



MECH 6v29.002 – Model Predictive Control

L7 – MPC Theory

- Stability
 - Lyapunov Stability
 - Closed-loop MPC Stability

- The **closed-loop stability** of a system under MPC is a complex function of the

- Model: A, B, C
- Control parameters:
 $N, Q \geq 0, R > 0, P \geq 0$
- Constraints: $u_{\min}, u_{\max}, y_{\min}, y_{\max}$

$$J_0^*(x_0) = \min_{U_0} \sum_{k=0}^{N-1} x_k^T Q x_k + u_k^T R u_k + x_N^T P x_N$$

s.t.

$$x_{k+1} = Ax_k + Bu_k, \quad k \in \{0, 1, \dots, N-1\}$$

$$u_{\min} \leq u_k \leq u_{\max}, \quad k \in \{0, 1, \dots, N-1\}$$

$$y_{\min} \leq Cx_k \leq y_{\max}, \quad k \in \{1, 2, \dots, N\}$$

$$x_0 = x(0)$$

- Multiple approaches to guarantee stability
 - **Terminal constraint:** $x_{k+N} = 0$
 - **Relaxed terminal constraint:** $x_{k+N} = \Omega$
 - **Terminal cost:** $x_N^T P x_N$, P satisfies DARE
 - **Contraction constraints:** $\|x_{k+1}\| \leq \alpha \|x_k\|$, $\alpha < 1$
- We will explore some of these

- While there are multiple ways of defining and analyzing stability, we will focus on **Lyapunov stability**
- Consider the generic, autonomous system

$$x_{k+1} = f(x_k)$$

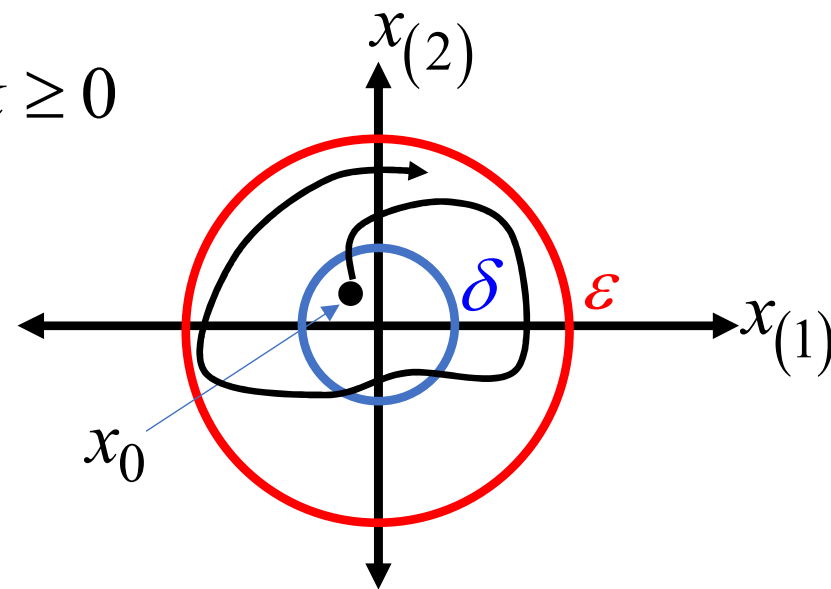
$$f(0) = 0$$

$x = 0$ is an equilibrium

- The equilibrium is

- **Stable** – if $\forall \varepsilon > 0, \exists \delta = \delta(\varepsilon) > 0$ s.t.

$$\|x_0\| < \delta \Rightarrow \|x_k\| < \varepsilon, \forall k \geq 0$$



- **Unstable** - if it is not stable

- Consider the generic, autonomous system

$$x_{k+1} = f(x_k)$$

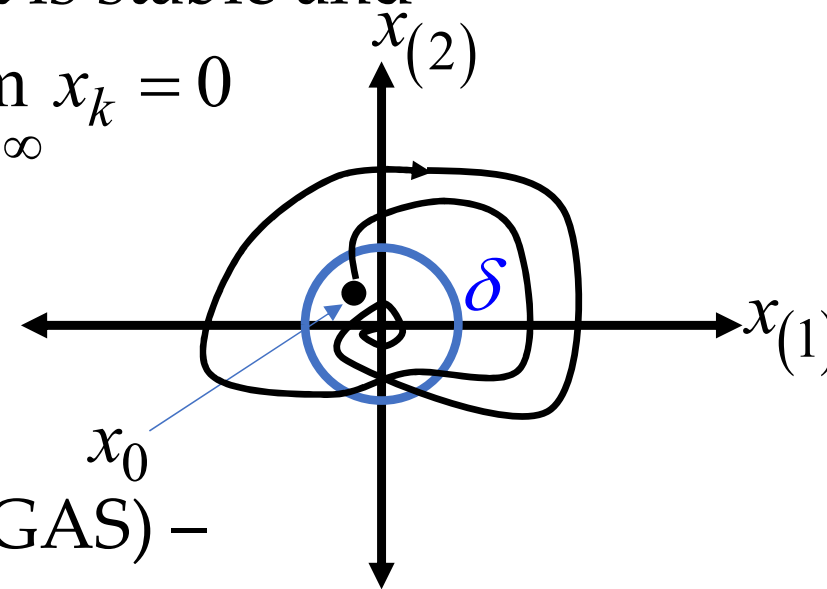
$$f(0) = 0$$

$x = 0$ is an equilibrium

- The equilibrium is

- Asymptotically stable** (AS) – if it is stable and

$$\exists \delta \text{ s.t. } \|x_0\| < \delta \Rightarrow \lim_{k \rightarrow \infty} x_k = 0$$



- Globally asymptotically stable** (GAS) –
if it is AS $\forall x_0 \in \mathbb{R}^n$

- Exponentially stable** – if it is stable and $\exists \alpha > 0, \gamma \in (0, 1)$ s.t.

$$\|x_0\| < \delta \Rightarrow \|x_k\| \leq \alpha \|x_0\| \gamma^k, \quad \forall k \geq 0$$

- Norm is bounded by an exponentially decaying envelope

- Lyapunov stability is determined based on the existence of a Lyapunov function
- **Lyapunov Stability Theorem**
 - Consider the equilibrium $x = 0$ of $x_{k+1} = f(x_k)$.
 - Let $\Omega \in \mathbb{R}^n$ be a closed and bounded set containing the origin.
 - Let $V : \mathbb{R}^n \rightarrow \mathbb{R}$ be a function, continuous at the origin, s.t.
 - 1) $V(0) = 0$
 - 2) $V(x) > 0, \forall x \in \Omega \setminus \{0\}$
 - 3) $V(x_{k+1}) < V(x_k), \forall x_k \in \Omega \setminus \{0\}$
 - Then $x = 0$ is asymptotically stable in Ω .
- A function $V(x)$ satisfying the above conditions is a **Lyapunov function** for our system

Lyapunov Stability Details

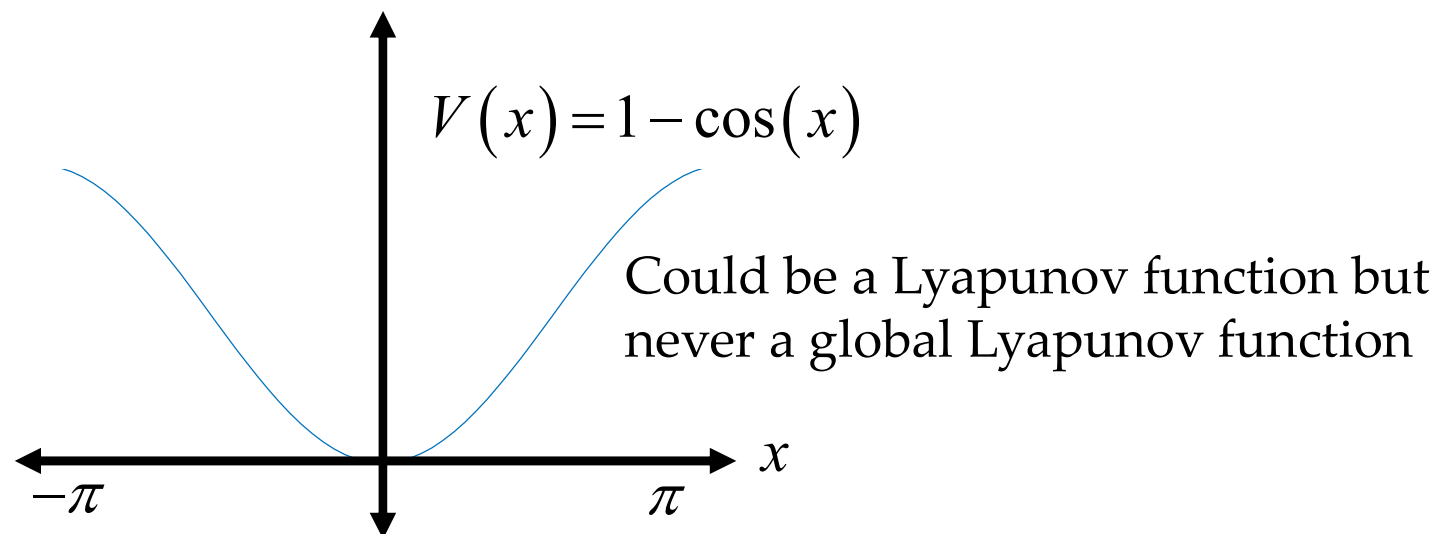


- Conceptually, it is often useful to think about the Lyapunov function as a **measure of energy in the system**
 - Energy must be zero at the equilibrium $V(0) = 0$
 - Energy must be positive $V(x) > 0, \forall x \in \Omega \setminus \{0\}$
 - Energy must decrease as the system evolves
$$V(x_{k+1}) < V(x_k), \forall x_k \in \Omega \setminus \{0\}$$
- The Lyapunov stability theory is only sufficient (not necessary)
 - just because you can't find a Lyapunov function, doesn't mean the system is not stable

- Global Lyapunov Stability Theorem

- Consider the equilibrium $x = 0$ of $x_{k+1} = f(x_k)$.
- Let $V : \mathbb{R}^n \rightarrow \mathbb{R}$ be a function, continuous at the origin, s.t.
 - 1) $V(0) = 0$
 - 2) $V(x) > 0, \forall x \neq 0$
 - 3) $V(x_{k+1}) < V(x_k), \forall x_k \neq 0$
 - 4) $\|x\| \rightarrow \infty \Rightarrow V(x) \rightarrow \infty$ ← Radially unbounded
- Then $x = 0$ is globally asymptotically stable.

- Example:



• Stability Theorem

- A linear system $x_{k+1} = Ax_k$ is globally asymptotically stable in the sense of Lyapunov if and only if all eigenvalues of A are inside the unit circle.
- Note that stable linear systems are always globally stable

• Proof:

- Try the candidate quadratic Lyapunov function

$$V(x) = x^T P x, \quad P > 0$$



1) $V(0) = 0$



2) $V(x) > 0, \forall x \neq 0$

3) $V(x_{k+1}) < V(x_k), \forall x_k \neq 0$



4) $\|x\| \rightarrow \infty \Rightarrow V(x) \rightarrow \infty$

$$V(x_{k+1}) < V(x_k)$$

$$V(x_{k+1}) - V(x_k) < 0$$

$$x_{k+1}^T P x_{k+1} - x_k^T P x_k < 0$$

$$x_k^T A^T P A x_k - x_k^T P x_k < 0$$

$$x_k^T (A^T P A - P) x_k < 0$$

$$-x_k^T Q x_k < 0 \Rightarrow x_k^T Q x_k > 0$$

Discrete-time Lyapunov equation

$$A^T P A - P = -Q, \quad Q > 0$$

Existence of P is guaranteed for any Q

MPC Stability Proof



- Use Lyapunov Stability to guarantee closed-loop asymptotic stability of MPC with terminal constraints

$$J_0^*(x_0) = \min_{U_0} \sum_{k=0}^{N-1} q(x_k, u_k) + p(x_N)$$

s.t.

$$x_{k+1} = f(x_k, u_k), \quad k \in \{0, 1, \dots, N-1\}$$

$$h(x_k, u_k) \leq 0, \quad k \in \{0, 1, \dots, N-1\}$$

$$x_N = 0$$

$$x_0 = x(0)$$

$$f(0) = 0$$

$x = 0$ is an equilibrium

- Assume feasible at time step 0
- Solve for $u_{0|0}^*$ based on $x(0)$
- Optimal (minimal) cost is $J_0^*(x_0)$
- System evolves to $x(1) = f(x(0), u_{0|0}^*)$

MPC Stability Proof

- At time step 1, consider the (suboptimal) input trajectory

$$U_1 = \begin{bmatrix} u_{1|0}^* \\ \vdots \\ u_{N-1|0}^* \\ \mathbf{0} \end{bmatrix}$$

- Cost of this candidate solution is

$$J_0^*(x_0) - q(x_0, u_{0|0}^*) + \underbrace{q(x_{N+1}, 0)}_{=0}$$

Because of terminal constraint at previous time step

$$J_0^*(x_0) = \min_{U_0} \sum_{k=0}^{N-1} q(x_k, u_k) + p(x_N)$$

s.t.

$$x_{k+1} = f(x_k, u_k), \quad k \in \{0, 1, \dots, N-1\}$$

$$h(x_k, u_k) \leq 0, \quad k \in \{0, 1, \dots, N-1\}$$

$$x_N = 0$$

$$x_0 = x(0)$$

- Since this is a suboptimal solution

$$J_1^*(x_1) \leq J_0^*(x_0) - q(x_0, u_{0|0}^*)$$

- Since the system dynamics and cost function are time invariant

$$J_1^*(x_1) = J_0^*(x_1) \Rightarrow J_0^*(x_1) \leq J_0^*(x_0) - q(x_0, u_{0|0}^*) \Rightarrow J_0^*(x_1) \leq J_0^*(x_0)$$

- Can now show that $J_0^*(x)$ is a Lyapunov function