

PSS 4 - Stability

Today: Exercise #1; Exercise 5.2

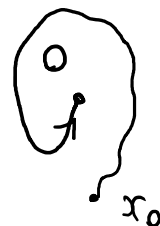
Summary:

plant: $x(k+1) = f(x(k))$, with $f(0) = 0$.

Stability?



Asymptotic stability:



Lyapunov functions: $V: \mathbb{R}^n \rightarrow \mathbb{R}^+$

V is a Lyapunov function if: (i) $\alpha_1(|x|) \leq V(x) \leq \alpha_2(|x|)$
(ii) $V(f(x)) - V(x) \leq -\alpha_3(|x|)$

where α_1, α_2 are k^∞ functions and α_3 is a positive definite function.

(to simplify, think of (i) and (ii) as: (i) $V(x) \geq 0$, $x \neq 0$, $V(0) = 0$.
(ii) $V(f(x)) - V(x) < 0$)

Prop 10.1: If $V(\cdot)$ is a Lyapunov function, then the origin is globally asymptotically stable.

What about linear systems? $x(k+1) = Ax(k)$

Easy: stability $\Leftrightarrow \text{eig}(A)$ inside the unit circle.

Still, we can build some Lyapunov functions:

$$V(x) = x^T S x, \text{ with } S \succ 0.$$

Is $V(\cdot)$ a Lyap function?

(i) OK

$$(ii) V(x_{k+1}) - V(x_k) = -x^T \underbrace{(S - A^T S A)}_Q x = -x^T Q x$$

$V(\cdot)$ is Lyap $\Leftrightarrow Q \succ 0$.

Lemma 2.7: A stable $\Leftrightarrow \forall Q \succ 0, \exists! S \succ 0$ such that the discrete time Lyapunov equation holds: $S - A^T S A = Q$

Exercise #1:

$$\text{Consider } x(k+1) = \begin{bmatrix} 0.5 & 1 \\ -0.1 & 0.2 \end{bmatrix} x(k)$$

a) Find a Lyap function $V(x) = x^T S x$

Hint: dlyap in Matlab.

first, A is asymptotically stable since $\text{eig}(A) = 0.35 \pm 0.28i$

Lemma 2.7 : pick $Q = I_2$, then solve Lyap equation.

$$S = \begin{bmatrix} 1.29 & 0.53 \\ 0.53 & 2.63 \end{bmatrix} \succ 0, \text{ so } V(x) = x^T S x \text{ is a Lyap function.}$$

! use A^T for A in dlyap (cf its doc).

b) Unconstrained ∞ horizon RHC:

$$V(u, x_0) = \sum_{i=0}^{\infty} (x(i)^T Q x(i) + u(i)^T R u(i)), \quad Q = I_2, R = 1$$

$$\text{s.t. } x(k+1) = Ax(k) + Bu(k), \quad A = \begin{bmatrix} 1 & 1 \\ -1 & 5 \end{bmatrix}, \quad B = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

Show that $V(u, x_0)$ is a Lyap function for the closed-loop system.

Closed-loop control by solving DARE: (use Matlab)

$$P = \begin{bmatrix} 13.66 & 7.39 \\ 7.39 & 39.44 \end{bmatrix} \quad K = \begin{bmatrix} -0.78 & 5.07 \end{bmatrix}$$

new use:

$$V(u, x_0) = x_0^T P x_0 \quad \text{since } \infty \text{ horizon control.}$$

To verify that this Lyap candidate is a Lyap function:

(i) $P \succ 0$?

(ii) $V(u, x_1) - V(u, x_0) = -x_0^T Q x_0$, is $Q \succ 0$?

(i) $\text{eig}(P) = \{11.4, 41.7\} \Rightarrow P \succ 0$

(ii)
$$\begin{aligned} V(u, x_1) - V(u, x_0) &= x_1^T P x_1 - x_0^T P x_0 \\ &= x_0^T (A+BK)^T P (A+BK) x_0 - x_0^T P x_0 \\ &= -x_0^T Q x_0 \end{aligned}$$

where $Q = P - (A+BK)^T P (A+BK) = \begin{bmatrix} 1.60 & -3.95 \\ -3.95 & 26.75 \end{bmatrix}$

$\text{eig}(Q) = \{1; 27, 35\}$

(i) and (ii) ~~and~~: $V(u, x_0)$ is a Lyapunov function.

5.2

* First, we prove this in the general case.

Here: $u(k+i) = 0, \forall i \geq M$.

assume: $z(k) = x(k)$.

$$V(k) = \underbrace{\sum_{i=M}^{\infty} \|z(k+i)\|_Q^2}_{\text{tail}} + \sum_{i=0}^{M-1} \left\{ \|z(k+i)\|_Q^2 + \|\Delta u(k+i)\|_R^2 + \|u(k+i)\|_S^2 \right\}$$

tail: $\sum_{i=M}^{\infty} \|z(k+i)\|_Q^2 = x(k+M)^T \bar{Q} x(k+M)$.

$= A x(k+M) + \underbrace{B u(k+M)}_{=0}$

How? $\sum_{i=M}^{\infty} \dots = x(k+M)^T Q x(k+M) + x(k+M+1)^T Q x(k+M+1) + \dots$

$$= x(k+M)^T \underbrace{\left(\sum_{i=0}^{\infty} (A^T)^i Q A^i \right)}_{=\bar{Q}} x(k+M)$$

How do we get rid of the infinite sum?
Trick: use Lyap equation!

$$A^T \bar{Q} A = \sum_{i=0}^{\infty} (A^T)^{i+1} Q A^{i+1} \quad \downarrow j=i+1$$

$$= \sum_{j=1}^{\infty} (A^T)^j Q A^j \quad \downarrow j=i+1$$

this is $\bar{Q}$ without its first term

$$A^T \bar{Q} A = \bar{Q} - Q \quad \text{Lyap equation. } (S \leftrightarrow \bar{Q})$$

Lemma 2.7: A stable $\Leftrightarrow \forall Q, \exists! \bar{Q}$.

We can now compute $\bar{Q}$ by solving Lyap equation.

$$\begin{aligned} \min V(k) &= \sum_{i=0}^{M-1} \{ \|x(k+i)\|^2 + \dots \} + z(k+M)^T \bar{Q} z(k+M) \quad (P) \\ \text{s.t. } x(k+1) &= Ax(k) + Bu(k) \end{aligned}$$

We have proved that $(P) \Leftrightarrow (P-I)$, if A stable.

(*) Second: assemble (P) as a QP with the given values.

$M=2$ so optimization variables: $w = [x(k+1) \ x(k+2) \ u(k) \ u(k+1)]^T$

$$H = \begin{bmatrix} Q & 0 & 0 & 0 \\ 0 & \bar{Q} & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \quad \text{solve: } (0.9^2 - 1) \bar{Q} = -1 \quad (\text{from part 1})$$

so QP: $\min_w V(k) = w^T H w$ can be solved!
s.t. $A_{eq} w = b_{eq}$