

Agenda



Introduction

Course Overview

Basic System Classes

First-Order Systems Second-Order Systems

Performance Specification

Poles and Zeros of State Space Models

Stability

Summary

Introduction

Curriculum for Reguleringsteknik (REG)



Matematiske og grafiske metoder til syntese af lineære tidsinvariante systemer: 1

- ► diskret og kontinuert tilstandsbeskrivelse
- analyse i tid og frekvens
- stabilitet, reguleringshastighed, følsomhed og fejl
- ► digitale PI, PID, LEAD og LAG regulatorer (serieregulatorer)
- ► tilstandsregulering, pole-placement og tilstands-estimering (observer)
- ► optimal regulering (least squares) og optimal tilstands-estimation (Kalman-filter)

Færdigheder:

Efter gennemførelse af kurset kan den succesfulde studerende:

 kunne analysere, dimensionere og implementere såvel kontinuert som tidsdiskret regulering af lineære tidsinvariante og stokastiske systemer

Kompetencer:

Efter gennemførelse af kurset kan den succesfulde studerende:

 anvende og implementere klassiske og moderne reguleringsteknikker for at kunne styre og regulere en robot hurtig og præcist

¹ Based on https://fagbesk.sam.sdu.dk/?fag_id=39673



The twelve lectures of the course are

- ► Lecture 1: Introduction to Linear Time-Invariant Systems
- ► Lecture 2: Stability and Performance Analysis
- ► Lecture 3: Introduction to Control
- ► Lecture 4: Design of PID Controllers
- ► Lecture 5: Root Locus
- ► Lecture 6: The Nyquist Plot
- ► Lecture 7: Dynamic Compensators and Stability Margins
- ► Lecture 8: Implementation
- ► Lecture 9: State Feedback
- ► Lecture 10: Observer Design
- ► Lecture 11: Optimal Control (Linear Quadratic Control)
- ► Lecture 12: The Kalman Filter

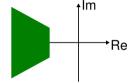


The performance of the system can be specified in both time and frequency.

Time Domain

- Rise time
- Settling time
- Overshoot

Frequency domain



Introduction

Overview of Today's Lecture

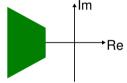


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Frequency domain



The stability of the dynamical system

$$\dot{x} = Ax + Bu$$
$$y = Cx$$

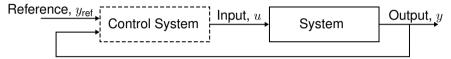
The stability of the transfer function

$$G(s) = \frac{Q(s)}{P(s)}$$

can be determined from the eigenvalues of $\$ can be determined from the poles of G(s). A.



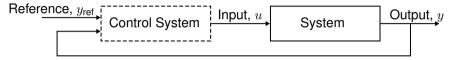
Task: Design a cruise control for a car.



- ightharpoonup Control Input: Throttle position u
- ► Measured Output: Velocity of the car y
- ightharpoonup Reference Input: Desired velocity of the car $y_{\rm ref}$

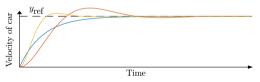


Task: Design a cruise control for a car.



Today, we answers to the following question:

► What is "good" performance?



Basic System Classes



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First-Order Systems Definition



A first-order system has one pole and is described by

$$H(s) = \frac{k}{\tau s + 1}$$

$$\begin{cases} \dot{x} = -\frac{1}{\tau}x + \frac{k}{\tau}u \\ y = x \end{cases}$$

where k is the **DC-gain** and τ is the **time-constant**.

First-Order Systems Definition



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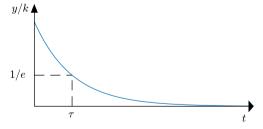
where k is the **DC-gain** and τ is the **time-constant**.

The system has a pole in $s=-\frac{1}{\tau},$ i.e., the smaller time-constant, the faster system response.



The impulse response of a first-order system is

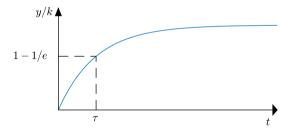
$$y(t) = ke^{-\frac{1}{\tau}t}$$





The step response of a first-order system is

$$y(t) = k(1 - e^{-\frac{1}{\tau}t})$$



Second-Order Systems Definition



The transfer function of a second-order system is

$$H(s) = \frac{k\omega_n^2}{s^2 + 2\zeta\omega_n s + \omega_n^2}$$

and is described by the gain k and two parameters: $\zeta>0$ and $\omega_n>0$.

Second-Order Systems Definition



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and is described by the gain k and two parameters: $\zeta > 0$ and $\omega_n > 0$.

The system has two poles, which are $s \in \mathbb{C}$ where

$$s^2 + 2\zeta\omega_n s + \omega_n^2 = 0$$

Second-Order Systems Poles



The poles of

$$H(s) = \frac{k\omega_n^2}{s^2 + 2\zeta\omega_n s + \omega_n^2}$$

$$s = \frac{-2\zeta\omega_n \pm \sqrt{(2\zeta\omega_n)^2 - 4\cdot 1\cdot \omega_n^2}}{2} = -\zeta\omega_n \pm \omega_n \sqrt{\zeta^2 - 1}.$$

Second-Order Systems Poles



The poles of

$$H(s) = \frac{k\omega_n^2}{s^2 + 2\zeta\omega_n s + \omega_n^2}$$

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1. If $0 < \zeta < 1$ then the poles of H(s) are complex.

Second-Order Systems Poles



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- **1.** If $0 < \zeta < 1$ then the poles of H(s) are complex. (**Underdamped case**)
- **2.** If $\zeta = 1$ then H(s) has a double pole in $s = -\zeta \omega_n$.



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Terminology

- ▶ The parameter ζ is called the **damping ratio**.
- ▶ The parameter ω_n is called the **undamped natural frequency**.

Second-Order Systems

Underdamped Second-Order System (1)



The damping ratio ζ of an underdamped second-order system is smaller than one; thus, it has a complex pole pair with poles at

$$-\underbrace{\zeta\omega_n}_{\sigma}\pm j\underbrace{\omega_n\sqrt{1-\zeta^2}}_{\omega_d}.$$

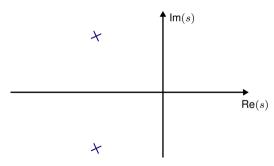
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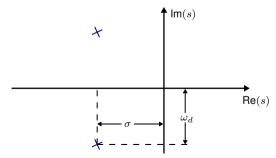


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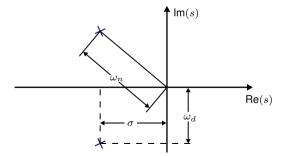


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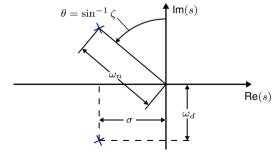
Second-Order Systems

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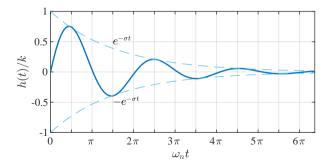
The impulse response of the system is

$$h(t) = k \frac{\omega_n}{\sqrt{1 - \zeta^2}} e^{-\sigma t} \sin(\omega_d t) 1(t).$$



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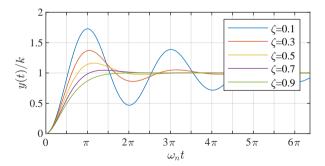
The step response of the system is

$$y(t) = k \left(1 - e^{-\sigma t} \left(\cos(\omega_d t) + \frac{\sigma}{\omega_d} \sin(\omega_d t) \right) \right)$$



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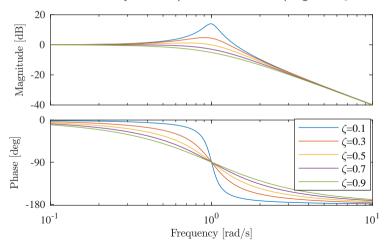
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Second-Order Systems
Underdamped Second-Order System (3)



The Bode plot of a second-order system depends on the damping ratio ζ .



Second-Order Systems Critically Damped Second-Order System (1)



A critically damped system has a damping ratio ζ that is one; thus, it has two poles at $s = -\zeta \omega_n$.

Second-Order Systems Critically Damped Second-Order System (1)



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The impulse response for the system is

$$h(t) = k\omega_n^2 t e^{-\omega_n t}$$

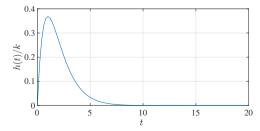
Second-Order Systems Critically Damped Second-Order System (1)



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Second-Order Systems

Critically Damped Second-Order System (2)



A critically damped system has a damping ratio ζ that is one; thus, it has two poles at $s=-\zeta\omega_n$.

The step response for the system is

$$y(t) = k \left(1 - e^{-\omega_n t} - \omega_n t e^{-\omega_n t} \right)$$

Second-Order Systems

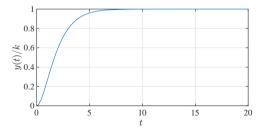
Critically Damped Second-Order System (2)



A critically damped system has a damping ratio ζ that is one; thus, it has two poles at $s=-\zeta\omega_n$.

The step response for the system is

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Second-Order Systems Overdamped Second-Order System (1)



An over damped system has a damping ratio ζ that is greater than one; thus, it has two distinct real poles λ_1 and λ_2 .

Second-Order Systems Overdamped Second-Order System (1)

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The impulse response for the system is

$$h(t) = k \frac{\lambda_1 \lambda_2}{\lambda_1 - \lambda_2} \left(e^{\lambda_1 t} - e^{\lambda_2 t} \right)$$

Second-Order Systems

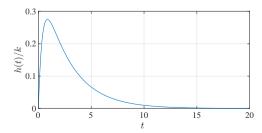
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Second-Order Systems

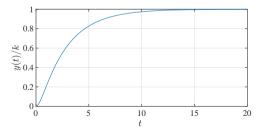
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Performance Specification



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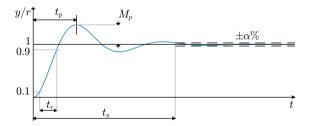
Summary

Performance Specification Time-Domain Specification



We consider three different performance measures of dynamical systems

- 1. The rise time t_r .
- 2. The settling time t_s .
- **3**. The **overshoot** M_p .

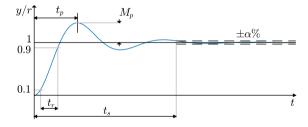


The **peak time** is denoted t_p .

Performance Specification



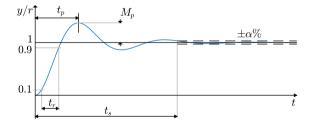
The rise time of a dynamical system is the time it takes to come from 0.1 times the setpoint value to reach 0.9 times the setpoint value.



Performance Specification



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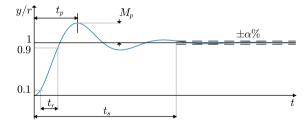
For a second-order system the rise time can be approximated as

$$t_r = \frac{1.8}{\omega_n}$$

Performance Specification Settling Time



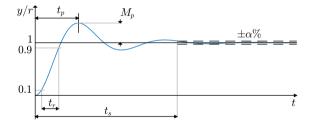
The $\alpha\%$ -settling time is the time it takes for the step response to reach $\pm \alpha\%$ of the setpoint value without leaving it again.



Performance Specification



The $\alpha\%$ -settling time is the time it takes for the step response to reach $\pm \alpha\%$ of the setpoint value without leaving it again.



For a second-order system the settling time can be approximated as

$$t_s = \frac{-\log(\alpha/100)}{\omega_n \zeta}$$



The peak time t_p can be found from the step response of the system

$$y(t) = 1 - e^{-\sigma t} \left(\cos(\omega_d t) + \frac{\sigma}{\omega_d} \sin(\omega_d t) \right)$$

At the maximum of y(t), we have $\dot{y}(t)=0$



The peak time t_p can be found from the step response of the system

$$y(t) = 1 - e^{-\sigma t} \left(\cos(\omega_d t) + \frac{\sigma}{\omega_d} \sin(\omega_d t) \right)$$

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$$\dot{y}(t) = e^{-\sigma t} \left(\omega_d \sin(\omega_d t) + \frac{\sigma^2}{\omega_d} \sin(\omega_d t) \right)$$

Thus, $\dot{y}(t) = 0$ for $\sin(\omega_d t)$, i.e.,

$$t_p = \frac{\pi}{\omega_d}.$$



The overshoot is computed from the step response at the peak time

$$y(t_p) = 1 - e^{-\sigma\pi/\omega_d} \left(\cos(\pi) + \frac{\sigma}{\omega_d} \sin(\pi) \right)$$
$$= 1 + M_p$$



The overshoot is computed from the step response at the peak time

$$y(t_p) = 1 - e^{-\sigma\pi/\omega_d} \left(\cos(\pi) + \frac{\sigma}{\omega_d} \sin(\pi) \right)$$
$$= 1 + M_p$$

This gives an expression for the overshoot

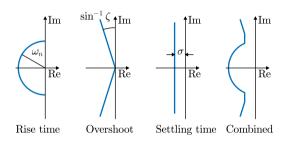
$$M_p = e^{-\pi\zeta/\sqrt{1-\zeta^2}}$$

for $0 \le \zeta \le 1$.

Performance Specification

Frequency and Time Specifications





To obtain a rise rise shorter than t_r

$$\omega_n \ge \frac{1.8}{t_r}$$

To obtain an overshoot that is smaller than M_n

$$\zeta \ge \sqrt{\frac{\left(\frac{\log(M_p)}{-\pi}\right)^2}{1 + \left(\frac{\log(M_p)}{p}\right)^2}}$$

To obtain an α %-settling time shorter than t_s

$$\sigma \ge \frac{-\log(\alpha/100)}{t_s}$$

Poles and Zeros of State Space Models



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Poles of State Space Models Definition



With

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

we have that:

$$G(s) \to \infty \text{ for } s \to p \quad \Rightarrow \quad \det(pI - A) = 0$$

Hence,

$$p$$
 is a pole for $G(s) \Rightarrow$

Poles of State Space Models Definition



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Hence,

p is a pole for $G(s) \ \Rightarrow \ p$ is an eigenvalue for A

Poles of State Space Models

Example: Mass-Spring-Damper



For the mass-spring-damper system, the *A* matrix was:

$$A = \begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix}$$

which has the characteristic polynomial:

$$\det(\lambda I - A) = \begin{vmatrix} \lambda & -1 \\ 2 & \lambda + 3 \end{vmatrix}$$
$$= \lambda^2 + 3\lambda + 2 = (\lambda + 1)(\lambda + 2)$$

Thus, the system has poles in $\{-1, -2\}$.

Zeros of State Space Models Definition



With

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

we have that:

$$G(z)u = 0 \Rightarrow C(zI - A)^{-1} Bu + Du = 0$$

$$\Rightarrow C\xi + Du = 0, \ \xi = (zI - A)^{-1} Bu$$

$$\Rightarrow C\xi + Du = 0, \ (A - zI)\xi + Bu = 0$$

$$\Rightarrow \begin{bmatrix} A - zI & B \\ C & D \end{bmatrix} \begin{bmatrix} \xi \\ u \end{bmatrix} = 0$$

Zeros of State Space Models Definition



$$G(z)u = 0 \Rightarrow C(zI - A)^{-1}Bu + Du = 0$$

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$$\Rightarrow C\xi + Du = 0, \ (A - zI)\xi + Bu = 0$$

$$\Rightarrow \begin{bmatrix} A - zI & B \\ C & D \end{bmatrix} \begin{bmatrix} \xi \\ u \end{bmatrix} = 0$$

Thus, z is a zero for $G(s) \Rightarrow$

$$\begin{bmatrix} A - zI & B \\ C & D \end{bmatrix}$$
 does not have full column rank

Zeros of State Space Models

Zeros of State Spa Example: Mass-Spring-Damper



For the mass-spring-damper system, zeros must satisfy:

$$\begin{vmatrix} A - zI & \mathbf{B} \\ C & D \end{vmatrix} = 0$$

or

$$\begin{vmatrix} -z & 1 & 0 \\ -2 & -3 - z & 1 \\ 0 & 1 & 0 \end{vmatrix} = \begin{vmatrix} -z & 1 \\ 0 & 1 \end{vmatrix} \cdot (-1) = z = 0$$

Hence, the system has a zero in the origin.

Stability



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Basic System Classes
First-Order Systems
Second-Order Systems

Performance Specification

Poles and Zeros of State Space Models

Stability

Summary

Stability



A linear discrete-time system

$$\mathbf{x}_{k+1} = \Phi \mathbf{x}_k$$

where $\mathbf{x}_k \in \mathbb{R}^n$ and $\Phi \in \mathbb{R}^{n \times n}$ is **asymptotically stable** if

$$\lim_{k\to\infty} \mathbf{x}_k = 0$$

for any $\mathbf{x}_0 \in \mathbb{R}^n$.



A linear discrete-time system described by the state equation

$$x_{k+1} = \Phi x_k$$

is asymptotically stable if and only if all eigenvalues have magnitude smaller than one (they are within the unit circle).



Assume that Φ is diagonalizable then by spectral decomposition Φ can be written as

$$\Phi = Q\Lambda Q^{-1}$$

where $\Lambda \in \mathbb{C}^{n \times n}$ is a diagonal matrix with the eigenvalues of Φ on the diagonal, and each column of Q is a (right) eigenvector of Φ .



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The state at sample k can be written as

$$\mathbf{x}_k = \Phi^k \mathbf{x}_0 = Q \Lambda^k Q^{-1} \mathbf{x}_0.$$



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The state at sample k can be written as

$$\mathbf{x}_k = \Phi^k \mathbf{x}_0 = Q \Lambda^k Q^{-1} \mathbf{x}_0.$$

Thus, the system is stable if and only if the magnitude of all diagonal elements of Λ is smaller than one.

Stability



A linear continuous-time system

$$\dot{x} = Ax$$

where $x \in \mathbb{R}^n$ and $A \in \mathbb{R}^{n \times n}$ is **asymptotically stable** if

$$\lim_{t \to \infty} \mathbf{x}(t) = 0$$

for any
$$\mathbf{x}(0) = \mathbf{x}_0 \in \mathbb{R}^n$$
.



A linear continuous-time system described by the state equation

$$\dot{x} = Ax$$

is asymptotically stable if and only if all eigenvalues have negative real part.



$$\mathbf{x}(t) = e^{At}\mathbf{x}(0)$$



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The solution can be rewritten by using the definition of the matrix exponential

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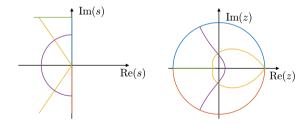
Thus, the system is stable if and only if the real parts of all diagonal elements of Λ are negative.



The equivalent characteristics in the z-plane are related to those in the s-plane by the expression

$$z = e^{sT}$$

where T is the sample period.



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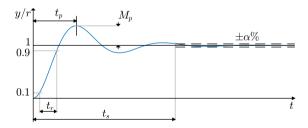
Stability

Summary



We consider three different performance measures of dynamical systems

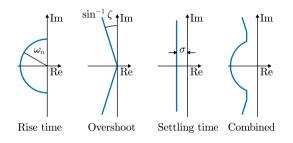
- 1. The rise time t_r .
- 2. The settling time t_s .
- **3.** The **overshoot** M_p .



The **peak time** is denoted t_p .

Summary Frequency-Domain Performance Specifications





To obtain a rise rise shorter than t_r

$$\omega_n \ge \frac{1.8}{t_r}$$

To obtain an overshoot that is smaller than M_n

$$\zeta \ge \sqrt{\frac{\left(\frac{\log(M_p)}{-\pi}\right)^2}{1 + \left(\frac{\log(M_p)}{-\pi}\right)^2}}$$

To obtain an α %-settling time shorter than t_s

$$\sigma \ge \frac{-\log(\alpha/100)}{t_c}$$

Summary



A linear continuous-time system described by the state equation

$$\dot{x} = Ax$$

is asymptotically stable if and only if all eigenvalues have negative real part.

A linear discrete-time system described by the state equation

$$x_{k+1} = \Phi x_k$$

is asymptotically stable if and only if all eigenvalues have magnitude smaller than one (they are within the unit circle).