

Integral Control and Optimal Control

Control Engineering (Reguleringsteknik)

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Agenda



Introduction

Anti-Windup

Optimal Control

Example: Optimal Control

Introducing Reference Signals

Zero Assignment

Example: Zero Assignment



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 regulerings teknikker** for at kunne styre og regulere en robot **hurtig og præcist**

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

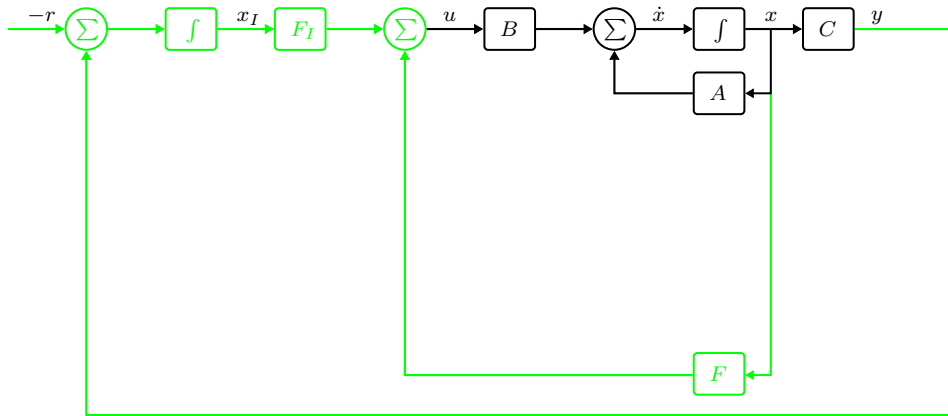


The twelve lectures of the course are

- ▶ **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

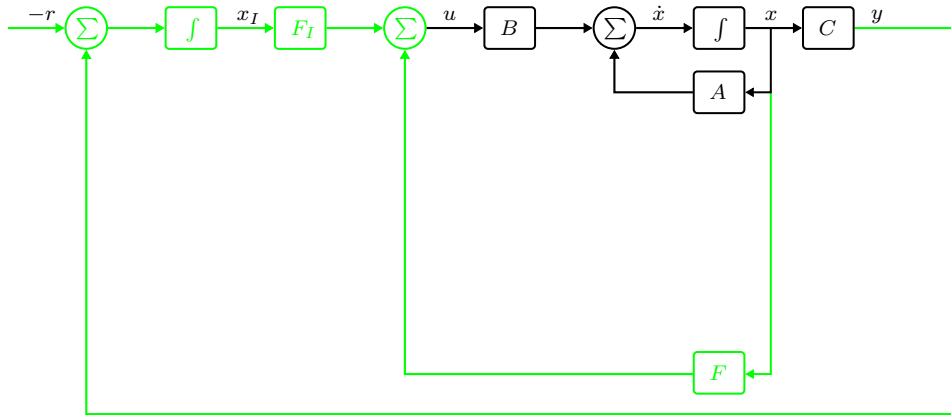
Integral Control

Block Diagram



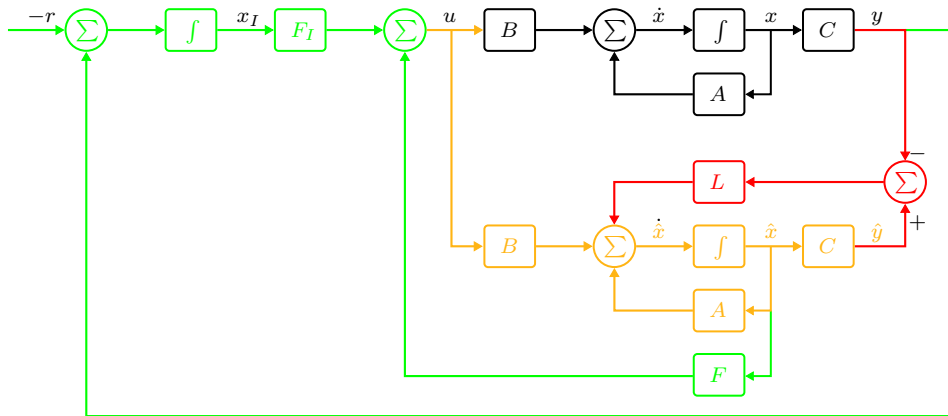
Integral Control

Block Diagram of Observer-Based Integral Control



Integral Control

Block Diagram of Observer-Based Integral Control





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Example: Optimal Control

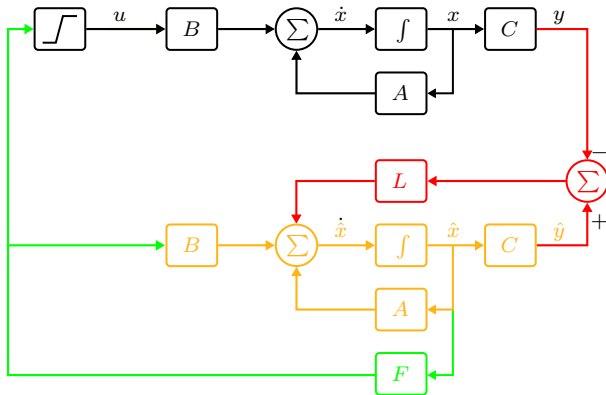
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Example: Zero Assignment

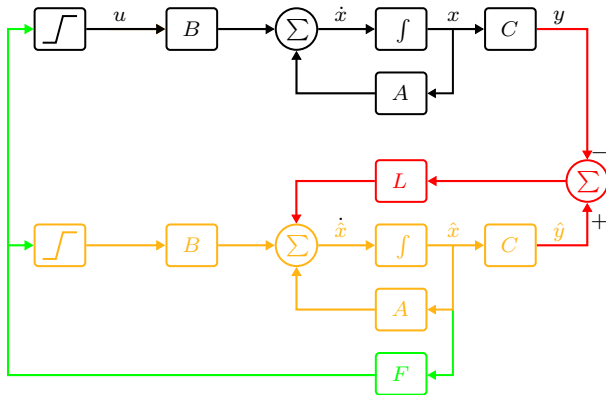
Anti-Windup

Architecture



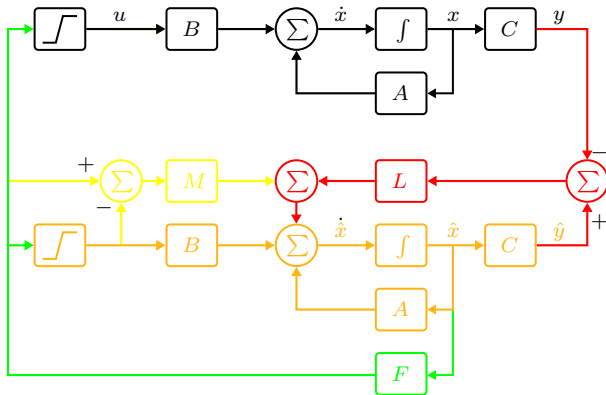
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Architecture



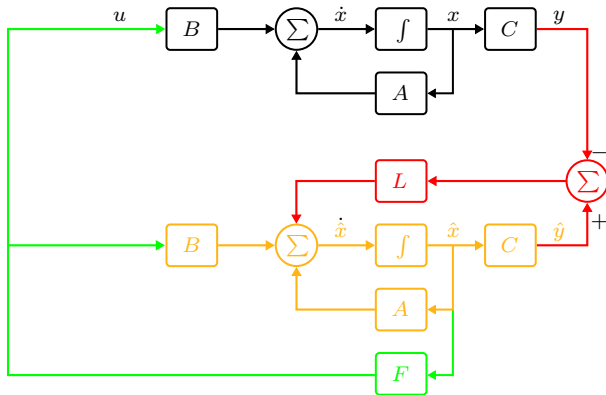
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Architecture



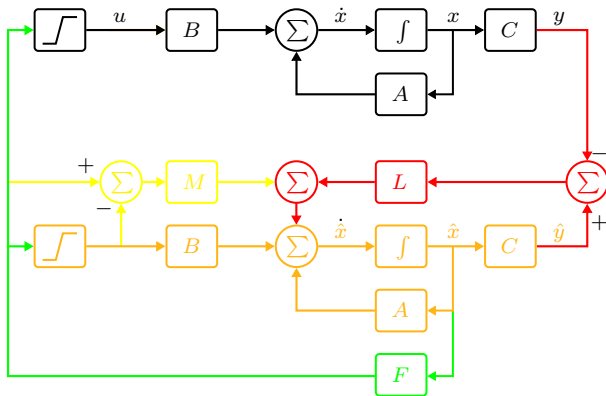
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Nominal



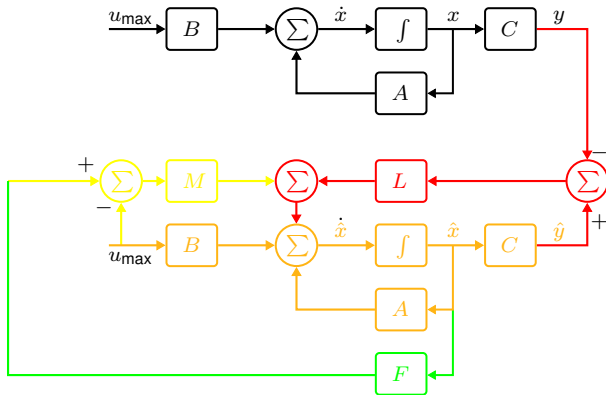
Anti-Windup

Saturated



Anti-Windup

Saturated





Dynamics of controller during saturation:

$$\dot{\hat{x}} = A\hat{x} + LC\hat{x} + MF\hat{x}$$

or

$$\dot{\hat{x}} = (A + LC + MF)\hat{x}$$



Dynamics of controller during saturation:

$$\dot{\hat{x}} = A\hat{x} + LC\hat{x} + MF\hat{x}$$

or

$$\dot{\hat{x}} = (A + LC + MF)\hat{x}$$

Determining M can be recognized as an observer gain design problem:

$$\dot{\hat{x}} = (\tilde{A} + \tilde{L}\tilde{C})\hat{x}$$

with $\tilde{A} = A + LC$, $\tilde{L} = M$, and $\tilde{C} = F$, from which the unknown $\tilde{L} = M$ can be chosen to assign any desired poles to the saturated controller.



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We consider a linear control system of the form:

$$\begin{aligned}\dot{x} &= Ax + Bu, \quad x(0) = x_0 \\ y &= Cx\end{aligned}$$

A control law for such a system is said to be *optimal*, if it minimizes the cost functional:

$$\mathcal{J} = \int_0^\infty x^T Q x + u^T R u \, dt$$

where $Q = Q^T$ is a positive semi-definite matrix and $R = R^T$ is a positive definite matrix.



An *Algebraic Riccati Equation* is a second order matrix equation in an indeterminate $P = P^T \in \mathbb{R}^{n \times n}$ of the form:

$$A^T P + P A - P B R^{-1} B^T P + Q = 0$$

where $A \in \mathbb{R}^{n \times n}$, $B \in \mathbb{R}^{n \times m}$ are matrices, $R = R^T \in \mathbb{R}^{m \times m}$ is a positive definite matrix, and $Q = Q^T \in \mathbb{R}^{n \times n}$ is a positive semidefinite matrix.



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P is called a *stabilizing solution* to the ARE, if it satisfies the equation, and further satisfies that the eigenvalues of $A - B R^{-1} B^T P$ are in the open left half plane.



THEOREM. Consider a linear system of the form:

$$\begin{aligned}\dot{x} &= Ax + Bu, \quad x(0) = x_0 \\ y &= Cx\end{aligned}$$

Let P be a stabilizing solution to the ARE:

$$A^T P + PA - PBR^{-1}B^T P + Q = 0$$

Then the optimal state feedback law is given by:

$$u = Fx \quad \text{where} \quad F = -R^{-1}B^T P$$



Introducing $y = Cx$ into a cost functional of the type

$$\mathcal{J} = \int_0^{\infty} \rho y^T y + u^T u \, dt, \quad \rho \in \mathbb{R}$$

this can be written as an optimal control problem

$$\begin{aligned} \mathcal{J} &= \int_0^{\infty} \rho y^T y + u^T u \, dt \\ &= \int_0^{\infty} \rho x^T C^T C x + u^T u \, dt \\ &= \int_0^{\infty} x^T Q x + u^T R u \, dt, \quad Q = \rho C^T C, R = I \end{aligned}$$



Alternatively, use a cost functional of the type

$$\mathcal{J} = \int_0^{\infty} x^T Q x + u^T R u \, dt$$

where Q and R are diagonal matrices with this can be written as an optimal control problem

$$Q_{ii} = \frac{1}{\text{maximum acceptable value of } x_i^2}$$
$$R_{jj} = \frac{1}{\text{maximum acceptable value of } u_j^2}.$$

Given the system

$$\begin{aligned}\dot{x} &= Ax + Bu + Gw \\ y &= Cx + Du + v\end{aligned}$$

with unbiased process noise w and measurement noise v with covariances

$$\mathcal{E}\{ww^T\} = Q, \quad \mathcal{E}\{vv^T\} = R$$

Then an optimal state estimator is given by:

$$\dot{\hat{x}} = A\hat{x} + Bu + L(C\hat{x} - y)$$



with unbiased process noise w and measurement noise v with covariances

$$\mathcal{E}\{ww^T\} = Q, \quad \mathcal{E}\{vv^T\} = R$$

Then an optimal state estimator is given by:

$$\dot{\hat{x}} = A\hat{x} + Bu + L(C\hat{x} - y)$$

where

$$L = -PC^T R^{-1}$$

P is a stabilizing solution to the ARE:

$$AP + PA^T - PC^T R^{-1} CP + Q = 0$$



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Example (1)



We consider once again the system

$$\begin{aligned}\dot{x} &= \begin{bmatrix} 2 & -3 \\ 4 & -5 \end{bmatrix} x + \begin{bmatrix} 2 \\ 3 \end{bmatrix} u \\ y &= \begin{bmatrix} -3 & 2 \end{bmatrix} x\end{aligned}$$

Computing an optimal state feedback for the cost functional:

$$\mathcal{J} = \int_0^{\infty} 800 y^T y + u^T u \, dt$$

can be done with the MATLAB command

$$F_{opt} = -\text{lqr}(A, B, 800 * C' * C, 1)$$



This yields the result:

$$F_{\text{opt}} = \begin{bmatrix} 69.3536 & -47.8542 \end{bmatrix}$$

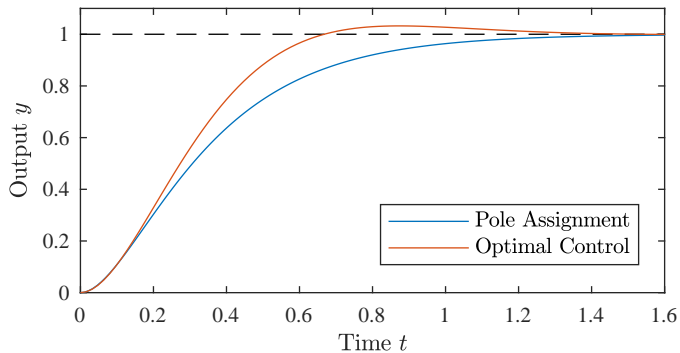
In comparison, a pole assignment with the poles $\{-4, -8\}$ leads to the gain:

$$F = \begin{bmatrix} 72 & -51 \end{bmatrix}$$

A first glance would suggest that the pole assignment with its larger gains would have faster dynamics. However, the optimal feedback assigns complex poles, giving a better rise-time.

Optimal Control

Example (3)



Introducing Reference Signals



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Example: Optimal Control

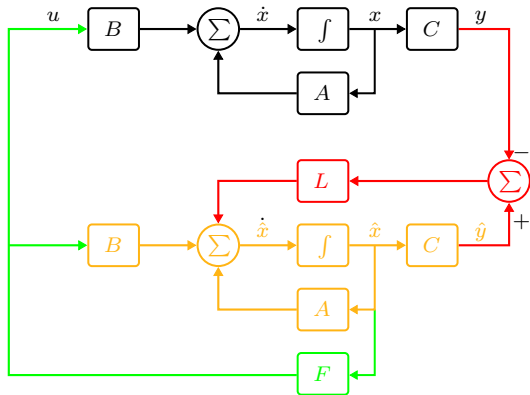
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Zero Assignment

Example: Zero Assignment

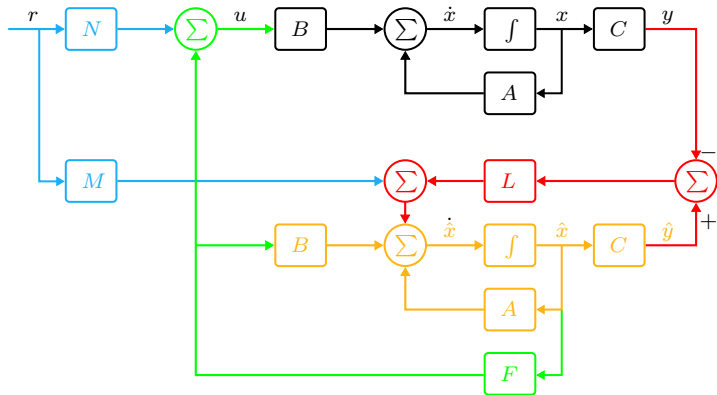
Introducing Reference Signals

Block Diagram



Introducing Reference Signals

Block Diagram



Introducing Reference Signals

System Description



System:

$$\dot{x} = Ax + B(F\hat{x} + Nr)$$

$$y = Cx$$

Observer:

$$\dot{\hat{x}} = A\hat{x} + BF\hat{x} + L(C\hat{x} - y) + Mr$$

System:

$$\begin{aligned}\dot{x} &= Ax + B(\textcolor{red}{F}\hat{x} + \textcolor{blue}{N}r) \\ y &= Cx\end{aligned}$$

Observer:

$$\dot{\hat{x}} = A\hat{x} + \textcolor{red}{B}\textcolor{red}{F}\hat{x} + \textcolor{red}{L}(C\hat{x} - y) + \textcolor{blue}{M}r$$

$$\begin{aligned}\begin{bmatrix} \dot{x} \\ \dot{\hat{x}} \end{bmatrix} &= \begin{bmatrix} A & \textcolor{red}{B}\textcolor{red}{F} \\ -\textcolor{red}{L}C & A + \textcolor{red}{B}\textcolor{red}{F} + \textcolor{red}{L}C \end{bmatrix} \begin{bmatrix} x \\ \hat{x} \end{bmatrix} + \begin{bmatrix} \textcolor{blue}{B}\textcolor{blue}{N} \\ \textcolor{blue}{M} \end{bmatrix} r \\ y &= \begin{bmatrix} C & 0 \end{bmatrix} \begin{bmatrix} x \\ \hat{x} \end{bmatrix}\end{aligned}$$

Introducing Reference Signals

Zeros of State Space Model



Recall from Lecture 2 how to find the zeros of a state space model.

LEMMA. A square (#inputs=#outputs) system with a state space model of the form

$$\begin{aligned}\dot{x} &= Ax + Bu \\ y &= Cx + Du\end{aligned}$$

has a zero with value $z \in \mathbb{C}$ only if

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

Introducing Reference Signals

Zeros of Closed-Loop System



$$\det \left(\begin{bmatrix} A_{cl} - zI & B_{cl} \\ C_{cl} & D_{cl} \end{bmatrix} \right) = 0$$

Introducing Reference Signals

Zeros of Closed-Loop System



$$\det \left(\begin{bmatrix} A_{cl} - zI & B_{cl} \\ C_{cl} & D_{cl} \end{bmatrix} \right) = 0$$

$$\det \left(\begin{bmatrix} A - zI & BF & BN \\ -LC & A + BF + LC - zI & M \\ C & 0 & 0 \end{bmatrix} \right) = 0$$

Introducing Reference Signals

Zeros of Closed-Loop System



$$\det \left(\begin{bmatrix} A_{cl} - zI & B_{cl} \\ C_{cl} & D_{cl} \end{bmatrix} \right) = 0$$

$$\det \left(\begin{bmatrix} A - zI & BF & BN \\ -LC & A + BF + LC - zI & M \\ C & 0 & 0 \end{bmatrix} \right) = 0$$

$$\det \left(\begin{bmatrix} A - zI & BF - BNN^{-1}F & BN \\ -LC & A + BF + LC - zI - MN^{-1}F & M \\ C & 0 & 0 \end{bmatrix} \right) = 0$$

Introducing Reference Signals

Zeros of Closed-Loop System



$$\det \begin{pmatrix} A_{cl} - zI & B_{cl} \\ C_{cl} & D_{cl} \end{pmatrix} = 0$$

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$$\det \begin{pmatrix} A - zI & BF - BNN^{-1}F & BN \\ -LC & A + BF + LC - zI - MN^{-1}F & M \\ C & 0 & 0 \end{pmatrix} = 0$$

$$\det \begin{pmatrix} A - zI & 0 & BN \\ -LC & A + BF + LC - MN^{-1}F - zI & M \\ C & 0 & 0 \end{pmatrix} = 0$$

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Zeros of Closed-Loop System



$$\det \begin{pmatrix} \begin{bmatrix} A - zI & BF & BN \\ -LC & A + BF + LC - zI & M \\ C & 0 & 0 \end{bmatrix} \end{pmatrix} = 0$$

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$$\det \begin{pmatrix} \begin{bmatrix} A - zI & 0 & BN \\ -LC & A + BF + LC - MN^{-1}F - zI & M \\ C & 0 & 0 \end{bmatrix} \end{pmatrix} = 0$$

$$\det \begin{pmatrix} \begin{bmatrix} A - zI & 0 & B \\ -LC & A + BF + LC - \tilde{M}F - zI & \tilde{M} \\ C & 0 & 0 \end{bmatrix} \end{pmatrix} = 0$$

Introducing Reference Signals

Zeros of Closed-Loop System



$$\det \begin{pmatrix} \begin{bmatrix} A-zI & BF-BNN^{-1}F & BN \\ -LC & A+BF+LC-zI-MN^{-1}F & M \\ C & 0 & 0 \end{bmatrix} \end{pmatrix} = 0$$

$$\det \begin{pmatrix} \begin{bmatrix} A-zI & 0 & BN \\ -LC & A+BF+LC-MN^{-1}F-zI & M \\ C & 0 & 0 \end{bmatrix} \end{pmatrix} = 0$$

$$\det \begin{pmatrix} \begin{bmatrix} A-zI & 0 & B \\ -LC & A+BF+LC-\tilde{M}F-zI & \tilde{M} \\ C & 0 & 0 \end{bmatrix} \end{pmatrix} = 0$$

$$\begin{cases} \det \left(\begin{bmatrix} A-zI & B \\ C & 0 \end{bmatrix} \right) = 0 & \text{or} \\ \det \left(A+BF+LC-\tilde{M}F-zI \right) = 0 \end{cases}$$

Zero Assignment



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Zero Assignment

Placement of Closed-Loop Zeros



LEMMA. If \tilde{M} is an 'observer gain' such that the characteristic polynomial of the matrix $A_{za} + \tilde{M}C_{za}$ has the characteristic polynomial

$$\det \left(sI - \left(A_{za} + \tilde{M}C_{za} \right) \right) = (s - z_1) \cdots (s - z_n)$$

with $A_{za} = A + BF + LC$ and $C_{za} = -F$, then the numbers z_1, \dots, z_n are all zeros of the closed loop transfer function from r to y .

Zero Assignment

Algorithm for Zero Assignment



1. Design \tilde{M} assigning zeros close to the cut-off frequency of the Bode plot, such that the 'horizontal' part is extended.

Zero Assignment

Algorithm for Zero Assignment



1. Design \tilde{M} assigning zeros close to the cut-off frequency of the Bode plot, such that the 'horizontal' part is extended.
2. Compute N such that the DC-value of the transfer function from r to y is unity:

$$N = - \left(C_{\text{cl}} A_{\text{cl}}^{-1} \tilde{B}_{\text{cl}} \right)^{-1}$$

where

$$A_{\text{cl}} = \begin{bmatrix} A & BF \\ -LC & A + BF + LC \end{bmatrix}, \quad \tilde{B}_{\text{cl}} = \begin{bmatrix} B \\ \tilde{M} \end{bmatrix}$$
$$C_{\text{cl}} = [C \quad 0]$$

Zero Assignment

Algorithm for Zero Assignment



1. Design \tilde{M} assigning zeros close to the cut-off frequency of the Bode plot, such that the 'horizontal' part is extended.
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$$C_{\text{cl}} = [C \quad 0]$$

3. Compute $M = MN^{-1}N = \tilde{M}N$.

Zero Assignment

Example



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Zero Assignment

Example (1)



We consider again the system

$$\begin{aligned}\dot{x} &= \begin{bmatrix} 2 & -3 \\ 4 & -5 \end{bmatrix} x + \begin{bmatrix} 2 \\ 3 \end{bmatrix} u \\ y &= \begin{bmatrix} -3 & 2 \end{bmatrix} x\end{aligned}$$

A state feedback F that assign poles in $\{-3, -4\}$ and an observer gain L that assigns poles in $\{-9, -12\}$ are given by:

$$F = \begin{bmatrix} 22 & -16 \end{bmatrix}, \quad L = \begin{bmatrix} -122 \\ -192 \end{bmatrix}$$

We would like to assign zeros from r to y in $\{-3, -4\}$ to cancel the poles from F .

Zero Assignment

Example (2)



With these values of F and L we obtain:

$$A_{za} = A + BF + LC = \begin{bmatrix} 412 & -279 \\ 646 & -437 \end{bmatrix}$$

$$C_{za} = -F = \begin{bmatrix} -22 & 16 \end{bmatrix}$$

An 'observer gain' that assigns poles in $\{-3, -4\}$ for $A_{za} + \tilde{M}C_{za}$ is

$$\tilde{M} = \begin{bmatrix} 7.0460 \\ 10.8133 \end{bmatrix}$$

Zero Assignment

Example (3)



N can be computed as:

$$N = - \left((C \quad 0) \begin{pmatrix} A & BF \\ -LC & A + BF + LC \end{pmatrix}^{-1} \begin{pmatrix} B \\ \tilde{M} \end{pmatrix} \right)^{-1} \\ = 108$$

M is obtained from:

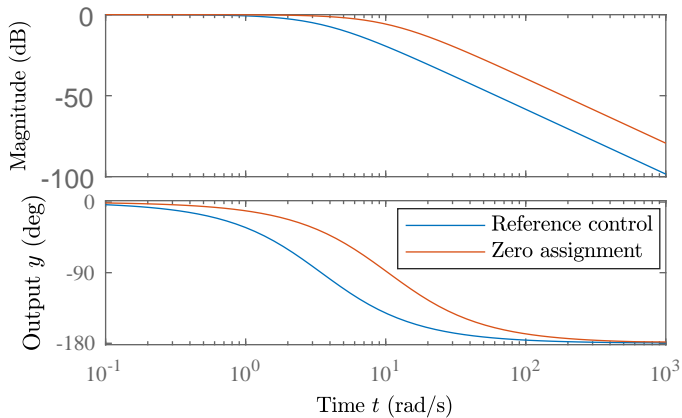
$$M = \tilde{M}N = \begin{bmatrix} 7.0460 \\ 10.8133 \end{bmatrix} \cdot 108 = \begin{bmatrix} 760.97 \\ 1167.84 \end{bmatrix}$$

Zero Assignment

Example: Bode Plot



Bode Diagram



Zero Assignment

Example: Step Response

