Stability Analysis of Discrete time Recurrent Neural Networks

Jayant Singh Advisor: Dr. Nikita Barabanov

> Department of Mathematics North Dakota State University Fargo,ND

> > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > < □ > <

- Problem Statement
- Applications of Recurrent Neural Networks (RNN)
- Absolute stability
- Novel Approach-Reduction of dissipativity domain

◆□▶ ◆□▶ ◆三▶ ◆三▶ 三三 のへぐ

Summary

Consider an example of discrete time RNN

$$\begin{aligned} x_{1}^{k+1} &= \tanh(W_{1}x_{1}^{k} + V_{n}x_{n}^{k} + b_{1}) \\ x_{2}^{k+1} &= \tanh(W_{2}x_{2}^{k} + V_{1}x_{1}^{k+1} + b_{2}) \\ \dots \\ x_{n}^{k+1} &= \tanh(W_{n}x_{n}^{k} + V_{n-1}x_{n-1}^{k+1} + b_{n}) \end{aligned}$$
(1)

where x_n^k is the state vector of *nth* layer at step k, W_n , V_n are weight matrices , and b_n represents the bias vector.

Problem 1.

Analyse the problem of global asymptotic stability of the RNN described above. (i.e. analyse the problem of global asymptotic stability of equilibrium point of the RNN.)

The applications of RNN include, but are not limited to

▲□▶ ▲圖▶ ▲臣▶ ▲臣▶ ―臣 … のへで

- Voice recognition,
- Pattern recognition, and
- Modelling of non-linear systems

Previous Stability Results Theory of Absolute Stability

• Consider a discrete time SISO system:

$$x^{k+1} = Ax^{k} + B\xi^{k}, \sigma^{k} = Cx^{k}$$

$$\xi^{k} = \varphi(\sigma^{k})$$
(2)

◆□▶ ◆□▶ ◆三▶ ◆三▶ 三三 のへぐ

where, A ,B, C are matrices , $\varphi(\cdot)$ satisfies some condition. • Analyze the stability of (2) • Consider $V(x) = x^*Hx$, where $H = H^* > 0$. $\Rightarrow V(x^{k+1}) - V(x^k) = (Ax^k + B\xi^k)^* H(Ax^k + B\xi^k) - (x^k)^* Hx^k$

• We want $V(x^{k+1}) - V(x^k) < 0$ for all $(x^k, \xi^k) \neq 0.$ s.t. $\xi^k = \varphi(x^k)$,and $\varphi(\cdot)$ satisfies some constraint.

Problem 2.

Suppose F is a quadratic function. Moreover, assume there exists matrix L s.t. A + BL is stable(i.e.(A,B) is stabilizable), and $F(x, Lx) \ge 0$. Find necessary and sufficient conditions for the existence of $H = H^* > 0$ s.t.

$$(Ax + B\xi)^* H(Ax + B\xi) - x^* Hx + F(x,\xi) < 0$$
(3)

for all $(x, \xi) \neq 0$.

Solution: Necessary condition is given by $\mathbb{R}e(F((e^{i\omega}I - A)^{-1}Bw, w)) < 0$ for all $\omega \in [0, \pi]$ and $w \neq 0$, called the Frequency domain condition. And sufficient condition is provided by Kalman Szegö Lemma which says that

Lemma 3.

Assume (A, B) is stabilizable. And $\mathbb{R}e(F((e^{i\omega}I - A)^{-1}Bw, w)) < 0$ for all $\omega \in [0, \pi]$ and $w \neq 0$. Then there exists $H = H^* > 0$ s.t.

 $(Ax+B\xi)^*H(Ax+B\xi)-x^*Hx+F(x,\xi)<0$

for all $(x,\xi) \neq 0$.

As a consequence, there exists $H = H^* > 0$ such that x^*Hx is a Lyapunov function.

- In case of $\mathsf{RNN}, \varphi(\cdot) = \mathsf{tanh}(\cdot),$
- $0 \leq \frac{\tanh(\sigma)}{\sigma} \leq 1$ (sector condition)
- $\varphi(\sigma)(\sigma \varphi(\sigma)) \ge 0$ is the quadratic function



Shortcomings in absolute stability approach

 A more general stability criteria should be developed.



- $x^{k+1} = \phi(x^k)$, $\phi(\cdot)$ is bounded non-linear function.
- Construct $\{D_k\}$ such that $D_{k+1} \subsetneq D_k, \phi(D_k) \subset D_{k+1}$ then $x^k \in D_k$, provided that $x^0 \in D_0$. Thus if $D_k \to 0$, then $x^k \to 0$, as $k \to \infty$.
- $D_{k+1} = \{x \in D_k : f_{k+1,j}(x) \le \alpha_{k+1,j}, j = 1 \dots m_{k+1}\}$ where m defines the number of constraints, $f_{k,j}$ defines the linear function, $\alpha_{k+1,j} := \max_{x \in D_k} f_{k,j}(\phi(x))$.

Convex Lyapunov function and constrained optimization problem

Theorem 4.

Define $\alpha_j^{k+1} = \max_{y \in D_k} (f_j(\phi(y)))$. Assume system $x^{k+1} = \phi(x^k)$ has a convex Lyapunov function. Then there exists linear functions $f_1, f_2, \ldots f_m$ such that $D_{k+1} = \{y : f_j(y) \le \alpha_j^{k+1}, j = 1 \ldots m\}$, and $\{D_k\} \to 0$.

▲□▶ ▲□▶ ▲□▶ ▲□▶ ▲□ ● ● ●

Construction of sets



Addition of new constraints



Addition of new constraints contd.



 We assume existence of convex Lyapunov function .On the other hand, in theory of absolute stability approach, we considered the problem of existence of quadratic Lyapunov function(a type of convex function).



Problem 5.

Given the function $f(x) = \sum_{i=1}^{n} c_i \phi(x_i)$, where $c_i \neq 0$ for all *i*. How to locate the points of local maxima for $f(\cdot)$ over a convex set(for our case it is a rectangle)?

▲□▶ ▲□▶ ▲□▶ ▲□▶ ▲□ ● ● ●

- The stability problem of discrete time RNN is studied
- Results from theory of absolute stability has been discussed
- The method of reduction of dissipativity domain has been presented

◆□▶ ◆□▶ ◆三▶ ◆三▶ 三三 のへぐ

• Some applications of RNN have been mentioned