

Fall 2007

線性系統 Linear Systems

Chapter 08 State Feedback & State Estimators (SISO)

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Materials used in these lecture notes are adopted from
“Linear System Theory & Design,” 3rd. Ed., by C.-T. Chen (1999)

Outline

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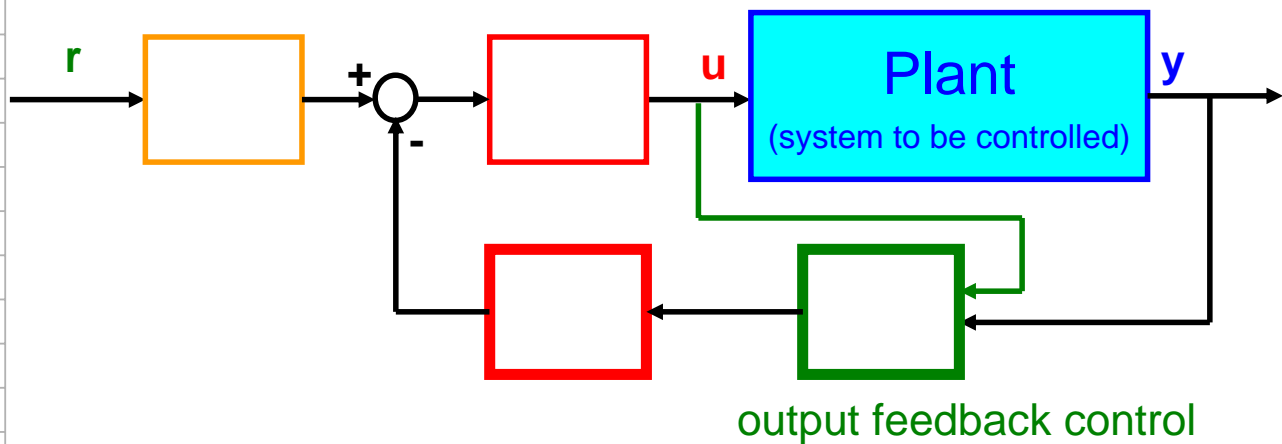
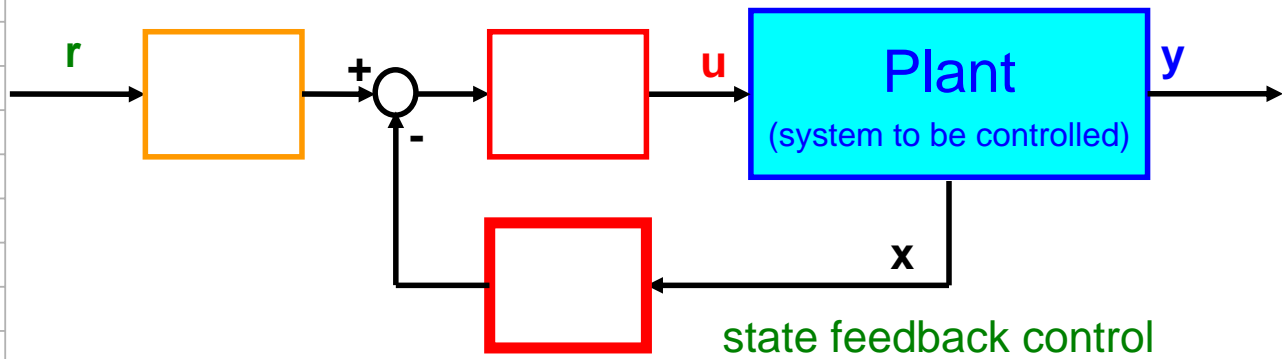
- State Feedback (8.2)
 - State Feedback in Controllable Canonical Form
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- Regulation & Tracking (8.3)
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 - Reduced-Dimensional State Estimator
- Feedback from Estimated States (8.5)

Basic Concept of Feedback Control (8.1)



r : reference signal

r: reference signal



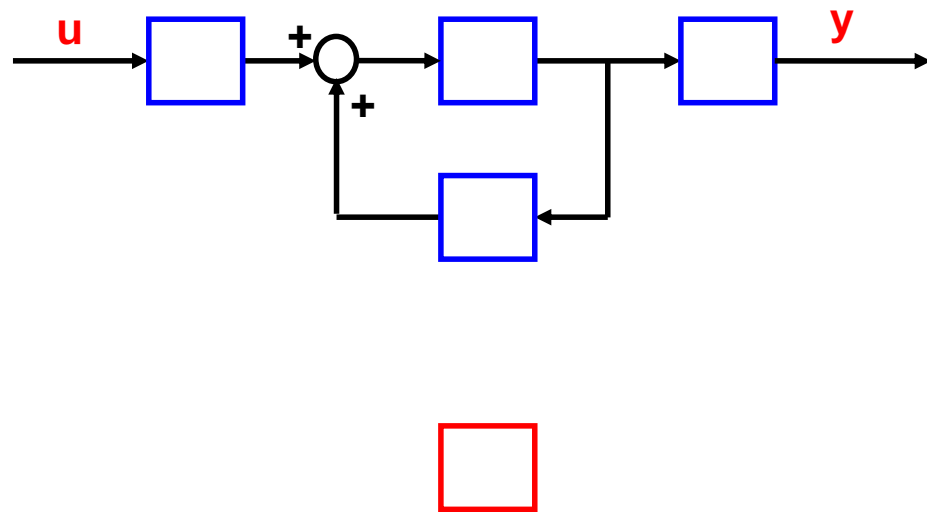
State Feedback (SISO Systems) (8.2)

$$\begin{cases} \dot{\mathbf{x}}(t) = \mathbf{A} \mathbf{x}(t) + \mathbf{b} \mathbf{u}(t) \\ \mathbf{y}(t) = \mathbf{c} \mathbf{x}(t) \end{cases}$$

$$\mathbf{u}(t) =$$

- Closed-loop system :

$$\begin{cases} \dot{\mathbf{x}}(t) = \mathbf{A} \mathbf{x}(t) + \mathbf{b} \mathbf{u}(t) = \\ \mathbf{y}(t) = \mathbf{c} \mathbf{x}(t) \end{cases}$$

**Theorem 8.1 (8.2)**Feng-Li Lian © 2007
NTUEE-LS8-DesignSISO-8**Theorem 8.1**

The pair $(\mathbf{A} - \mathbf{b}\mathbf{k}, \mathbf{b})$, for any $1 \times n$ real constant vector \mathbf{k} , is controllable if and only if (\mathbf{A}, \mathbf{b}) is controllable.

Proof:

Example 8.1 (8.2)

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$$\dot{\mathbf{x}} = \begin{bmatrix} 1 & 2 \\ 3 & 1 \end{bmatrix} \mathbf{x} + \begin{bmatrix} 0 \\ 1 \end{bmatrix} u$$

$$y = \begin{bmatrix} 1 & 2 \end{bmatrix} \mathbf{x}$$

Controllability: $\mathcal{C} = \begin{bmatrix} \\ \end{bmatrix}$

Observability: $\mathcal{O} = \begin{bmatrix} \\ \end{bmatrix}$

controllable and observable

State feedback: $u = r - \begin{bmatrix} \\ \end{bmatrix} \mathbf{x}$

Feedback system:

Example 8.1 – 2

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$$\dot{\mathbf{x}} = \begin{bmatrix} 1 & 2 \\ 3 & 1 \end{bmatrix} \mathbf{x} + \begin{bmatrix} 0 \\ 1 \end{bmatrix} \left(r - \begin{bmatrix} \\ \end{bmatrix} \mathbf{x} \right)$$

$$= \begin{bmatrix} 1 & 2 \\ 3 & 1 \end{bmatrix} \mathbf{x} + \begin{bmatrix} \\ \end{bmatrix} \mathbf{x} + \begin{bmatrix} 0 \\ 1 \end{bmatrix} r$$

$$= \begin{bmatrix} \\ \end{bmatrix} \mathbf{x} + \begin{bmatrix} 0 \\ 1 \end{bmatrix} r$$

Controllability: $\mathcal{C}_f = \begin{bmatrix} \\ \end{bmatrix}$

Observability: $\mathcal{O}_f = \begin{bmatrix} \\ \end{bmatrix}$

- Observability may change after state feedback

$$\dot{\mathbf{x}} = \begin{bmatrix} 1 & 3 \\ 3 & 1 \end{bmatrix} \mathbf{x} + \begin{bmatrix} 1 \\ 0 \end{bmatrix} u$$

Characteristic Polynomial:

$$\det \left(s \mathbf{I} - \begin{bmatrix} 1 & 3 \\ 3 & 1 \end{bmatrix} \right) =$$

State feedback:

$$u = r - \begin{bmatrix} k_1 & k_2 \end{bmatrix} \mathbf{x}$$

Feedback system:

$$\dot{\mathbf{x}} = \begin{bmatrix} 1 & 3 \\ 3 & 1 \end{bmatrix} \mathbf{x} + \begin{bmatrix} 1 \\ 0 \end{bmatrix} \left(r - \begin{bmatrix} k_1 & k_2 \end{bmatrix} \mathbf{x} \right)$$

$$= \begin{bmatrix} 1 & 3 \\ 3 & 1 \end{bmatrix} \mathbf{x} + \begin{bmatrix} -k_1 & -k_2 \\ 0 & 0 \end{bmatrix} \mathbf{x} + \begin{bmatrix} 1 \\ 0 \end{bmatrix} r$$

$$= \begin{bmatrix} 1 - k_1 & 3 - k_2 \\ 3 & 1 \end{bmatrix} \mathbf{x} + \begin{bmatrix} 1 \\ 0 \end{bmatrix} r$$

Characteristic Polynomial:

$$\det \left(s \mathbf{I} - \begin{bmatrix} 1 - k_1 & 3 - k_2 \\ 3 & 1 \end{bmatrix} \right) =$$

$$\{A_1, b_1, c_1, d_1\} \iff \{A_2, b_2, c_2, d_2\}$$

$$x_2 = P x_1$$

$$\dot{x}_1 = A_1 x_1 + b_1 u$$

$$y = c_1 x_1 + d_1 u$$



$$\dot{x}_2 = A_2 x_2 + b_2 u$$

$$y = c_2 x_2 + d_2 u$$

$$A_2 = P A_1 P^{-1}$$

$$b_2 = P b_1$$

$$c_2 = c_1 P^{-1}$$

State Feedback in Controllable Canonical Form – 2

$$\dot{x} = Ax + bu$$

If $\{A, b\}$ controllable

$$y = cx + du$$

$$\Delta(s) = s^4 + a_1 s^3 + a_2 s^2 + a_3 s + a_4$$

$$C = \begin{bmatrix} b & Ab & A^2b & A^3b \end{bmatrix} \text{ is invertible}$$

$$P_1 = C^{-1} = \begin{bmatrix} b & Ab & A^2b & A^3b \end{bmatrix}^{-1}$$

$$\begin{aligned} A \begin{bmatrix} b & Ab & A^2b & A^3b \end{bmatrix} &= \begin{bmatrix} b & Ab & A^2b & A^3b \end{bmatrix} \\ &= \begin{bmatrix} Ab & AAb & AA^2b & AA^3b \end{bmatrix} = \end{aligned}$$

$$\mathbf{P}_1 \mathbf{A} \mathbf{P}_1^{-1} = \begin{bmatrix} 0 & 0 & 0 & -a_4 \\ 1 & 0 & 0 & -a_3 \\ 0 & 1 & 0 & -a_2 \\ 0 & 0 & 1 & -a_1 \end{bmatrix} =: \mathbf{A}_1$$

$$\mathbf{b} = \begin{bmatrix} \mathbf{b} & \mathbf{A}\mathbf{b} & \mathbf{A}^2\mathbf{b} & \mathbf{A}^3\mathbf{b} \end{bmatrix} \begin{bmatrix} \\ \\ \\ \end{bmatrix}$$

$$\mathbf{P}_1 \mathbf{b} = \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} =: \mathbf{b}_1$$

$$\mathbf{c} \mathbf{P}_1^{-1} = \mathbf{c} \begin{bmatrix} \mathbf{b} & \mathbf{A}\mathbf{b} & \mathbf{A}^2\mathbf{b} & \mathbf{A}^3\mathbf{b} \end{bmatrix} = \begin{bmatrix} \mathbf{c}\mathbf{b} & \mathbf{c}\mathbf{A}\mathbf{b} & \mathbf{c}\mathbf{A}^2\mathbf{b} & \mathbf{c}\mathbf{A}^3\mathbf{b} \end{bmatrix} =: \mathbf{c}_1$$

$$\mathbf{P}_2 = \begin{bmatrix} 1 & a_1 & a_2 & a_3 \\ 0 & 1 & a_1 & a_2 \\ 0 & 0 & 1 & a_1 \\ 0 & 0 & 0 & 1 \end{bmatrix}^{-1}$$

$$\begin{bmatrix} 1 & a_1 & a_2 & a_3 \\ 0 & 1 & a_1 & a_2 \\ 0 & 0 & 1 & a_1 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} -a_1 & -a_2 & -a_3 & -a_4 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} = \begin{bmatrix} 0 & 0 & 0 & -a_4 \\ 1 & 0 & 0 & -a_3 \\ 0 & 1 & 0 & -a_2 \\ 0 & 0 & 1 & -a_1 \end{bmatrix} \begin{bmatrix} 1 & a_1 & a_2 & a_3 \\ 0 & 1 & a_1 & a_2 \\ 0 & 0 & 1 & a_1 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

$$\begin{bmatrix} 1 & a_1 & a_2 & a_3 \\ 0 & 1 & a_1 & a_2 \\ 0 & 0 & 1 & a_1 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\mathbf{A}_2 = \mathbf{P}_2 \mathbf{A}_1 \mathbf{P}_2^{-1}$$

$$\mathbf{b}_2 = \mathbf{P}_2 \mathbf{b}_1$$

$$\mathbf{c}_1 \mathbf{P}_2^{-1} = \begin{bmatrix} \mathbf{c}\mathbf{b} & \mathbf{c}\mathbf{A}\mathbf{b} & \mathbf{c}\mathbf{A}^2\mathbf{b} & \mathbf{c}\mathbf{A}^3\mathbf{b} \end{bmatrix} \begin{bmatrix} 1 & a_1 & a_2 & a_3 \\ 0 & 1 & a_1 & a_2 \\ 0 & 0 & 1 & a_1 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

$$= \begin{bmatrix} (\mathbf{c}\mathbf{b}) & (\mathbf{a}_1 \mathbf{c}\mathbf{b} + \mathbf{c}\mathbf{A}\mathbf{b}) & \cdots \end{bmatrix} = \begin{bmatrix} b_1 & b_2 & b_3 & b_4 \end{bmatrix} =: \mathbf{c}_2$$

$$\{A, b, c, d\}$$

$$P_1 = [b \quad Ab \quad A^2b \quad A^3b]^{-1}$$

$$\Leftrightarrow \left\{ \begin{bmatrix} 0 & 0 & 0 & -a_4 \\ 1 & 0 & 0 & -a_3 \\ 0 & 1 & 0 & -a_2 \\ 0 & 0 & 1 & -a_1 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}, [cb \quad cAb \quad cA^2b \quad cA^3b], d \right\}$$

$$P_2 = \begin{bmatrix} 1 & a_1 & a_2 & a_3 \\ 0 & 1 & a_1 & a_2 \\ 0 & 0 & 1 & a_1 \\ 0 & 0 & 0 & 1 \end{bmatrix}^{-1}$$

$$P = P_2 P_1$$

$$\Leftrightarrow \left\{ \begin{bmatrix} -a_1 & -a_2 & -a_3 & -a_4 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}, [b_1 \quad b_2 \quad b_3 \quad b_4], d \right\}$$

$$\dot{x}_c = A_c x_c + b_c u = \begin{bmatrix} -a_1 & -a_2 & -a_3 & -a_4 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} x_c + \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} u$$

$$\Delta(s) = s^4 + a_1 s^3 + a_2 s^2 + a_3 s + a_4$$

$$\Delta_d(s) = s^4 + d_1 s^3 + d_2 s^2 + d_3 s + d_4$$

$$\dot{x}_c = A_d x_c + b_d u = \begin{bmatrix} -d_1 & -d_2 & -d_3 & -d_4 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} x_c + \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} u$$

$$u = r - [k_1 \ k_2 \ k_3 \ k_4] \mathbf{x}_c$$

$$\dot{\mathbf{x}}_c = \begin{bmatrix} -a_1 & -a_2 & -a_3 & -a_4 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \mathbf{x}_c + \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} (r - [k_1 \ k_2 \ k_3 \ k_4] \mathbf{x}_c)$$

$$= \left(\begin{bmatrix} -a_1 & -a_2 & -a_3 & -a_4 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} + \begin{bmatrix} -k_1 & -k_2 & -k_3 & -k_4 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \right) \mathbf{x}_c + \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} r$$

$$= \begin{bmatrix} (-a_1 - k_1) & (-a_2 - k_2) & (-a_3 - k_3) & (-a_4 - k_4) \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \mathbf{x}_c + \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} r$$

$$\dot{\mathbf{x}}_c = \begin{bmatrix} -d_1 & -d_2 & -d_3 & -d_4 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \mathbf{x}_c + \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} r$$

$$\Delta(s) = s^4 + a_1 s^3 + a_2 s^2 + a_3 s + a_4$$

$$\Delta_d(s) = s^4 + d_1 s^3 + d_2 s^2 + d_3 s + d_4$$

$$u = r - [k_1 \ k_2 \ k_3 \ k_4] \mathbf{x}_c$$

$$k_i = (d_i - a_i)$$

$$\{A, b, c, d\} \iff \left\{ \begin{bmatrix} -a_1 & -a_2 & -a_3 & -a_4 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}, c_c, d \right\}$$

$$u = r - \mathbf{k}\mathbf{x}$$

$$\mathbf{P}^{-1} = \begin{bmatrix} \mathbf{b} & \mathbf{A}\mathbf{b} & \mathbf{A}^2\mathbf{b} & \mathbf{A}^3\mathbf{b} \end{bmatrix} \begin{bmatrix} 1 & a_1 & a_2 & a_3 \\ 0 & 1 & a_1 & a_2 \\ 0 & 0 & 1 & a_1 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

$$\Delta(s) = s^4 + a_1s^3 + a_2s^2 + a_3s + a_4$$

$$\iff \left\{ \begin{bmatrix} -d_1 & -d_2 & -d_3 & -d_4 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}, c_c - d\mathbf{k}_c, d \right\}$$

$$\mathbf{k} = \mathbf{k}_c\mathbf{P}$$

$$u = r - \mathbf{k}_c\mathbf{x}_c$$

$$\Delta_d(s) = s^4 + d_1s^3 + d_2s^2 + d_3s + d_4$$

In Summary – 1

$$\dot{\mathbf{x}} = \mathbf{A}\mathbf{x} + \mathbf{b}u$$

$$u = r - \mathbf{k}\mathbf{x}$$

$$\iff \mathbf{x}_c = \mathbf{P}\mathbf{x} \quad \mathbf{P}^{-1} = \begin{bmatrix} \mathbf{b} & \mathbf{A}\mathbf{b} & \mathbf{A}^2\mathbf{b} & \mathbf{A}^3\mathbf{b} \end{bmatrix} \begin{bmatrix} 1 & a_1 & a_2 & a_3 \\ 0 & 1 & a_1 & a_2 \\ 0 & 0 & 1 & a_1 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

$$\dot{\mathbf{x}}_c = \mathbf{A}_c\mathbf{x}_c + \mathbf{b}_cu = \begin{bmatrix} -a_1 & -a_2 & -a_3 & -a_4 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \mathbf{x}_c + \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} u$$

$$\Delta(s) = s^4 + a_1s^3 + a_2s^2 + a_3s + a_4$$

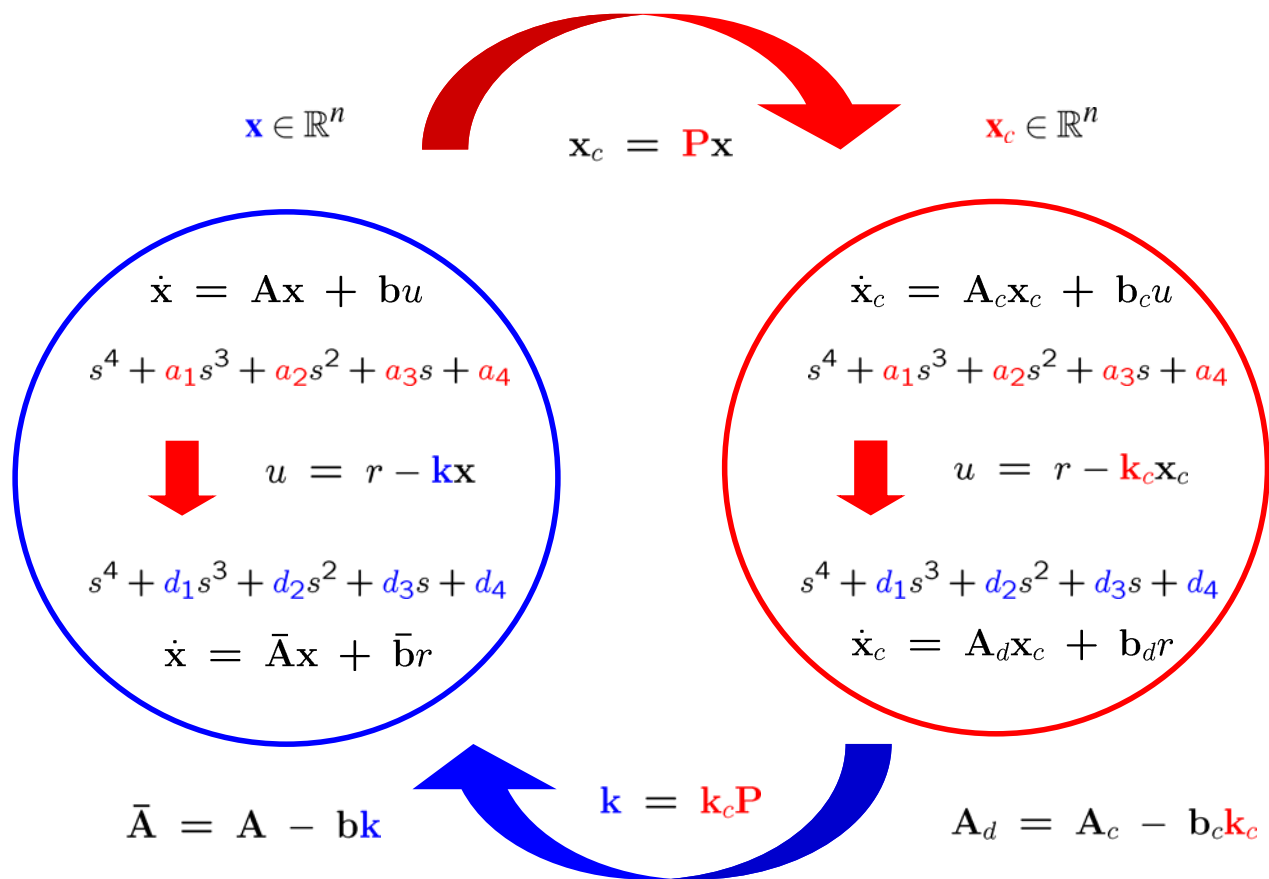
$$\iff u = r - \mathbf{k}_c\mathbf{x}_c$$

$$\dot{\mathbf{x}}_c = \mathbf{A}_d\mathbf{x}_c + \mathbf{b}_du = \begin{bmatrix} -d_1 & -d_2 & -d_3 & -d_4 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \mathbf{x}_c + \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} r$$

$$\Delta_d(s) = s^4 + d_1s^3 + d_2s^2 + d_3s + d_4$$

$$\iff u = r - \mathbf{k}_c\mathbf{x}_c = r - \mathbf{k}_c\mathbf{P}\mathbf{x} = r - \mathbf{k}\mathbf{x}$$

$$\mathbf{k} = \mathbf{k}_c\mathbf{P}$$

**Theorem 8.2 (8.2)****Theorem 8.2**

Consider the state equation $(\mathbf{A}, \mathbf{b}, \mathbf{c})$ with $n = 4$ and the characteristic polynomial

$$\Delta(s) = \det(s\mathbf{I} - \mathbf{A}) = s^4 + \alpha_1s^3 + \alpha_2s^2 + \alpha_3s + \alpha_4$$

If (\mathbf{A}, \mathbf{b}) is controllable, then it can be transformed by the transformation $\bar{\mathbf{x}} = \mathbf{P}\mathbf{x}$ with

$$\mathbf{Q} := \mathbf{P}^{-1} = [\mathbf{b} \ \mathbf{A}\mathbf{b} \ \mathbf{A}^2\mathbf{b} \ \mathbf{A}^3\mathbf{b}] \begin{bmatrix} 1 & \alpha_1 & \alpha_2 & \alpha_3 \\ 0 & 1 & \alpha_1 & \alpha_2 \\ 0 & 0 & 1 & \alpha_1 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

into the controllable canonical form

$$\dot{\bar{\mathbf{x}}} = \bar{\mathbf{A}}\bar{\mathbf{x}} + \bar{\mathbf{b}}u = \begin{bmatrix} -\alpha_1 & -\alpha_2 & -\alpha_3 & -\alpha_4 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \bar{\mathbf{x}} + \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} u$$

$$y = \bar{\mathbf{c}}\bar{\mathbf{x}} = [\beta_1 \ \beta_2 \ \beta_3 \ \beta_4]\bar{\mathbf{x}}$$

Furthermore, the transfer function of $(\mathbf{A}, \mathbf{b}, \mathbf{c})$ with $n = 4$ equals

$$\hat{g}(s) = \frac{\beta_1s^3 + \beta_2s^2 + \beta_3s + \beta_4}{s^4 + \alpha_1s^3 + \alpha_2s^2 + \alpha_3s + \alpha_4}$$

Proof:

If with the transformation \mathbf{Q} , the state equation $(\mathbf{A}, \mathbf{b}, \mathbf{c})$ is transformed into $(\bar{\mathbf{A}}, \bar{\mathbf{b}}, \bar{\mathbf{c}})$, then $\hat{g}(s)$ is as shown.

$$\text{Let } \mathbf{Q}_1 = [\mathbf{b} \quad \mathbf{A}\mathbf{b} \quad \mathbf{A}^2\mathbf{b} \quad \mathbf{A}^3\mathbf{b}],$$

$$\mathbf{Q}_2 = \begin{bmatrix} 1 & \alpha_1 & \alpha_2 & \alpha_3 \\ 0 & 1 & \alpha_1 & \alpha_2 \\ 0 & 0 & 1 & \alpha_1 \\ 0 & 0 & 0 & 1 \end{bmatrix},$$

$$\mathbf{x} = \mathbf{Q}_1 \tilde{\mathbf{x}} = \mathbf{Q}_1 \mathbf{Q}_2 \bar{\mathbf{x}} = \mathbf{Q} \bar{\mathbf{x}}$$

$$\text{Then } \tilde{\mathbf{A}} = \mathbf{Q}_1^{-1} \mathbf{A} \mathbf{Q}_1 = \begin{bmatrix} 0 & 0 & 0 & -\alpha_4 \\ 1 & 0 & 0 & -\alpha_3 \\ 0 & 1 & 0 & -\alpha_2 \\ 0 & 0 & 1 & -\alpha_1 \end{bmatrix}, \quad \tilde{\mathbf{b}} = \mathbf{Q}_1^{-1} \mathbf{b} = \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix},$$

$$\tilde{\mathbf{c}} = \mathbf{c} \mathbf{Q}_1 = [\mathbf{c}\mathbf{b} \quad \mathbf{c}\mathbf{A}\mathbf{b} \quad \mathbf{c}\mathbf{A}^2\mathbf{b} \quad \mathbf{c}\mathbf{A}^3\mathbf{b}]$$

Theorem 8.2 – 3

$$\text{And } \mathbf{Q}_2 \bar{\mathbf{A}} = \tilde{\mathbf{A}} \mathbf{Q}_2 = \begin{bmatrix} 0 & 0 & 0 & -\alpha_4 \\ 1 & \alpha_1 & \alpha_2 & 0 \\ 0 & 1 & \alpha_1 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix}$$

$$\Rightarrow \bar{\mathbf{A}} = \begin{bmatrix} -\alpha_1 & -\alpha_2 & -\alpha_3 & -\alpha_4 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix}$$

$$\mathbf{Q}_2 \bar{\mathbf{b}} = \tilde{\mathbf{b}} = [1 \ 0 \ 0 \ 0]'$$

$$\Rightarrow \bar{\mathbf{b}} = [1 \ 0 \ 0 \ 0]', \quad \bar{\mathbf{c}} = \tilde{\mathbf{c}} \mathbf{Q}_2 = [\beta_1 \ \beta_2 \ \beta_3 \ \beta_4]$$

Note: (1) $\beta_1 = \mathbf{c} \mathbf{b}$, $\beta_2 = \alpha_1 \mathbf{c} \mathbf{b} + \mathbf{c} \mathbf{A} \mathbf{b}$, ..., but not so important here

(2) $\mathbf{Q}_1 = \mathbf{e}$, $\mathbf{Q}_2 = \bar{\mathbf{e}}^{-1}$.

Theorem 8.3

If the n -dimensional state equation $(\mathbf{A}, \mathbf{b}, \mathbf{c})$ is controllable, then by state feedback $u = r - \mathbf{k}\mathbf{x}$, where \mathbf{k} is a $1 \times n$ real constant vector, the eigenvalues of $\mathbf{A} - \mathbf{b}\mathbf{k}$ can arbitrarily be assigned provided that complex conjugate eigenvalues are assigned in pairs.

Proof:

$$\dot{\bar{\mathbf{x}}} = \bar{\mathbf{A}} \bar{\mathbf{x}} + \bar{\mathbf{b}} u \quad \longleftrightarrow \quad \dot{\mathbf{x}} = \mathbf{A} \mathbf{x} + \mathbf{b} u$$

$$u = r - \mathbf{k} \mathbf{x} =$$

$$\bar{\mathbf{A}} - \bar{\mathbf{b}} \bar{\mathbf{k}} =$$

Theorem 8.3 – 2

Choose \mathbf{P} such that

$$\dot{\bar{\mathbf{x}}} = \bar{\mathbf{A}} \bar{\mathbf{x}} + \bar{\mathbf{b}} u = \begin{bmatrix} -\alpha_1 & -\alpha_2 & -\alpha_3 & -\alpha_4 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \bar{\mathbf{x}} + \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} u$$

Use the state feedback gain

$$\bar{\mathbf{k}} = [\bar{\alpha}_1 - \alpha_1 \quad \bar{\alpha}_2 - \alpha_2 \quad \bar{\alpha}_3 - \alpha_3 \quad \bar{\alpha}_4 - \alpha_4]$$

Can obtain any desired characteristic polynomial after state feedback

$$\Delta_f(s) = s^4 + \bar{\alpha}_1 s^3 + \bar{\alpha}_2 s^2 + \bar{\alpha}_3 s + \bar{\alpha}_4$$

And get

$$\dot{\bar{\mathbf{x}}} = (\bar{\mathbf{A}} - \bar{\mathbf{b}}\bar{\mathbf{k}})\bar{\mathbf{x}} + \bar{\mathbf{b}}r = \begin{bmatrix} -\bar{\alpha}_1 & -\bar{\alpha}_2 & -\bar{\alpha}_3 & -\bar{\alpha}_4 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \bar{\mathbf{x}} + \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} r$$

Thus $\mathbf{k} = \bar{\mathbf{k}}\mathbf{P} = \bar{\mathbf{k}}\bar{\mathbf{C}}\mathbf{C}^{-1}$

Double check:

$$\begin{aligned} \Delta_f(s) &= \det(s\mathbf{I} - \mathbf{A} + \mathbf{b}\mathbf{k}) = \det((s\mathbf{I} - \mathbf{A})[\mathbf{I} + (s\mathbf{I} - \mathbf{A})^{-1}\mathbf{b}\mathbf{k}]) \\ &= \det(s\mathbf{I} - \mathbf{A})\det[\mathbf{I} + (s\mathbf{I} - \mathbf{A})^{-1}\mathbf{b}\mathbf{k}] \\ &= \Delta(s)[1 + \mathbf{k}(s\mathbf{I} - \mathbf{A})^{-1}\mathbf{b}] \end{aligned}$$

$$\Delta_f(s) - \Delta(s) = \Delta(s)\mathbf{k}(s\mathbf{I} - \mathbf{A})^{-1}\mathbf{b} = \Delta(s)\underbrace{\bar{\mathbf{k}}(s\mathbf{I} - \bar{\mathbf{A}})^{-1}\bar{\mathbf{b}}}_{\frac{\bar{k}_1s^3 + \bar{k}_2s^2 + \bar{k}_3s + \bar{k}_4}{\Delta(s)}} = \bar{k}_1s^3 + \bar{k}_2s^2 + \bar{k}_3s + \bar{k}_4$$

- Zeros are not affected by state feedback

- Before state feedback: $\dot{\bar{\mathbf{x}}} = \bar{\mathbf{A}}\bar{\mathbf{x}} + \bar{\mathbf{b}}u = \begin{bmatrix} -\alpha_1 & -\alpha_2 & -\alpha_3 & -\alpha_4 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \bar{\mathbf{x}} + \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} u$

$$y = \bar{\mathbf{c}}\bar{\mathbf{x}} = [\beta_1 \ \beta_2 \ \beta_3 \ \beta_4]\bar{\mathbf{x}}$$

$$\hat{g}(s) = \mathbf{c}(s\mathbf{I} - \mathbf{A})^{-1}\mathbf{b} = \frac{\beta_1s^3 + \beta_2s^2 + \beta_3s + \beta_4}{s^4 + \alpha_1s^3 + \alpha_2s^2 + \alpha_3s + \alpha_4}$$

- After state feedback: $\dot{\bar{\mathbf{x}}} = (\bar{\mathbf{A}} - \bar{\mathbf{b}}\bar{\mathbf{k}})\bar{\mathbf{x}} + \bar{\mathbf{b}}r = \begin{bmatrix} -\bar{\alpha}_1 & -\bar{\alpha}_2 & -\bar{\alpha}_3 & -\bar{\alpha}_4 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \bar{\mathbf{x}} + \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} r$

$$y = [\beta_1 \ \beta_2 \ \beta_3 \ \beta_4]\bar{\mathbf{x}}$$

$$\hat{g}_f(s) = \bar{\mathbf{c}}(s\mathbf{I} - \bar{\mathbf{A}} + \bar{\mathbf{b}}\bar{\mathbf{k}})^{-1}\bar{\mathbf{b}} = \frac{\beta_1s^3 + \beta_2s^2 + \beta_3s + \beta_4}{s^4 + \bar{\alpha}_1s^3 + \bar{\alpha}_2s^2 + \bar{\alpha}_3s + \bar{\alpha}_4}$$

- Observability may change with possible new pole-zero relation

$$\dot{\mathbf{x}} = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & -1 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 5 & 0 \end{bmatrix} \mathbf{x} + \begin{bmatrix} 0 \\ 1 \\ 0 \\ -2 \end{bmatrix} u \quad \Delta(s) = s^2(s^2 - 5) = s^4 + 0 \cdot s^3 - 5s^2 + 0 \cdot s + 0$$

$$y = [1 \ 0 \ 0 \ 0] \mathbf{x}$$

$$\mathbf{P}^{-1} = C\bar{C}^{-1} = \begin{bmatrix} 0 & 1 & 0 & 2 \\ 1 & 0 & 2 & 0 \\ 0 & -2 & 0 & -10 \\ -2 & 0 & -10 & 0 \end{bmatrix} \begin{bmatrix} 1 & 0 & -5 & 0 \\ 0 & 1 & 0 & -5 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 & -3 \\ 1 & 0 & -3 & 0 \\ 0 & -2 & 0 & 0 \\ -2 & 0 & 0 & 0 \end{bmatrix}$$

If the desired eigenvalues are $-1.5 \pm 0.5j$ and $-1 \pm j$, then

$$\begin{aligned} \Delta_f(s) &= (s + 1.5 - 0.5j)(s + 1.5 + 0.5j)(s + 1 - j)(s + 1 + j) \\ &= s^4 + 5s^3 + 10.5s^2 + 11s + 5 \end{aligned}$$

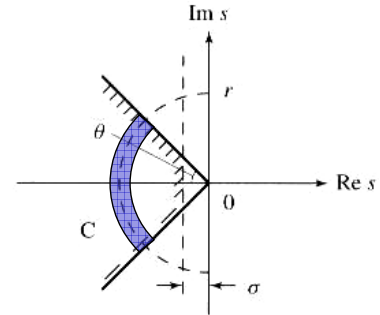
$$\bar{\mathbf{k}} = [5 - 0 \ 10.5 + 5 \ 11 - 0 \ 5 - 0] = [5 \ 15.5 \ 11 \ 5]$$

$$\mathbf{k} = \bar{\mathbf{k}}\mathbf{P} = [-\frac{5}{3} \quad -\frac{11}{3} \quad -\frac{103}{12} \quad -\frac{13}{3}]$$

- Guidelines for selecting “desired eigenvalues”
 - More negative real parts of eigenvalues
 - ➔ Faster response $y(t)$,
 - Larger system bandwidth (noise problem), and
 - Larger input $u(t)$ (actuator saturation problem)
 - Clustered eigenvalues
 - ➔ Response $y(t)$ sensitive to parameter change, and
 - Larger input $u(t)$

- Suggested methods:

- (1) Let the eigenvalues **uniformly spread** over the crescent region



- (2) Find **k** to minimize the “**performance index**” **J**
— **Optimal Control**

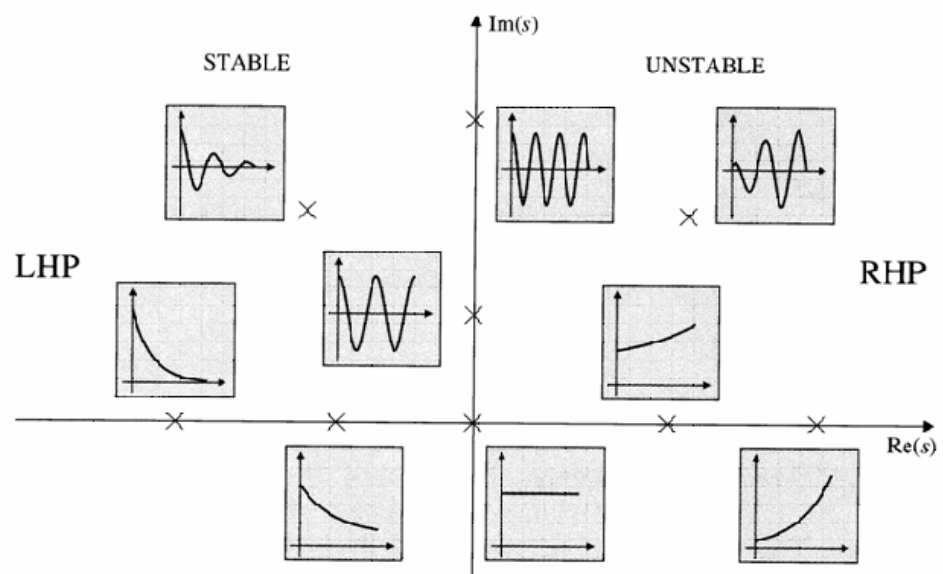
$$J = \int_0^{\infty} [\mathbf{x}'(t)\mathbf{Q}\mathbf{x}(t) + \mathbf{u}'(t)\mathbf{R}\mathbf{u}(t)] dt$$

weighting matrices

Design Guidelines – 3

Figure 3.20

Time functions associated with points in the s -plane (LHP, left half-plane; RHP, right half-plane)



State Feedback of Discrete-Time Systems (SISO Systems) (8.2)

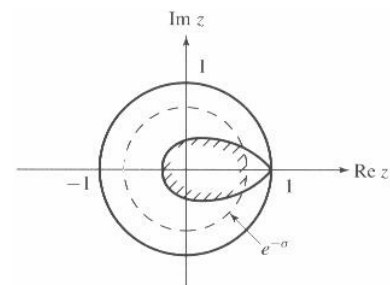
$$\mathbf{x}[k+1] = \mathbf{A} \mathbf{x}[k] + \mathbf{b} u[k]$$

$$u[k] = r[k] - \mathbf{k} \mathbf{x}[k]$$

$$\Rightarrow \mathbf{x}[k+1] = (\mathbf{A} - \mathbf{b}\mathbf{k}) \mathbf{x}[k] + \mathbf{b} r[k]$$

Thus **everything is the same**,
except that the **desired eigenvalues**
are different from the C.T. case.

For example,
the **hatched area** in the figure
is suggested.



Eigenvalue Assignment by Solving the Lyapunov Equation (8.2.1)

- A different method of computing state feedback gain for eigenvalue assignment
- Restriction: Different sets of eigenvalues wrt $\text{eig}(\mathbf{A})$

Procedure 8.1

Consider controllable (\mathbf{A}, \mathbf{b}) , where \mathbf{A} is $n \times n$ and \mathbf{b} is $n \times 1$. Find a $1 \times n$ real \mathbf{k} such that $(\mathbf{A} - \mathbf{b}\mathbf{k})$ has any set of desired eigenvalues that contains no eigenvalues of \mathbf{A} .

1. Select an $n \times n$ matrix \mathbf{F} that has the set of desired eigenvalues. The form of \mathbf{F} can be chosen arbitrarily and will be discussed later.
2. Select an arbitrary $1 \times n$ vector $\bar{\mathbf{k}}$ such that $(\mathbf{F}, \bar{\mathbf{k}})$ is observable.
3. Solve the unique \mathbf{T} in the Lyapunov equation $\mathbf{A}\mathbf{T} - \mathbf{T}\mathbf{F} = \mathbf{b}\bar{\mathbf{k}}$.
4. Compute the feedback gain $\mathbf{k} = \bar{\mathbf{k}}\mathbf{T}^{-1}$.

Note that:

In the procedure if a solution **T** exists and is nonsingular,
Then from steps 3 and 4, we have

$$(\mathbf{A} - \mathbf{b}\mathbf{k})\mathbf{T} = \mathbf{T}\mathbf{F} \quad \text{or} \quad \mathbf{A} - \mathbf{b}\mathbf{k} = \mathbf{T}\mathbf{F}\mathbf{T}^{-1}$$

Thus **A-bk** and **F** have the same (assigned) eigenvalues.

A sufficient condition for **T** to exist is that

A and **F** have no common eigenvalues

(i.e., every eigenvalue must move).

Theorem 8.4 (8.2.1)

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Theorem 8.4

If **A** and **F** have no eigenvalues in common, then the unique solution **T** of $\mathbf{A}\mathbf{T} - \mathbf{T}\mathbf{F} = \mathbf{b}\bar{\mathbf{k}}$ is nonsingular if and only if **(A, b)** is controllable and **(F, k̄)** is observable.

Proof: (for $n = 4$ only)

Let the characteristic polynomial of **A** be

$$\Delta(s) = s^4 + \alpha_1 s^3 + \alpha_2 s^2 + \alpha_3 s + \alpha_4$$

Then
$$\Delta(\mathbf{A}) = \mathbf{A}^4 + \alpha_1 \mathbf{A}^3 + \alpha_2 \mathbf{A}^2 + \alpha_3 \mathbf{A} + \alpha_4 \mathbf{I} = \mathbf{0}$$

And $\Delta(\bar{\lambda}_i) \neq 0$ for all eigenvalues λ_i of **F**

Thus
$$\Delta(\mathbf{F}) := \mathbf{F}^4 + \alpha_1 \mathbf{F}^3 + \alpha_2 \mathbf{F}^2 + \alpha_3 \mathbf{F} + \alpha_4 \mathbf{I}$$

has nonzero eigenvalues and is nonsingular

Also, observe the recursive relation

$$\mathbf{I} \mathbf{T} - \mathbf{T} \mathbf{I} = 0 \quad \times$$

$$\mathbf{A} \mathbf{T} - \mathbf{T} \mathbf{F} = \mathbf{b} \bar{\mathbf{k}} \quad \times$$

$$\mathbf{A} \mathbf{T} - \mathbf{T} \mathbf{F} = \mathbf{b} \bar{\mathbf{k}}$$

$$\mathbf{A} \mathbf{T} - \mathbf{T} \mathbf{F} = \mathbf{b} \bar{\mathbf{k}}$$

$$\mathbf{A}(\mathbf{A} \mathbf{T}) - (\mathbf{T} \mathbf{F}) \mathbf{F} = \mathbf{A}^2 \mathbf{T} - \mathbf{T} \mathbf{F}^2 = \quad \times$$

$$\mathbf{A}(\mathbf{A}^2 \mathbf{T}) - (\mathbf{T} \mathbf{F}^2) \mathbf{F} = \mathbf{A}^3 \mathbf{T} - \mathbf{T} \mathbf{F}^3 = \quad \times$$

$$\mathbf{A}(\mathbf{A}^3 \mathbf{T}) - (\mathbf{T} \mathbf{F}^3) \mathbf{F} = \mathbf{A}^4 \mathbf{T} - \mathbf{T} \mathbf{F}^4 = \quad (+)$$

$$\Delta(\mathbf{A})\mathbf{T} - \mathbf{T}\Delta(\mathbf{F}) = [\mathbf{b} \ \mathbf{A}\mathbf{b} \ \mathbf{A}^2\mathbf{b} \ \mathbf{A}^3\mathbf{b}] \begin{bmatrix} \alpha_3 & \alpha_2 & \alpha_1 & 1 \\ \alpha_2 & \alpha_1 & 1 & 0 \\ \alpha_1 & 1 & 0 & 0 \\ 1 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} \bar{\mathbf{k}} \\ \bar{\mathbf{k}}\mathbf{F} \\ \bar{\mathbf{k}}\mathbf{F}^2 \\ \bar{\mathbf{k}}\mathbf{F}^3 \end{bmatrix}$$

$\therefore \mathbf{T}$ is

Gain Selection (8.2.1)

Given a set of desired eigenvalues, how to set \mathbf{F} and $\bar{\mathbf{k}}$?

(1) Using the observable canonical form,

set \mathbf{F} to the companion form

with the desired eigenvalues, and let $\bar{\mathbf{k}} = [1 \ 0 \ \cdots \ 0]$;

$$\begin{bmatrix} -\alpha_1 & 1 & 0 & 0 \\ -\alpha_2 & 0 & 1 & 0 \\ -\alpha_3 & 0 & 0 & 1 \\ -\alpha_4 & 0 & 0 & 0 \end{bmatrix}$$

(2) Using the modal form,

set \mathbf{F} to the diagonal form

with the desired eigenvalues, and

Let $\bar{\mathbf{k}}$ have at least one nonzero element

associated with each diagonal block of \mathbf{F} ,

such as $\bar{\mathbf{k}} = [1 \ 1 \ 0 \ 1 \ 0], [1 \ 1 \ 0 \ 0 \ 1], [1 \ 1 \ 1 \ 1 \ 1]$.

$$\begin{bmatrix} \lambda_1 & 0 & 0 & 0 & 0 \\ 0 & \alpha_1 & \beta_1 & 0 & 0 \\ 0 & -\beta_1 & \alpha_1 & 0 & 0 \\ 0 & 0 & 0 & \alpha_2 & \beta_2 \\ 0 & 0 & 0 & -\beta_2 & \alpha_2 \end{bmatrix}$$

(the same as Example 8.3)

$$\dot{\mathbf{x}} = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & -1 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 5 & 0 \end{bmatrix} \mathbf{x} + \begin{bmatrix} 0 \\ 1 \\ 0 \\ -2 \end{bmatrix} u$$

$$y = [1 \ 0 \ 0 \ 0] \mathbf{x}$$

The desired eigenvalues are $-1.5 \pm 0.5j$ and $-1 \pm j$, thus set

$$\mathbf{F} = \begin{bmatrix} -1 & 1 & 0 & 0 \\ -1 & -1 & 0 & 0 \\ 0 & 0 & -1.5 & 0.5 \\ 0 & 0 & -0.5 & -1.5 \end{bmatrix}$$

No matter $\bar{\mathbf{k}}$ is set to $[1 \ 0 \ 1 \ 0]$ or $[1 \ 1 \ 1 \ 1]$,
the resulting \mathbf{k} is the same as the one obtained in Example 8.3
(different $\bar{\mathbf{k}}$, different \mathbf{T} , but the same $\bar{\mathbf{k}}\mathbf{T}^{-1}$).

Theorem 8.4 – 3

For single-input systems,

the state feedback gain \mathbf{k} corresponding to

a set of pre-assigned eigenvalues is unique.

$$\begin{cases} \dot{\mathbf{x}} = \mathbf{A}\mathbf{x} + \begin{bmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \end{bmatrix} u \\ y = [c_1 \ c_2 \ c_3 \ c_4] \mathbf{x} \end{cases}$$

$$s^4 + a_1 s^3 + a_2 s^2 + a_3 s + a_4$$

$$\lambda_1, \lambda_2, \lambda_3, \lambda_4,$$

$$u = r - [k_1 \ k_2 \ k_3 \ k_4] \mathbf{x}$$

$$\left(\mathbf{A} - \begin{bmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \end{bmatrix} [k_1 \ k_2 \ k_3 \ k_4] \right)$$

$$s^4 + d_1 s^3 + d_2 s^2 + d_3 s + d_4$$

$$\lambda_1^c, \lambda_2^c, \lambda_3^c, \lambda_4^c,$$

$$s^4 + a_1 s^3 + a_2 s^2 + a_3 s + a_4$$

$$\lambda_1, \lambda_2, \lambda_3, \lambda_4,$$

$$\begin{cases} \dot{\mathbf{x}} = \mathbf{A}\mathbf{x} + \mathbf{b}u \\ y = \mathbf{c}\mathbf{x} \end{cases}$$

$$\begin{cases} \dot{\mathbf{x}} = \mathbf{A}\mathbf{x} + \begin{bmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \end{bmatrix} u \\ y = \begin{bmatrix} c_1 & c_2 & c_3 & c_4 \end{bmatrix} \mathbf{x} \end{cases}$$

$$\dot{\hat{\mathbf{x}}} = (\mathbf{A} - \mathbf{l}\mathbf{c})\hat{\mathbf{x}} + \mathbf{b}u + \mathbf{l}y$$

$$s^4 + c_1 s^3 + c_2 s^2 + c_3 s + c_4$$

$$\lambda_1^o, \lambda_2^o, \lambda_3^o, \lambda_4^o,$$

$$\left(\mathbf{A} - \begin{bmatrix} l_1 \\ l_2 \\ l_3 \\ l_4 \end{bmatrix} \begin{bmatrix} c_1 & c_2 & c_3 & c_4 \end{bmatrix} \right)$$

$$u = r - \mathbf{k} \hat{\mathbf{x}}$$

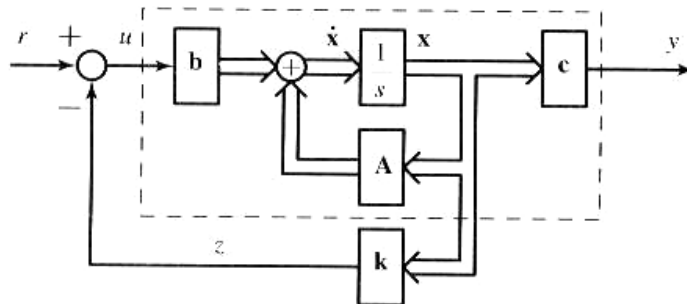
$$(\mathbf{A} - \mathbf{b}\mathbf{k})$$

$$\left(\mathbf{A} - \begin{bmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \end{bmatrix} \begin{bmatrix} k_1 & k_2 & k_3 & k_4 \end{bmatrix} \right)$$

$$s^4 + b_1 s^3 + b_2 s^2 + b_3 s + b_4$$

$$\lambda_1^c, \lambda_2^c, \lambda_3^c, \lambda_4^c,$$

Regulation and Tracking (8.3)



$$\begin{cases} \dot{\mathbf{x}}(t) = \mathbf{A}\mathbf{x}(t) + \mathbf{b}u(t) \\ y(t) = \mathbf{c}\mathbf{x}(t) \end{cases}$$

Regulation:

$$y(t) \text{ to follow } r(t) = 0, \forall t \geq 0$$

Tracking:

$$y(t) \text{ to follow } r(t) = a, \text{ a constant, } \forall t \geq 0$$

Servomechanism problem:

$$y(t) \text{ to follow } r(t), \text{ a non-constant reference signal, } \forall t \geq 0$$

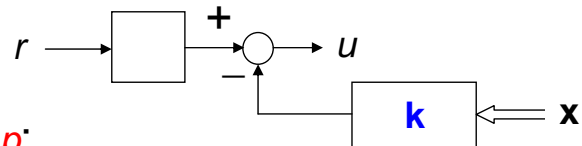
To solve the regulation problem is to find a state feedback gain \mathbf{k} such that $\mathbf{A} - \mathbf{b}\mathbf{k}$ is a stable matrix,

$$r(t) = 0$$

$$u(t) = -\mathbf{k} \mathbf{x}(t)$$

$$y(t) = \mathbf{c} e^{(\mathbf{A}-\mathbf{b}\mathbf{k})t} \mathbf{x}(0)$$

To solve the tracking problem, we need an extra feedforward gain p :



$$r(t) = a$$

$$u(t) = p r(t) - \mathbf{k} \mathbf{x}(t)$$

$$y(t) = \mathbf{c} \left(e^{(\mathbf{A}-\mathbf{b}\mathbf{k})t} \mathbf{x}(0) + \int_0^t e^{(\mathbf{A}-\mathbf{b}\mathbf{k})(t-\tau)} \mathbf{b} p r(\tau) d\tau \right)$$

With the feedforward gain and state feedback, the transfer function is

$$\hat{g}_f(s) = \frac{\hat{y}(s)}{\hat{r}(s)} = p \frac{\beta_1 s^3 + \beta_2 s^2 + \beta_3 s + \beta_4}{s^4 + \bar{\alpha}_1 s^3 + \bar{\alpha}_2 s^2 + \bar{\alpha}_3 s + \bar{\alpha}_4}$$

Thus for $y(t)$ to follow $r(t) = a$, a constant, choose \mathbf{k} such that $\mathbf{A} - \mathbf{b}\mathbf{k}$ is stable, and p such that

$$1 = \hat{g}_f(0) = p \frac{\beta_4}{\bar{\alpha}_4} \quad \text{or} \quad p = \frac{\bar{\alpha}_4}{\beta_4}$$

Which requires $\beta_4 \neq 0$, i.e., the plant has no zeros at $s = 0$

In practical applications, there will be **disturbance**,
i.e., exogenous input affecting the system, and
there will be **uncertainty** regarding the exact values
of **A**, **b**, and **c**.

Robust Tracking:

Tracking with **parameter uncertainty**
(β_4 and $\bar{\alpha}_4$ not known exactly)

Disturbance Rejection:

Elimination of the effect by **disturbance**

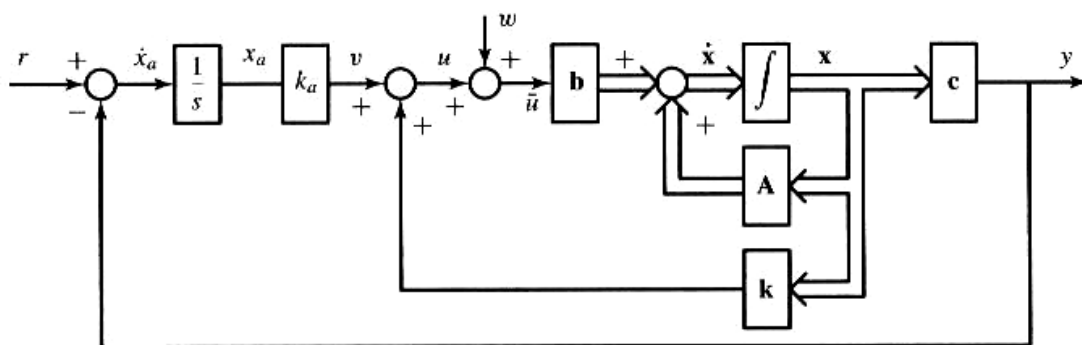
Robust Tracking and Disturbance Rejection – 2

Plant nominal equation:

$$\dot{\mathbf{x}} = \mathbf{A}\mathbf{x} + \mathbf{b}u + \mathbf{b}w$$

$$y = \mathbf{c}\mathbf{x}$$

← **constant disturbance**



An **internal model** control configuration with state feedback

$$\dot{x}_a = r - y = r - \mathbf{c}\mathbf{x}$$

$$u = [\mathbf{k} \ k_a] \begin{bmatrix} \mathbf{x} \\ x_a \end{bmatrix}$$



$$\begin{bmatrix} \dot{\mathbf{x}} \\ \dot{x}_a \end{bmatrix} = \begin{bmatrix} \mathbf{A} + \mathbf{b}\mathbf{k} & \mathbf{b}k_a \\ -\mathbf{c} & 0 \end{bmatrix} \begin{bmatrix} \mathbf{x} \\ x_a \end{bmatrix} + \begin{bmatrix} \mathbf{0} \\ 1 \end{bmatrix} r + \begin{bmatrix} \mathbf{b} \\ 0 \end{bmatrix} w$$

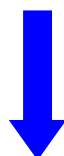
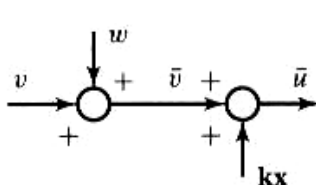
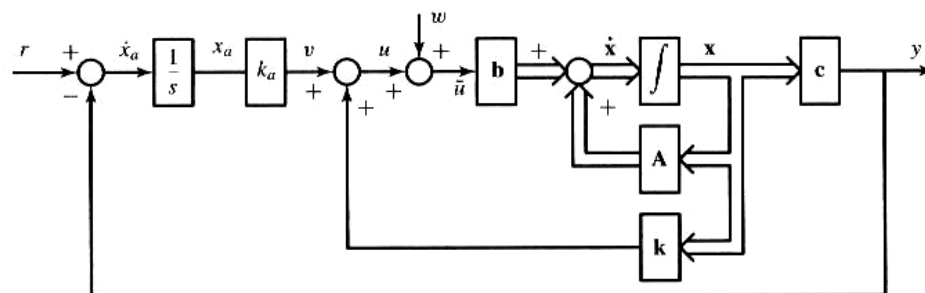
$$y = [\mathbf{c} \ 0] \begin{bmatrix} \mathbf{x} \\ x_a \end{bmatrix}$$

Theorem 8.5

If (\mathbf{A}, \mathbf{b}) is controllable and if $\hat{g}(s) = \mathbf{c}(s\mathbf{I} - \mathbf{A})^{-1}\mathbf{b}$ has no zero at $s = 0$, then all eigenvalues of the closed-loop system can be assigned arbitrarily by selecting a feedback gain $[\mathbf{k} \quad k_d]$.

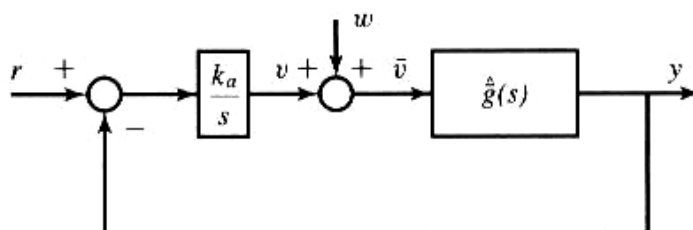
Proof:

Block diagram manipulation:



$$\frac{\bar{N}(s)}{\bar{D}(s)} := \mathbf{c}(s\mathbf{I} - \mathbf{A} - \mathbf{b}\mathbf{k})^{-1}\mathbf{b}$$

$$\bar{D}(s) = \det(s\mathbf{I} - \mathbf{A} - \mathbf{b}\mathbf{k})$$



Taking **determinants** of the identity

$$\begin{bmatrix} \mathbf{I} & \mathbf{0} \\ -\mathbf{c}(s\mathbf{I} - \mathbf{A} - \mathbf{b}\mathbf{k})^{-1} & 1 \end{bmatrix} \begin{bmatrix} s\mathbf{I} - \mathbf{A} - \mathbf{b}\mathbf{k} & -\mathbf{b}k_a \\ \mathbf{c} & s \end{bmatrix} \\ = \begin{bmatrix} s\mathbf{I} - \mathbf{A} - \mathbf{b}\mathbf{k} & -\mathbf{b}k_a \\ 0 & s + \mathbf{c}(s\mathbf{I} - \mathbf{A} - \mathbf{b}\mathbf{k})^{-1}\mathbf{b}k_a \end{bmatrix}$$

$$\rightarrow 1 \cdot \Delta_f(s) = \bar{D}(s) \left(s + \frac{\bar{N}(s)}{\bar{D}(s)} k_a \right) \quad \text{i.e.,} \quad \Delta_f(s) = s\bar{D}(s) + k_a\bar{N}(s)$$

$$\rightarrow \hat{g}_{yw}(s) = \frac{\frac{\bar{N}(s)}{\bar{D}(s)}}{1 + \frac{k_a\bar{N}(s)}{s\bar{D}(s)}} = \frac{s\bar{N}(s)}{s\bar{D}(s) + k_a\bar{N}(s)} = \frac{s\bar{N}(s)}{\Delta_f(s)}$$

For **constant** (step-type) **disturbance** $\hat{w}(s) = \bar{w}/s$

$$\hat{y}_w(s) = \frac{s\bar{N}(s)}{\Delta_f(s)} \frac{\bar{w}}{s} = \frac{\bar{w}\bar{N}(s)}{\Delta_f(s)}$$

\therefore Even with **(small) parameter uncertainty**,

$$\lim_{t \rightarrow \infty} y_w(t) = 0,$$

as long as **roots** of $\Delta_f(s)$ all have **negative real parts**

➔ Robust constant-disturbance rejection

Also,

$$\hat{g}_{yr}(s) = \frac{\frac{k_a \bar{N}(s)}{s \bar{D}(s)}}{1 + \frac{k_a \bar{N}(s)}{s \bar{D}(s)}} = \frac{k_a \bar{N}(s)}{s \bar{D}(s) + k_a \bar{N}(s)} = \frac{k_a \bar{N}(s)}{\Delta_f(s)}$$

Thus even with parameter uncertainty

$$\hat{g}_{yr}(0) = \frac{k_a \bar{N}(0)}{0 \cdot \bar{D}(0) + k_a \bar{N}(0)} = \frac{k_a \bar{N}(0)}{k_a \bar{N}(0)} = 1$$

➡ robust tracking for constant reference

Stabilization (8.3.2)

$$\begin{cases} \dot{\mathbf{x}}(t) = \mathbf{A}\mathbf{x}(t) + \mathbf{b}u(t) \\ y(t) = \mathbf{c}\mathbf{x}(t) \end{cases}$$

- Stable systems
- Stabilizable systems

- Without loss of generality, consider

$$\begin{bmatrix} \dot{\bar{\mathbf{x}}_c} \\ \dot{\bar{\mathbf{x}}_{\bar{c}}} \end{bmatrix} = \begin{bmatrix} \bar{\mathbf{A}}_c & \bar{\mathbf{A}}_{12} \\ \mathbf{0} & \bar{\mathbf{A}}_{\bar{c}} \end{bmatrix} \begin{bmatrix} \bar{\mathbf{x}}_c \\ \bar{\mathbf{x}}_{\bar{c}} \end{bmatrix} + \begin{bmatrix} \bar{\mathbf{b}}_c \\ \mathbf{0} \end{bmatrix} u$$

$$u = r - \mathbf{k}\mathbf{x} = r - \bar{\mathbf{k}}\bar{\mathbf{x}} = r - [\bar{\mathbf{k}}_1 \ \bar{\mathbf{k}}_2] \begin{bmatrix} \bar{\mathbf{x}}_c \\ \bar{\mathbf{x}}_{\bar{c}} \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} \dot{\bar{\mathbf{x}}_c} \\ \dot{\bar{\mathbf{x}}_{\bar{c}}} \end{bmatrix} = \begin{bmatrix} \bar{\mathbf{A}}_c - \bar{\mathbf{b}}_c \bar{\mathbf{k}}_1 & \bar{\mathbf{A}}_{12} - \bar{\mathbf{b}}_c \bar{\mathbf{k}}_2 \\ \mathbf{0} & \bar{\mathbf{A}}_{\bar{c}} \end{bmatrix} \begin{bmatrix} \bar{\mathbf{x}}_c \\ \bar{\mathbf{x}}_{\bar{c}} \end{bmatrix} + \begin{bmatrix} \bar{\mathbf{b}}_c \\ \mathbf{0} \end{bmatrix} r$$

\therefore Uncontrollable systems are stabilizable by state feedback if and only if the uncontrollable part of the system is stable

State Estimator (Observer) for SISO Systems (8.4)

$$\begin{cases} \dot{\mathbf{x}}(t) = \mathbf{A}\mathbf{x}(t) + \mathbf{b}u(t) \\ y(t) = \mathbf{c}\mathbf{x}(t) \end{cases}$$

Given \mathbf{A} , \mathbf{b} , \mathbf{c} ,

Can we estimate $\mathbf{x}(t)$

by measuring $y(\tau)$ and $u(\tau)$ for $\tau \in [0, t]$?

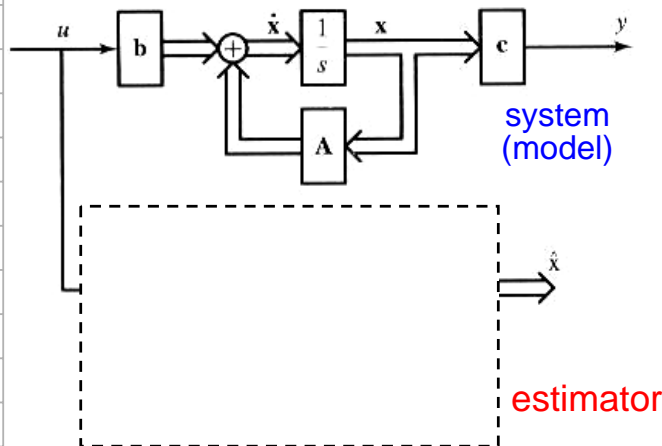
• Open-Loop Estimator:

$$\begin{cases} \dot{\mathbf{x}}(t) = \mathbf{A}\mathbf{x}(t) + \mathbf{b}u(t) \\ y(t) = \mathbf{c}\mathbf{x}(t) \end{cases}$$

and use $y(\tau)$ and $u(\tau)$ for $\tau \in [0, t]$

$$\dot{\hat{\mathbf{x}}}(t) = \mathbf{A}\hat{\mathbf{x}}(t) + \mathbf{b}u(t)$$

to compute $\hat{\mathbf{x}}(0) = \mathbf{x}(0)$



Disadvantage 1:

Any **inaccuracy** in the determination of $\mathbf{x}(0)$ will make $\lim_{t \rightarrow \infty} \|\hat{\mathbf{x}}(t) - \mathbf{x}(t)\| = \infty$ for **unstable A**

Disadvantage 2:

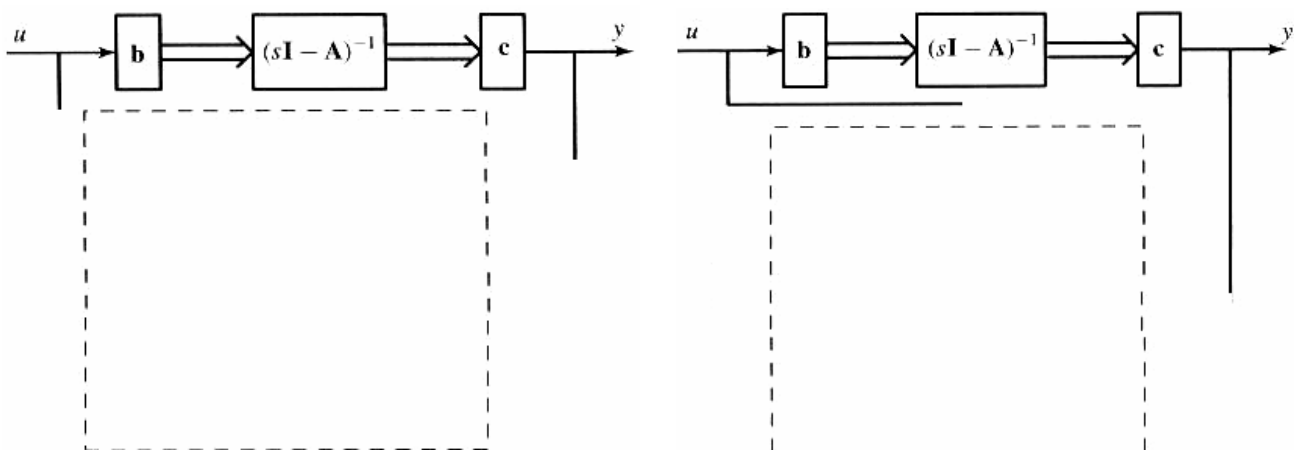
Very **sensitive** to **parameter uncertainty** in **A** and **b**

• Closed-Loop Estimator:

$$\begin{cases} \dot{\mathbf{x}}(t) = \mathbf{A}\mathbf{x}(t) + \mathbf{b}u(t) \\ y(t) = \mathbf{c}\mathbf{x}(t) \end{cases}$$

$$\dot{\hat{\mathbf{x}}}(t) = \mathbf{A}\hat{\mathbf{x}}(t) + \mathbf{b}u(t)$$

=



- Estimator Error:

$$\mathbf{e}(t) := \mathbf{x}(t) - \hat{\mathbf{x}}(t)$$

$$\begin{aligned} \rightarrow \dot{\mathbf{e}}(t) &= \dot{\mathbf{x}}(t) - \dot{\hat{\mathbf{x}}}(t) \\ &= \end{aligned}$$

Theorem 8.O3 (8.4)

Theorem 8.O3

Consider the pair (\mathbf{A}, \mathbf{c}) . All eigenvalues of $(\mathbf{A} - \mathbf{l}\mathbf{c})$ can be assigned arbitrarily by selecting a real constant vector \mathbf{l} if and only if (\mathbf{A}, \mathbf{c}) is observable.

Proof:

• A different method for designing state estimators:

Procedure 8.O1

1. Select an arbitrary $n \times n$ stable matrix \mathbf{F} that has no eigenvalues in common with those of \mathbf{A} .
2. Select an arbitrary $n \times 1$ vector \mathbf{l} such that (\mathbf{F}, \mathbf{l}) is controllable.
3. Solve the unique \mathbf{T} in the Lyapunov equation $\mathbf{T}\mathbf{A} - \mathbf{F}\mathbf{T} = \mathbf{l}\mathbf{c}$. This \mathbf{T} is nonsingular following the dual of Theorem 8.4.
4. Then the state equation

$$\begin{aligned}\dot{\mathbf{z}} &= \mathbf{F}\mathbf{z} + \mathbf{T}\mathbf{b}u + \mathbf{l}y \\ \hat{\mathbf{x}} &= \mathbf{T}^{-1}\mathbf{z}\end{aligned}$$

generates an estimate of \mathbf{x} .

$$\begin{aligned}\mathbf{e} &:= \mathbf{z} - \mathbf{T}\mathbf{x} \quad \Rightarrow \quad \dot{\mathbf{e}} = \dot{\mathbf{z}} - \mathbf{T}\dot{\mathbf{x}} \\ &= \mathbf{F}\mathbf{z} + \mathbf{T}\mathbf{b}u + \mathbf{l}\mathbf{c}\mathbf{x} - \mathbf{T}\mathbf{A}\mathbf{x} - \mathbf{T}\mathbf{b}u \\ &= \mathbf{F}\mathbf{z} + \mathbf{l}\mathbf{c}\mathbf{x} - (\mathbf{F}\mathbf{T} + \mathbf{l}\mathbf{c})\mathbf{x} \\ &= \mathbf{F}(\mathbf{z} - \mathbf{T}\mathbf{x}) \\ &= \mathbf{F}\mathbf{e}\end{aligned}$$

Reduced-Dimensional State Estimator (8.4.1)

The equation $y(t) = \mathbf{c}\mathbf{x}(t)$ already has one-dimensional information about $\mathbf{x}(t)$, so only an $(n-1)$ -dimensional state estimator is needed to estimate the remaining information about $\mathbf{x}(t)$

Procedure 8.R1

1. Select an arbitrary $(n-1) \times (n-1)$ stable matrix \mathbf{F} that has no eigenvalues in common with those of \mathbf{A} .
2. Select an arbitrary $(n-1) \times 1$ vector \mathbf{l} such that (\mathbf{F}, \mathbf{l}) is controllable.
3. Solve the unique \mathbf{T} in the Lyapunov equation $\mathbf{T}\mathbf{A} - \mathbf{F}\mathbf{T} = \mathbf{l}\mathbf{c}$. Note that \mathbf{T} is an $(n-1) \times n$ matrix.
4. Then the $(n-1)$ -dimensional state equation

$$\begin{aligned}\dot{\mathbf{z}} &= \mathbf{F}\mathbf{z} + \mathbf{T}\mathbf{b}u + \mathbf{l}y \\ \hat{\mathbf{x}} &= \begin{bmatrix} \mathbf{c} \\ \mathbf{T} \end{bmatrix}^{-1} \begin{bmatrix} y \\ \mathbf{z} \end{bmatrix}\end{aligned}$$

is an estimate of \mathbf{x} .

Theorem 8.6

If \mathbf{A} and \mathbf{F} have no common eigenvalues, then the square matrix

$$\mathbf{P} = \begin{bmatrix} \mathbf{c} \\ \mathbf{T} \end{bmatrix}$$

where \mathbf{T} is the unique solution of $\mathbf{TA} - \mathbf{FT} = \mathbf{lc}$, is nonsingular if and only if (\mathbf{A}, \mathbf{c}) is observable and (\mathbf{F}, \mathbf{l}) is controllable.

Theorem 8.6 – 2

Proof: (necessity for general n)

If (\mathbf{F}, \mathbf{l}) not controllable

If (\mathbf{A}, \mathbf{c}) not observable

Proof: (sufficiency for $n = 4$)As in **Theorem 8.4**,

Let $\Delta(s) = \det(s\mathbf{I} - \mathbf{A}) = s^4 + \alpha_1 s^3 + \alpha_2 s^2 + \alpha_3 s + \alpha_4$

Then

$$-\Delta(\mathbf{F})\mathbf{T} = [\mathbf{I} \ \mathbf{F} \ \mathbf{F}^2 \ \mathbf{F}^3] \begin{bmatrix} \alpha_3 & \alpha_2 & \alpha_1 & 1 \\ \alpha_2 & \alpha_1 & 1 & 0 \\ \alpha_1 & 1 & 0 & 0 \\ 1 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} \mathbf{c} \\ \mathbf{cA} \\ \mathbf{cA}^2 \\ \mathbf{cA}^3 \end{bmatrix}$$

$$\therefore \mathbf{T} = -\Delta^{-1}(\mathbf{F})\mathbf{C}_4\mathbf{\Lambda}\mathbf{O}$$

$$\mathbf{P} = \begin{bmatrix} \mathbf{c} \\ \mathbf{T} \end{bmatrix} = \begin{bmatrix} \mathbf{c} \\ -\Delta^{-1}(\mathbf{F})\mathbf{C}_4\mathbf{\Lambda}\mathbf{O} \end{bmatrix} = \begin{bmatrix} 1 & \mathbf{0} \\ \mathbf{0} & -\Delta^{-1}(\mathbf{F}) \end{bmatrix} \begin{bmatrix} \mathbf{c} \\ \mathbf{C}_4\mathbf{\Lambda}\mathbf{O} \end{bmatrix}$$

If there is a **nonzero** vector \mathbf{r} , such that $\mathbf{Pr} = \mathbf{0}$, i.e.,

$$\begin{bmatrix} \mathbf{c} \\ \mathbf{C}_4\mathbf{\Lambda}\mathbf{O} \end{bmatrix} \mathbf{r} = \begin{bmatrix} \mathbf{c}\mathbf{r} \\ \mathbf{C}_4\mathbf{\Lambda}\mathbf{O}\mathbf{r} \end{bmatrix} = \mathbf{0}$$

Then consider

$$\mathbf{a} := \mathbf{\Lambda}\mathbf{O}\mathbf{r} = \begin{bmatrix} a_1 \\ a_2 \\ a_3 \\ a_4 \end{bmatrix} = \begin{bmatrix} \alpha_3 & \alpha_2 & \alpha_1 & 1 \\ \alpha_2 & \alpha_1 & 1 & 0 \\ \alpha_1 & 1 & 0 & 0 \\ 1 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} \mathbf{c}\mathbf{r} \\ \mathbf{cA}\mathbf{r} \\ \mathbf{cA}^2\mathbf{r} \\ \mathbf{cA}^3\mathbf{r} \end{bmatrix} = \begin{bmatrix} x \\ x \\ x \\ \mathbf{c}\mathbf{r} \end{bmatrix}$$

$$\mathbf{C}_4\mathbf{\Lambda}\mathbf{O}\mathbf{r} = \mathbf{C}_4\mathbf{a} = \mathbf{C}\bar{\mathbf{a}} = \mathbf{0} \quad \Rightarrow \quad \bar{\mathbf{a}} = \mathbf{0} \quad \Rightarrow \quad \mathbf{a} = \mathbf{0} \quad \Rightarrow \quad \mathbf{r} = \mathbf{0}.$$

$$[\mathbf{I} \ \mathbf{F} \ \mathbf{F}^2 \ \mathbf{F}^3]$$

$$[\mathbf{I} \ \mathbf{F} \ \mathbf{F}^2]$$

 (\mathbf{F}, \mathbf{I}) controllable (\mathbf{A}, \mathbf{c}) observable
and $\mathbf{\Lambda}$ nonsingular

In the **reduced-dimensional** state estimator

$$\begin{cases} \dot{\mathbf{z}} = \mathbf{F}\mathbf{z} + \mathbf{T}\mathbf{b}u + \mathbf{l}y \\ \hat{\mathbf{x}} = \begin{bmatrix} \mathbf{c} \\ \mathbf{T} \end{bmatrix}^{-1} \begin{bmatrix} y \\ \mathbf{z} \end{bmatrix} \end{cases}$$

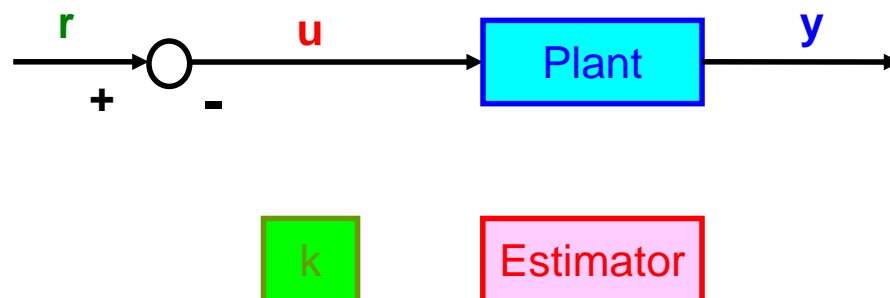
Define the **error signal** $\mathbf{e} = \mathbf{z} - \mathbf{T}\mathbf{x}$, then

$$\dot{\mathbf{e}} = \dot{\mathbf{z}} - \mathbf{T}\dot{\mathbf{x}} = \mathbf{F}\mathbf{z} + \mathbf{T}\mathbf{b}u + \mathbf{l}c\mathbf{x} - \mathbf{T}\mathbf{A}\mathbf{x} - \mathbf{T}\mathbf{b}u = \mathbf{F}\mathbf{e}$$

Thus $\lim_{t \rightarrow \infty} \mathbf{e}(t) = 0$ if \mathbf{F} is chosen to be **stable**, and

$$\lim_{t \rightarrow \infty} \hat{\mathbf{x}}(t) = \lim_{t \rightarrow \infty} \begin{bmatrix} \mathbf{c} \\ \mathbf{T} \end{bmatrix}^{-1} \begin{bmatrix} y(t) \\ \mathbf{z}(t) \end{bmatrix} = \begin{bmatrix} \mathbf{c} \\ \mathbf{T} \end{bmatrix}^{-1} \begin{bmatrix} \mathbf{c}\mathbf{x}(t) \\ \mathbf{T}\mathbf{x}(t) \end{bmatrix} = \begin{bmatrix} \mathbf{c} \\ \mathbf{T} \end{bmatrix}^{-1} \begin{bmatrix} \mathbf{c} \\ \mathbf{T} \end{bmatrix} \mathbf{x}(t) = \mathbf{x}(t)$$

Feedback from Estimated States (8.5)



$$\begin{cases} \dot{\mathbf{x}} = \mathbf{A}\mathbf{x} + \mathbf{b}u \\ y = \mathbf{c}\mathbf{x} \end{cases}$$

$$\dot{\hat{\mathbf{x}}} = (\mathbf{A} - \mathbf{l}\mathbf{c})\hat{\mathbf{x}} + \mathbf{b}u + \mathbf{l}y$$

$$u = r - \mathbf{k}\hat{\mathbf{x}}$$

$$\begin{cases} \dot{\mathbf{x}} = \mathbf{A}\mathbf{x} - \mathbf{b}\mathbf{k}\hat{\mathbf{x}} + \mathbf{b}r \\ \dot{\hat{\mathbf{x}}} = (\mathbf{A} - \mathbf{l}\mathbf{c})\hat{\mathbf{x}} + \mathbf{b}(r - \mathbf{k}\hat{\mathbf{x}}) + \mathbf{l}\mathbf{c}\mathbf{x} \end{cases}$$

$$\begin{bmatrix} \dot{\mathbf{x}} \\ \dot{\hat{\mathbf{x}}} \end{bmatrix} = \begin{bmatrix} \mathbf{A} & -\mathbf{b}\mathbf{k} \\ \mathbf{l}\mathbf{c} & \mathbf{A} - \mathbf{l}\mathbf{c} - \mathbf{b}\mathbf{k} \end{bmatrix} \begin{bmatrix} \mathbf{x} \\ \hat{\mathbf{x}} \end{bmatrix} + \begin{bmatrix} \mathbf{b} \\ \mathbf{b} \end{bmatrix} r$$

$$y = \begin{bmatrix} \mathbf{c} & \mathbf{0} \end{bmatrix} \begin{bmatrix} \mathbf{x} \\ \hat{\mathbf{x}} \end{bmatrix}$$

Consider the equivalence transformation

$$\begin{bmatrix} \mathbf{x} \\ \mathbf{e} \end{bmatrix} = \begin{bmatrix} \mathbf{x} \\ \mathbf{x} - \hat{\mathbf{x}} \end{bmatrix} = \underbrace{\begin{bmatrix} \mathbf{I} & \mathbf{0} \\ \mathbf{I} & -\mathbf{I} \end{bmatrix}}_{\mathbf{P} = \mathbf{P}^{-1}} \begin{bmatrix} \mathbf{x} \\ \hat{\mathbf{x}} \end{bmatrix}$$

We see

$$\begin{bmatrix} \dot{\mathbf{x}} \\ \dot{\mathbf{e}} \end{bmatrix} = \begin{bmatrix} \mathbf{A} - \mathbf{b}\mathbf{k} & \mathbf{b}\mathbf{k} \\ \mathbf{0} & \mathbf{A} - \mathbf{l}\mathbf{c} \end{bmatrix} \begin{bmatrix} \mathbf{x} \\ \mathbf{e} \end{bmatrix} + \begin{bmatrix} \mathbf{b} \\ \mathbf{0} \end{bmatrix} r$$

$$y = \begin{bmatrix} \mathbf{c} & \mathbf{0} \end{bmatrix} \begin{bmatrix} \mathbf{x} \\ \mathbf{e} \end{bmatrix}$$

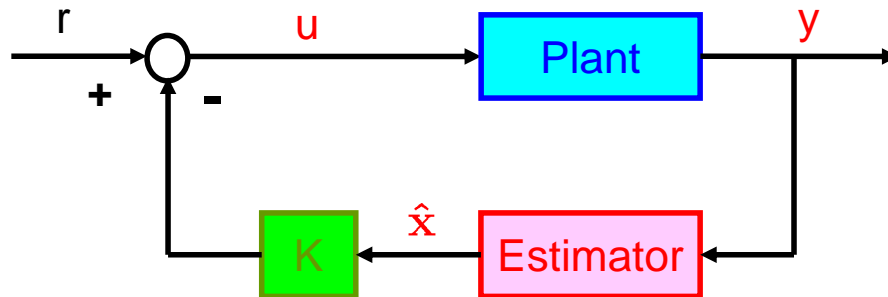
$$\hat{g}_f(s) =$$

$$\dot{\mathbf{x}} = (\mathbf{A} - \mathbf{b}\mathbf{k})\mathbf{x} + \mathbf{b}r$$

$$y = \mathbf{c}\mathbf{x}$$

$$\begin{cases} \dot{\mathbf{x}} = \mathbf{A}\mathbf{x} + \mathbf{b}u \\ \mathbf{y} = \mathbf{c}\mathbf{x} \end{cases}$$

$$s^4 + a_1s^3 + a_2s^2 + a_3s + a_4$$



$$u = r - \mathbf{k} \hat{\mathbf{x}}$$

$$\dot{\hat{\mathbf{x}}} = (\mathbf{A} - \mathbf{l}\mathbf{c})\hat{\mathbf{x}} + \mathbf{b}u + \mathbf{l}y$$

$$s^4 + b_1s^3 + b_2s^2 + b_3s + b_4$$

$$s^4 + c_1s^3 + c_2s^2 + c_3s + c_4$$

$$\begin{bmatrix} \dot{\mathbf{x}} \\ \dot{\mathbf{e}} \end{bmatrix} = \begin{bmatrix} \mathbf{A} - \mathbf{b}\mathbf{k} & \mathbf{b}\mathbf{k} \\ \mathbf{0} & \mathbf{A} - \mathbf{l}\mathbf{c} \end{bmatrix} \begin{bmatrix} \mathbf{x} \\ \mathbf{e} \end{bmatrix} + \begin{bmatrix} \mathbf{b} \\ \mathbf{0} \end{bmatrix} r$$

$$\mathbf{e} = \mathbf{x} - \hat{\mathbf{x}}$$

State Feedback and State Estimation (MIMO) (8.6 & 8.7)

$$s^4 + a_1s^3 + a_2s^2 + a_3s + a_4$$

$$\lambda_1, \lambda_2, \lambda_3, \lambda_4,$$

$$\begin{cases} \dot{\mathbf{x}} = \mathbf{A}\mathbf{x} + \mathbf{b}u \\ \mathbf{y} = \mathbf{c}\mathbf{x} \end{cases}$$

$$\dot{\mathbf{x}} = \mathbf{A}\mathbf{x} + \begin{bmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \end{bmatrix} u$$

$$\mathbf{y} = \begin{bmatrix} c_1 & c_2 & c_3 & c_4 \end{bmatrix} \mathbf{x}$$

$$\dot{\hat{\mathbf{x}}} = (\mathbf{A} - \mathbf{l}\mathbf{c})\hat{\mathbf{x}} + \mathbf{b}u + \mathbf{l}y$$

$$s^4 + c_1s^3 + c_2s^2 + c_3s + c_4$$

$$\lambda_1^o, \lambda_2^o, \lambda_3^o, \lambda_4^o,$$

$$\left(\mathbf{A} - \begin{bmatrix} l_1 \\ l_2 \\ l_3 \\ l_4 \end{bmatrix} \begin{bmatrix} c_1 & c_2 & c_3 & c_4 \end{bmatrix} \right)$$

$$u = r - \mathbf{k} \hat{\mathbf{x}}$$

$$(\mathbf{A} - \mathbf{b}\mathbf{k})$$

$$\left(\mathbf{A} - \begin{bmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \end{bmatrix} \begin{bmatrix} k_1 & k_2 & k_3 & k_4 \end{bmatrix} \right)$$

$$s^4 + b_1s^3 + b_2s^2 + b_3s + b_4$$

$$\lambda_1^c, \lambda_2^c, \lambda_3^c, \lambda_4^c,$$

$$s^4 + a_1 s^3 + a_2 s^2 + a_3 s + a_4 \quad \lambda_1, \lambda_2, \lambda_3, \lambda_4,$$

$$\begin{cases} \dot{\mathbf{x}} = \mathbf{A}\mathbf{x} + \mathbf{B}\mathbf{u} \\ \mathbf{y} = \mathbf{C}\mathbf{x} \end{cases}$$

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

$$s^4 + c_1 s^3 + c_2 s^2 + c_3 s + c_4 \quad \lambda_1^o, \lambda_2^o, \lambda_3^o, \lambda_4^o,$$

$$\mathbf{u} = \mathbf{r} - \mathbf{K}\hat{\mathbf{x}}$$

$$(\mathbf{A} - \mathbf{B}\mathbf{K})$$

$$s^4 + b_1 s^3 + b_2 s^2 + b_3 s + b_4 \quad \lambda_1^c, \lambda_2^c, \lambda_3^c, \lambda_4^c,$$

$$s^4 + a_1 s^3 + a_2 s^2 + a_3 s + a_4 \quad \lambda_1, \lambda_2, \lambda_3, \lambda_4,$$

$$\begin{cases} \dot{\mathbf{x}} = \mathbf{A}\mathbf{x} + \mathbf{B}\mathbf{u} \\ \mathbf{y} = \mathbf{C}\mathbf{x} \end{cases} \quad \dot{\mathbf{x}} = \mathbf{A}\mathbf{x} + \begin{bmatrix} b_{11} & b_{12} & b_{13} \\ b_{21} & b_{22} & b_{23} \\ b_{31} & b_{32} & b_{33} \\ b_{41} & b_{42} & b_{43} \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \\ u_3 \end{bmatrix}$$

$$\begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = \begin{bmatrix} c_{11} & c_{12} & c_{13} & c_{14} \\ c_{21} & c_{22} & c_{23} & c_{24} \end{bmatrix} \mathbf{x}$$

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

$$s^4 + c_1 s^3 + c_2 s^2 + c_3 s + c_4 \quad \lambda_1^o, \lambda_2^o, \lambda_3^o, \lambda_4^o,$$

$$\left(\mathbf{A} - \begin{bmatrix} l_{11} & l_{12} \\ l_{21} & l_{22} \\ l_{31} & l_{32} \\ l_{41} & l_{42} \end{bmatrix} \begin{bmatrix} c_{11} & c_{12} & c_{13} & c_{14} \\ c_{21} & c_{22} & c_{23} & c_{24} \end{bmatrix} \right)$$

$$\mathbf{u} = \mathbf{r} - \mathbf{K}\hat{\mathbf{x}}$$

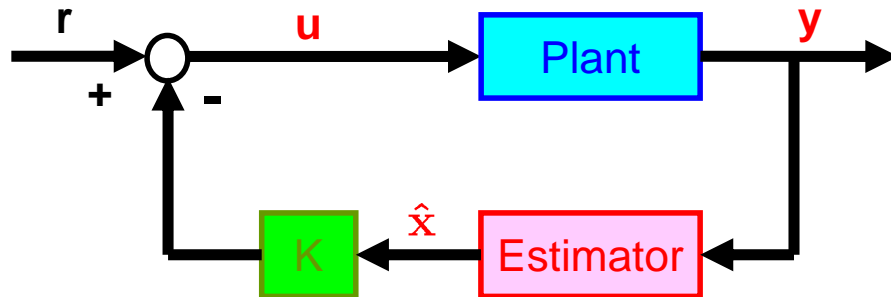
$$\left(\mathbf{A} - \begin{bmatrix} b_{11} & b_{12} & b_{13} \\ b_{21} & b_{22} & b_{23} \\ b_{31} & b_{32} & b_{33} \\ b_{41} & b_{42} & b_{43} \end{bmatrix} \begin{bmatrix} k_{11} & k_{12} & k_{13} & k_{14} \\ k_{21} & k_{22} & k_{23} & k_{24} \\ k_{31} & k_{32} & k_{33} & k_{34} \end{bmatrix} \right)$$

$$(\mathbf{A} - \mathbf{B}\mathbf{K})$$

$$s^4 + b_1 s^3 + b_2 s^2 + b_3 s + b_4 \quad \lambda_1^c, \lambda_2^c, \lambda_3^c, \lambda_4^c,$$

$$\begin{cases} \dot{\mathbf{x}} = \mathbf{A}\mathbf{x} + \mathbf{B}\mathbf{u} \\ \mathbf{y} = \mathbf{C}\mathbf{x} \end{cases}$$

$$s^4 + a_1 s^3 + a_2 s^2 + a_3 s + a_4$$



$$\mathbf{u} = \mathbf{r} - \mathbf{K} \hat{\mathbf{x}}$$

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

$$s^4 + b_1 s^3 + b_2 s^2 + b_3 s + b_4$$

$$s^4 + c_1 s^3 + c_2 s^2 + c_3 s + c_4$$

$$\begin{bmatrix} \dot{\mathbf{x}} \\ \dot{\mathbf{e}} \end{bmatrix} = \begin{bmatrix} \mathbf{A} - \mathbf{B}\mathbf{K} & \mathbf{B}\mathbf{K} \\ \mathbf{0} & \mathbf{A} - \mathbf{L}\mathbf{C} \end{bmatrix} \begin{bmatrix} \mathbf{x} \\ \mathbf{e} \end{bmatrix} + \begin{bmatrix} \mathbf{B} \\ \mathbf{0} \end{bmatrix} \mathbf{r}$$

$$\mathbf{e} = \mathbf{x} - \hat{\mathbf{x}}$$