

# **ECE 68000: MODERN AUTOMATIC CONTROL**

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Stability of dynamical systems:  
The Lyapunov perspective

# Lyapunov Function Definition

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A function that allows one to deduce stability is termed a Lyapunov function

# Lyapunov Function Properties for Continuous Time Systems

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- Continuous-time system

$$\dot{x}(t) = f(x(t))$$

- Equilibrium state of interest

$$x_e$$

# Three Properties of a Lyapunov Function

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We seek an aggregate summarizing function  $V$

- $V$  is continuous
- $V$  has a unique minimum with respect to all other points in some neighborhood of the equilibrium of interest
- Along any trajectory of the system, the value of  $V$  never increases

# Lyapunov Theorem for Continuous Systems

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- Continuous-time system

$$\dot{x}(t) = f(x(t))$$

- Equilibrium state of interest

$$x_e = 0$$

# Lyapunov Theorem---Negative Rate of Increase of $V$

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- If  $\mathbf{x}(t)$  is a trajectory, then  $V(\mathbf{x}(t))$  represents the corresponding values of  $V$  along the trajectory
- In order for  $V(\mathbf{x}(t))$  not to increase, we require

$$\dot{V}(\mathbf{x}(t)) \leq 0$$

# The Lyapunov Derivative

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- Use the chain rule to compute the derivative of  $V(\mathbf{x}(t))$

$$\dot{V}(x(t)) = \nabla V(x)^T \dot{x}$$

- Use the plant model to obtain

$$\dot{V}(x(t)) = \nabla V(x)^T f(x)$$

- Recall

$$\nabla V(x) = \left[ \frac{\partial V}{\partial x_1} \quad \frac{\partial V}{\partial x_2} \quad \dots \quad \frac{\partial V}{\partial x_n} \right]^T$$

# Lyapunov Theorem for LTI Systems

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The system  $d\mathbf{x}/dt=\mathbf{A}\mathbf{x}$  is asymptotically stable, that is, the equilibrium state  $\mathbf{x}_e=\mathbf{0}$  is asymptotically stable (a.s), if and only if any solution converges to  $\mathbf{x}_e=\mathbf{0}$  as  $t$  tends to infinity for any initial  $\mathbf{x}_0$



# Lyapunov Theorem Interpretation

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- View the vector  $\mathbf{x}(t)$  as defining the coordinates of a point in an  $n$ -dimensional state space
- In an a.s. system the point  $\mathbf{x}(t)$  converges to  $\mathbf{x}_e = \mathbf{0}$

# Lyapunov Theorem for $n=2$

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If a trajectory is converging to  $\mathbf{x}_e = \mathbf{0}$ , it should be possible to find a nested set of closed curves  $V(x_1, x_2) = c$ ,  $c \geq 0$ , such that decreasing values of  $c$  yield level curves shrinking in on the equilibrium state  $\mathbf{x}_e = \mathbf{0}$

# Lyapunov Theorem and Level Curves

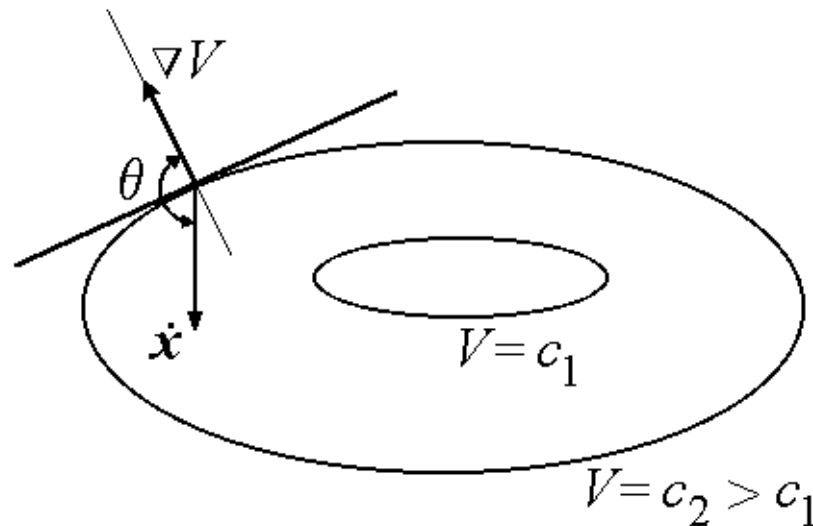
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- The limiting level curve  $V(x_1, x_2) = V(\mathbf{0}) = 0$  is 0 at the equilibrium state  $\mathbf{x}_e = \mathbf{0}$
- The trajectory moves through the level curves by cutting them in the inward direction ultimately ending at  $\mathbf{x}_e = \mathbf{0}$

# The trajectory is moving in the direction of decreasing $V$

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Note that  $\dot{V} = \|\nabla V\| \|\dot{\mathbf{x}}\| \cos \theta < 0$



# Level Sets

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- The level curves can be thought of as contours of a cup-shaped surface
- For an a.s. system, that is, for an a.s. equilibrium state  $\mathbf{x}_e = \mathbf{0}$ , each trajectory falls to the bottom of the cup

# Positive Definite Function---General Definition

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The function  $V$  is positive definite in  $S$ , with respect to  $\mathbf{x}_e$ , if  $V$  has continuous partials,  $V(\mathbf{x}_e)=0$ , and  $V(\mathbf{x})>0$  for all  $\mathbf{x}$  in  $S$ , where  $\mathbf{x}\neq\mathbf{x}_e$

# Positive Definite Function With Respect to the Origin

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Assume, for simplicity,  $\mathbf{x}_e = \mathbf{0}$ , then the function  $V$  is positive definite in  $S$  if  $V$  has continuous partials,  $V(\mathbf{0}) = 0$ , and  $V(\mathbf{x}) > 0$  for all  $\mathbf{x}$  in  $S$ , where  $\mathbf{x} \neq \mathbf{0}$

# Example: Positive Definite Function

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Positive definite function of two variables

$$\begin{aligned} V(x_1, x_2) &= 2x_1^2 + 3x_2^2 \\ &= \begin{bmatrix} x_1 & x_2 \end{bmatrix} \begin{bmatrix} 2 & 0 \\ 0 & 3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \\ &= \mathbf{x}^T \mathbf{P} \mathbf{x} \\ &> 0 \quad \text{for all} \quad \mathbf{x} \neq \mathbf{0} \end{aligned}$$



# Positive Semi-Definite Function---

## General Definition

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The function  $V$  is positive semi-definite in  $S$ , with respect to  $\mathbf{x}_e$ , if  $V$  has continuous partials,  $V(\mathbf{x}_e)=0$ , and  $V(\mathbf{x})\geq 0$  for all  $\mathbf{x}$  in  $S$

# Positive Semi-Definite Function With Respect to the Origin

---

Assume, for simplicity,  $\mathbf{x}_e = \mathbf{0}$ , then the function  $V$  is positive semi-definite in  $S$  if  $V$  has continuous partials,  $V(\mathbf{0}) = 0$ , and  $V(\mathbf{x}) \geq 0$  for all  $\mathbf{x}$  in  $S$

# Example: Positive Semi-Definite Function

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An example of positive semi-definite function of two variables

$$\begin{aligned} V(x_1, x_2) &= x_1^2 \\ &= \begin{bmatrix} x_1 & x_2 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \\ &= \mathbf{x}^T \mathbf{P} \mathbf{x} \\ &\geq 0 \quad \text{in } \mathbb{R}^2 \end{aligned}$$