

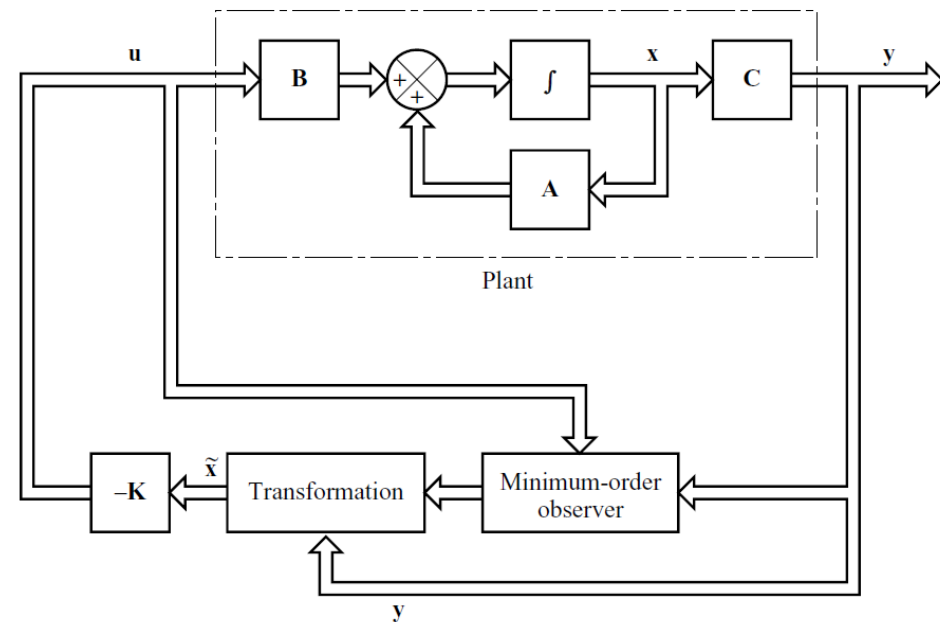
# Reduced-Order Observers

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EPD 30.114 ADVANCED FEEDBACK & CONTROL

# Choosing States to Observer

- Full State Observers are designed to reconstruct all state variables. In practice some of these variables can be accurately measured using sensors (position, speed, temperature, pressure, etc)
  - Such accurately measurable state variables need not be estimated
- For an  $n$ -dimension state vector, and  $m$ -dimension output vector, due to formulation of state-space, the output  $m$  variables are linear combinations of state variables
  - $m$  state measurements do not need to be estimated
  - Only need to estimate  $n-m$  state variables (Reduced-Order Observer)
- If only  $n-m$  states are estimated, the reduced-order observer becomes an  $(n-m)^{\text{th}}$  order observer and in this case is also called a **Minimum-Order Observer (MOO)**



# Conceiving Minimum-Order Observer

- Consider a system with  $n$  state variables ( $\mathbf{x}$ ) and 1 output variable ( $y$ )

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

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

- We can partition the state vector into 2 parts
  - $x_a$  : (scalar) is the measurable portion of the state vector (related to output)
  - $\mathbf{x}_b$ : ( $n-1$  vector) is the unmeasurable portion of the state vector

- The partitioned state and output equations are:

$$\begin{bmatrix} \dot{x}_a \\ \dot{\mathbf{x}}_b \end{bmatrix} = \begin{bmatrix} A_{aa} & \mathbf{A}_{ab} \\ \mathbf{A}_{ba} & \mathbf{A}_{bb} \end{bmatrix} \begin{bmatrix} x_a \\ \mathbf{x}_b \end{bmatrix} + \begin{bmatrix} B_a \\ \mathbf{B}_b \end{bmatrix} u$$

$$y = \begin{bmatrix} 1 & \mathbf{0} \end{bmatrix} \begin{bmatrix} x_a \\ \mathbf{x}_b \end{bmatrix}$$

- Rearranging  $x_a$  state equation

$$\dot{x}_a = A_{aa}x_a + \mathbf{A}_{ab}\mathbf{x}_b + B_a u$$

$$\dot{x}_a - A_{aa}x_a - B_a u = \mathbf{A}_{ab}\mathbf{x}_b$$

Can be measured

Cannot be measured

# Conceiving Minimum-Order Observer

- The state equation for the unmeasurable state:  $\dot{\mathbf{x}}_b = \mathbf{A}_{ba}x_a + \mathbf{A}_{bb}\mathbf{x}_b + \mathbf{B}_bu$
- Design by analogy: Full-order observer Minimum-order observer

State Equation

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

$$\dot{\mathbf{x}}_b = \mathbf{A}_{bb}\mathbf{x}_b + \mathbf{A}_{ba}x_a + \mathbf{B}_bu$$

Output Equation

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

$$\dot{x}_a - A_{aa}x_a - B_a u = \mathbf{A}_{ab}\mathbf{x}_b$$

Full-Order State Observer	Minimum-Order State Observer
$\tilde{\mathbf{x}}$	$\tilde{\mathbf{x}}_b$
$\mathbf{A}$	$\mathbf{A}_{bb}$
$\mathbf{B}u$	$\mathbf{A}_{ba}x_a + \mathbf{B}_bu$
$y$	$\dot{x}_a - A_{aa}x_a - B_a u$
$\mathbf{C}$	$\mathbf{A}_{ab}$
$\mathbf{K}_e$ ( $n \times 1$ matrix)	$\mathbf{K}_e$ [ $(n - 1) \times 1$ matrix]

Observer Dynamics

$$\dot{\tilde{\mathbf{x}}} = (\mathbf{A} - \mathbf{K}_e\mathbf{C})\tilde{\mathbf{x}} + \mathbf{B}u + \mathbf{K}_e y$$

$$\dot{\tilde{\mathbf{x}}}_b = (\mathbf{A}_{bb} - \mathbf{K}_e\mathbf{A}_{ab})\tilde{\mathbf{x}}_b + \mathbf{A}_{ba}x_a + \mathbf{B}_bu + \mathbf{K}_e (\dot{x}_a - A_{aa}x_a - B_a u)$$

# Improving Minimum-Order Observer

- The following expression for the minimum-order observer requires the derivative of  $x_a$ :  $\dot{\tilde{\mathbf{x}}}_b = (\mathbf{A}_{bb} - \mathbf{K}_e \mathbf{A}_{ab}) \tilde{\mathbf{x}}_b + \mathbf{A}_{ba} x_a + \mathbf{B}_b u + \mathbf{K}_e (\dot{x}_a - A_{aa} x_a - B_a u)$ 
  - This is usually not desirable as differentiation amplifies noise

- Eliminate it by re-expressing into

$$y = x_a$$

$$\begin{aligned} \dot{\tilde{\mathbf{x}}}_b - \mathbf{K}_e \dot{x}_a &= (\mathbf{A}_{bb} - \mathbf{K}_e \mathbf{A}_{ab}) \tilde{\mathbf{x}}_b + (\mathbf{A}_{ba} - \mathbf{K}_e A_{aa}) y + (\mathbf{B}_b - \mathbf{K}_e B_a) u \\ &= (\mathbf{A}_{bb} - \mathbf{K}_e \mathbf{A}_{ab}) (\tilde{\mathbf{x}}_b - \mathbf{K}_e y) + (\mathbf{A}_{bb} - \mathbf{K}_e \mathbf{A}_{ab}) \mathbf{K}_e y + (\mathbf{A}_{ba} - \mathbf{K}_e A_{aa}) y + (\mathbf{B}_b - \mathbf{K}_e B_a) u \\ &= (\mathbf{A}_{bb} - \mathbf{K}_e \mathbf{A}_{ab}) (\tilde{\mathbf{x}}_b - \mathbf{K}_e y) + [(\mathbf{A}_{bb} - \mathbf{K}_e \mathbf{A}_{ab}) \mathbf{K}_e + (\mathbf{A}_{ba} - \mathbf{K}_e A_{aa})] y + (\mathbf{B}_b - \mathbf{K}_e B_a) u \end{aligned}$$

- Defining:

$$\begin{aligned} \mathbf{x}_b - \mathbf{K}_e y &= \mathbf{x}_b - \mathbf{K}_e x_a = \boldsymbol{\eta} \\ \tilde{\mathbf{x}}_b - \mathbf{K}_e y &= \tilde{\mathbf{x}}_b - \mathbf{K}_e x_a = \tilde{\boldsymbol{\eta}} \end{aligned}$$

$$\dot{\tilde{\boldsymbol{\eta}}} = (\mathbf{A}_{bb} - \mathbf{K}_e \mathbf{A}_{ab}) \tilde{\boldsymbol{\eta}} + [(\mathbf{A}_{bb} - \mathbf{K}_e \mathbf{A}_{ab}) \mathbf{K}_e + (\mathbf{A}_{ba} - \mathbf{K}_e A_{aa})] y + (\mathbf{B}_b - \mathbf{K}_e B_a) u$$

- Or:

$$\dot{\tilde{\boldsymbol{\eta}}} = \hat{\mathbf{A}} \tilde{\boldsymbol{\eta}} + \hat{\mathbf{B}} y + \hat{\mathbf{F}} u$$

$$\hat{\mathbf{A}} = \mathbf{A}_{bb} - \mathbf{K}_e \mathbf{A}_{ab}$$

$$\hat{\mathbf{B}} = \hat{\mathbf{A}} \mathbf{K}_e + (\mathbf{A}_{ba} - \mathbf{K}_e A_{aa})$$

$$\hat{\mathbf{F}} = \mathbf{B}_b - \mathbf{K}_e B_a$$

# Minimum-Order Observer Error Dynamics

- Recalling  $y$  and estimate of  $x$ :

$$y = \begin{bmatrix} 1 & \mathbf{0} \end{bmatrix} \begin{bmatrix} x_a \\ \mathbf{x}_b \end{bmatrix}$$

$$\tilde{\mathbf{x}} = \begin{bmatrix} x_a \\ \tilde{\mathbf{x}}_b \end{bmatrix} = \begin{bmatrix} y \\ \tilde{\mathbf{x}}_b \end{bmatrix} = \begin{bmatrix} \mathbf{0} \\ \mathbf{I}_{n-1} \end{bmatrix} [\tilde{\mathbf{x}}_b - \mathbf{K}_e y] + \begin{bmatrix} 1 \\ \mathbf{K}_e \end{bmatrix} y = \begin{bmatrix} \mathbf{0} \\ \mathbf{I}_{n-1} \end{bmatrix} \tilde{\boldsymbol{\eta}} + \begin{bmatrix} 1 \\ \mathbf{K}_e \end{bmatrix} y$$

- Making some definitions:

- This describes the relationship between  $\boldsymbol{\eta}$  and  $\mathbf{x}$ :

$$\tilde{\mathbf{x}} = \hat{\mathbf{C}} \tilde{\boldsymbol{\eta}} + \hat{\mathbf{D}} y$$

$$\hat{\mathbf{C}} = \begin{bmatrix} \mathbf{0} \\ \mathbf{I}_{n-1} \end{bmatrix} \quad \hat{\mathbf{D}} = \begin{bmatrix} 1 \\ \mathbf{K}_e \end{bmatrix}$$

- Minimum-order error dynamics:

- Recall these 2 equations and combining:

$$\dot{x}_a - A_{aa}x_a - B_a u = \mathbf{A}_{ab} \mathbf{x}_b$$

$$\dot{\tilde{\mathbf{x}}}_b = (\mathbf{A}_{bb} - \mathbf{K}_e \mathbf{A}_{ab}) \tilde{\mathbf{x}}_b + \mathbf{A}_{ba} x_a + \mathbf{B}_b u + \mathbf{K}_e (\dot{x}_a - A_{aa}x_a - B_a u)$$

$$\boxed{\dot{\tilde{\mathbf{x}}}_b = (\mathbf{A}_{bb} - \mathbf{K}_e \mathbf{A}_{ab}) \tilde{\mathbf{x}}_b + \mathbf{A}_{ba} x_a + \mathbf{B}_b u + \mathbf{K}_e \mathbf{A}_{ab} \mathbf{x}_b}$$

- And the unmeasurable state portion:

$$\dot{\mathbf{x}}_b = \mathbf{A}_{ba} x_a + \mathbf{A}_{bb} \mathbf{x}_b + \mathbf{B}_b u$$

- Creating the error vector:  $\mathbf{e} = \mathbf{x}_b - \tilde{\mathbf{x}}_b = \boldsymbol{\eta} - \tilde{\boldsymbol{\eta}}$

$$\dot{\mathbf{x}}_b - \dot{\tilde{\mathbf{x}}}_b = \mathbf{A}_{ba} x_a + \mathbf{A}_{bb} \mathbf{x}_b + \mathbf{B}_b u - ((\mathbf{A}_{bb} - \mathbf{K}_e \mathbf{A}_{ab}) \tilde{\mathbf{x}}_b + \mathbf{A}_{ba} x_a + \mathbf{B}_b u + \mathbf{K}_e \mathbf{A}_{ab} \mathbf{x}_b)$$

$$\boxed{\dot{\mathbf{x}}_b - \dot{\tilde{\mathbf{x}}}_b = (\mathbf{A}_{bb} - \mathbf{K}_e \mathbf{A}_{ab}) (\mathbf{x}_b - \tilde{\mathbf{x}}_b)}$$

$$\boxed{\dot{\mathbf{e}} = (\mathbf{A}_{bb} - \mathbf{K}_e \mathbf{A}_{ab}) \mathbf{e}}$$

# Minimum-Order Observer Error Dynamics

- The Error Dynamics is:

$$\dot{\mathbf{e}} = (\mathbf{A}_{bb} - \mathbf{K}_e \mathbf{A}_{ab}) \mathbf{e}$$

- $\mathbf{e}$  is a  $(n-1)$  vector

- Again, similarly to the design of the full-order state observer, the **minimum-order system** needs to be completely observable

$$(\mathbf{A}_{bb} - \mathbf{K}_e \mathbf{A}_{ab})^* = (\mathbf{A}_{bb}^* - \mathbf{A}_{ab}^* \mathbf{K}_e^*)$$

- In order to place poles at arbitrary locations:

$$\mathbf{O}_{BM} = \begin{bmatrix} \mathbf{A}_{ab}^* & \mathbf{A}_{bb}^* \mathbf{A}_{ab}^* & \cdots & (\mathbf{A}_{bb}^*)^{n-2} \mathbf{A}_{ab}^* \end{bmatrix} \text{ should be rank } \mathbf{n-1}$$

- Characteristic equation for minimum-order observer:

- c.e. is now  $n-1$  order

$$\begin{aligned} |s\mathbf{I} - (\mathbf{A}_{bb} - \mathbf{K}_e \mathbf{A}_{ab})| &= |s\mathbf{I} - (\mathbf{A}_{bb}^* - \mathbf{A}_{ab}^* \mathbf{K}_e^*)| = |s\mathbf{I} - (\mathbf{A}_{bb}^* - \mathbf{A}_{ab}^* \mathbf{K}_e^*)| \\ &= s^{n-1} + \hat{\alpha}_1 s^{n-2} + \cdots + \hat{\alpha}_{n-2} s + \hat{\alpha}_{n-1} = 0 \end{aligned}$$

- Use Pole-Placement Techniques to determine  $\mathbf{K}_e$

# State Feedback Control with MOO

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- Recall the Separation Principle for state feedback control with full-order observer
  - The closed-loop poles of the system consist of the poles of the state feedback system as well as the reduced-order observer
  - State feedback design and reduced observer design are **independent** of each other
  - Can be designed **separately** and combined
- Combined MOO & system characteristic equation:

$$\left| s\mathbf{I} - \mathbf{A} + \mathbf{BK} \right| \left| s\mathbf{I} - \mathbf{A}_{bb} + \mathbf{K}_e \mathbf{A}_{ab} \right| = 0$$

$$\left| s\mathbf{I} - \mathbf{A} + \mathbf{BK} \right| \left| s\mathbf{I} - \mathbf{A}_{bb}^* + \mathbf{A}_{ab}^* \mathbf{K}_e^* \right| = 0$$



# State Feedback Control with MOO

- The state-space equation for a system under MOO state feedback is:

$$\begin{aligned}
 \dot{\mathbf{x}} &= \mathbf{A}\mathbf{x} + \mathbf{B}u = \mathbf{A}\mathbf{x} + \mathbf{B}(-\mathbf{K}\tilde{\mathbf{x}}) = \mathbf{A}\mathbf{x} - \mathbf{BK} \begin{bmatrix} x_a \\ \tilde{\mathbf{x}}_b \end{bmatrix} = \mathbf{A}\mathbf{x} - \mathbf{BK} \begin{bmatrix} x_a \\ \mathbf{x}_b - \mathbf{e} \end{bmatrix} & u = -\mathbf{K}\tilde{\mathbf{x}} \\
 &= \mathbf{A}\mathbf{x} - \mathbf{BK} \left[ \begin{bmatrix} x_a \\ \mathbf{x}_b \end{bmatrix} - \begin{bmatrix} 0 \\ \mathbf{e} \end{bmatrix} \right] = \mathbf{A}\mathbf{x} - \mathbf{BK} \left[ \mathbf{x} - \begin{bmatrix} 0 \\ \mathbf{e} \end{bmatrix} \right] & = -[K_a \quad \mathbf{K}_b] \begin{bmatrix} x_a \\ \tilde{\mathbf{x}}_b \end{bmatrix} \\
 &= (\mathbf{A} - \mathbf{BK})\mathbf{x} + \mathbf{B}[K_a \quad \mathbf{K}_b] \begin{bmatrix} 0 \\ \mathbf{e} \end{bmatrix}
 \end{aligned}$$

- Remember the error dynamics for MOO:  $\dot{\mathbf{e}} = (\mathbf{A}_{bb} - \mathbf{K}_e \mathbf{A}_{ab})\mathbf{e}$
- Combining and augmenting:

$$\begin{bmatrix} \dot{\mathbf{x}} \\ \dot{\mathbf{e}} \end{bmatrix} = \begin{bmatrix} \mathbf{A} - \mathbf{BK} & \mathbf{BK}_b \\ \mathbf{0} & \mathbf{A}_{bb} - \mathbf{K}_e \mathbf{A}_{ab} \end{bmatrix} \begin{bmatrix} \mathbf{x} \\ \mathbf{e} \end{bmatrix}$$

- This describes the system under minimum-order observer based state feedback control and the MOO's error dynamics

# Transfer Function of MOO Based Controller

- The minimum-order observer dynamics is given by:

$$\dot{\tilde{\eta}} = (\mathbf{A}_{bb} - \mathbf{K}_e \mathbf{A}_{ab}) \tilde{\eta} + [(\mathbf{A}_{bb} - \mathbf{K}_e \mathbf{A}_{ab}) \mathbf{K}_e + (\mathbf{A}_{ba} - \mathbf{K}_e \mathbf{A}_{aa})] y + (\mathbf{B}_b - \mathbf{K}_e \mathbf{B}_a) u$$

- The equations that govern an MOO + SF Control:

$$\begin{cases} \dot{\tilde{\eta}} = \hat{\mathbf{A}} \tilde{\eta} + \hat{\mathbf{B}} y + \hat{\mathbf{F}} u \\ \tilde{\eta} = \tilde{\mathbf{x}}_b - \mathbf{K}_e y \\ u = -\mathbf{K} \tilde{\mathbf{x}} \end{cases}$$

$$\begin{aligned} \hat{\mathbf{A}} &= \mathbf{A}_{bb} - \mathbf{K}_e \mathbf{A}_{ab} \\ \hat{\mathbf{B}} &= \hat{\mathbf{A}} \mathbf{K}_e + (\mathbf{A}_{ba} - \mathbf{K}_e \mathbf{A}_{aa}) \\ \hat{\mathbf{F}} &= \mathbf{B}_b - \mathbf{K}_e \mathbf{B}_a \end{aligned}$$

- Expanding the state feedback and expressing in terms of  $\eta$

$$u = -\mathbf{K} \tilde{\mathbf{x}} = -\begin{bmatrix} K_a & \mathbf{K}_b \end{bmatrix} \begin{bmatrix} y \\ \tilde{\mathbf{x}}_b \end{bmatrix} = -K_a y - \mathbf{K}_b \tilde{\mathbf{x}}_b = -\mathbf{K}_b \tilde{\eta} - (K_a + \mathbf{K}_b \mathbf{K}_e) y$$

- Inserting back to observer dynamics:

$$\dot{\tilde{\eta}} = \hat{\mathbf{A}} \tilde{\eta} + \hat{\mathbf{B}} y + \hat{\mathbf{F}} (-\mathbf{K}_b \tilde{\eta} - (K_a + \mathbf{K}_b \mathbf{K}_e) y)$$

$$\dot{\tilde{\eta}} = (\hat{\mathbf{A}} - \hat{\mathbf{F}} \mathbf{K}_b) \tilde{\eta} + [\hat{\mathbf{B}} - \hat{\mathbf{F}} (K_a + \mathbf{K}_b \mathbf{K}_e)] y$$

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

'traditional' SS system!

$$\tilde{\mathbf{A}} = (\hat{\mathbf{A}} - \hat{\mathbf{F}} \mathbf{K}_b)$$

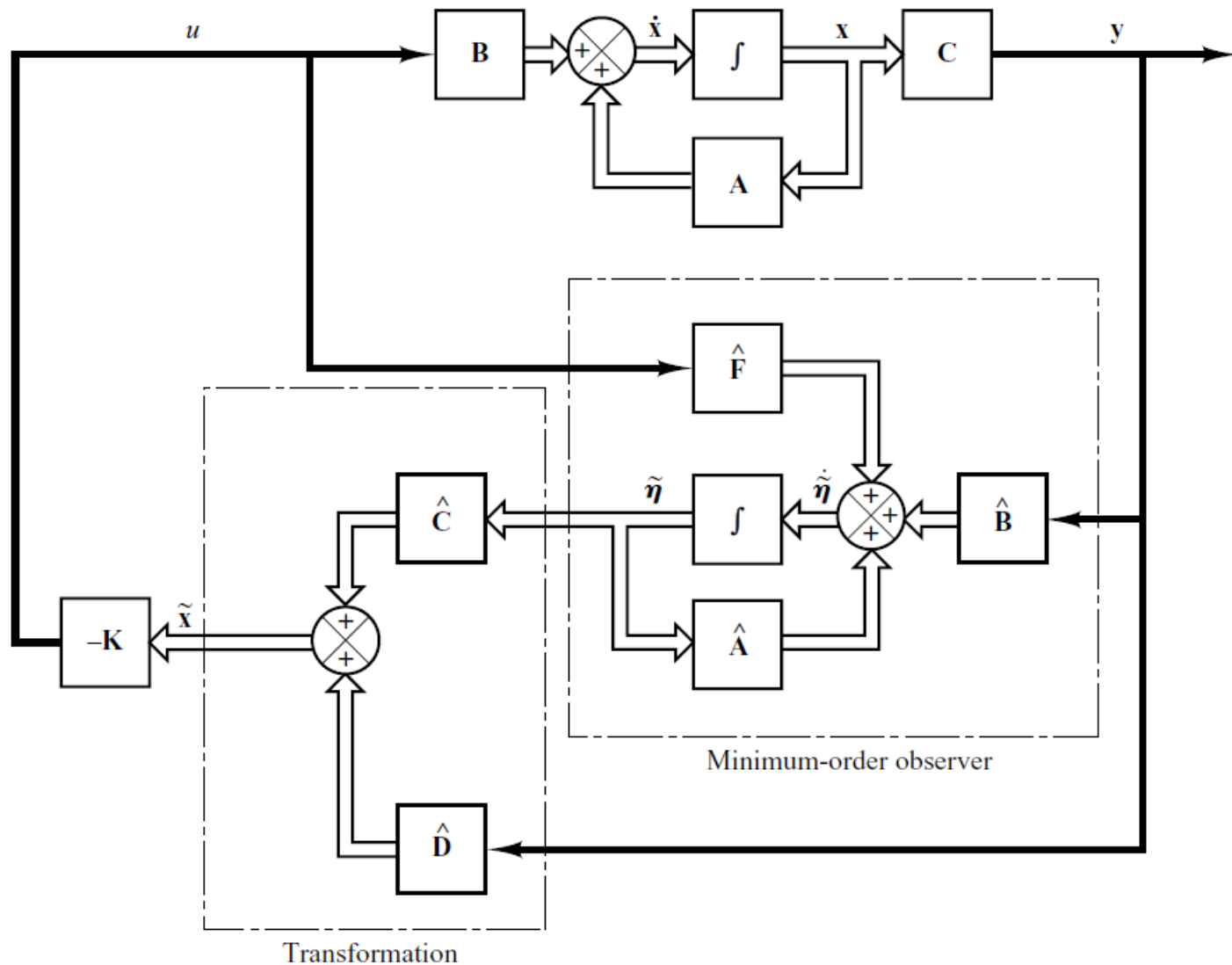
$$\tilde{\mathbf{B}} = [\hat{\mathbf{B}} - \hat{\mathbf{F}} (K_a + \mathbf{K}_b \mathbf{K}_e)]$$

$$\tilde{\mathbf{C}} = -\mathbf{K}_b$$

$$\tilde{D} = -(K_a + \mathbf{K}_b \mathbf{K}_e)$$

TF of MOO controller: 
$$\frac{U(s)}{-Y(s)} = -\left[ \tilde{\mathbf{C}} (s\mathbf{I} - \tilde{\mathbf{A}})^{-1} \tilde{\mathbf{B}} + \tilde{D} \right]$$

# Block Diagram of State Feedback with Minimum-Order Observer



# Exercise

- For the following system, design a state feedback controller that utilizes a minimum-order observer.
  - Desired observer poles:  $s = -10, s = -10$
  - Desired state feedback poles:  $s = -2 + j2\sqrt{3}, s = -2 - j2\sqrt{3}, s = -6$
  - Construct a Block Diagram schematic of the entire system

$$\begin{aligned} \dot{\mathbf{x}} &= \mathbf{A}\mathbf{x} + \mathbf{B}u \\ y &= \mathbf{C}\mathbf{x} \end{aligned} \quad \mathbf{A} = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ -6 & -11 & -6 \end{bmatrix}, \mathbf{B} = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}, \mathbf{C} = [1 \quad 0 \quad 0]$$

$$\begin{bmatrix} \dot{x}_a \\ \dot{\mathbf{x}}_b \end{bmatrix} = \begin{bmatrix} \mathbf{A}_{aa} & \mathbf{A}_{ab} \\ \mathbf{A}_{ba} & \mathbf{A}_{bb} \end{bmatrix} \begin{bmatrix} x_a \\ \mathbf{x}_b \end{bmatrix} + \begin{bmatrix} \mathbf{B}_a \\ \mathbf{B}_b \end{bmatrix} u$$

$$y = [1 \quad \mathbf{0}] \begin{bmatrix} x_a \\ \mathbf{x}_b \end{bmatrix}$$

$$A_{aa} = 0, \mathbf{A}_{ab} = [1 \quad 0], \mathbf{A}_{ba} = \begin{bmatrix} 0 \\ -6 \end{bmatrix}, \mathbf{A}_{bb} = \begin{bmatrix} 0 & 1 \\ -11 & -6 \end{bmatrix}, B_a = 0, \mathbf{B}_b = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

Check state controllability

$$\begin{aligned} \mathbf{C}_o &= [\mathbf{B} \quad \mathbf{AB} \quad \mathbf{A}^2\mathbf{B}] \\ &= \begin{bmatrix} 0 & 0 & 1 \\ 0 & 1 & -6 \\ 1 & -6 & 25 \end{bmatrix} \end{aligned}$$

Controllability matrix is rank 3. (State vector is 3 dimension). System is completely state controllable

Check minimum observability

$$\begin{aligned} \mathbf{O}_{BM} &= [\mathbf{A}_{ab}^* \quad \mathbf{A}_{bb}^* \mathbf{A}_{ab}^*] \\ &= \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \end{aligned}$$

Minimum Observability matrix is rank 2. (unmeasurable state vector is 2 dimension). System is completely minimum-order observable

# Exercise (Minimum-Order Observer Design)

■ Desired c.e:  $(s+10)(s+10) = s^2 + 20s + 100 = 0$   $|s\mathbf{I} - \mathbf{A}_{bb}^* + \mathbf{A}_{ab}^* \mathbf{K}_e^*| = 0$

Using Ackermann's Formula:

$$\mathbf{K}_e^* = [0 \quad 1] [\mathbf{A}_{ab}^* \quad \mathbf{A}_{bb}^* \mathbf{A}_{ab}^*]^{-1} \phi(\mathbf{A}_{bb}^*)$$

$$\mathbf{K}_e = \phi(\mathbf{A}_{bb}) \begin{bmatrix} \mathbf{A}_{ab} \\ \mathbf{A}_{ab} \mathbf{A}_{bb} \end{bmatrix}^{-1} \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

$$\phi(\mathbf{A}_{bb}) = (\mathbf{A}_{bb})^2 + 20\mathbf{A}_{bb} + 100\mathbf{I} = \begin{bmatrix} 89 & 14 \\ -154 & 5 \end{bmatrix}$$

$$\begin{bmatrix} \mathbf{A}_{ab} \\ \mathbf{A}_{ab} \mathbf{A}_{bb} \end{bmatrix}^{-1} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}^{-1} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$\boxed{\mathbf{K}_e = \phi(\mathbf{A}_{bb}) \begin{bmatrix} \mathbf{A}_{ab} \\ \mathbf{A}_{ab} \mathbf{A}_{bb} \end{bmatrix}^{-1} \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 89 & 14 \\ -154 & 5 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 14 \\ 5 \end{bmatrix}}$$

# Exercise (State Feedback Design)

- Desired c.e:  $(s+2-j2\sqrt{3})(s+2+j2\sqrt{3})(s+6) = (s^2+4s+16)(s+6) = s^3+10s^2+40s+96$

$$\mathbf{A} = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ -6 & -11 & -6 \end{bmatrix} \quad \begin{array}{l} a_1 = 6 \\ a_2 = 11 \\ a_3 = 6 \end{array}$$

$$\alpha_1 = 10$$

$$\alpha_2 = 40$$

$$\alpha_3 = 96$$

$$\begin{aligned} \mathbf{K} &= [\alpha_3 - a_3 \quad \alpha_2 - a_2 \quad \alpha_1 - a_1] \\ &= [96 - 6 \quad 40 - 11 \quad 10 - 6] \\ &= [90 \quad 29 \quad 4] \end{aligned}$$

# Block Diagram

