

# **ECE 602: LUMPED LINEAR SYSTEMS**

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Stability of Continuous-Time LTI Systems

# Stability of CT Autonomous Linear Systems

Continuous-time linear system has an **equilibrium point** at  $x_e = 0$

$$\dot{x}(t) = A(t)x(t) \quad (1)$$

## Definition (Asymptotic Stability)

System (1) is called **asymptotically stable** at  $x_e = 0$  if its solution  $x(t)$  starting from any initial condition  $x(0) \in \mathbb{R}^n$  satisfies  $x(t) \rightarrow 0$  as  $t \rightarrow \infty$

## Definition (Exponential Stability)

System (1) is called **exponentially stable** at  $x_e = 0$  if its solution  $x(t)$  starting from any initial condition  $x(0) \in \mathbb{R}^n$  satisfies

$$\|x(t)\| \leq Ke^{-rt}\|x(0)\|, \quad \forall t \geq 0,$$

for some constants  $K, r > 0$ .

- Exponential stability  $\Rightarrow$  asymptotic stability

# Internal Stability vs Input/Output Stability

For CT LTI system  $\dot{x} = Ax + Bu$ ,  $y = Cx + Du$ , its transfer function

$$H(s) = C(sI - A)^{-1}B + D$$

**Input/Output Stability:** poles of  $H(s)$  all have negative real parts

- Assume  $x(0) = 0$ , and input  $u(\cdot)$  is arbitrary
- Equivalently, bounded input  $u(\cdot)$  results in bounded output  $y(\cdot)$  (**BIBO stability**)

**Internal Stability:** stability notion to be studied in this lecture

- Assume  $u(\cdot) \equiv 0$ , and  $x(0)$  is arbitrary
- Focus on state solution  $x(\cdot)$  rather than output  $y(\cdot)$
- A stronger notion of stability (more on this later)

# Characterizing Stability of CT LTI Systems

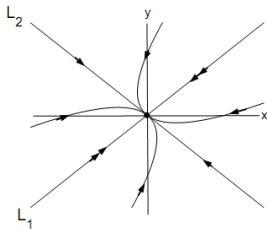
## Theorem

For CT LTI system  $\dot{x} = Ax$ , the following statements are equivalent

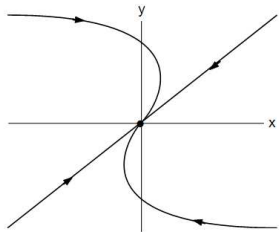
- 1 System is asymptotically stable
- 2 System is exponentially stable
- 3 All eigenvalues of  $A$  are in the *open* left half of the complex plane  $\mathbb{C}$

**Proof:** Solution  $x(t) = e^{At}x(0)$  is a linear combination of modes whose entries are  $p(t)e^{\lambda_i t}$ , where  $\lambda_i$  is an eigenvalue of  $A$  and  $p(t)$  is a polynomial of  $t$ . ■

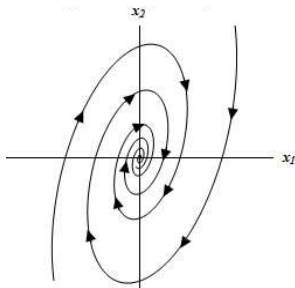
# Phase Portraits of Stable 2D LTI Systems



Stable node



Stable degenerate node



Stable focus

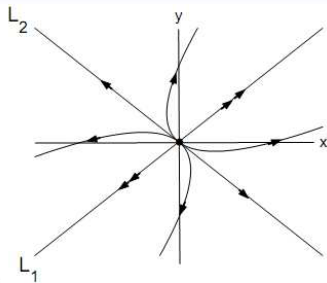
# Unstable Systems

LTI system  $\dot{x} = Ax$  is **unstable** if  $\|x(t)\|$  is unbounded for **some**  $x(0)$

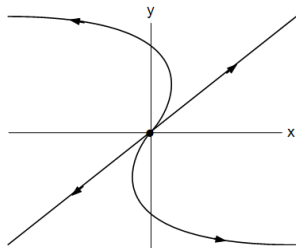
This is the case if **either** of the following holds:

- ①  $A$  has eigenvalues on the open right half plane of  $\mathbb{C}$
- ②  $A$  has **defective** eigenvalues on the  $j\omega$ -axis
  - An eigenvalue is defective if it has a Jordan block of size  $> 1$

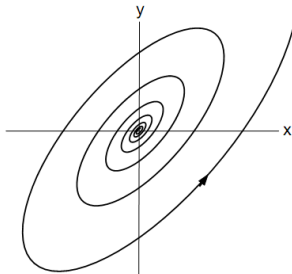
# Phase Portraits of Unstable 2D Systems



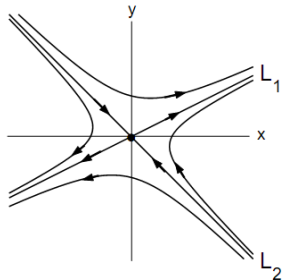
Unstable node



Unstable degenerate node



Unstable focus



Saddle

## Marginally Stable Systems

LTI system  $\dot{x} = Ax$  is **marginally stable** if all solutions  $x(t)$  are bounded and at least one solution  $x(t)$  does not converge to zero

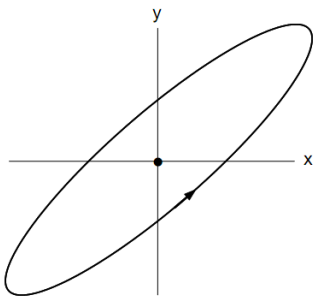
This is the case if **both** of the following hold:

- 1  $A$  has no eigenvalues on the open right half of  $\mathbb{C}$
- 2  $A$  has eigenvalues on the  $j\omega$ -axis, all being non-defective

Note: some textbooks classify marginally stable systems as unstable

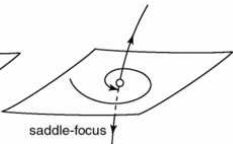
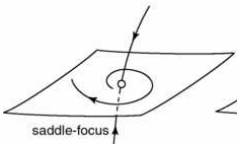
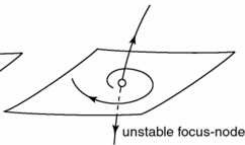
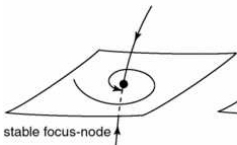
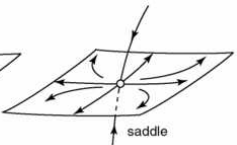
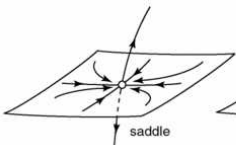
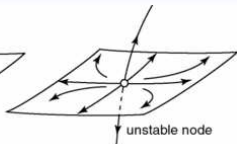
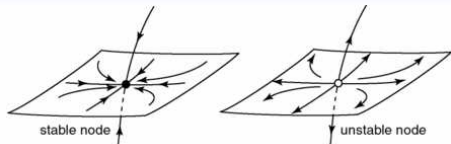


## Phase Portraits of Marginally Stable 2D Systems



$$\dot{x} = Ax \text{ with } A = \underbrace{\begin{bmatrix} v_1 & v_2 \end{bmatrix}}_T \begin{bmatrix} 0 & \omega \\ -\omega & 0 \end{bmatrix} T^{-1}$$

# Phase Portraits of 3D Systems



LTI system  $\dot{x} = Ax$ , where  $A \in \mathbb{R}^{3 \times 3}$

- System has three modes
- Each mode could be stable or unstable