## ECE 586 BH: Interplay between Control and Machine Learning Fall 2023 Homework 2

Instructor: Bin Hu

Due date: Oct 10, 2023

1. (60 points) Consider  $x_{k+1} = Ax_k + B_k \sigma (C_k x_k + b_k)$ , where  $\sigma$  is the ReLU activation function. We know that if there exist a positive definite diagonal matrix  $\Lambda_k$  such that the following matrix inequality holds

$$\begin{bmatrix} A_k^{\mathsf{T}} A_k - I & A_k^{\mathsf{T}} B_k \\ B_k^{\mathsf{T}} A_k & B_k^{\mathsf{T}} B_k \end{bmatrix} + \begin{bmatrix} C_k & 0 \\ 0 & I \end{bmatrix}^{\mathsf{T}} \begin{bmatrix} 0 & \Lambda_k \\ \Lambda_k & -2\Lambda_k \end{bmatrix} \begin{bmatrix} C_k & 0 \\ 0 & I \end{bmatrix} \le 0,$$
(1)

then we have  $||x'_{k+1} - x_{k+1}|| \le ||x'_k - x_k||$ . Apply the above condition to show

(a) The convex potential layer  $x_{k+1} = x_k - \frac{2}{\|W_k\|^2} W_k \sigma(W_k^{\mathsf{T}} x + b_k)$  is 1-Lipschitz.

(b) The layer  $x_{k+1} = x_k - 2W_k\sigma(W_k^{\mathsf{T}}x_k + b_k)$  with  $W_k^{\mathsf{T}}W_k = I$  is 1-Lipschitz.

(c) The layer  $x_{k+1} = x_k - 2W_k \operatorname{diag}(\sum_{j=1}^n |W_k^\mathsf{T} W_k|_{ij})^{-1} \sigma(W_k^\mathsf{T} x_k + b_k)$  is 1-Lipschitz (here we assume  $\sum_{j=1}^n |W_k^\mathsf{T} W_k|_{ij} \neq 0$  for all *i*, and *n* is the column dimension of  $W_k^\mathsf{T} W_k$ ).

(d) The layer  $x_{k+1} = \sqrt{2}M_k^{\mathsf{T}}\Psi_k\sigma(\sqrt{2}\Psi_k^{-1}N_kx_k + b_k)$  with  $\Psi_k$  being any positive definite diagonal matrix and  $(M_k, N_k)$  being any matrix pair satisfying  $M_kM_k^{\mathsf{T}} + N_kN_k^{\mathsf{T}} = I$  is 1-Lipschitz.

**2**. (10 points) Consider the deep equilibrium model  $z = \sigma(Wz + Ux + b_z)$  with  $\sigma$  being ReLU. Suppose the model is set up in a way that there is a unique solution z for any input x (so we assume the deep equilibrium model is well-posed). Derive a matrix inequality condition to ensure  $||z' - z|| \leq L||x' - x||$  for any (x', x). (Notice that z is the output of x, and z' is the output of x'.)

Hint: Since  $\sigma$  is ReLU, we must have

$$\begin{bmatrix} y' - y \\ \sigma(y') - \sigma(y) \end{bmatrix}^{\mathsf{T}} \begin{bmatrix} 0 & \Lambda_k \\ \Lambda_k & -2\Lambda_k \end{bmatrix} \begin{bmatrix} y' - y \\ \sigma(y') - \sigma(y) \end{bmatrix} \ge 0$$

where  $\Lambda_k$  is any positive definite diagonal matrix. Use the above quadratic constraint to derive the matrix inequality condition.

**3**. Consider a control affine system  $\dot{x} = f(x) + g(x)u$ .

(a) (15 points) Suppose we want the system to avoid the area defined by  $\{x : ||x - x*|| \le 3\}$  where  $x^*$  is a prescribed point. We assume that the perfect state measurement is available. In addition, we also assume that a baseline controller is given by u = K(x), where K is some neural network policy. How can we apply the control barrier function (CBF) approach to solve this problem?

(b) (15 points) Now suppose only a state estimate  $\hat{x}$  is provided. A prescribed error bound is given, i.e. we always have  $||x(t) - \hat{x}(t)|| \leq r$  for our system. We still want the system to avoid the area defined by  $\{x : ||x - x^*|| \leq 3\}$ . How can we achieve this via a CBF-based approach? You can make assumptions on Lipschitz properties as needed.