

8. State Feedback and State Estimators

- ✓ State Feedback Controller Design
- ✓ Regulation and Tracking
- ✓ State Estimator Design
- ✓ Feedback from Estimated States
- ✓ State Feedback-Multivariable Case
- ✓ State Estimator-Multivariable Case
- ✓ Feedback from Estimated States-Multivariable Case

State Feedback Controller Design

State Feedback

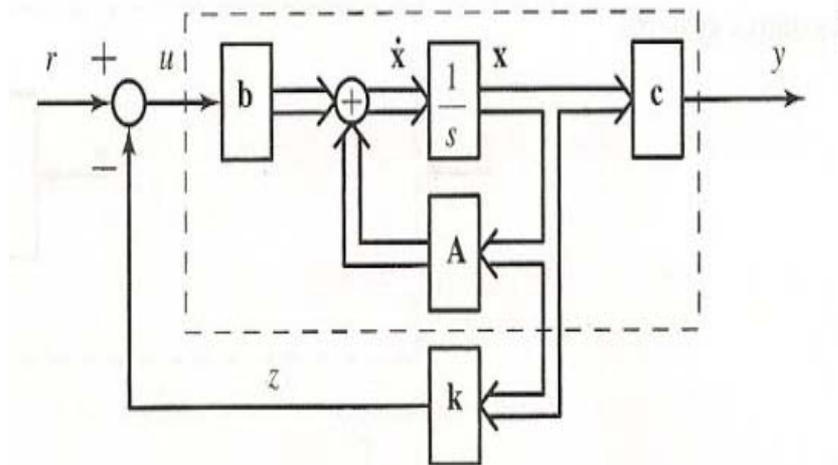
$$\dot{\mathbf{x}} = \mathbf{A}\mathbf{x} + \mathbf{b}u$$

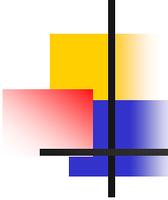
$$y = \mathbf{c}\mathbf{x}$$

$$u = r - \mathbf{k}\mathbf{x} = r - [k_1 \quad k_2 \quad \cdots \quad k_n]\mathbf{x} = r - \sum_{i=1}^n k_i x_i$$

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

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





State Feedback Controller Design

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.

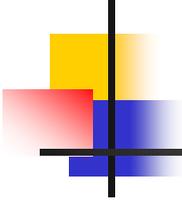
Proof : We show the theorem for $n = 4$.

$$C = [\mathbf{b} \quad \mathbf{A}\mathbf{b} \quad \mathbf{A}^2\mathbf{b} \quad \mathbf{A}^3\mathbf{b}]$$

and

$$C_f = [\mathbf{b} \quad (\mathbf{A} - \mathbf{b}\mathbf{k})\mathbf{b} \quad (\mathbf{A} - \mathbf{b}\mathbf{k})^2\mathbf{b} \quad (\mathbf{A} - \mathbf{b}\mathbf{k})^3\mathbf{b}]$$
$$C_f = C \begin{bmatrix} 1 & -\mathbf{k}\mathbf{b} & -\mathbf{k}(\mathbf{A} - \mathbf{b}\mathbf{k})\mathbf{b} & -\mathbf{k}(\mathbf{A} - \mathbf{b}\mathbf{k})^2\mathbf{b} \\ 0 & 1 & -\mathbf{k}\mathbf{b} & -\mathbf{k}(\mathbf{A} - \mathbf{b}\mathbf{k})\mathbf{b} \\ 0 & 0 & 1 & -\mathbf{k}\mathbf{b} \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

$\rightarrow \rho C_f = \rho C \rightarrow$ Controllability is invariant.



State Feedback Controller Design

Example 8.1

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

$$y = [1 \quad 2] \mathbf{x}$$

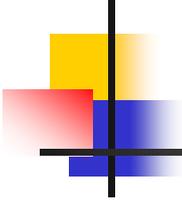
$$C = \begin{bmatrix} 0 & 2 \\ 1 & 1 \end{bmatrix}, \quad O = \begin{bmatrix} 1 & 2 \\ 7 & 4 \end{bmatrix} : \textit{controllable \& observable}$$

$$u = r - [3 \quad 1] \mathbf{x}$$

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

$$y = [1 \quad 2] \mathbf{x}$$

$$C_f = \begin{bmatrix} 0 & 2 \\ 1 & 0 \end{bmatrix}, \quad O_f = \begin{bmatrix} 1 & 2 \\ 1 & 2 \end{bmatrix} : \textit{controllable \& not observable}$$



State Feedback Controller Design

Example 8.2

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

$$\Delta(s) = (s-1)^2 - 9 = s^2 - 2s - 8 = (s-4)(s+2) : \textit{unstable}$$

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

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

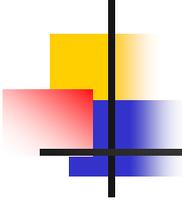
$$\Delta_f(s) = (s-1+k_1)(s-1) - 3(3-k_2) = s^2 + (k_1-2)s + (3k_2 - k_1 - 8)$$

If we want to place the eigenvalues at $-1 \pm j2$,

$$\Delta_f(s) = (s-1-j2)(s-1+j2) = s^2 + 2s + 5$$

$$\rightarrow k_1 = 4, k_2 = 17/3.$$

\rightarrow *Stabilized.*



State Feedback Controller Design

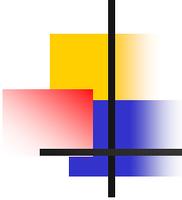
Theorem 8.2

Consider the state equation in (8.1) with $n = 4$
and the characteristic polynomial

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

If (8.1) is controllable, then it can be transformed
by the transformation $\bar{\mathbf{x}} = \mathbf{P}\mathbf{x}$ with

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



State Feedback Controller Design

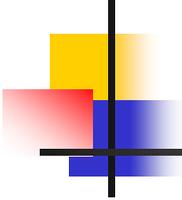
into the controllable canonical form

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

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

Furthermore, the transfer function of (8.1) with $n = 4$ equals

$$g(s) = \frac{\beta_1 s^3 + \beta_2 s^2 + \beta_3 s + \beta_4}{s^4 + \alpha_1 s^3 + \alpha_2 s^2 + \alpha_3 s + \alpha_4}$$



State Feedback Controller Design

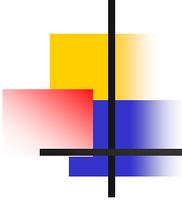
Proof :

C & \bar{C} is nonsingular since the eq. is controllable and
 $\bar{C} = \mathbf{P}C$.

$$\mathbf{P} = \bar{C}C^{-1} \quad \text{or} \quad \mathbf{Q} := \mathbf{P}^{-1} = C\bar{C}^{-1}$$

$$\bar{C} = \begin{bmatrix} 1 & -\alpha_1 & \alpha_1^2 - \alpha_2 & -\alpha_3^3 + 2\alpha_1\alpha_2 - \alpha_3 \\ 0 & 1 & -\alpha_1 & \alpha_1^2 - \alpha_2 \\ 0 & 0 & 1 & -\alpha_1 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

$$\bar{C}^{-1} = \begin{bmatrix} 1 & \alpha_1 & \alpha_2 & \alpha_3 \\ 0 & 1 & \alpha_1 & \alpha_2 \\ 0 & 0 & 1 & \alpha_1 \\ 0 & 0 & 0 & 1 \end{bmatrix} \text{ is assumed. Proved by multiplying.}$$



State Feedback Controller Design

Theorem 8.3

If the n -dimensional state equation in (8.1) is controllable, then by the 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.

State Feedback Controller Design

Proof :

$$u = r - \mathbf{k}\mathbf{x} = r - \mathbf{k}\mathbf{P}^{-1}\bar{\mathbf{x}} =: r - \bar{\mathbf{k}}\bar{\mathbf{x}}$$

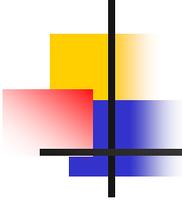
$$\Delta_f(s) = s^4 + \bar{\alpha}_1 s^3 + \bar{\alpha}_2 s^2 + \bar{\alpha}_3 s + \bar{\alpha}_4$$

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

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

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

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



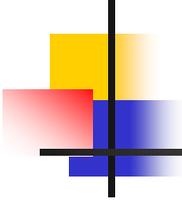
State Feedback Controller Design

Alternative derivation of $\bar{\mathbf{k}}$:

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

$$\Delta_f(s) = \Delta(s) \left[1 + \mathbf{k}(s\mathbf{I} - \mathbf{A})^{-1}\mathbf{b}\right]$$

$$\Delta_f(s) - \Delta(s) = \Delta(s)\mathbf{k}(s\mathbf{I} - \mathbf{A})^{-1}\mathbf{b} = \Delta(s)\bar{\mathbf{k}}(s\mathbf{I} - \bar{\mathbf{A}})^{-1}\bar{\mathbf{b}}$$



State Feedback Controller Design

$$\bar{\mathbf{c}}(s\mathbf{I} - \bar{\mathbf{A}})^{-1} \bar{\mathbf{b}} = \frac{\beta_1 s^3 + \beta_2 s^2 + \beta_3 s + \beta_4}{\Delta(s)}$$

$$\bar{\mathbf{k}}(s\mathbf{I} - \bar{\mathbf{A}})^{-1} \bar{\mathbf{b}} = \frac{\bar{k}_1 s^3 + \bar{k}_2 s^2 + \bar{k}_3 s + \bar{k}_4}{\Delta(s)}$$

Then

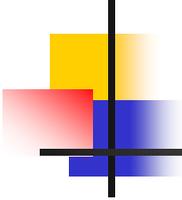
$$\Delta_f(s) - \Delta(s) = \bar{k}_1 s^3 + \bar{k}_2 s^2 + \bar{k}_3 s + \bar{k}_4$$

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

Feedback Transfer Function

$$g(s) = \mathbf{c}(s\mathbf{I} - \mathbf{A})^{-1} \mathbf{b} = \frac{\beta_1 s^3 + \beta_2 s^2 + \beta_3 s + \beta_4}{s^4 + \alpha_1 s^3 + \alpha_2 s^2 + \alpha_3 s + \alpha_4}$$

$$g_f(s) = \mathbf{c}(s\mathbf{I} - \mathbf{A} + \mathbf{b}\mathbf{k})^{-1} \mathbf{b} = \frac{\beta_1 s^3 + \beta_2 s^2 + \beta_3 s + \beta_4}{s^4 + \bar{\alpha}_1 s^3 + \bar{\alpha}_2 s^2 + \bar{\alpha}_3 s + \bar{\alpha}_4}$$



State Feedback Controller Design

Example 8.3

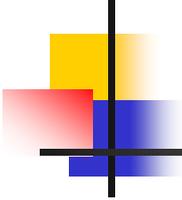
$$\dot{\mathbf{x}} = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & -1 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 5 & 0 \end{bmatrix} \mathbf{x} + \begin{bmatrix} 0 \\ 1 \\ 0 \\ -2 \end{bmatrix} u$$

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

→ *Controllable*

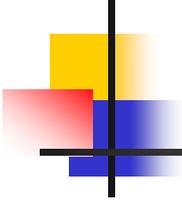
→ its eigenvalues can be assigned arbitrarily.

$$\Delta(s) = s^2(s^2 - 5) = s^4 + 0 \cdot s^3 - 5s^2 + 0 \cdot s + 0$$



State Feedback Controller Design

$$\begin{aligned}\mathbf{P}^{-1} = \mathbf{C}\bar{\mathbf{C}}^{-1} &= \begin{bmatrix} 0 & 1 & 0 & 2 \\ 1 & 0 & 2 & 0 \\ 0 & -2 & 0 & -10 \\ -2 & 0 & -10 & 0 \end{bmatrix} \begin{bmatrix} 1 & 0 & -5 & 0 \\ 0 & 1 & 0 & -5 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \\ &= \begin{bmatrix} 0 & 1 & 0 & -3 \\ 1 & 0 & -3 & 0 \\ 0 & -2 & 0 & 0 \\ -2 & 0 & 0 & 0 \end{bmatrix} \\ \mathbf{P} &= \begin{bmatrix} 0 & 0 & 0 & -\frac{1}{2} \\ 0 & 0 & -\frac{1}{2} & 0 \\ 0 & -\frac{1}{3} & 0 & -\frac{1}{6} \\ -\frac{1}{3} & 0 & -\frac{1}{6} & 0 \end{bmatrix}\end{aligned}$$



State Feedback Controller Design

Let the desired eigenvalues be $-1.5 \pm 0.5j$ and $-1 \pm j$

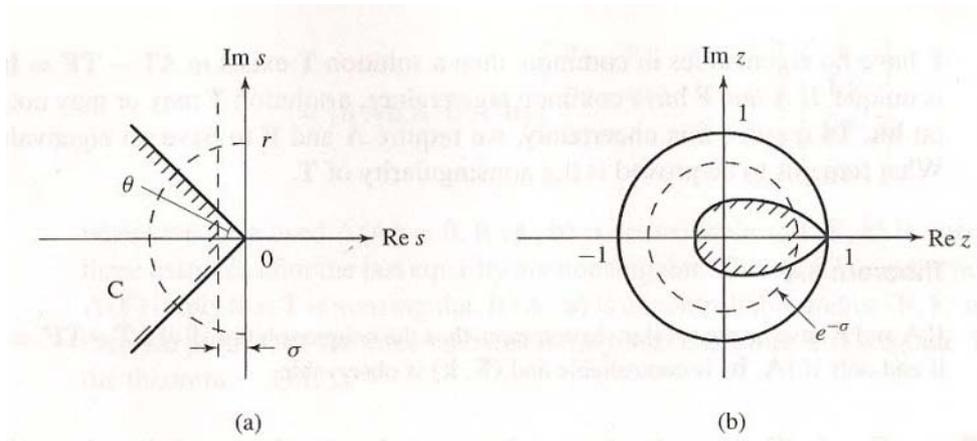
$$\begin{aligned}\Delta_f(s) &= (s + 1.5 - 0.5j)(s + 1.5 + 0.5j)(s + 1 - j)(s + 1 + j) \\ &= s^4 + 5s^3 + 10.5s^2 + 11s + 5\end{aligned}$$

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

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

State Feedback Controller Design

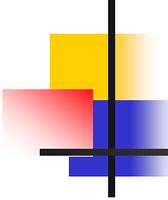
How to determine the desired eigenvalues?



Find \mathbf{k} to minimize the objective function

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

→ *Optimal Control Theory*



State Feedback Controller Design

Solving the Lyapunov Equation

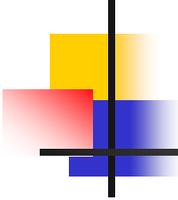
Procedure 8.1

Consider controllable (\mathbf{A}, \mathbf{b}) . Find \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.
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:

$$(\mathbf{A} - \mathbf{b}\mathbf{k})\mathbf{T} = \mathbf{T}\mathbf{F} \quad \text{or} \quad \mathbf{A} - \mathbf{b}\mathbf{k} = \mathbf{T}\mathbf{F}\mathbf{T}^{-1} \rightarrow \mathbf{A} - \mathbf{b}\mathbf{k} \text{ is similar to } \mathbf{F}.$$



State Feedback Controller Design

Theorem 8.4

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

Proof :

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

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

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

Since \mathbf{A} and \mathbf{F} have no common eigenvalues, $\Delta(\bar{\lambda}_i) \neq 0$.

If $\bar{\lambda}_i$ is eigenvalue of \mathbf{F} , $\Delta(\bar{\lambda}_i)$ is eigenvalue of $\Delta(\mathbf{F})$ (Problem 3.19)

$$\det \Delta(\mathbf{F}) = \prod_i \Delta(\bar{\lambda}_i) \neq 0 \rightarrow \Delta(\mathbf{F}) \text{ is nonsingular.}$$

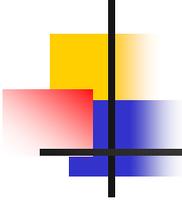
State Feedback Controller Design

Substituting $\mathbf{AT} = \mathbf{TF} + \mathbf{bk}$ into $\mathbf{A}^2\mathbf{T} - \mathbf{AF}^2$ yields
 $\mathbf{A}^2\mathbf{T} - \mathbf{TF}^2 = \mathbf{A}(\mathbf{TF} + \mathbf{bk}) - \mathbf{TF}^2 = \mathbf{Abk} + (\mathbf{AT} - \mathbf{TF})\mathbf{F}$
 $= \mathbf{Abk} + \mathbf{bkF}$

$$\begin{array}{r}
 \mathbf{IT} - \mathbf{TI} = \mathbf{0} \qquad \qquad \qquad \times \alpha_4 \\
 \mathbf{AT} - \mathbf{TF} = \mathbf{bk} \qquad \qquad \qquad \times \alpha_3 \\
 \mathbf{A}^2\mathbf{T} - \mathbf{TF}^2 = \mathbf{Abk} + \mathbf{bkF} \qquad \qquad \times \alpha_2 \\
 \mathbf{A}^3\mathbf{T} - \mathbf{TF}^3 = \mathbf{A}^2\mathbf{bk} + \mathbf{AbkF} + \mathbf{bkF}^2 \qquad \times \alpha_1 \\
 + \mathbf{A}^4\mathbf{T} - \mathbf{TF}^4 = \mathbf{A}^3\mathbf{bk} + \mathbf{A}^2\mathbf{bkF} + \mathbf{AbkF}^2 + \mathbf{bkF}^3 \qquad \times 1 \\
 \hline
 \Delta(\mathbf{A})\mathbf{T} - \Delta(\mathbf{F}) = \Delta(\mathbf{F})\Delta(\mathbf{A}) \text{ since } = 0
 \end{array}$$

$$= \begin{bmatrix} \mathbf{b} & \mathbf{Ab} & \mathbf{A}^2\mathbf{b} & \mathbf{A}^3\mathbf{b} \end{bmatrix} \begin{bmatrix} \alpha_3 & \alpha_2 & \alpha_1 & 1 \\ \alpha_2 & \alpha_1 & 1 & 0 \\ \alpha_1 & 1 & 0 & 0 \\ 1 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} \mathbf{k} \\ \mathbf{kF} \\ \mathbf{kF}^2 \\ \mathbf{kF}^3 \end{bmatrix}$$

→ \mathbf{T} is nonsingular.



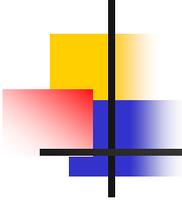
State Feedback Controller Design

Selection of observable pair $\{\mathbf{F}, \bar{\mathbf{k}}\}$

$$\mathbf{F} = \begin{bmatrix} 0 & 0 & 0 & 0 & \alpha_5 \\ 1 & 0 & 0 & 0 & \alpha_4 \\ 0 & 1 & 0 & 0 & \alpha_3 \\ 0 & 0 & 1 & 0 & \alpha_2 \\ 0 & 0 & 0 & 1 & \alpha_1 \end{bmatrix}, \bar{\mathbf{k}} = [1 \quad 0 \quad 0 \quad 0 \quad 0]$$

Desired Eigenvalues: $\lambda_1, \alpha_1 \pm j\beta_1, \alpha_2 \pm j\beta_2$

$$\mathbf{F} = \begin{bmatrix} \lambda_1 & 0 & 0 & 0 & 0 \\ 0 & \alpha_1 & \beta_1 & 0 & 0 \\ 0 & -\beta_1 & \alpha_1 & 0 & 0 \\ 0 & 0 & 0 & \alpha_2 & \beta_2 \\ 0 & 0 & 0 & -\beta_2 & \alpha_2 \end{bmatrix}, \begin{array}{l} \bar{\mathbf{k}} = [1 \quad 1 \quad 0 \quad 1 \quad 0] \\ \bar{\mathbf{k}} = [1 \quad 1 \quad 0 \quad 0 \quad 1] \\ \bar{\mathbf{k}} = [1 \quad 1 \quad 1 \quad 1 \quad 1] \end{array} \quad \text{(Problem 6.16)}$$



HW 8-1

1. Find the state feedback gain for the state equation

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

so that the resulting system has eigenvalues -2 and $-1 \pm j$.

Use the method you think is the simplest by hand to carry out.

2. Consider a system with transfer function

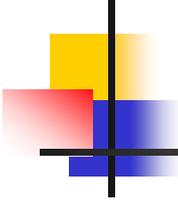
$$g(s) = \frac{(s-1)(s+2)}{(s+1)(s-2)(s+3)}.$$

Is it possible to change the transfer function to

$$g_f(s) = \frac{1}{s+3}$$

by the state feedback? Is the resulting system BIBO stable?

Is it asymptotically stable?



Regulation and Tracking

Regulation and Tracking

Regulation : to find out the state feedback gain
so that the output decay to zero.

Tracking : to find out the state feedback control $u(t)$
so that $y(t)$ approaches to $r(t) = a$

By stabilizing control $u = -\mathbf{kx}$ for $r(t) = 0$,

Zero input response

$$y(t) = ce^{(\mathbf{A}-\mathbf{bk})t} \mathbf{x}(0)$$

will decay to zero. The regulation can be easily achieved.

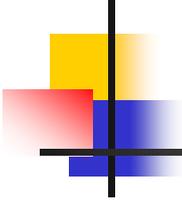
For tracking , we need a feedforward gain p as

$$u(t) = pr(t) - \mathbf{kx}.$$

$$\lim_{t \rightarrow \infty} y(t) = \lim_{s \rightarrow 0} sy(s) = \lim_{s \rightarrow 0} sg_f(s)r(s)$$

$$= \lim_{s \rightarrow 0} sp \frac{\beta_1 s^3 + \beta_2 s^2 + \beta_3 s + \beta_4}{s^4 + \bar{\alpha}_1 s^3 + \bar{\alpha}_2 s^2 + \bar{\alpha}_3 s + \bar{\alpha}_4} \frac{a}{s} = \frac{p\beta_4}{\bar{\alpha}_4} a = a$$

$$\rightarrow p = \frac{\bar{\alpha}_4}{\beta_4}, \text{ where } \beta_4 \text{ should not be zero.}$$



Regulation and Tracking

Robust Tracking and Disturbance Rejection

Robust tracking:

When the parameter of the transfer function is perturbed, the feedforward gain p may not yield the exact tracking.

→ nonrobust → robust design is required.

Disturbance rejection:

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

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

To design the controller so that the output track the step response even with the presence of a disturbance $w(t)$.

Regulation and Tracking

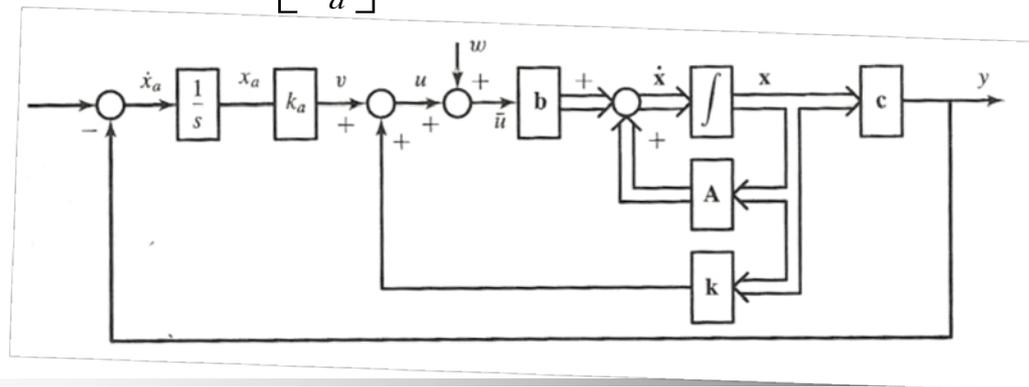
Robust Tracking and Disturbance Rejection

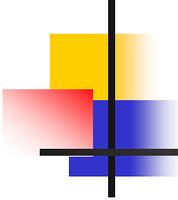
$$\dot{x}_a = r - y = r - \mathbf{c}\mathbf{x}$$

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

$$\begin{bmatrix} \dot{\mathbf{x}} \\ \dot{x}_a \end{bmatrix} = \begin{bmatrix} \mathbf{A} + \mathbf{b}\mathbf{k} & \mathbf{b}k_a \\ -\mathbf{c} & 0 \end{bmatrix} \begin{bmatrix} \mathbf{x} \\ x_a \end{bmatrix} + \begin{bmatrix} \mathbf{0} \\ 1 \end{bmatrix} r + \begin{bmatrix} \mathbf{b} \\ 0 \end{bmatrix} w \quad \dots\dots (8-29)$$

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





Regulation and Tracking

Theorem 8.5

If (\mathbf{A}, \mathbf{b}) is controllable and if $g(s) = \mathbf{c}(s\mathbf{I} - \mathbf{A})^{-1}\mathbf{b}$ has no zero at $s = 0$, then all eigenvalues of the \mathbf{A} -matrix in (8.29) can be assigned arbitrarily by selecting a feedback gain $[\mathbf{k} \quad k_a]$

Proof :

$$\begin{bmatrix} \dot{\mathbf{x}} \\ \dot{x}_a \end{bmatrix} = \begin{bmatrix} \mathbf{A} + \mathbf{b}\mathbf{k} & \mathbf{b}k_a \\ -\mathbf{c} & 0 \end{bmatrix} \begin{bmatrix} \mathbf{x} \\ x_a \end{bmatrix} + \begin{bmatrix} \mathbf{0} \\ 1 \end{bmatrix} r + \begin{bmatrix} \mathbf{b} \\ 0 \end{bmatrix} w \quad \text{..... (8-29)}$$

can be expressed as

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

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

Regulation and Tracking

Proof of Theorem 8.5

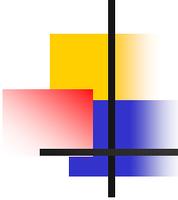
It is enough to show that

$$\begin{bmatrix} \mathbf{A} & \mathbf{0} \\ -\mathbf{c} & 0 \end{bmatrix} \begin{bmatrix} \mathbf{b} \\ 0 \end{bmatrix}$$

is controllable if only if $\beta_4 \neq 0$ (the plant has no zero at $s = 0$).

We prove for $n = 4$.

$$\begin{bmatrix} \mathbf{b} & \mathbf{A}\mathbf{b} & \mathbf{A}^2\mathbf{b} & \mathbf{A}^3\mathbf{b} & \mathbf{A}^4\mathbf{b} \\ 0 & -\mathbf{c}\mathbf{b} & -\mathbf{c}\mathbf{A}\mathbf{b} & -\mathbf{c}\mathbf{A}^2\mathbf{b} & -\mathbf{c}\mathbf{A}^3\mathbf{b} \end{bmatrix} = \begin{bmatrix} 1 & -\alpha_1 & \alpha_1^2 - \alpha_2 & -\alpha_1(\alpha_1^2 - \alpha_2) + \alpha_2\alpha_1 - \alpha_3 & a_{15} \\ 0 & 1 & -\alpha_1 & \alpha_1^2 - \alpha_2 & a_{25} \\ 0 & 0 & 1 & -\alpha_1 & a_{35} \\ 0 & 0 & 0 & 1 & a_{45} \\ 0 & -\beta_1 & \beta_1\alpha_1 - \beta_2 & -\beta_1(\alpha_1^2 - \alpha_2) + \beta_2\alpha_1 - \beta_3 & a_{55} \end{bmatrix}$$



Regulation and Tracking

The rank of a matrix does not change by elementary operations. Adding the second row multiplied β_1 to the last row, and adding the third row multiplied β_2 to the last row, and adding the fourth row multiplied β_3 to the last row, we obtain

$$\begin{bmatrix} 1 & -\alpha_1 & \alpha_1^2 - \alpha_2 & -\alpha_1(\alpha_1^2 - \alpha_2) + \alpha_2\alpha_1 - \alpha_3 & a_{15} \\ 0 & 0 & -\alpha_1 & \alpha_1^2 - \alpha_2 & a_{25} \\ 0 & 0 & 1 & -\alpha_1 & a_{35} \\ 0 & 0 & 0 & 1 & a_{45} \\ 0 & 0 & 0 & 0 & -\beta_4 \end{bmatrix}$$

which is nonsingular.

Regulation and Tracking

Characteristic polynomial of overall system

$$\Delta_f(s) = \det \begin{bmatrix} s\mathbf{I} - \mathbf{A} - \mathbf{b}\mathbf{k} & -\mathbf{b}k_a \\ \mathbf{c} & s \end{bmatrix}$$

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

$$1 \cdot \Delta_f(s) = \bar{D}(s) \left(s + \frac{\bar{N}(s)}{\bar{D}(s)} k_a \right)$$

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

\Rightarrow

$$\Delta_f(s) = s\bar{D}(s) + k_a\bar{N}(s)$$

Regulation and Tracking

Disturbance Rejection

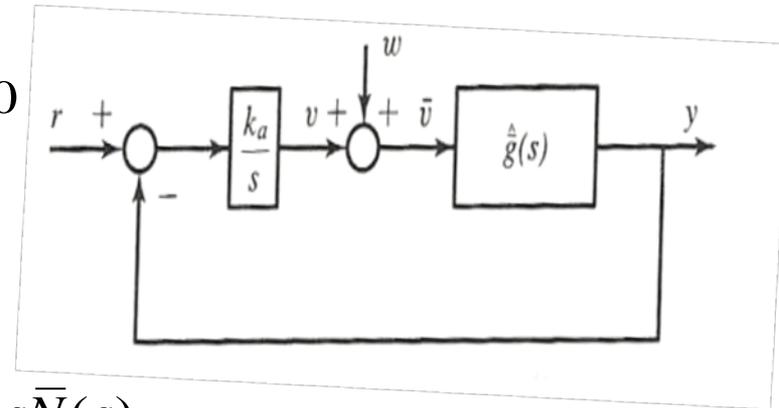
$$y = \frac{\bar{N}(s)}{\bar{D}(s)}(v + w), \quad v = \frac{k_a}{s}(r - y), \quad r = 0$$

$$\left(1 + \frac{k_a \bar{N}(s)}{s \bar{D}(s)}\right) y = \frac{\bar{N}(s)}{\bar{D}(s)} w$$

$$g_{yw} = \frac{\frac{\bar{N}(s)}{\bar{D}(s)}}{1 + \frac{k_a \bar{N}(s)}{s \bar{D}(s)}} = \frac{s \bar{N}(s)}{s \bar{D}(s) + k_a \bar{N}(s)} = \frac{s \bar{N}(s)}{\Delta_f(s)}$$

If $w(t) = \bar{w}$, $\mathcal{L}(w) = \bar{w}/s$,

$$y_w(s) = \frac{s \bar{N}(s)}{\Delta_f(s)} \frac{\bar{w}}{s} = \frac{\bar{w} \bar{N}(s)}{\Delta_f(s)} \rightarrow 0 \text{ by assigning poles in LHP.}$$



Regulation and Tracking

Tracking to step reference

$$y = \frac{\bar{N}(s)}{\bar{D}(s)}(v + w), \quad v = \frac{k_a}{s}(r - y), \quad w = 0$$

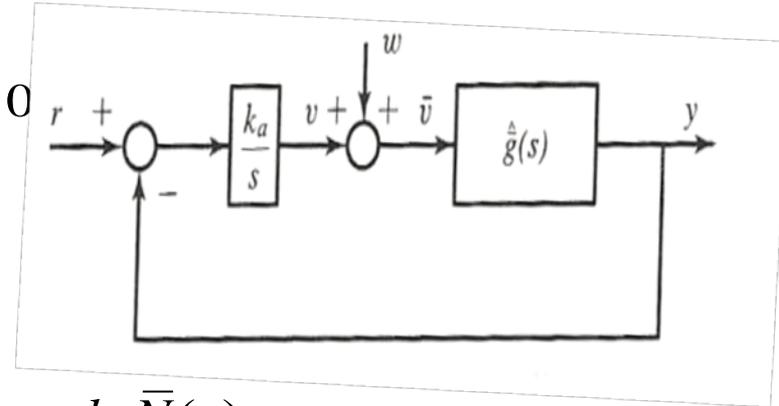
$$\left(1 + \frac{k_a \bar{N}(s)}{s \bar{D}(s)}\right) y = \frac{k_a \bar{N}(s)}{s \bar{D}(s)} r$$

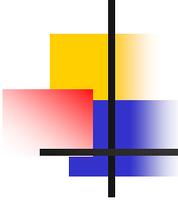
$$g_{yr}(s) = \frac{\frac{k_a \bar{N}(s)}{s \bar{D}(s)}}{1 + \frac{k_a \bar{N}(s)}{s \bar{D}(s)}} = \frac{k_a \bar{N}(s)}{s \bar{D}(s) + k_a \bar{N}(s)} = \frac{k_a \bar{N}(s)}{\Delta_f(s)}$$

If $r(t) = \bar{r}$, $\mathcal{L}(w) = \bar{r} / s$,

$$y_r(s) = \frac{k_a \bar{N}(s) \bar{r}}{\Delta_f(s) s},$$

$$\lim_{t \rightarrow \infty} y(t) = \lim_{s \rightarrow 0} s y_r(s) = \frac{k_a \bar{N}(0) \bar{r}}{0 \cdot \bar{D}(s) + k_a \bar{N}(0)} = \frac{k_a \bar{N}(0) \bar{r}}{k_a \bar{N}(0)} = \bar{r}$$





Regulation and Tracking

Stabilization

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

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

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

$\bar{\mathbf{A}}_{\bar{c}}$ is not affected by state feedback.

If $\bar{\mathbf{A}}_{\bar{c}}$ is stable and $(\bar{\mathbf{A}}_c, \bar{\mathbf{b}}_c)$ is controllable,

(\mathbf{A}, \mathbf{b}) is said to be **stabilizable**.

State Estimator Design

State Estimator

Open-loop state Estimator

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

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

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

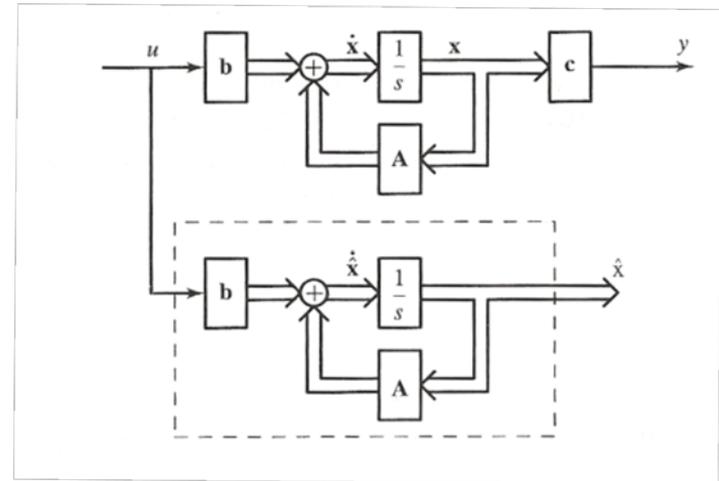
Two problems

1. Initial state must be computed $\leftarrow \{ \mathbf{A}, \mathbf{c} \}$ is observable
2. If \mathbf{A} is unstable, the estimate error may diverge.

$$\dot{\mathbf{x}} - \dot{\hat{\mathbf{x}}} = \mathbf{A}(\mathbf{x} - \hat{\mathbf{x}}),$$

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

$$\dot{\mathbf{e}} = \mathbf{A}\mathbf{e}, \quad \mathbf{e} \rightarrow \infty \quad \text{if } \text{Re}(\lambda(\mathbf{A})) > 0.$$



State Estimator Design

Closed-loop State Estimator

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

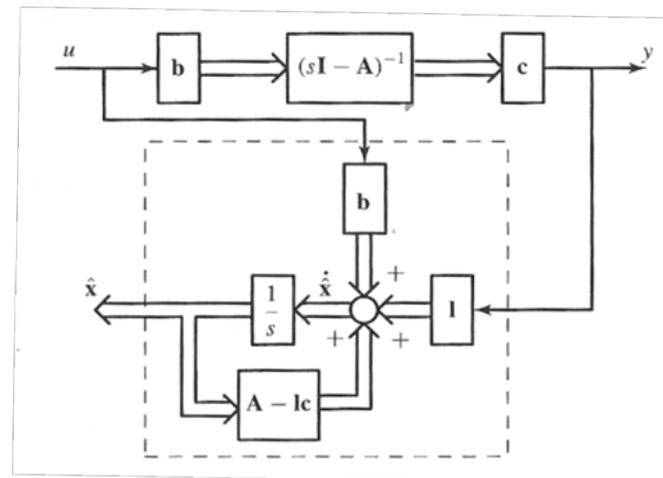
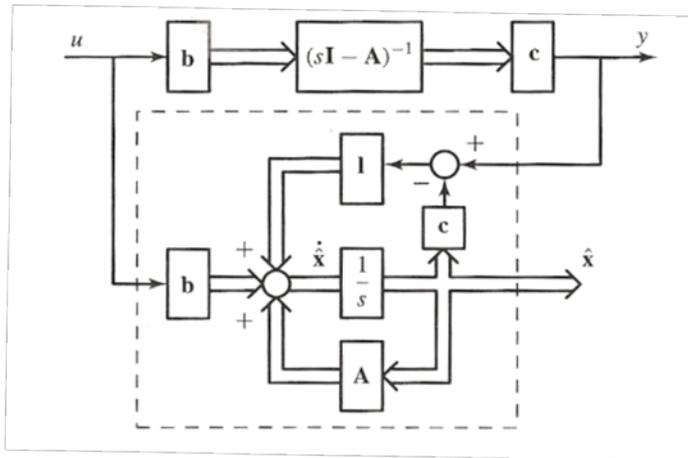
$$\dot{\hat{\mathbf{x}}} \hat{=} (\mathbf{A} - \mathbf{l}\mathbf{c})\mathbf{x} + \mathbf{b}u + \mathbf{l}y$$

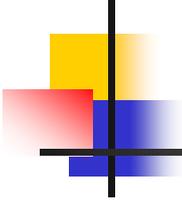
$$\mathbf{e}(t) := \mathbf{x}(t) - \hat{\mathbf{x}}(t)$$

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

$$= (\mathbf{A} - \mathbf{l}\mathbf{c})\mathbf{x} - (\mathbf{A} - \mathbf{l}\mathbf{c})\hat{\mathbf{x}} \hat{=} (\mathbf{A} - \mathbf{l}\mathbf{c})(\mathbf{x} - \hat{\mathbf{x}})$$

$$\dot{\mathbf{e}} = (\mathbf{A} - \mathbf{l}\mathbf{c})\mathbf{e}$$





State Estimator Design

Theorem 8.03

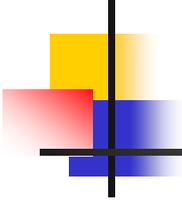
Consider the pair (\mathbf{A}, \mathbf{c}) . All eigenvalues of $(\mathbf{A} - \mathbf{l}\mathbf{c})$ can be assigned arbitrarily by selecting a real constant vector \mathbf{l} if and only if (\mathbf{A}, \mathbf{c}) is observable.

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

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

Procedure 8.01

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.



State Estimator Design

3. Solve the unique \mathbf{T} in the Lyapunov equation $\mathbf{TA} - \mathbf{FT} = \mathbf{lc}$. This \mathbf{T} is nonsingular following the dual of Theorem 8.4.
4. Then the state equation

$$\dot{\mathbf{z}} = \mathbf{Fz} + \mathbf{Tbu} + \mathbf{ly}$$

$$\hat{\mathbf{x}} = \mathbf{T}^{-1}\mathbf{z}$$

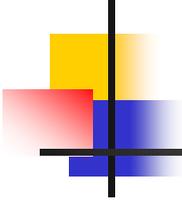
generates an estimate of \mathbf{x} .

Verify :

$$\mathbf{e} := \mathbf{z} - \mathbf{T}\mathbf{x}$$

$$\begin{aligned}\dot{\mathbf{e}} &= \dot{\mathbf{z}} - \mathbf{T}\dot{\mathbf{x}} = \mathbf{Fz} + \mathbf{Tbu} + \mathbf{lcx} - \mathbf{T}\mathbf{Ax} - \mathbf{Tbu} \\ &= \mathbf{Fz} + \mathbf{lcx} - (\mathbf{FT} + \mathbf{lc})\mathbf{x} = \mathbf{F}(\mathbf{z} - \mathbf{T}\mathbf{x}) = \mathbf{F}\mathbf{e}\end{aligned}$$

$$\lim_{t \rightarrow \infty} \mathbf{e} = 0 \rightarrow \lim_{t \rightarrow \infty} \mathbf{z} = \mathbf{T}\mathbf{x} \rightarrow \lim_{t \rightarrow \infty} \mathbf{T}^{-1}\mathbf{z} = \mathbf{x}$$



State Estimator Design

Reduced-Dimensional State Estimator

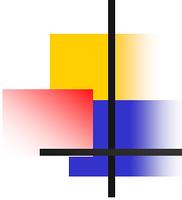
Procedure 8.R1

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 $(n-1) \times 1$ matrix.
4. Then the $(n-1)$ -dimensional state equation

$$\dot{\mathbf{z}} = \mathbf{F}\mathbf{z} + \mathbf{T}\mathbf{b}u + \mathbf{l}y$$

$$\hat{\mathbf{x}} = \begin{bmatrix} \mathbf{c} \\ \mathbf{T} \end{bmatrix}^{-1} \begin{bmatrix} y \\ \mathbf{z} \end{bmatrix}$$

is an estimate of \mathbf{x} .



State Estimator Design

$$\begin{bmatrix} y \\ \mathbf{z} \end{bmatrix} = \begin{bmatrix} \mathbf{c} \\ \mathbf{T} \end{bmatrix} \hat{\mathbf{x}} =: \mathbf{P}\mathbf{x}$$

$$\mathbf{e} = \mathbf{z} - \mathbf{T}\mathbf{x}$$

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

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 $\mathbf{T}\mathbf{A} - \mathbf{F}\mathbf{T} = \mathbf{l}\mathbf{c}$,
is nonsingular if and only if (\mathbf{A}, \mathbf{c}) is observable
and (\mathbf{F}, \mathbf{l}) is controllable.

State Estimator Design

Proof :

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

Dual to (8.22),

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

$$:= \mathbf{C}_4 \mathbf{\Lambda} \mathbf{O}$$

$\Delta(\mathbf{F})$ is nonsingular if \mathbf{A} and \mathbf{F} have no common eigenvalues.

Then $\mathbf{T} = \mathbf{A}\mathbf{R}(\mathbf{F})^{-1} \mathbf{C}_4 \mathbf{O}$ and becomes

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

If $(\mathbf{F}, \mathbf{1})$ is not controllable, \mathbf{C}_4 has rank 2 at most and \mathbf{P} is singular.

If (\mathbf{A}, \mathbf{c}) is not observable, \mathbf{O} has at least 1-dim. null space and there exists $\mathbf{r} \neq \mathbf{0}$ such that $\mathbf{O}\mathbf{r} = \mathbf{0}$ which implies $\mathbf{c}\mathbf{r} = \mathbf{0}$ and $\mathbf{P}\mathbf{r} = \mathbf{0}$.

Thus \mathbf{P} is singular. This is proof of the necessity part.

State Estimator Design

The sufficiency is proved by contraction.

Suppose P is singular, then there exists $\mathbf{r} \neq 0$ such that

$$\begin{bmatrix} \mathbf{c} \\ C_4 \Lambda \mathbf{O} \mathbf{r} \end{bmatrix} \mathbf{r} = \begin{bmatrix} \mathbf{c} \mathbf{r} \\ C_4 \mathbf{O} \end{bmatrix} = \mathbf{0}$$

Define $\mathbf{a} = \mathbf{A} \mathbf{r} \mathbf{O} = [a_1 \ a_2 \ a_3 \ a_4]'$, $\mathbf{a} = [\quad a_4]$

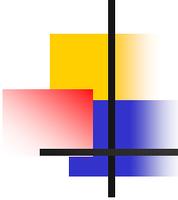
$$\begin{bmatrix} a_1 \\ a_2 \\ a_3 \\ a_4 \end{bmatrix} = \begin{bmatrix} \alpha_3 & \alpha_2 & \alpha_1 & 1 \\ \alpha_2 & \alpha_1 & 1 & 0 \\ \alpha_1 & 1 & 0 & 0 \\ 1 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} \mathbf{c} \mathbf{r} \\ \mathbf{c} \mathbf{A} \mathbf{r} \\ \mathbf{c} \mathbf{A}^2 \mathbf{r}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

$C_4 \Lambda \mathbf{O} = \mathbf{O}_4 = \mathbf{a}' \mathbf{O} \rightarrow \mathbf{a}' = 0$ iff $\{ \mathbf{F}, \mathbf{l} \}$ is controllable.

$\rightarrow \mathbf{a} = 0$ if $\{ \mathbf{F}, \mathbf{l} \}$ is controllable.

$\rightarrow \mathbf{a} \mathbf{A} \mathbf{r} \mathbf{O} = \mathbf{O} \mathbf{r} \rightarrow = \mathbf{O} \mathbf{A} \mathbf{r} \mathbf{O} \rightarrow \{ \mathbf{A}, \mathbf{c} \}$ is observable.

\rightarrow *contradict.*



HW 8-2

3. Find a reduced dimensional state estimator for the state equation and
Verify the validity of the designed estimator through Matlab simulation

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

$$y = [1 \quad 1 \quad 1] \mathbf{x}.$$

Select the eigenvalues ~~$-\beta$~~ and $-3 \pm 2j$.

State Feedback from Estimated States

Feedback from Estimated States

The state equation:

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

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

If the state is not measurable,

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

we use the estimated state for feedback

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

Then

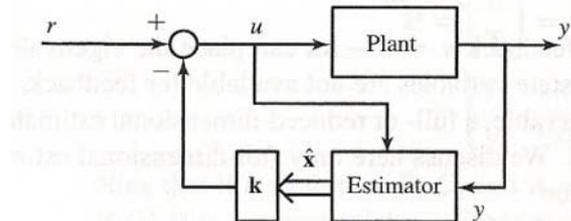
$$\dot{\mathbf{x}} = \mathbf{A}\mathbf{x} - \mathbf{b}\mathbf{k}\hat{\mathbf{x}} + \mathbf{b}r$$

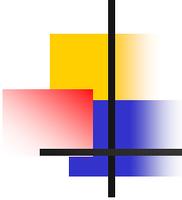
$$\dot{\hat{\mathbf{x}}} = (\mathbf{A} - \mathbf{l}\mathbf{c})\hat{\mathbf{x}} + \mathbf{b}(r - \mathbf{k}\mathbf{x}) + \mathbf{l}\mathbf{c}\mathbf{x}$$

Matrix form:

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

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





State Feedback from Estimated States

Separation Property

By selecting a equivalence transformation

$$\begin{bmatrix} \mathbf{x} \\ \mathbf{e} \end{bmatrix} = \begin{bmatrix} \mathbf{x} \\ \mathbf{x} - \hat{\mathbf{x}} \end{bmatrix} = \begin{bmatrix} \mathbf{I} & \mathbf{0} \\ \mathbf{I} & -\mathbf{I} \end{bmatrix} \begin{bmatrix} \mathbf{x} \\ \mathbf{x} \end{bmatrix} =: \mathbf{P} \begin{bmatrix} \mathbf{x} \\ \mathbf{x} \end{bmatrix}$$

The equivalent eq. is

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

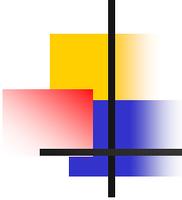
$$y = [\mathbf{c} \quad \mathbf{0}] \begin{bmatrix} \mathbf{x} \\ \mathbf{e} \end{bmatrix}$$

Controllable part is given by

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

Overall transfer function becomes

$$g_f(s) = \mathbf{c}(s\mathbf{I} - \mathbf{A} + \mathbf{b}\mathbf{k})^{-1}\mathbf{b}$$



State Feedback-Multivariable Case

State Feedback-Multivariable Case

Theorem 8.M1

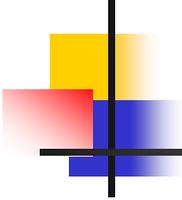
The pair $(\mathbf{A} - \mathbf{BK}, \mathbf{B})$, for any $p \times n$ real constant matrix \mathbf{K} , is controllable if and only if (\mathbf{A}, \mathbf{B}) is controllable.

Proof :

The proof follows the proof of Theorem 8.1.

The only difference is that (8.4) is modified as:

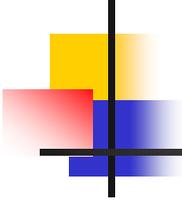
$$C_f = C \begin{bmatrix} \mathbf{I}_p & -\mathbf{KB} & -\mathbf{K}(\mathbf{A} - \mathbf{BK})\mathbf{B} & -\mathbf{K}(\mathbf{A} - \mathbf{BK})^2\mathbf{B} \\ \mathbf{0} & \mathbf{I}_p & -\mathbf{KB} & -\mathbf{K}(\mathbf{A} - \mathbf{BK})\mathbf{B} \\ \mathbf{0} & \mathbf{0} & \mathbf{I}_p & -\mathbf{KB} \\ \mathbf{0} & \mathbf{0} & \mathbf{0} & \mathbf{I}_p \end{bmatrix}$$



State Feedback-Multivariable Case

Theorem 8.M3

All eigenvalues ($\mathbf{A} - \mathbf{BK}$) can be assigned arbitrarily (provided complex conjugate eigenvalues are assigned in pairs) by selecting a real constant \mathbf{K} if and only if (\mathbf{A}, \mathbf{B}) is controllable.



State Feedback-Multivariable Case

Cyclic Design

Theorem 8.7

If the n -dimensional p -input pair (\mathbf{A}, \mathbf{B}) is controllable and if \mathbf{A} is cyclic, then for almost any $p \times 1$ vector \mathbf{v} , the single-input pair $(\mathbf{A}, \mathbf{B}\mathbf{v})$ is controllable.

\mathbf{A} is cyclic if its characteristic polynomial equals its minimal polynomial.

\mathbf{A} is cyclic iff its Jordan form has only one Jordan block for each distinct eigenvalue.

State Feedback-Multivariable Case

Intuitive Validation :

Controllability is invariant under any equivalence transformation, thus we assume \mathbf{A} to be Jordan form

$$\mathbf{A} = \begin{bmatrix} 2 & 1 & 0 & 0 & 0 \\ 0 & 2 & 1 & 0 & 0 \\ 0 & 0 & 2 & 0 & 0 \\ 0 & 0 & 0 & -1 & 1 \\ 0 & 0 & 0 & 0 & -1 \end{bmatrix} \quad \mathbf{B} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \\ 1 & 2 \\ 4 & 3 \\ 1 & 0 \end{bmatrix} \quad \mathbf{B}\mathbf{v} = \mathbf{B} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = \begin{bmatrix} x \\ x \\ \alpha \\ x \\ \beta \end{bmatrix}$$

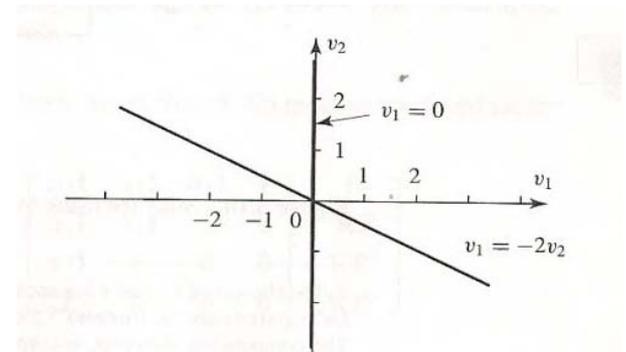
→ $(\mathbf{A}, \mathbf{B}\mathbf{v})$ is controllable iff $\alpha \neq 0, \beta \neq 0$.

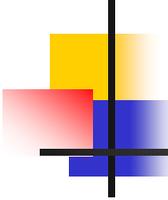
← $\alpha = v_1 + 2v_2 \neq 0$ and $\beta = v_1 \neq 0$

→ *Almost controllable*

$$\mathbf{A} = \begin{bmatrix} 2 & 1 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 2 \end{bmatrix} \quad \mathbf{B} = \begin{bmatrix} 2 & 1 \\ 0 & 2 \\ 1 & 0 \end{bmatrix}$$

→ \exists no \mathbf{v} s.t. $(\mathbf{A}, \mathbf{B}\mathbf{v})$ is controllable





State Feedback-Multivariable Case

Theorem 8.8

If (\mathbf{A}, \mathbf{B}) is controllable, then for almost any $p \times n$ real constant matrix \mathbf{K} , the matrix $(\mathbf{A} - \mathbf{BK})$ has only distinct eigenvalues and is, consequently, cyclic.

Intuitive Verification :

$(\mathbf{A} - \mathbf{BK})$ has

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

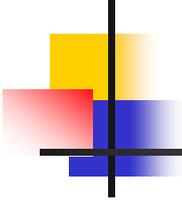
where a_i are functions of the entries of \mathbf{K} . By differentiation

$$\Delta'_f(s) = 4s^3 + 3a_1s^2 + 2a_2s + a_3$$

If $\Delta_f(s)$ has repeated roots, then $\Delta_f(s)$ and $\Delta'_f(s)$ are not coprime.

Then \exists a coprime fraction $\bar{\Delta}'_f(s) / \bar{\Delta}_f(s)$ such that

$$\Delta'_f(s) / \Delta_f(s) = \bar{\Delta}'_f(s) / \bar{\Delta}_f(s)$$



State Feedback-Multivariable Case

The sufficient and necessary condition is

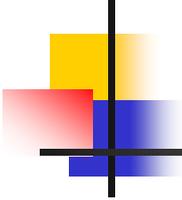
Sylvester resultant is singular

$$\det \begin{bmatrix} a_4 & a_3 & 0 & 0 & 0 & 0 & 0 & 0 \\ a_3 & 2a_2 & a_4 & a_3 & 0 & 0 & 0 & 0 \\ a_2 & 3a_1 & a_3 & 2a_2 & a_4 & a_3 & 0 & 0 \\ a_1 & 4 & a_2 & 3a_1 & a_3 & 2a_2 & a_4 & a_3 \\ 1 & 0 & a_1 & 4 & a_2 & 3a_1 & a_3 & 2a_2 \\ 0 & 0 & 1 & 0 & a_1 & 4 & a_2 & 3a_1 \\ 0 & 0 & 0 & 0 & 1 & 0 & a_1 & 4 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 \end{bmatrix} = b(k_{ij}) = 0$$

The solution space is a line(a very small portion)
in high dimensional space.

Thus Sylvester resultant is almost nonsingular, and

A – BK is almost cyclic.



State Feedback-Multivariable Case

If \mathbf{A} is not cyclic, we can choose $\mathbf{u} = \mathbf{w} - \mathbf{K}_1\mathbf{x}$, then

$$\dot{\mathbf{x}} = (\mathbf{A} - \mathbf{BK}_1)\mathbf{x} + \mathbf{Bw} =: \bar{\mathbf{A}}\mathbf{x} + \mathbf{Bw}$$

where $\bar{\mathbf{A}}$ is cyclic. $(\bar{\mathbf{A}}, \mathbf{B})$ is controllable since (\mathbf{A}, \mathbf{B}) is controllable.

Thus $\exists \mathbf{v}$ such that $(\bar{\mathbf{A}}, \mathbf{Bv})$ is controllable. By choosing

$$\mathbf{w} = \mathbf{r} - \mathbf{K}_2\mathbf{x}, \text{ with } \mathbf{K}_2 = \mathbf{vk}.$$

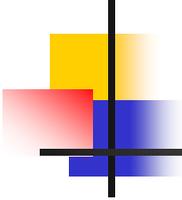
Then the system becomes

$$\dot{\mathbf{x}} = (\bar{\mathbf{A}} - \mathbf{BK}_2)\mathbf{x} + \mathbf{Br} = (\bar{\mathbf{A}} - \mathbf{Bvk})\mathbf{x} + \mathbf{Br}$$

The resulting state feedback control becomes

$$\mathbf{u} = \mathbf{r} - (\mathbf{K}_1 + \mathbf{K}_2)\mathbf{x} =: \mathbf{r} - \mathbf{Kx}.$$

The $\mathbf{K} = \mathbf{K}_1 + \mathbf{K}_2$ can achieve arbitrary eigenvalue assignment.



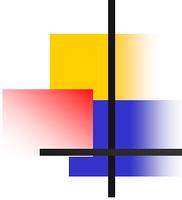
State Feedback-Multivariable Case

Lyapunov-Equation Method

1. Select an $n \times n$ matrix \mathbf{F} with a set of desired eigenvalues that contains no eigenvalues of \mathbf{A} .
2. Select an arbitrary $p \times n$ matrix $\bar{\mathbf{K}}$ such that $(\mathbf{F}, \bar{\mathbf{K}})$ is observable.
3. Solve the unique \mathbf{T} in the Lyapunov equation $\mathbf{AT} - \mathbf{TF} = \mathbf{B}\bar{\mathbf{K}}$.
4. If \mathbf{T} is singular, select a different $\bar{\mathbf{K}}$ and repeat the process. If \mathbf{T} is nonsingular, we compute $\mathbf{K} = \bar{\mathbf{K}}\mathbf{T}^{-1}$, and $(\mathbf{A} - \mathbf{BK})$ has the set of desired eigenvalues.

The Lyapunov equation becomes

$$(\mathbf{A} - \mathbf{BK})\mathbf{T} = \mathbf{TF} \quad \text{or} \quad \mathbf{A} - \mathbf{BK} = \mathbf{TFT}^{-1}$$



State Feedback-Multivariable Case

Theorem 8.M4

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

Proof :

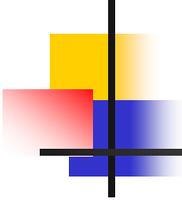
The proof of Theorem 8.4 applies here except that (8.22) must be modified as

$$-\mathbf{T}\Delta(\mathbf{F}) = \begin{bmatrix} \mathbf{B} & \mathbf{AB} & \mathbf{A}^2\mathbf{B} & \mathbf{A}^3\mathbf{B} \end{bmatrix} \begin{bmatrix} \alpha_3\mathbf{I} & \alpha_2\mathbf{I} & \alpha_1\mathbf{I} & \mathbf{I} \\ \alpha_2\mathbf{I} & \alpha_1\mathbf{I} & \mathbf{I} & \mathbf{0} \\ \alpha_1\mathbf{I} & \mathbf{I} & \mathbf{0} & \mathbf{0} \\ \mathbf{I} & \mathbf{0} & \mathbf{0} & \mathbf{0} \end{bmatrix} \begin{bmatrix} \bar{\mathbf{K}} \\ \bar{\mathbf{K}}\mathbf{F} \\ \bar{\mathbf{K}}\mathbf{F}^2 \\ \bar{\mathbf{K}}\mathbf{F}^3 \end{bmatrix}$$

$$-\mathbf{T}\Delta(\mathbf{F}) = \mathbf{C}\Sigma\mathbf{O}$$

where $\Delta(\mathbf{F})$ is nonsingular. If (\mathbf{A}, \mathbf{B}) is uncontrollable or $(\mathbf{F}, \bar{\mathbf{K}})$ is unobservable, then \mathbf{T} is singular.

The contraction statement is true.



State Feedback-Multivariable Case

Canonical Form Method

$$\dot{\bar{\mathbf{x}}} = \begin{bmatrix} -\alpha_{111} & -\alpha_{112} & -\alpha_{113} & -\alpha_{114} & \vdots & -\alpha_{121} & -\alpha_{122} \\ 1 & 0 & 0 & 0 & \vdots & 0 & 0 \\ 0 & 1 & 0 & 0 & \vdots & 0 & 0 \\ 0 & 0 & 1 & 0 & \vdots & 0 & 0 \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ -\alpha_{211} & -\alpha_{212} & -\alpha_{213} & -\alpha_{214} & \vdots & -\alpha_{221} & -\alpha_{222} \\ 0 & 0 & 0 & 0 & \vdots & 1 & 0 \end{bmatrix} \bar{\mathbf{x}} + \begin{bmatrix} 1 & b_{12} \\ 0 & 0 \\ 0 & 0 \\ 0 & 0 \\ \dots & \dots \\ 0 & 1 \\ 0 & 0 \end{bmatrix} \mathbf{u}$$

$$y = \begin{bmatrix} \beta_{111} & \beta_{112} & \beta_{113} & \beta_{114} & \beta_{121} & \beta_{122} \\ \beta_{211} & \beta_{212} & \beta_{213} & \beta_{214} & \beta_{221} & \beta_{222} \end{bmatrix} \bar{\mathbf{x}}$$

State Feedback-Multivariable Case

Effects on Transfer Matrices

$$\mathbf{G}(s) = \mathbf{N}(s)\mathbf{D}^{-1}(s)$$

$$\mathbf{y}(s) = \mathbf{N}(s)\mathbf{D}^{-1}(s)\mathbf{u}(s)$$

$$\mathbf{D}(s)\mathbf{v}(s) = \mathbf{u}(s)$$

$$\mathbf{y}(s) = \mathbf{N}(s)\mathbf{v}(s)$$

$$\mathbf{x}(s) = \mathbf{L}(s)\mathbf{v}(s)$$

$$\mathbf{u}(s) = \mathbf{r}(s) - \mathbf{K}\mathbf{x}(s) = \mathbf{r}(s) - \mathbf{K}\mathbf{L}(s)\mathbf{v}(s)$$

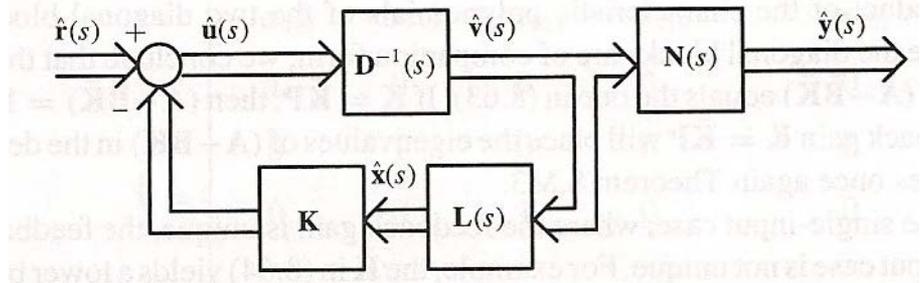
$$\mathbf{D}(s) = \mathbf{D}_{hc}\mathbf{H}(s) + \mathbf{D}_{lc}\mathbf{L}(s)$$

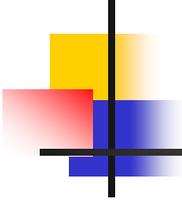
$$[\mathbf{D}_{hc}\mathbf{H}(s) + \mathbf{D}_{lc}\mathbf{L}(s)]\mathbf{v}(s) = \mathbf{r}(s) - \mathbf{K}\mathbf{L}(s)\mathbf{v}(s)$$

$$[\mathbf{D}_{hc}\mathbf{H}(s) + (\mathbf{D}_{lc} + \mathbf{K})\mathbf{L}(s)]\mathbf{v}(s) = \mathbf{r}(s)$$

$$\mathbf{y}(s) = \mathbf{N}(s)[\mathbf{D}_{hc}\mathbf{H}(s) + (\mathbf{D}_{lc} + \mathbf{K})\mathbf{L}(s)]^{-1}\mathbf{r}(s)$$

$$\mathbf{G}_f(s) = \mathbf{N}(s)[\mathbf{D}_{hc}\mathbf{H}(s) + (\mathbf{D}_{lc} + \mathbf{K})\mathbf{L}(s)]^{-1} = \mathbf{N}(s)[\mathbf{D}(s) + \mathbf{K}\mathbf{L}(s)]^{-1}$$





State Estimators-Multivariable Case

State Estimators-Multivariable Case

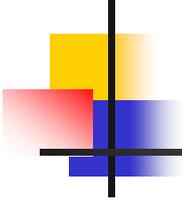
$$\dot{\mathbf{x}} = \mathbf{A}\mathbf{x} + \mathbf{B}\mathbf{u}$$

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

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

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

$$\dot{\mathbf{e}} = (\mathbf{A} - \mathbf{L}\mathbf{C})\mathbf{e}$$



State Estimators-Multivariable Case

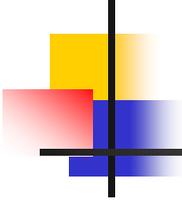
Procedure 8.MR1

Consider the n -dimensional q -output observable pair (\mathbf{A}, \mathbf{C}) . It is assumed that \mathbf{C} has rank q .

1. Select an arbitrary $(n - q) \times (n - q)$ stable matrix \mathbf{F} that has no eigenvalues in common with those \mathbf{A} .
2. Select an arbitrary $(n - q) \times q$ matrix \mathbf{L} such that (\mathbf{F}, \mathbf{L}) is controllable.
3. Solve the unique $(n - q) \times n$ matrix \mathbf{T} in the Lyapunov equation $\mathbf{T}\mathbf{A} - \mathbf{F}\mathbf{T} = \mathbf{L}\mathbf{C}$.
4. If the square matrix of order n

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

is singular, go back to Step 2 and repeat the process.



State Estimators-Multivariable Case

If \mathbf{P} is nonsingular, then the $(n - q)$ -dimensional state equation

$$\dot{\mathbf{z}} = \mathbf{Fz} + \mathbf{TBu} + \mathbf{Ly}$$

$$\mathbf{x} = \begin{bmatrix} \mathbf{C} \\ \mathbf{T} \end{bmatrix}^{-1} \begin{bmatrix} \mathbf{y} \\ \mathbf{z} \end{bmatrix}$$

generates an estimate of \mathbf{x} .

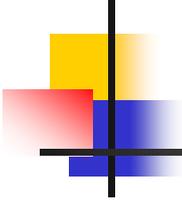
To justify the procedure,

$$\begin{bmatrix} \mathbf{y} \\ \mathbf{z} \end{bmatrix} = \begin{bmatrix} \mathbf{C} \\ \mathbf{T} \end{bmatrix} \mathbf{x}$$

$$\mathbf{e} := \mathbf{z} - \mathbf{T}\mathbf{x}$$

$$\begin{aligned} \dot{\mathbf{e}} &= \dot{\mathbf{z}} - \mathbf{T}\dot{\mathbf{x}} = \mathbf{Fz} + \mathbf{TBu} + \mathbf{LCx} - \mathbf{TAx} - \mathbf{TBu} \\ &= \mathbf{Fz} + (\mathbf{LC} - \mathbf{TA})\mathbf{x} = \mathbf{F}(\mathbf{z} - \mathbf{T}\mathbf{x}) = \mathbf{F}\mathbf{e}. \end{aligned}$$

Since \mathbf{F} is selected as stable, $\mathbf{e} \rightarrow 0$ as $t \rightarrow \infty$.



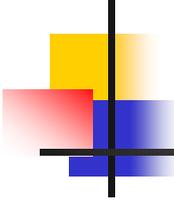
State Estimators-Multivariable Case

Theorem 8.M6

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 $\mathbf{T}\mathbf{A} - \mathbf{F}\mathbf{T} = \mathbf{L}\mathbf{C}$,
is nonsingular only if (\mathbf{A}, \mathbf{C}) is observable and
 (\mathbf{F}, \mathbf{L}) is controllable.



Feedback from Estimated States- Multivariable Case

Feedback from Estimated States-Multivariable Case

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

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

$$\begin{bmatrix} \mathbf{Q}_1 & \mathbf{Q}_2 \end{bmatrix} \begin{bmatrix} \mathbf{C} \\ \mathbf{T} \end{bmatrix} = \mathbf{Q}_1\mathbf{C} + \mathbf{Q}_2\mathbf{T} = \mathbf{I}$$

$$\dot{\mathbf{z}} = \mathbf{F}\mathbf{z} + \mathbf{T}\mathbf{B}\mathbf{u} + \mathbf{L}\mathbf{y}$$

$$\mathbf{x} = \mathbf{Q}_1\mathbf{y} + \mathbf{Q}_2\mathbf{z}$$

$$\mathbf{u} = \mathbf{r} - \mathbf{K}\mathbf{x} = \mathbf{r} - \mathbf{K}\mathbf{Q}_1\mathbf{y} - \mathbf{K}\mathbf{Q}_2\mathbf{z}$$

$$\dot{\mathbf{x}} = \mathbf{A}\mathbf{x} + \mathbf{B}(\mathbf{r} - \mathbf{K}\mathbf{Q}_1\mathbf{C}\mathbf{x} - \mathbf{K}\mathbf{Q}_2\mathbf{z})$$

$$= (\mathbf{A} - \mathbf{B}\mathbf{K}\mathbf{Q}_1\mathbf{C})\mathbf{x} - \mathbf{B}\mathbf{K}\mathbf{Q}_2\mathbf{z} + \mathbf{B}\mathbf{r}$$

$$\dot{\mathbf{z}} = \mathbf{F}\mathbf{z} + \mathbf{T}\mathbf{B}(\mathbf{r} - \mathbf{K}\mathbf{Q}_1\mathbf{C}\mathbf{x} - \mathbf{K}\mathbf{Q}_2\mathbf{z}) + \mathbf{L}\mathbf{C}\mathbf{x}$$

$$= (\mathbf{L}\mathbf{C} - \mathbf{T}\mathbf{B}\mathbf{K}\mathbf{Q}_1\mathbf{C})\mathbf{x} + (\mathbf{F} - \mathbf{T}\mathbf{B}\mathbf{K}\mathbf{Q}_2)\mathbf{z} + \mathbf{T}\mathbf{B}\mathbf{r}$$

Feedback from Estimated States- Multivariable Case

$$\begin{bmatrix} \dot{\mathbf{x}} \\ \dot{\mathbf{z}} \end{bmatrix} = \begin{bmatrix} \mathbf{A} - \mathbf{BKQ}_1\mathbf{C} & -\mathbf{BKQ}_2 \\ \mathbf{LC} - \mathbf{TBKQ}_1\mathbf{C} & \mathbf{F} - \mathbf{TBKQ}_2 \end{bmatrix} \begin{bmatrix} \mathbf{x} \\ \mathbf{z} \end{bmatrix} + \begin{bmatrix} \mathbf{B} \\ \mathbf{TB} \end{bmatrix} \mathbf{r}$$

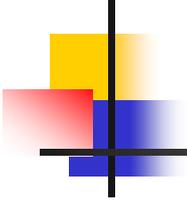
$$\mathbf{y} = [\mathbf{C} \quad \mathbf{0}] \begin{bmatrix} \mathbf{x} \\ \mathbf{z} \end{bmatrix}$$

$$\begin{bmatrix} \mathbf{x} \\ \mathbf{e} \end{bmatrix} = \begin{bmatrix} \mathbf{x} \\ \mathbf{z} - \mathbf{T}\mathbf{x} \end{bmatrix} = \begin{bmatrix} \mathbf{I}_n & \mathbf{0} \\ -\mathbf{T} & \mathbf{I}_{n-q} \end{bmatrix} \begin{bmatrix} \mathbf{x} \\ \mathbf{z} \end{bmatrix}$$

$$\begin{bmatrix} \dot{\mathbf{x}} \\ \dot{\mathbf{e}} \end{bmatrix} = \begin{bmatrix} \mathbf{A} - \mathbf{BK} & -\mathbf{BKQ}_2 \\ \mathbf{0} & \mathbf{F} \end{bmatrix} \begin{bmatrix} \mathbf{x} \\ \mathbf{e} \end{bmatrix} + \begin{bmatrix} \mathbf{B} \\ \mathbf{0} \end{bmatrix} \mathbf{r}$$

$$y = [\mathbf{C} \quad \mathbf{0}] \begin{bmatrix} \mathbf{x} \\ \mathbf{e} \end{bmatrix}$$

$$\mathbf{G}_f(s) = \mathbf{C}(s\mathbf{I} - \mathbf{A} + \mathbf{BK})^{-1}\mathbf{B}$$



HW 8-3

4. Given

$$\mathbf{A} = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ -3 & 1 & 2 & 3 \\ 2 & 1 & 0 & 0 \end{bmatrix}, \quad \mathbf{B} = \begin{bmatrix} 0 & 0 \\ 0 & 0 \\ 1 & 2 \\ 0 & 2 \end{bmatrix}$$

Find two different constant matrices \mathbf{K} such that $(\mathbf{A} - \mathbf{BK})$ has eigenvalues $-4 \pm 3j$ and $-5 \pm 4j$.