

# 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  $x_1 = 0 \neq 1$ .

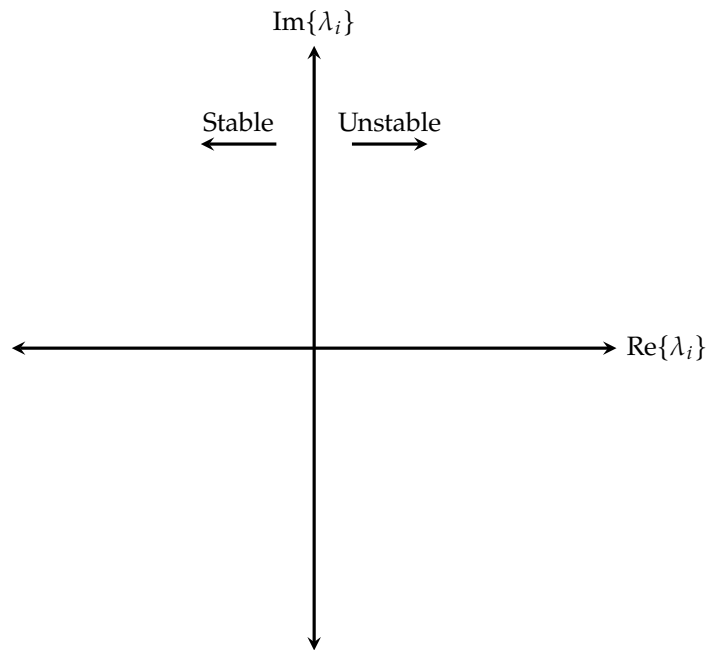
## 2 Stability

### 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.



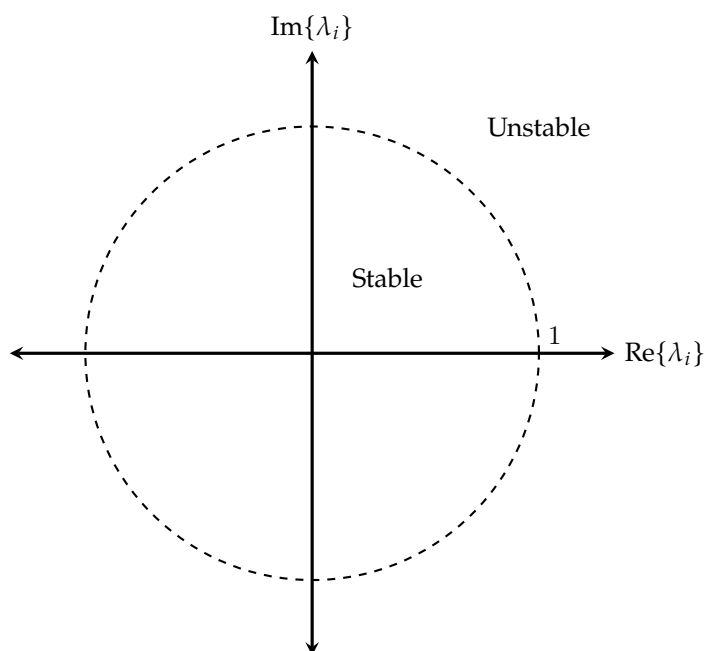
If  $\text{Re}\{\lambda_i\} \geq 0$ , the system is unstable in the context of BIBO stability.

## 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  $f: \mathbb{R}^2 \mapsto \mathbb{R}^3$

$$\frac{d}{du} \ln(u) = \frac{dy}{u}$$

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

↑ natural log or  $\ln$

Calculate its Jacobian.

Given  $p$  functions and  $n$  variables, the Jacobian matrix will be of size  $p \times n$

Here: 3 functions, 2 variables,  $J \in \mathbb{R}^{3 \times 2}$

$$J = \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} \\ \frac{\partial f_2}{\partial x_1} & \frac{\partial f_2}{\partial x_2} \\ \frac{\partial f_3}{\partial x_1} & \frac{\partial f_3}{\partial x_2} \end{bmatrix} \quad \nabla_{\vec{x}} = \begin{bmatrix} \frac{\partial f}{\partial x_1} \\ \frac{\partial f}{\partial x_2} \end{bmatrix}$$

↑  
partials of  $x_1$

↑  
partials of  $x_2$

$$J = \begin{bmatrix} 2x_1 & -2x_2 e^{x_2^2} \\ 2x_1 + x_2^2 \cos x_1 & \sin(x_1) 2x_2 \\ \frac{2x_1}{1+x_1^2} & 0 \end{bmatrix}$$

$$\frac{\partial f_1}{\partial x_1} = 2x_1 \quad \frac{\partial f_1}{\partial x_2} = -e^{x_2^2} \cdot 2x_2$$

$$\frac{\partial f_2}{\partial x_1} = 2x_1 + x_2^2 \cos x_1$$

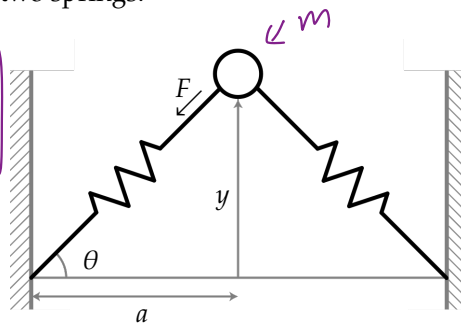
$$\frac{\partial f_2}{\partial x_2} = \sin(x_1) 2x_2$$

$$\frac{\partial f_3}{\partial x_1} = \frac{2x_1}{1+x_1^2}$$

$$\frac{\partial f_3}{\partial x_2} = \frac{0}{1+x_1^2} = 0$$

#### 4 Linearization

Consider a mass attached to two springs:



General Advice:  
For an  $n^{\text{th}}$  order diff-eq  
pick  $x_1 = y, x_2 = \frac{dy}{dt}, \dots$   
 $x_n = \frac{d^{n-1}y}{dt^{n-1}}$

$$k > 0 \quad m > 0$$

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  $\frac{d^2 y}{dt^2} = -\frac{2k}{m} \left( y - X_0 \frac{y}{\sqrt{y^2 + a^2}} \right)$ .

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

$$x_1 = y, \quad x_2 = \frac{dy}{dt}$$

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

$$\frac{dx_1}{dt} = \frac{dy}{dt} = x_2$$

$$\frac{dx_2}{dt} = \frac{d^2 y}{dt^2} = -\frac{2k}{m} \left( y - X_0 \frac{y}{\sqrt{y^2 + a^2}} \right) = -\frac{2k}{m} \left( x_1 - X_0 \frac{x_1}{\sqrt{x_1^2 + a^2}} \right)$$

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

$$\text{Find } \vec{x}^*, \text{ s.t. } f(\vec{x}^*) = 0 \quad \frac{d}{dt} \vec{x} = \vec{0}$$

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

First row:  $x_2 = 0$

eq point:  $\vec{x}^* = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$

Second Row:  $-\frac{2k}{m} \left( x_1 - X_0 \frac{x_1}{\sqrt{x_1^2 + a^2}} \right) = 0$

$$x_1 \left( 1 - \frac{X_0}{\sqrt{x_1^2 + a^2}} \right) = 0$$

Case 1:  $x_1 = 0$

Case 2:  $1 = \frac{X_0}{\sqrt{x_1^2 + a^2}} \rightarrow$

$$x_1^2 + a^2 = X_0^2 \quad x_1^2 = X_0^2 - a^2 < 0$$

$x_1$  is imaginary  
Contradiction!

$$\sqrt{x_1^2 + a^2} = (x_1^2 + a^2)^{1/2}$$

c) Linearize your model about the equilibrium.

no input

Find Jacobian  $J_{\vec{x}} = \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} \\ \frac{\partial f_2}{\partial x_1} & \frac{\partial f_2}{\partial x_2} \end{bmatrix}$

$$\frac{\partial f_1}{\partial x_1} = 0$$

$$\vec{f} = \begin{bmatrix} x_2 \\ -\frac{2k}{m} \left( x_1 - x_0 \frac{x_1}{\sqrt{x_1^2 + a^2}} \right) \end{bmatrix}$$

$$\frac{\partial f_1}{\partial x_2} = 1$$

$$\frac{\partial f_2}{\partial x_1} = -\frac{2k}{m} \left( 1 - x_0 \frac{\sqrt{x_1^2 + a^2} \cdot 1 - x_1 \cdot \frac{1}{2} (x_1^2 + a^2)^{-1/2} \cdot (2x_1)}{(x_1^2 + a^2)} \right)$$

$$\frac{\partial f_2}{\partial x_2} = 0$$

Evaluate at  $x_1^* = x_2^* = 0$

$$\frac{\partial f_2}{\partial x_1} (0,0) = -\frac{2k}{m} \left( 1 - \frac{x_0}{a} \right)$$

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

Write out Linearized System

$$\frac{d}{dt} \vec{x}_e = \begin{bmatrix} 0 & 1 \\ -\frac{2k}{m} \left( 1 - \frac{x_0}{a} \right) & 0 \end{bmatrix} \vec{x}_e$$

$J_{\vec{x}}$

## 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)

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

b)

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

c)

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