

# Systems with a Small Number of Variables

Course: Complex Systems Modeling

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# Outline

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# Continuous-Time Models with Differential Equations 1/2

- Continuous-time models are described using differential equations.
- These models are prevalent in science and engineering due to their ability to represent natural phenomena occurring smoothly over time (e.g., motion of objects, flow of electric current).

## General Formulation of a First-Order Continuous-Time Model

$$\frac{dx}{dt} = F(x, t)$$

- $x$ : State of the system (scalar or vector variable).
- $\frac{dx}{dt}$ : Time derivative of  $x$ , formally defined as:

$$\frac{dx}{dt} = \lim_{\delta t \rightarrow 0} \frac{x(t + \delta t) - x(t)}{\delta t}$$

# Continuous-Time Models with Differential Equations 2/2

- Integrating the model over time  $t$  provides the system's trajectory.
  - Integration can sometimes be performed algebraically, but computational simulation (numerical integration) is often used in practice.
  - A fundamental assumption:
    - Trajectories of the system's state are smooth everywhere in the phase space.
    - Instantaneous abrupt changes, possible in discrete-time models, do not occur in continuous-time models.
- Recall discussion on the assumptions of the predator-prey model, e.g. 1.4 rabbit, 2.5 wolves. Also small vs. large number of entities - continuum assumption.

# Classifications of Model Equations

- We may distinct between linear and nonlinear systems, as well as autonomous and non-autonomous systems.
  - First-order systems: Involve only first-order derivatives ( $\frac{dx}{dt}$ ).
  - Higher-order systems: Involve higher-order derivatives ( $\frac{d^2x}{dt^2}, \frac{d^3x}{dt^3}, \dots$ ).
- Non-autonomous, higher-order systems can always be transformed into:
  - Autonomous, first-order forms.
  - This is achieved by introducing additional state variables.

# Classification of Systems: Linear vs. Nonlinear

## Linear Systems

- $\dot{x} = Ax + Bu$
- Scalar:  $\dot{x} = ax + bu$
- Superposition principle holds

## Nonlinear Systems

- $\dot{x} = f(x, u)$
- Scalar:  $\dot{x} = ax^2 + bu$
- Complex dynamics: chaos, bifurcations, limit cycles

# Classification of Systems: Autonomous vs. Non-Autonomous

## Autonomous Systems

- Evolution depends only on the state, not explicitly on time.
- **Matrix form:**

$$\dot{x} = Ax$$

- **Scalar example:**

$$\frac{dx}{dt} = ax$$

- System is *time-invariant* → suitable for phase-space or equilibrium analysis.

## Non-Autonomous Systems

- Explicit time-dependence in the equations or inputs.
- **Matrix form:**

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

- **Scalar example:**

$$\frac{dx}{dt} = a(t)x + b(t)$$

- Require special treatment for time-dependent forcing or varying parameters.

# Example: Pendulum Dynamics

**Second-order (nonlinear) equation of motion:** **Variables**

$$\frac{d^2\theta}{dt^2} = -\frac{g}{L} \sin \theta$$

**Interpretation:**

- The restoring torque is proportional to  $-\sin \theta$ .
- Equation is second-order and nonlinear.
- For large angles, motion deviates from simple harmonic behavior.

- $\theta$ : angular position
- $g$ : gravitational acceleration
- $L$ : pendulum length

**Remarks**

- Nonlinear due to  $\sin \theta$
- Linearization possible for small angles



# Converting to a First-Order System

Introduce angular velocity:

$$\omega = \frac{d\theta}{dt}$$

Substitute into the second-order equation:

$$\begin{aligned}\dot{\theta} &= \omega, \\ \dot{\omega} &= -\frac{g}{L} \sin \theta\end{aligned}$$

Compactly written as:

$$\dot{\mathbf{x}} = \begin{bmatrix} \omega \\ -\frac{g}{L} \sin \theta \end{bmatrix}, \quad \mathbf{x} = \begin{bmatrix} \theta \\ \omega \end{bmatrix}$$

## Notes

- Two coupled first-order ODEs
- Nonlinear due to  $\sin \theta$
- Enables phase-plane analysis and simulation

# Linearized Pendulum (Small Angles)

For small angles, approximate  $\sin \theta \approx \theta$ :

$$\begin{aligned}\dot{\theta} &= \omega, \\ \dot{\omega} &= -\frac{g}{L}\theta\end{aligned}$$

**Matrix form:**

$$\dot{\mathbf{x}} = \underbrace{\begin{bmatrix} 0 & 1 \\ -\frac{g}{L} & 0 \end{bmatrix}}_A \mathbf{x}$$

## Remarks

- Linear time-invariant (LTI) system
- Represents a *simple harmonic oscillator*
- Solution:

$$\theta(t) = \theta_0 \cos(\sqrt{g/L} t)$$

- Basis for control and small-signal stability analysis

# Conversion of Higher-Order Equations

- The technique for converting higher-order equations to first-order forms works for:
  - 3rd/4th/.../-order equations,
  - Any higher-order equations (as long as the highest order is finite).
- This method ensures equations are represented as autonomous, first-order systems for analysis and simulation.

## Key Advantage

Autonomous, first-order forms are universal and cover all dynamics of higher-order and non-autonomous systems.

# A Non-Autonomous Driven Pendulum

## Second-Order Non-Autonomous Equation

$$\frac{d^2\theta}{dt^2} = -\frac{g}{L} \sin \theta + k \sin(2\pi ft + \phi) \quad (6.7)$$

- Describes a driven pendulum with:
  - $\frac{g}{L}$ : Gravitational term.
  - $k \sin(2\pi ft + \phi)$ : Periodic external force, e.g., from an electromagnet.
- This equation is nonlinear and non-autonomous due to explicit time-dependence ( $t$ ).
- Goal: Convert it into a first-order, autonomous system.

# Step 1: First-Order Conversion

Introduce  $\omega = \frac{d\theta}{dt}$

$$\frac{d\theta}{dt} = \omega \quad (6.8)$$

$$\frac{d\omega}{dt} = -\frac{g}{L} \sin \theta + k \sin(2\pi ft + \phi) \quad (6.9)$$

- $\omega$ : Angular velocity of the pendulum.
- The system is now in first-order form but still non-autonomous due to explicit  $t$  in the sine term.

# Step 2: Eliminating Time Dependency

**Introduce a clock variable  $\tau$ :**

$$\frac{d\tau}{dt} = 1, \quad \tau(0) = 0$$

- Ensures  $\tau(t) = t$ .
- Replaces explicit time dependence with state variable  $\tau$ .
- Converts a non-autonomous system into an autonomous one.

**Original (time-dependent) term:**

$$k \sin(2\pi f t + \phi)$$

becomes

$$k \sin(2\pi f \tau + \phi)$$

**Resulting autonomous system:**

$$\begin{aligned}\dot{\theta} &= \omega, \\ \dot{\omega} &= -\frac{g}{L} \sin \theta + k \sin(2\pi f \tau + \phi), \\ \dot{\tau} &= 1\end{aligned}$$

**Notes**

- System now autonomous in  $(\theta, \omega, \tau)$ .
- Useful for phase-space analysis and numerical integration.

# Key Insights from the Conversion Process

- Any higher-order, non-autonomous system can be converted into a first-order, autonomous system by:
  - Introducing new state variables for higher-order derivatives.
  - Replacing explicit time dependencies with a clock variable ( $\tau$ ).
- The resulting equations:
  - Are always first-order.
  - Have no explicit time dependency (autonomous).

## General Utility

This method ensures compatibility with standard tools for analyzing first-order, autonomous systems.

# Exercise 1: Convert to First-Order Form

**Given Equation:**

$$\frac{d^2x}{dt^2} - x \frac{dx}{dt} + x^2 = 0$$

**Introduce a new variable:**

$$v = \frac{dx}{dt}$$

Substitute and rewrite:

$$\begin{aligned}\dot{x} &= v, \\ \dot{v} &= xv - x^2\end{aligned}$$

**First-Order System:**

$$\begin{aligned}\dot{x} &= v, \\ \dot{v} &= xv - x^2\end{aligned}$$

**Remarks:**

- Nonlinear due to product term  $xv$ .
- Expressed as a system of two coupled ODEs.
- Easier to simulate or visualize in the phase plane.



# Exercise 2: Convert to Autonomous, First-Order Form

**Given Equation:**

$$\frac{d^2x}{dt^2} - a \cos(bt) = 0$$

**Step 1: Introduce**  $v = \frac{dx}{dt}$

$$\dot{x} = v, \quad \dot{v} = a \cos(bt)$$

**Step 2: Eliminate explicit time dependency:**

$$\frac{d\tau}{dt} = 1, \quad \tau(0) = 0$$

Replace  $t \rightarrow \tau$  in the forcing term:

$$\cos(bt) \rightarrow \cos(b\tau)$$

**Autonomous, First-Order Form:**

$$\begin{aligned}\dot{x} &= v, \\ \dot{v} &= a \cos(b\tau), \\ \dot{\tau} &= 1\end{aligned}$$

**Remarks:**

- System is now autonomous in  $(x, v, \tau)$ .
- Clock variable replaces explicit time.
- Technique generalizes to many driven systems.

# Summary: Differential Equations and System Behavior

## 1. Typical Behaviors of Linear Systems

- Exponential growth or decay
- Periodic oscillations
- Stationary (equilibrium) states
- Combined forms, e.g. exponentially damped oscillations

## 2. Extended Behaviors (Non-Diagonalizable Systems)

- Solutions may include
  - Polynomials in time
  - Products of polynomials and exponentials
- Occurs when the coefficient matrix cannot be fully diagonalized

## 3. Solvability and Methods

- Linear ODEs: typically analytically solvable
- Nonlinear ODEs: often require numerical integration
- Many nonlinear systems exhibit complex phenomena:
  - Chaos, bifurcations, limit cycles

### Key Insight

Understanding these distinctions guides model selection and the choice of analytical vs. numerical solution methods.