EE/AA547-W22

Topic: Lyapunov Stability

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1 Overview

Definition 1 (Stable Equilibrium). The following are characterizations of stability (in the sense of Lyapunov).

a. Marginally Stable: Consider the equilibrium point $x^* = 0$.

 x^* is stable $\iff \forall x_0 \in \mathbb{R}^n, \forall t_0 \in \mathbb{R}^n, t \mapsto x(t) = \Phi(t, t_0) x_0$ is bounded $\forall t \ge t_0$.

Note: the effect of initial conditions does not grow unbounded with time (but it may grow temporarily during a transient phase).

b. Asymptotic Stability. Consider the equilibrium point $x^* = 0$.

 $x^* = 0$ is asymptotically stable $\iff x_0 = 0$ is stable and $x(t) = \Phi(t, t_0) x_0 \longrightarrow 0$ as $t \to \infty$.

Note: the effect of initial conditions eventually disappears with time.

c. Exponential Stability. Consider the equilibrium point $x^* = 0$.

 $x^* = 0$ is exponentially stable $\iff \exists M, \alpha > 0 : ||x(t)|| \le M \exp(-\alpha(t-t_0))||x_0||$

Spectral Conditions for Stability.

Proposition 2 (Continuous Time). Consider the differential equation $\dot{x} = Ax$, $x(0) = x_0$. From the above expression:

$$\{\exp(At) \to 0 \text{ as } t \to \infty\} \iff \{\forall \lambda_k \in \operatorname{spec}(A), \operatorname{Re}(\lambda_k) < 0\}$$

and

$$\{t \mapsto \exp(At) \text{ is bounded on } \mathbb{R}_+\} \iff \left\{ \begin{array}{ll} \forall \lambda_k \in \operatorname{spec}(A), & \operatorname{Re}(\lambda_k) \le 0 \& \\ m_k = 1 \text{ when } & \operatorname{Re}(\lambda_k) = 0 \end{array} \right\}$$

Claim 1.

$$\dot{x} = Ax$$
 is exponentially stable \iff spec $(A) \subset \mathbb{C}_{-}^{\circ}$

Linearized System Stability. Consider a general non-linear system

$$\dot{x} = f(x), \ x \in \mathbb{R}^{n}$$

with an equilibrium point x^* such that $f(x^*) = 0$. Recall that the local linearization around x^* is given by

$$\tilde{x} = A\tilde{x}$$

with $\tilde{x} = x - x^*$ and $A := Df(x^*)$. The following theorem is the celebrated Hartman-Grobman theorem which states that trajectories of the nonlinear system are "equivalent" to trajectories of the linearization in a neighborhood of an equilibrium, and hence we can assess (local) stability of the nonlinear system by assessing stability of the linearized system.¹

¹If you are interested in learning more about nonlinear systems, I suggest Shankar Sastry's book "Nonlinear Systems" [sas-try2013nonlinear].

Theorem 3 (Hartman-Grobman). Consider a nonlinear dynamical system $\dot{x} = f(x)$ with an equilibrium point x^* (i.e. $f(x^*) = 0$). If the linearization of the system $A := D_x f(x)|_{x=x^*}$ has no zero or purely imaginary eigenvalues then there exists a homeomorphism (i.e., a continuous map with a continuous inverse) from a neighborhood U of x^* into \mathbb{R}^n ,

$$h: U \to \mathbb{R}^n$$
.

taking trajectories of the nonlinear system $\dot{x} = f(x)$ and mapping them onto those of $\tilde{x} = A\tilde{x}$. In particular, we have that x^* maps to the equilibrium of the linearized system—i.e., $h(x^*) = 0$.

The above theorem directly translates to the following corollary.

Corollary 4. Suppose that $f \in C^2(\mathbb{R}^n, \mathbb{R})$. If the linearized system is exponentially stable, then there exists a ball $B \subset \mathbb{R}^n$ around x^* and constants $c, \lambda > 0$ such that for every solution x(t) to the nonlinear system that starts at $x(t_0) \in B$, we have

$$||x(t) - x^*|| \le ce^{-\lambda(t-t_0)} ||x(t_0) - x^*||.$$

This means that the properties of the linearized system are preserved in the nonlinear system.

Numerical Integration. Spectral stability properties are also important for numerical integration schemes such as forward and backward Euler. In particular choosing the step size for the integration scheme determines the stability of the discrete time update equation, and the choice of step size depends on the eigenvalues of A.

2 Problems

Problem 1. (Stability and Eigenvalues). Consider the system

Is the system asymptotically stable? Is the system stable?

solution. The system is stable but not asymptotically stable since all its eigenvalues are in the CLHP and the ones on the $j\omega$ -axis have Jordan sub-blocks of size 1.

We can immediately see from the structure of A that the eigenvalues are in the closed left-half plane and that the ones on the $j\omega$ -axis have Jordan sub-blocks of size 1, so its (marginally) stable but not asymptotically stable.

Another way to see this is the following: A system is considered asymptotically stable if and only if $\Phi(t, 0) \to 0$ as $t \to \infty$ (cf. lecture notes). Using functions of matrices, we have

$$e^{At} = \begin{bmatrix} e^{-3t} & -te^{-3t} & -\frac{1}{2}t^2e^{-3t} & 0 & 0 & 0 & 0\\ 0 & e^{-3t} & -te^{-3t} & 0 & 0 & 0 & 0\\ 0 & 0 & e^{-3t} & 0 & 0 & 0 & 0\\ 0 & 0 & 0 & e^{-4t} & -te^{-4t} & 0 & 0\\ 0 & 0 & 0 & 0 & e^{-4t} & 0 & 0\\ 0 & 0 & 0 & 0 & 0 & 1 & 0\\ 0 & 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix}$$

Problem 2. (Stiff Differential Equations.) In the simulation of several engineering systems we encounter parasitic elements which result in the differential equation becoming "stiff". A problem is stiff if it contains widely varying time scales, i.e., some components of the solution decay much more rapidly than others. For example, parasitic capacitances and inductances in electronic circuits. This has consequences not just in terms of the behavior of the actual system (which can experience thigns like hysterisis etc) but also for numerical simulation.

Fact 5. A stiff differential equation is numerically unstable unless the step size is extremely small. Stiff differential equations are characterized as those whose exact solution has a term of the form $e^{-\alpha t}$ where α is a large positive constant. And, large derivatives of $e^{-\alpha t}$ give error terms that are dominating in the solution.

This results in some state variables changing much more rapidly than the others. To represent this, consider the system with x_1 representing the "slow" variables and x_2 the "fast" variables given by

$$\dot{x}_1 = A_{11}x_1 + A_{12}x_2
\epsilon \dot{x}_2 = A_{21}x_1 + A_{22}x_2$$

with $x_1 \in \mathbb{R}^n, x_2 \in \mathbb{R}^m$ and A_{22} non-singular.

a. Prove the following claim:

Claim 2. For the system above, *m* of the eigenvalues tend to ∞ like $\frac{\sigma(A_{22})}{\epsilon}$, and the other *n* eigenvalues tend to $\operatorname{spec}(A_{11} - A_{12}A_{22}^{-1}A_{21})$.

b. The system

$$\dot{x}_1 = A_{11}x_1 + A_{12}x_2 0 = A_{21}x_1 + A_{22}x_2$$

is referred to as the singularly perturbed or low frequency approximation.

In electronic circuits, we also have in addition to parasitic (small) capacitances, coupling (large) capacitances. These are modeled by

with $\epsilon > 0$ small and $\mu > 0$ large. A *mid frequency model* takes $\epsilon = 0, \mu = \infty$, a *low frequency model* takes $\epsilon = 0$ and sets $\mu = \infty$ in the $\tau = \frac{t}{\mu}$ time scale and a *high frequency model* sets $\mu = \infty$ and then sets $\epsilon = 0$ in the time scale $\tau = \frac{t}{\epsilon}$. Find the relationship between the eigenvalues of each if these models.

Solution.

Claim 3. For the system above, *m* of the eigenvalues tend to ∞ like $\frac{\sigma(A_{22})}{\epsilon}$, and the other *n* eigenvalues tend to $\operatorname{spec}(A_{11} - A_{12}A_{22}^{-1}A_{21})$.

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Nummerical Simulation Facts for Stiff Differential Equations.

- A problem is stiff if the stepsize is dictated by stability requirements rather than by accuracy requirements.
- A problem is stiff if explicit methods don't work, or work only extremely slowly.
- A linear problem is stiff if all of its eigenvalues have negative real part, and the stiffness ratio (the ratio of the magnitudes of the real parts of the largest and smallest eigenvalues) is large.

Problem 3. (Linearization of Nonlinear Systems.) Consider the nonlinear systems given below. Find the equilibrium points and determine the type of stability they exhibit.

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a. x \in \mathbb{R}: \dot{x} = ax - x^3 for arbitrary a \in \mathbb{R}
Solution.
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b. Let a be some arbitrary parameter for the system

$$\begin{array}{rcl} \dot{x} & = & x^2 + y \\ \dot{y} & = & x - y + a \end{array}$$

where $(x, y) \in \mathbb{R} \times \mathbb{R}$. Solution.

3 Lyapunov

Theorem 6 (Lyapynov Theorem). Consider the dynamical system defined by $f \in C^1(\mathbb{R}^n, \mathbb{R})$. Let W be an open subset of \mathbb{R}^n containing the equilibrium point x^* —i.e., $f(x^*) = 0$. Suppose that there exists a realvalued function $V \in C^1$ such that $V(x^*) = 0$ and V(x) > 0 when $x \neq x^*$. Then, the following implications hold:

 $\begin{array}{lll} \mathbf{a}. & \dot{V}(x) \leq 0, \, \forall x \in W \implies & x^* \text{ is stable.} \\ \mathbf{b}. & \dot{V}(x) < 0, \, \forall x \in W \backslash \{x^*\} \implies & x^* \text{ is asymptotically stable.} \\ \mathbf{c}. & \dot{V}(x) > 0, \, \forall x \in W \backslash \{x^*\} \implies & x^* \text{ is unstable.} \end{array}$

For linear systems, it turns out that Lyapunov functions take the form

$$V(z) = z^{\top} P z$$

for some positive definite symmetric matrix $P \succ 0$.

For a linear system $\dot{x} = Ax$, if

$$V(z) = z^{\top} P z$$

then

$$\dot{V}(z) = (Az)^{\top} P z + z^{\top} P (Az) = -z^{\top} Q z$$

That is, if $z^{\top}Pz$ is the (generalized) energy, then $z^{\top}Qz$ is the associated (generalized) dissipation.

If $P \succ 0$, then the sublevel sets² of this function are ellipsoids and bounded. Further, we have that

$$V(z) = z^{\top} P z = 0 \iff z = 0.$$

If $P \succ 0, Q \succeq 0$, then all the trajectories of $\dot{x} = Ax$ are bounded (i.e., $\operatorname{Re}(\lambda_i) \leq 0$ and if $\operatorname{Re}(\lambda_i) = 0$, then λ_i corresponds to a Jordan block of size one). Further, the ellipsoids $\{z \mid z^\top Pz \leq a\}$ are invariant sets.

Theorem 7. The following conditions are equivalent:

- **a**. The system $\dot{x} = Ax$ is asymptotically (equivalently exponentially) stable.
- b. All the eigenvalues of A have strictly negative real parts.
- c. For every symmetric positive definite matrix $Q = Q^{\top} \succ 0$, there exists a unique solution P to the Lyapunov equation

$$A^{\top}P + PA = -Q$$

Moreover, P is symmetric and positive-definite—i.e., $P = P^{\top} \succ 0$ —and is given by

$$P = \int_0^\infty e^{A^\top t} Q e^{At} dt$$

d. There exists a symmetric positive-definite matrix $P = P^{\top} \succ 0$ for which the following Lyapunov matrix inequality holds:

$$A^{+}P + PA < 0$$

3.1 Problems

Problem 4. (Lyapunov Equation.) Consider the linear map $\mathcal{L} : \mathbb{R}^{n \times n} \to \mathbb{R}^{n \times n}$ defined by $\mathcal{L}(P) = A^{\top}P + PA$. Show that if $\lambda_i + \bar{\lambda}_j \neq 0$, for all $\lambda_i, \lambda_j \in \sigma(A)$, the equation

$$A^{\top}P + PA = Q$$

²i.e., $\{x \mid V(x) < a\}$

has a unique symmetric solution for given symmetric Q. Solution.

To prove this we need to show the following technical lemma.

Lemma 8. Any injective linear map $A : \mathbb{R}^n \to \mathbb{R}^n$ is also surjective.

Now towards solving the problem, we show that $\mathcal{L}(P) = Q$ has a unique solution. It suffices to show that $\mathcal{L}(P) \mapsto A^{\top}P + PA$ is bijective or equivalently $\operatorname{Ker}(\mathcal{L}) = \{0\}$.

Assume A is diagonalizable. What follows can be generalized even if this is not possible - the qualitative results still hold.

Lemma 9. The eigenvalues of $\mathcal{L}(P) \mapsto A^{\top}P + PA$ are $\lambda_i + \bar{\lambda}_j$ for all $\lambda_i, \lambda_j \in \operatorname{spec}(A)$.