

# Nonlinear and linear network theory

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## 1. Lyapunov function for a symmetric network

(a) Show that the function

$$L(x_1, \dots, x_N) = \sum_i \bar{F}(x_i) - \sum_i b_i x_i - \frac{1}{2} \sum_{ij} W_{ij} x_i x_j \quad (1)$$

is nonincreasing under the dynamics

$$\frac{dx_i}{dt} + x_i = f \left( b_i + \sum_j W_{ij} x_j \right) \quad (2)$$

assuming that  $W_{ij} = W_{ji}$ ,  $f$  is monotonic, and  $\bar{F}' = f^{-1}$ . Why does the proof break down if  $W$  is not symmetric? (Hint: use the chain rule to show that  $dL/dt \leq 0$ ).

(b) Show that  $dL/dt = 0$  only at steady states of the dynamics.

(c) Consider the special case of  $f(x) = \tanh(x)$ . Show directly from the dynamics (2) that if the  $x_i$  are all initialized in the range  $[-1, 1]$ , then they must all stay within that range for all time.

(d) What is  $\bar{F}(x)$  for this case? Show that  $L$  is lower bounded in the domain  $[-1, 1]^N$ . Combined with the fact that  $L$  is everywhere decreasing, except at steady states where it is stationary, it follows that  $L$  is a Lyapunov function. (The condition of radial unboundedness is not necessary because the domain is bounded).

## 2. The flip-flop

(a) Simulate the flip-flop dynamics

$$\dot{x}_1 + x_1 = f(-\beta x_2) \quad (3)$$

$$\dot{x}_2 + x_2 = f(-\beta x_1) \quad (4)$$

where  $f(x) = \tanh(x)$  using XPP, MATLAB, or whatever software you choose. Experiment with different values of  $\beta$  and different initial conditions. How many steady states are there for small  $\beta$ ? How many are there for large  $\beta$ ? Which are stable?

(b) Perform a linear stability analysis of the symmetric steady state  $x_1 = x_2 = 0$ . Find the critical value of  $\beta$  at which multistability emerges.

(c) Write down a Lyapunov function for the dynamics using the general result given in the previous problem. Try to visualize the function for different values of  $\beta$ .

## 3. Lateral excitation and inhibition in a linear network. Consider the $N$ -neuron linear network

$$\tau \frac{dx_i}{dt} + x_i = b_i + \sum_{j=1}^N W_{ij} x_j \quad (5)$$

where  $W$  is a *circulant* matrix

$$W = \begin{pmatrix} w_0 & w_{N-1} & \cdot & \cdot & w_1 \\ w_1 & w_0 & w_{N-1} & \cdot & w_2 \\ \cdot & w_1 & w_0 & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot & w_{N-1} \\ w_{N-1} & w_{N-2} & \cdot & w_1 & w_0 \end{pmatrix} \quad (6)$$

This form means that the interaction strength between two neurons depends only on their separation, if the neurons are placed on a ring. The vector  $b$  consists of  $N$  inputs to the neurons, and the vector  $x$  consists of  $N$  neural activities.

In matrix-vector form, the fixed point equations are

$$x = b + Wx \quad (7)$$

- (a) Show that the Fourier modes are eigenvectors of  $W$ , and find the eigenvalues of these modes in terms of the entries in the circulant matrix. Make sure that you find all  $N$  Fourier modes (there is a finite number because of periodic boundary conditions on the ring).
  - (b) Consider the case where  $w_1 = w_{N-1} = c$  and the rest of the entries are zero. For what values of  $c$  is the network stable? In the following questions, restrict  $c$  to these values.
  - (c) Specialize to the inhibitory case  $c < 0$ .
    - i. Solve for the steady-state  $x = (I - W)^{-1}b$  when there is a single input on, and all the rest are zero (impulse response). Describe verbally and sketch graphically the qualitative features of the solution.
    - ii. Sketch the eigenvalue spectrum of the gain matrix  $(I - W)^{-1}$ . Which modes are amplified and which are attenuated?
  - (d) Specialize to the excitatory case  $c > 0$ .
    - i. Solve for the steady-state  $x = (I - W)^{-1}b$  when there is a single input on, and all the rest are zero (impulse response). Describe verbally and sketch graphically the qualitative features of the solution.
    - ii. Sketch the eigenvalue spectrum of the gain matrix  $(I - W)^{-1}$ . Which modes are amplified and which are attenuated?
4. **Linear network integrator** Consider the same linear network as in Eq. (5), but this time with weight matrix diagonalized as

$$W_{ij} = \sum_{\mu} \tilde{W}_{\mu} \xi_i^{\mu} \eta_j^{\mu} \quad (8)$$

where the right ( $\xi$ ) and left eigenvectors ( $\eta$ ) satisfy the orthogonality conditions

$$\delta_{ij} = \sum_{\mu} \xi_i^{\mu} \eta_j^{\mu} \quad (9)$$

$$\delta_{\mu\nu} = \sum_i \xi_i^{\mu} \eta_i^{\nu} \quad (10)$$

- (a) **Conditions for a line attractor** Show that the conditions for a line attractor are that
  - A single eigenvalue must be tuned to exactly one.
  - The real parts of the other eigenvalues must be less than one.
  - The left eigenvector of the unity eigenvalue must be orthogonal to  $b$ .

- (b) **Conditions for the VOR** Suppose that  $b$  contains vestibular input, so that it is modulated by head velocity  $\dot{H}$

$$b_i = b_i^0 + \dot{H}\zeta_i \quad (11)$$

And let eye position be read out by the oculomotor plant according to

$$\tau_E \dot{E} + E = \sum_i \rho_i x_i \quad (12)$$

Analyze the linear transformation from  $H$  to  $E$ . What are the conditions that the system must satisfy to have a good VOR ( $\dot{E} = -\dot{H}$ )? Hint: in general, this can be satisfied only approximately. It might be good to start with the rank one case ( $W_{ij} = \xi_i \eta_j$ ) and then generalize.