticklabels(around(linspace(y_min, y_max, 3),1),__

¬fontsize=12);
      label_text = 'Osc ' + str(osc+1)
      ax[osc].set_ylabel(label_text, fontsize=14)
  ax[osc].set_xticks(linspace(par_min, par_max, 6));
  ax[osc].set_xticklabels(around(linspace(par_min, par_max, 6), 3),__

¬fontsize=12);
  if par_set[-1] < par_set[0]:</pre>
      chars = 'Right'
  else:
      chars = 'Left'
  title_chars = 'Starts from: ' + chars
  fig.suptitle(title_chars, fontsize=14)
  fig.tight_layout()
  return fig, ax
```

#### 1.4 Time Series

```
[176]: # Number of oscillators
N = 2

# Excitatory input parameter
h_ex_0 = -3.3
h_in_0 = -4

eps = 0.0001
SEED = 1234

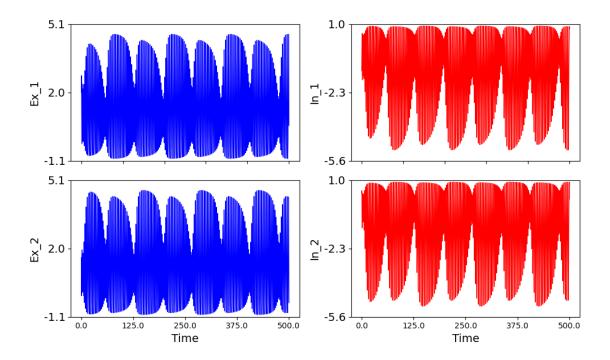
rng = default_rng(SEED)

h_ex_rand = h_ex_0 + eps*rng.normal(0,1,size=N)
h_in_rand = h_in_0 + eps*rng.normal(0,1,size=N)

pars = (1, 1, 4, 6, 6, 0)
```

```
# Coupling matrices
coupling_matrix_E_ini = ones(shape=(N, N))
fill_diagonal(coupling_matrix_E_ini, 0)
coupling_strength_E = 0.3
coupling_matrix_E = coupling_strength_E * coupling_matrix_E_ini
# Initial conditions
SEED = 12
# y_ini = rng.uniform(size=2*N)
y_{ini} = y[-1, :]
# Time array
time\_stop = 500
         = 1000
time
          = linspace(start=0, stop=time_stop, num=time_stop*sr)
# Simulation
y= odeint(func=N_oscillators, y0=y_ini, t=time,
          args=(N, h_ex_rand, h_in_rand, coupling_matrix_E, pars),
          hmax=0.1)
# Show final values of all variables
print('End of run:', list(around(y[-1,:],3)))
print('')
End of run: [0.507, 0.81, -0.441, 0.107]
```

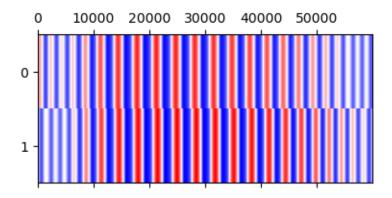
 ${\tt figs/SuperHopf\_Timeseries\_coup\_E\_0.3\_h\_ex-3.3.png}$ 

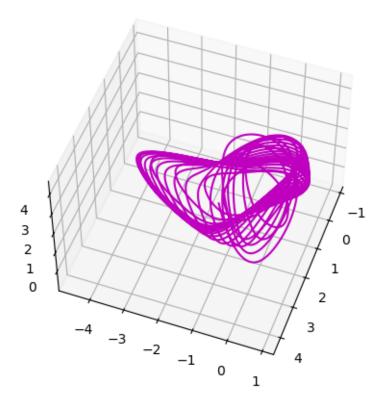


```
[]:
```

```
fig, ax = subplots(figsize=(4.5,2))
stop = 60000
ax.matshow(y[:stop, ::2].T, aspect='auto', cmap='bwr');
show()

fig, ax = subplots(figsize=(6, 4))
ax.set_visible(False)
ax = axes(projection='3d')
ax.plot3D(y[:stop, 0], y[:stop, 1], y[:stop, 2], color='m');
ax.view_init(50, 20);
fig.tight_layout()
show()
```





```
[]:
```

# 1.5 Calculate Correlation

```
[168]: data_corr = corrcoef(y[:stop, 0], y[:stop, 2], rowvar='False')
data_corr[0, 1]
```

#### 1.6 Bifurcation Diagram

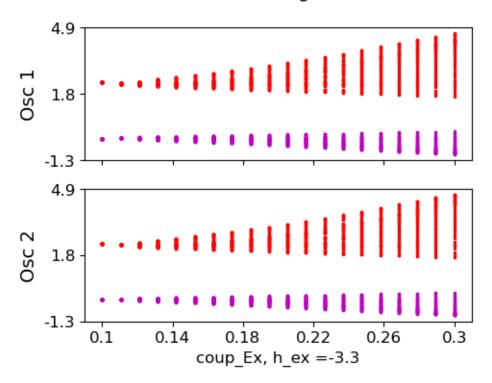
```
[169]: # Initial conditions
       # y_ini = uniform(size=2*N)
       y_{ini} = y[-1, :]
       run_correlation = 'yes' # 'yes' or 'no'
       # Bifurcation parameter range
       steps = 20
       par_min, par_max = 0.30, 0.1
      par_set = linspace(par_min, par_max, steps)
       # Stop time
       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
       data_corrs = list()
       # Simulation "forward"
       for par in par_set:
           coupling_matrix_E_par = par * coupling_matrix_E_ini
           y_f = odeint(func=N_oscillators, y0=y_ini, t=time,
                    args=(N, h_ex_rand, h_in_rand, coupling_matrix_E_par, pars),
                    hmax=0.1)
           for num, series in enumerate(y_f[rows//2:,:-1:2].T):
              if var(series) < 0.00005:</pre>
                   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)
                         = series[y_f_max_inds[0]]
           y_f_maxs
            y_f_min_inds = find_peaks(-series, distance=100)
                         = series[y_f_min_inds[0]]
           y_f_mins
            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 run_correlation == 'yes':
        data_corr = corrcoef(y_f[rows//2:, 0], y_f[rows//2:, 2], rowvar='False')
        data_corrs.append(data_corr[0, 1])
   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):
   coupling_matrix_E_par = par * coupling_matrix_E_ini
   y_b = odeint(func=N_oscillators, y0=y_ini, t=time,
         args=(N, h_ex_rand, h_in_rand, coupling_matrix_E_par, pars),
        hmax=0.1)
   for num, series in enumerate(y_b[rows//2:,:-1:2].T):
        if var(series) < 0.00005:</pre>
            if num not in results_max_b:
                                   = [series[-1]]
                results_max_b[num]
                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)
```

Complete! [2.538, -2.487, 1.936, -3.327]

# Starts from: Right

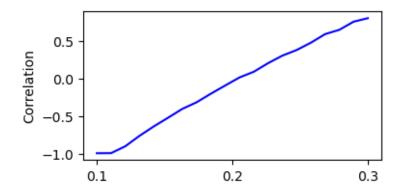


[187]: 'figs/SuperHopf\_Bifs\_coup\_Ex\_h\_ex=-3.3\_back.png'

#### 1.7 Correlation as a Function of Coupling

```
[171]: fig, ax = subplots(figsize=(4, 2))
    ax.plot(flip(par_set), flip(data_corrs), color='b');
    ax.set_xticks(linspace(par_max, par_min, 3))
    ax.set_xticklabels(around(linspace(par_max, par_min, 3), 2));
    ax.set_ylabel('Correlation')
    show()

    title_chars = 'figs/SuperHopf_Corrs_coup_Ex_' + '_h_ex' + str(h_ex_0) + direct
    # fig.savefig(title_chars, format='png')
    title_chars
```



#### [171]: 'figs/SuperHopf\_Corrs\_coup\_Ex\_\_h\_ex-3.3\_back.png'

The correlation between corresponding variables in the two oscillators is not always a simple (e.g. continuously growing) function of the coupling constant.

## 2 Notes on the Reading

# 2.1 Coupled oscillators approach in analysis of bivariate data, M Rosenblum et al.

This paper discribes the behaviour of nonlinear coupled oscillators. It derives the phase relationships as they vary with increasing coupling strength. The proposed method to quantify phase relationships is applicable to experimental data. It is argued, that in reverse, experimentally obtained values of phase relationships allow some conclusion about the strength of the coupling between interacting oscillators.

[]: