

Problem 1

Use stability definitions to determine whether the following systems are Lyapunov stable (i.e. stable in the sense of Lyapunov), asymptotically stable, globally asymptotically stable, or none. The first two systems are in \mathbb{R}^2 and the last one is in \mathbb{R} .

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|--|--|
| <p>(a) $\dot{x}_1 = 0$
 $\dot{x}_2 = -x_2$</p> | <p>(c) $\dot{x} = 0$ if $x > 1$
 $\dot{x} = -x$ if $x \leq 1$</p> |
| <p>(b) $\dot{x}_1 = -x_2$
 $\dot{x}_2 = 0$</p> | |

Carefully justify your answers, using only the definitions of stability. (Do not use eigenvalue methods or Lyapunov's method.)

Solution

- (a) Equilibrium set: $\Omega = \{x : x_2 = 0\}$. Starting from $x^0 = [x_1^0, x_2^0]$, $x(t) = [x_1^0, e^{-t}x_2^0] \rightarrow [x_1^0, 0]$. For any $x_e \in \Omega$, $\forall \epsilon > 0$, there exists $\delta = \epsilon$ such that if $\|x_0 - x_e\|^2 < \delta$, $\|x(t) - x_e\|^2 < \epsilon$, $\forall t \geq 0$. Therefore all equilibrium points are stable in sense of Lyapunov. However, $\forall \delta > 0$, there exists x_0 such that $\|x_0 - x_e\| < \delta$ but $x(t)$ doesn't converge to x_e , and thus all equilibrium points are not asymptotically stable.
- (b) Equilibrium set: $\Omega = \{x : x_2 = 0\}$. Starting from $x^0 = [x_1^0, x_2^0]$, $x(t) = [x_1^0 - x_2^0 t, x_2^0]$. For any $x_e \in \Omega$, $\forall \delta > 0$, there exists x_0 such that $\|x_0 - x_e\|^2 < \delta$, but $x_2^0 > 0$ and $x(t)$ goes to $[-\infty, x_2^0]$. Therefore all equilibrium points are not stable in sense of Lyapunov.
- (c) Equilibrium set: $\Omega = \{x : |x| > 1 \text{ or } x = 0\}$. Starting from $\{x_0 : |x_0| \leq 1\}$, $x(t) = e^{-t}x_0 \rightarrow 0$. Starting from $\{x_0 : |x_0| > 1\}$, $x(t) = x_0$. For $x_e = 0$, $\forall \epsilon > 0$, there exists $\delta = \min\{\epsilon, 1\}$ such that if $\|x_0 - x_e\|^2 < \delta$, $\|x(t) - x_e\|^2 < \epsilon$, $\forall t \geq 0$. Therefore equilibrium point 0 is stable in sense of Lyapunov. Moreover, let $\delta = 1$, then if $\|x_0 - 0\| < \delta$, $x(t)$ converges to 0. Therefore, 0 is a asymptotically state equilibrium point. However, δ cannot be greater than 1, and thus 0 is not global asymptotically stable. For $x_e \in \{x : |x| > 1\}$, $\forall \epsilon > 0$, there exists $\delta = \min\{\epsilon, \|+1 - x_e\|^2, \|-1 - x_e\|^2\}$ such that if $\|x_0 - x_e\|^2 < \delta$, $\|x(t) - x_e\|^2 = \|x_0 - x_e\|^2 < \epsilon$, $\forall t \geq 0$. Therefore, x_e is stable in sense of Lyapunov. However, $\forall \delta > 0$, there exists x_0 such that $\|x_0 - x_e\| < \delta$ but $x(t)$ doesn't converge to x_e , and thus the equilibrium points are not asymptotically stable.

Problem 2

Prove the variation-of-constants formula for linear time-varying ordinary differential equations, stated in Section 3.7 of the course reader and repeated here:

$$x(t) = \phi(t, t_0)x_0 + \int_{t_0}^t \phi(t, s)B(s)u(s)ds \quad (1)$$

Solution

For $t = t_0$, $x(t) = x_0$. Differentiate both sides of the equation with the Leibniz Rule, then we have

$$\begin{aligned} \dot{x}(t) &= \dot{\phi}(t, t_0) x_0 + \int_{t_0}^t \dot{\phi}(t, s) B(s) u(s) ds + \phi(t, t) B(t) u(t) \\ &= A(t) \phi(t, t_0) x_0 + A(t) \int_{t_0}^t \phi(t, s) B(s) u(s) ds + B(t) u(t) = A(t) x(t) + B(t) u(t) \end{aligned}$$

Therefore, this equation is a solution of the LTV model.

Problem 3

Let M be a symmetric real-valued $n \times n$ matrix. Show that the following three statements are equivalent:

- (a) M is positive definite.
- (b) All eigenvalues of M are positive definite.
- (c) $M = N^T N$ for some nonsingular $n \times n$ matrix N .

Solution

- . (a) \implies (b). We proceed by contradiction: suppose there exists a eigenvalue $\lambda \leq 0$ of M and the corresponding eigenvector $v \neq \mathbf{0}$, then we have $v^* M v = v^* \lambda v = \lambda |v|^2 \leq 0$, which is contradictive to (a). Therefore all eigenvalues of M are positive definite.
- . (b) \implies (c). Since M is a symmetric real-valued matrix, there exists orthogonal matrix P which can diagonalize M , i.e. $M = P \Lambda P^T$. Since all eigenvalues of M is positive, we have $N = P \sqrt{\Lambda} P^T$. N is a nonsingular matrix, and $N^T N = P \sqrt{\Lambda} P^T P \sqrt{\Lambda} P^T = M$.
- . (c) \implies (a). (I can only give the proof by supposing N is real-valued matrix, and thus $N^* = N^T$.) $\forall x \in \mathbb{C}$ and $x \neq \mathbf{0}$, $x^* M x = x^* N^T N x = (N x)^* (N x) = \|N x\|^2$. Because N is non-singular, $N x \neq \mathbf{0}$, and thus $x^* M x > 0$. Therefore, M is positive definite.

Problem 4

Suppose $V(x) = x^T A x$ for a real matrix A . Prove that if the function V is positive definite then the matrix $M = \frac{1}{2}(A + A^T)$ is positive definite. Provide an example of a positive definite V where the matrix A itself is not symmetric.

Solution

Because A is a real matrix, $M^* = M^T = M$, and thus M is Hermitian. $\forall x \in \mathbb{C}$, let $\Re(x)$ and $\Im(x)$ denote the real and imaginary part of x respectively. $\forall x \in \mathbb{C}$ and $x \neq \mathbf{0}$, $x^* M x = \frac{1}{2}(x^* A x + x^* A^T x) = \frac{1}{2}(x^* A x + \overline{x^* A x}) = \Re(x^* A x)$. Since $x^* A x = (\Re(x)^T - j \Im(x)^T) A (\Re(x) + j \Im(x)) = \Re(x)^T A \Re(x) + \Im(x)^T A \Im(x) + j(\dots)$, $\Re(x^* A x) = \Re(x)^T A \Re(x) + \Im(x)^T A \Im(x) = V(\Re(x)) + V(\Im(x)) > 0$. Therefore, M is positive definite.

$A = \begin{bmatrix} 1 & 10 \\ -10 & 1 \end{bmatrix}$, and we have $\frac{A+A^T}{2} = I$ which is positive definite. Thus $V(x)$ is positive definite.

Problem 5

Consider the LTI system described by:

$$\dot{x} = \begin{bmatrix} -1 & 1 \\ -2 & 3 \end{bmatrix} x$$

Investigate asymptotic stability using the Lyapunov equation:

$$A^T P + P A = -Q \quad Q = I$$

Can you arrive at any definite conclusion?

Solution

Solve the Lyapunov equation, and we have $P = \begin{bmatrix} 3 & -\frac{5}{4} \\ -\frac{5}{4} & \frac{1}{4} \end{bmatrix}$. P is not positive definite. We cannot arrive at any definite conclusion.

Problem 6

In this problem, we will investigate Lyapunov functions for discrete-time systems. Consider the discrete-time system:

$$x[k+1] = f(x[k]) \quad k = 0, 1, \dots$$

- Suppose we have a Lyapunov function $V(x) = x^T P x$ for some symmetric positive definite P . What is $V(f(x)) - V(x)$? This is analogous to $\dot{V}(x(t))$ in continuous time which is given by $\nabla V \cdot f$.
- In the LTI case where $f(x) = Ax$ show that the origin is asymptotically stable if for some $Q > 0$ there exists $P > 0$ that solves:

$$A^T P A - P = -Q$$

Hint: Try to bound the norm $|x_n|^2$ as a function of n . You may also need the linear algebra fact that for a symmetric matrix M :

$$\lambda_{\min}|x|^2 \leq x^T M x \leq \lambda_{\max}|x|^2$$

where λ_{\min} is the smallest eigenvalue of M and λ_{\max} is the largest eigenvalue of M . **Note:** For LTI systems in discrete-time, the equivalent of the state transition matrix e^{At} is A^k . In other words:

$$x[n] = A^n x[0] + \sum_{k=0}^{n-1} A^{n-k} B u[k]$$

Solution

(a)

$$\begin{aligned}
V(f(x[k])) - V(x[k]) &= V(x[k+1]) - V(x[k]) \\
&= (x[k+1] + x[k])^T P (x[k+1] - x[k]) = \nabla V \left(\frac{x[k+1] + x[k]}{2} \right) (x[k+1] - x[k])
\end{aligned}$$

(b) Let the Lyapunov function be $V(x) = x^T P x$. Suppose there exist positive matrix P and Q such that $A^T P A - P = -Q$. Let $\lambda_{Q,min}$ be the smallest eigenvalue of Q , and $\lambda_{P,max}$ be the largest eigenvalue of P . Pick a $\epsilon \in (0, \frac{\lambda_{Q,min}}{\lambda_{P,max}})$, then we have $Q - \epsilon P \succ 0$ according to the Hint. (We can do that because P and Q do not have zero eigenvalues.) Then, we have $A^T P A - P + \epsilon P \prec 0$, and thus $x[k]^T (A^T P A - P + \epsilon P) x[k] < 0$, $\forall x[k] \neq 0$. Thus, we have

$$\begin{aligned}
x[k]^T (A^T P A - P + \epsilon P) x[k] &= x[k+1]^T P x[k+1] - x[k]^T P x[k] + \epsilon x[k]^T P x[k] \\
&= V(x[k+1]) - V(x[k]) + \epsilon V(x[k]) < 0 \quad (2)
\end{aligned}$$

Because $V(x[k+1]) - V(x[k]) < -\epsilon V(x[k])$, we have $V(x[k]) < -\epsilon \sum_{i=0}^{k-1} V(x[i]) + V(x[0])$.

Suppose that $V(x[k])$ doesn't converge to 0, then $\sum_{i=0}^{k-1} V(x[i])$ goes to ∞ , and thus $V(x[k])$ goes to $-\infty$. So, we have a contradiction, and thus we must have $V(x[k]) \rightarrow 0$ and $x[k] \rightarrow 0$. So, origin is asymptotically stable.