

# Problem Set 4 (due March 3)

## Lyapunov functions

March 1, 2005

In lecture, you were told that the stability of

$$\dot{x}_i + x_i = \left[ b_i + \sum_j W_{ij} x_j \right]^+$$

could be analyzed in the nonnegative orthant using the Lyapunov function

$$L(\mathbf{x}) = \frac{1}{2} \mathbf{x}^T (\mathbf{I} - \mathbf{W}) \mathbf{x} - \mathbf{b}^T \mathbf{x}.$$

In this problem you will prove that  $L$  is a Lyapunov function for the specific case of the “winner-take-all” network

$$\dot{x}_i + x_i = \left[ b_i + \alpha x_i - \beta \sum_j x_j \right]^+.$$

1. Lyapunov function for the WTA network.

- Specialize the above general expression for  $L$  to the WTA dynamics.
- Prove that  $L$  is nonincreasing ( $dL/dt \leq 0$ ) on trajectories of the dynamics, with equality only at steady states of the dynamics.  
*Hint:*  $\frac{dL}{dt} = \sum_i \frac{\partial L}{\partial x_i} \dot{x}_i$
- Prove that  $L$  is lower bounded if  $\alpha < 1 + \beta$  in the nonnegative orthant, and is not lower bounded if  $\alpha > 1 + \beta$ .
- Prove that  $L$  is *radially unbounded* ( $L(c\mathbf{x}) \rightarrow \infty$  as  $c \rightarrow \infty$ ) for  $\alpha < 1 + \beta$  and  $x$  in the nonnegative orthant. This completes the proof that  $L$  is a Lyapunov function of the network dynamics.

*Note:* In class we talked about co-positivity of the  $I - W$  matrix which is a sufficient condition for  $L$  to be radially unbounded.

2. 2-neurons network

Remember what you have learned in class for the general case of  $N$  neurons. This network architecture has three distinct regimes: (a) Weak excitation ( $\alpha < 1$ ) could lead to  $k$  active neurons depending on how well-separated the inputs are. (b) Strong excitation ( $\alpha > 1$ ) can lead to a winner-take-all operation and a single active neuron for well-separated inputs. (c) There exists a third case which is called the integration regime for  $\alpha = 1$  for which only the maximally activated neuron is active.

- Specialize the general expression for  $L$  to the WTA dynamics with two neurons. You will have to refer to this equation later.

- (b) Make an input phase-diagram for each network regime, *i.e.* for each of the conditions  $\alpha < 1$ ,  $\alpha = 1$ ,  $1 < \alpha < 1 + \beta$ ,  $\alpha > 1 + \beta$ , make a plot of the input space ( $b_1$  vs  $b_2$ ) and characterize the different regions for which  $x_1$  and/or  $x_2$  can be active.
- (c) Study the Lyapunov function associated with each of those regimes and show how they relate to the phase-diagram you drew previously. Explore 'representative' values of  $\alpha$  and  $\beta$  and in each case, describe the shape of  $L$ , explain how many possible steady-states there are and if those are stable or not. Show the effect of changing the network inputs on  $L$ . We expect you to submit plots containing the Lyapunov function  $L$  and the trajectories followed by the network dynamics (you can use the 'mesh' command to plot  $L$  and the 'hold' command to superimpose the two plots). You should also submit the parameters you used to simulate your particular dynamics.
- (d) You showed in problem 1 that  $L$  is not a Lyapunov function of the dynamics if  $\alpha > 1 + \beta$ . What happens to the network in this case? Justify your answer using both a plot as well as mathematical arguments. .

### 3. Unconditional MAX behavior

In the following you will show that when the network activities are all initialized to 0 (*i.e.*  $x(0) = 0$ ), the network **always** selects the unit that receives the maximum input as the winner for  $\alpha > 1$ .

- (a) Show that by changing variables, the network equation can be written as:

$$\dot{u}_i + u_i = b_i + \sum_j W_{ij} [u_j]^+$$

- (b) Show that if  $u_i = u_j$  then  $\frac{d}{dt}(u_i - u_j) = b_i - b_j$ .
- (c) Show that the only possible 'winner' has to be the one receiving the maximal input.