

## 1 Conditions for Equilibria

### Continuous-Time Systems

Let us take a closer look at the conditions for a linear system represented by the differential equation

$$\frac{d}{dt}\vec{x}(t) = A\vec{x}(t) + B\vec{u}(t) \quad (1)$$

From the get-go we see that  $(\vec{x}^*, \vec{u}^*) = (\vec{0}, \vec{0})$  must be an equilibrium point. This is since the system is at rest. Now if we put in a constant input  $\vec{u}^*$  then to solve for equilibria, we get the following system of equations

$$A\vec{x} + B\vec{u}^* = \vec{0} \quad (2)$$

To solve for the states  $\vec{x}$  in which the system would be in equilibrium, our analysis boils down to whether the square matrix  $A$  is invertible <sup>1</sup>.

- a) If  $A$  is invertible, then there is a unique equilibrium point  $\vec{x}^* = -A^{-1}B\vec{u}^*$ .
- b) If  $A$  is non-invertible, depending on the range of  $A$ , we have two scenarios.
  - If  $B\vec{u}^* \in \text{Col}(A)$  then we will have infinitely many equilibrium points.
  - If  $B\vec{u}^* \notin \text{Col}(A)$  then the system has no solution and we will have no equilibrium points.

### Discrete-Time Systems

Now let's take a look at the discrete-time system

$$\vec{x}(t+1) = A\vec{x}(t) + B\vec{u}(t) \quad (3)$$

Again we see that  $(\vec{0}, \vec{0})$  is an equilibrium point but notice that the conditions for equilibria are different for discrete-time systems. A system is in equilibrium if it is not changing. In other words, this means that  $\vec{x}^*(t+1) = \vec{x}^*(t)$  therefore, for a constant input  $\vec{u}^*$  we get the following system of equations

$$\vec{x} = A\vec{x} + B\vec{u}^* \implies (I - A)\vec{x} = B\vec{u}^* \quad (4)$$

The conditions for equilibria now depend on the matrix  $I - A$  being invertible instead of the matrix  $A$ .

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<sup>1</sup>This should be review from 16A/54, but we restate it here since it isn't quite obvious when  $A$  is singular or non-invertible. Normally a singular matrix has infinite solutions but take the system  $A\vec{x} = \vec{b}$  with  $A = \begin{bmatrix} 1 & 0 \\ 1 & 0 \end{bmatrix}$  and  $\vec{b} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$ . This leads to a contradiction that  $x_1 = 0 \neq 1$ .

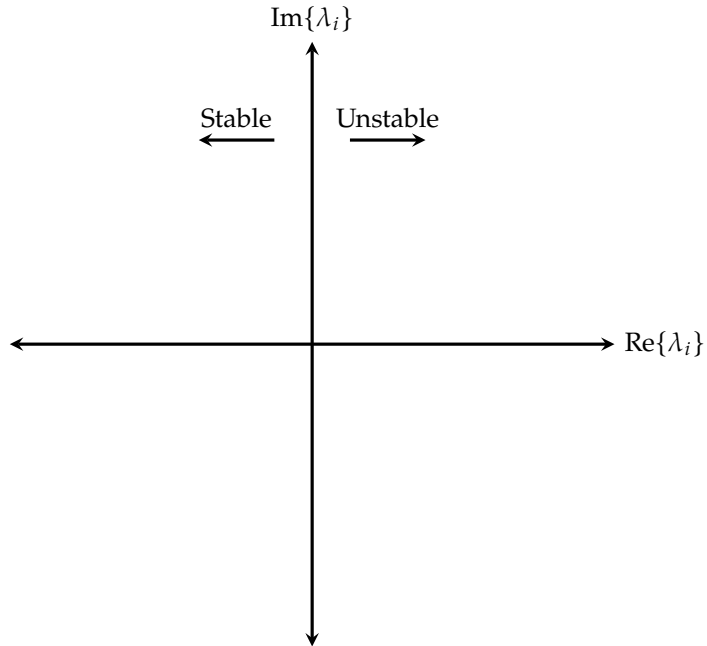
## 2 Stability

### Continuous time systems

A continuous time system is of the form:

$$\frac{d\vec{x}}{dt}(t) = A\vec{x}(t) + B\vec{u}(t)$$

This system is stable if  $\text{Re}\{\lambda_i\} < 0$  for all  $\lambda_i$ , where  $\lambda_i$ 's are the eigenvalues of  $A$ . If we plot all  $\lambda_i$  for  $A$  on the complex plane, if all  $\lambda_i$  lie to the left of  $\text{Re}\{\lambda_i\} = 0$ , then the system is stable.



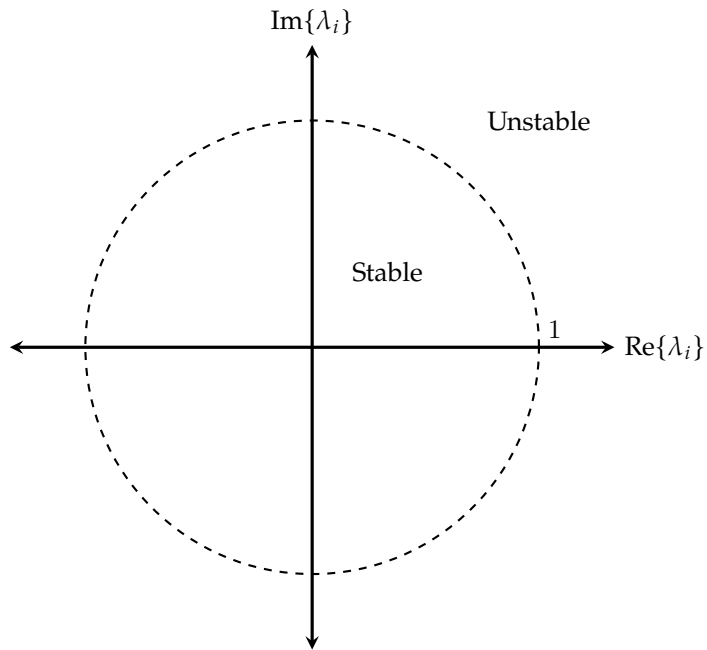
If  $\text{Re}\{\lambda_i\} \geq 0$ , the system is unstable in the context of BIBO stability.

## Discrete time systems

A discrete time system is of the form:

$$\vec{x}(t+1) = A\vec{x}(t) + B\vec{u}(t)$$

This system is stable if  $|\lambda_i| < 1$  for all  $\lambda_i$ , where  $\lambda_i$ 's are the eigenvalues of  $A$ . If we plot all  $\lambda_i$  for  $A$  on the complex plane, if all  $\lambda_i$  lie within (not on) the unit circle, then the system is stable.



If  $|\lambda| \geq 1$ , we say the system is unstable in the context of Bounded-Input Bounded-Output (BIBO) stability.

### 3 Jacobian Warm-Up

Consider the following function  $f : \mathbb{R}^2 \mapsto \mathbb{R}^3$

$$f(x_1, x_2) = \begin{bmatrix} f_1(x_1, x_2) \\ f_2(x_1, x_2) \\ f_3(x_1, x_2) \end{bmatrix} = \begin{bmatrix} x_1^2 - e^{x_2^2} \\ x_1^2 + \sin(x_1)x_2^2 \\ \log(1 + x_1^2) \end{bmatrix}$$

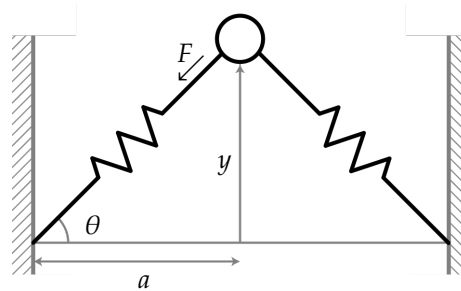
Calculate its Jacobian.

**Answer**

$$\begin{aligned} \frac{df}{d\vec{x}} &= \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} \\ \frac{\partial f_2}{\partial x_1} & \frac{\partial f_2}{\partial x_2} \\ \frac{\partial f_3}{\partial x_1} & \frac{\partial f_3}{\partial x_2} \end{bmatrix} \\ &= \begin{bmatrix} 2x_1 & -2x_2e^{x_2^2} \\ 2x_1 + \cos(x_1)x_2^2 & 2\sin(x_1)x_2 \\ \frac{2x_1}{1+x_1^2} & 0 \end{bmatrix} \end{aligned}$$

## 4 Linearization

Consider a mass attached to two springs:



We assume that each spring is linear with spring constant  $k$  and resting length  $X_0$ . We want to build a state space model that describes how the displacement  $y$  of the mass from the spring base evolves. The differential equation modeling this system is  $\frac{d^2y}{dt^2} = -\frac{2k}{m}(y - X_0 \frac{y}{\sqrt{y^2+a^2}})$ .

- a) Write this model in state space form  $\dot{x} = f(x)$ .

### Answer

We introduce states  $x_1 = y$  and  $x_2 = \dot{y}$ . Writing the model in state space form gives

$$\begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} x_2 \\ \frac{-2k}{m} \left( x_1 - X_0 \frac{x_1}{\sqrt{x_1^2 + a^2}} \right) \end{bmatrix}.$$

- b) Find the equilibrium of the state-space model. You can assume  $X_0 < a$ .

### Answer

We find the equilibrium by solving  $0 = \dot{x} = f(x)$ :

$$\begin{bmatrix} 0 \\ 0 \end{bmatrix} = \begin{bmatrix} x_2 \\ \frac{-2k}{m} \left( x_1 - X_0 \frac{x_1}{\sqrt{x_1^2 + a^2}} \right) \end{bmatrix}.$$

The unique solution is the equilibrium at  $(x_1, x_2) = (0, 0)$ .

- c) Linearize your model about the equilibrium.

### Answer

$$\begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} \\ \frac{\partial f_2}{\partial x_1} & \frac{\partial f_2}{\partial x_2} \end{bmatrix} \Bigg|_{x=(0,0)} = \begin{bmatrix} 0 & 1 \\ -\frac{2k}{m} \left( 1 - X_0 \frac{a^2}{(x_1^2 + a^2)^{3/2}} \right) & 0 \end{bmatrix} \Bigg|_{x=(0,0)} = \begin{bmatrix} 0 & 1 \\ -\frac{2k}{m} \left( 1 - \frac{X_0}{a} \right) & 0 \end{bmatrix}$$

So the linearized system is

$$\dot{x} = \begin{bmatrix} 0 & 1 \\ -\frac{2k}{m} \left(1 - \frac{X_0}{a}\right) & 0 \end{bmatrix} x.$$

d) Compute the eigenvalues of your linearized model. Is this equilibrium stable?

**Answer**

To compute the eigenvalues, we solve

$$0 = \det(A - \lambda I) = \det \left( \begin{bmatrix} -\lambda & 1 \\ -\frac{2k}{m} \left(1 - \frac{X_0}{a}\right) & -\lambda \end{bmatrix} \right) = \lambda^2 + \frac{2k}{m} \left(1 - \frac{X_0}{a}\right).$$

Since  $X_0 < a$ , this means that  $\left(1 - \frac{X_0}{a}\right) > 0$ . So we have a pair of imaginary eigenvalues

$$\lambda = \pm \sqrt{\frac{2k}{m} \left(1 - \frac{X_0}{a}\right)} j.$$

Since the linearized system has purely imaginary eigenvalues that are not repeated, their real parts are zero. Therefore the equilibrium is unstable.

## 5 Stability in discrete time system

Determine which values of  $\alpha$  and  $\beta$  will make the following discrete-time state space models stable. Assume,  $\alpha$  and  $\beta$  are real numbers and  $b \neq 0$ .

a)

$$x(t+1) = \alpha x(t) + bu(t)$$

**Answer**

$$|\alpha| < 1$$

b)

$$\vec{x}(t+1) = \begin{bmatrix} \alpha & -\beta \\ \beta & \alpha \end{bmatrix} \vec{x}(t) + b\vec{u}(t)$$

**Answer**

The eigenvalues of this system are:

$$\lambda = \alpha \pm j\beta$$

$$|\lambda| = \sqrt{\alpha^2 + \beta^2}$$

For this system to be stable,  $|\lambda| < 1$ , so

$$\alpha^2 + \beta^2 < 1$$

c)

$$\vec{x}(t+1) = \begin{bmatrix} 1 & \alpha \\ 0 & 1 \end{bmatrix} \vec{x}(t) + b\vec{u}(t)$$

**Answer**

The eigenvalues of this system are

$$\lambda = 1, 1$$

This means that regardless of  $\alpha$ , this system is always unstable since  $|\lambda| \geq 1$ .