

State Observers

EPD 30.114 ADVANCED FEEDBACK & CONTROL

State Observation & Estimation

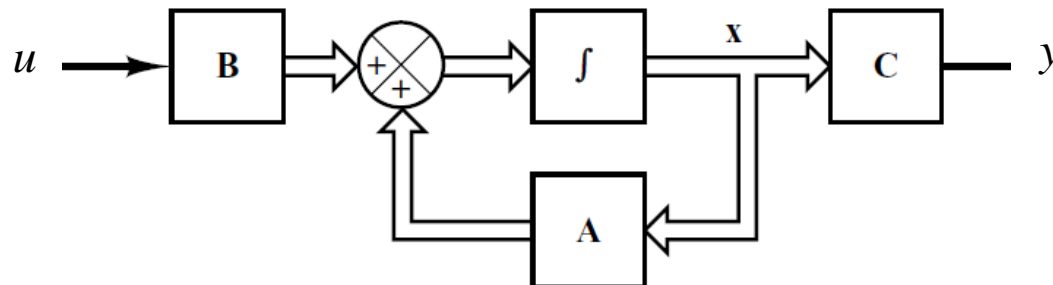
- When designing feedback controllers (Pole-Placement), all state variables were assumed to be present
- In practice, not all state variables are available for feedback or they might be noisy or unreliable
 - Need to estimate unavailable or unknown state variables
- **OBSERVATION:**
 - Process of estimating state variables in the *deterministic* case (e.g. Luenberger Observers)
- **ESTIMATION:**
 - Process of estimate state variables in the *stochastic* case (e.g. Kalman Filters)
- If the state observer observes ALL state variables, it is a **Full-Order State Observer**
 - Many times this is unnecessary as observation is needed only for the unmeasurable states
- An observer that estimates fewer than the dimension of the state vector is a **Reduced-Order State Observer**
 - If the order of the reduced-order state observer is the minimum possible, it is called the **Minimum-Order State Observer**

State Observer (Luenberger Observer)

- The state observer estimates the state variables based on the measurements of the output and control variables
 - It is also called Luenberger Observer in honour of David Luenberger who formulated the concept in 1964
 - Recall the concept of Observability. A state observer can only be designed if and only if the observability condition is satisfied (will be shown)
- Consider the standard plant: $\dot{\mathbf{x}} = \mathbf{A}\mathbf{x} + \mathbf{B}u$
 $y = \mathbf{C}\mathbf{x}$
 - All feedback controllers we are explored so far require the state vector \mathbf{x} to be known
 - Observers attempt to reconstruct the state vector \mathbf{x} through measurement of output y and control input u with help of a 'duplicated' mathematical model of the system
 - They are designed to be asymptotically stable so that the observation error is zero at steady state

$$\mathbf{O}_B = \begin{bmatrix} \mathbf{C}^* & \mathbf{A}^* \mathbf{C}^* & \dots & (\mathbf{A}^*)^{n-1} \mathbf{C}^* \end{bmatrix}$$

Must be at least rank n



Full-Order State Observer

- In addition to duplicating the dynamics of the system, we would like the observer to compensate for inaccuracies in **A** and **B** as well as lack of knowledge of the initial error $e(0)$
 - Let's define $\tilde{\mathbf{x}}$ as the state observation/estimation of the observer
 - An observer gain \mathbf{K}_e is used as a weighting matrix to the observation error: $y - \tilde{y} = y - \mathbf{C}\tilde{\mathbf{x}}$
 - It continuously corrects the model output and improves performance of the observer

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

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

Observer dynamics (inputs: u & y)

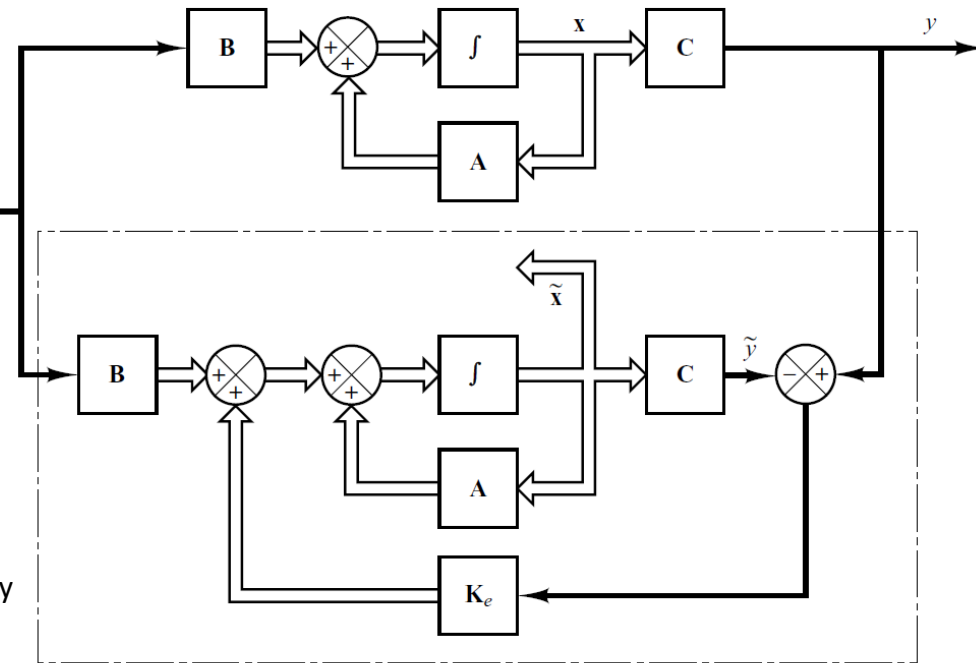
System dynamics u

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

$$= (\mathbf{A} - \mathbf{K}_e \mathbf{C})(\mathbf{x} - \tilde{\mathbf{x}})$$

Defining: $\mathbf{e} = \mathbf{x} - \tilde{\mathbf{x}}$

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



Full-order state observer

- Dynamic performance of the error dynamics governed by eigenvalues of $\mathbf{A} - \mathbf{K}_e \mathbf{C}$
- If $\mathbf{A} - \mathbf{K}_e \mathbf{C}$ is stable, the error will converge to zero for any initial error

Determining Observer Gain

- Now that we have the error dynamics for the full-state observer system, how to solve for \mathbf{K}_e ? $\dot{\mathbf{e}} = (\mathbf{A} - \mathbf{K}_e \mathbf{C}) \mathbf{e}$

- From properties of matrix conjugate transpose:

$$(\mathbf{A} + \mathbf{B})^* = \mathbf{A}^* + \mathbf{B}^*$$

$$(\mathbf{A} - \mathbf{K}_e \mathbf{C})^* = (\mathbf{A}^* - \mathbf{C}^* \mathbf{K}_e^*)$$

$$(\mathbf{AB})^* = \mathbf{B}^* \mathbf{A}^*$$

- In addition, eigenvalues are not affected by conjugate transpose

- Eigenvalues of $(\mathbf{A} - \mathbf{K}_e \mathbf{C})$ are the same with $(\mathbf{A} - \mathbf{K}_e \mathbf{C})^* = (\mathbf{A}^* - \mathbf{C}^* \mathbf{K}_e^*)$
 - And more importantly the characteristic equations are the same:

$$\begin{aligned} |s\mathbf{I} - (\mathbf{A} - \mathbf{K}_e \mathbf{C})| &= |s\mathbf{I} - (\mathbf{A} - \mathbf{K}_e \mathbf{C})^*| \\ &= |s\mathbf{I} - (\mathbf{A}^* - \mathbf{C}^* \mathbf{K}_e^*)| \end{aligned}$$

- Recall that we have been expertly solving using the Pole-Placement to solve equations of the form: $\dot{\mathbf{x}} = (\mathbf{A} - \mathbf{BK}) \mathbf{x}$

- This means you can use the normal **Pole-Placement techniques** to determine the required \mathbf{K}_e for a specified eigenvalues for the observer

- You will have to make the following substitutions: $\dot{\mathbf{e}}^* = \mathbf{e}^* (\mathbf{A}^* - \mathbf{C}^* \mathbf{K}_e^*)$

$$\mathbf{A} = \mathbf{A}^* \quad , \quad \mathbf{B} = \mathbf{C}^* \quad , \quad \mathbf{K} = \mathbf{K}_e^*$$

Going One Step Even Further

- This is a startling result! But we can go deeper!
- Recall when solving $\dot{\mathbf{x}} = (\mathbf{A} - \mathbf{BK})\mathbf{x}$, there was mention on when we could find a \mathbf{K} to place the eigenvalues anywhere we wanted?
 - It was called **State Controllability** and the controllability matrix $\mathbf{C}_o = [\mathbf{B} \quad \mathbf{AB} \quad \dots \quad \mathbf{A}^{n-1}\mathbf{B}]$
- If this analysis was applied to the error dynamics of the observer:

$$\dot{\mathbf{z}} = (\mathbf{A}^* - \mathbf{C}^* \mathbf{K}_e^*) \mathbf{z}$$

- A generic state space system under state feedback that produces this equation is:

$$\dot{\mathbf{z}} = \mathbf{A}^* \mathbf{z} + \mathbf{C}^* \mathbf{v}$$

$$\mathbf{w} = \mathbf{B}^* \mathbf{z}$$

$$\mathbf{v} = -\mathbf{K}_e^* \mathbf{z}$$

- In order to place eigenvalues at specific locations in $\mathbf{A}^* - \mathbf{C}^* \mathbf{K}_e^*$, this system must be completely state controllable
 - The condition is that the matrix should be rank n $\mathbf{C}_o = [\mathbf{C}^* \quad \mathbf{A}^* \mathbf{C}^* \quad \dots \quad (\mathbf{A}^*)^{n-1} \mathbf{C}^*]$
 - This is incidentally the condition for **complete observability** of the ORIGINAL System
 - Hence, in order to design a full state observer for a system, it must be completely observable
 - Recall the concept of Duality. The above generic system is the dual system of the original system

Controllable & Observable Canonical Forms

- Hence, it is also the reason why CCF and OCF are so closely related

$$\begin{aligned}
 \begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \\ \vdots \\ \dot{x}_{n-1} \\ \dot{x}_n \end{bmatrix} &= \begin{bmatrix} 0 & 1 & 0 & \cdots & 0 \\ 0 & 0 & 1 & \cdots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \cdots & 1 \\ -a_n & -a_{n-1} & -a_{n-2} & \cdots & -a_1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_{n-1} \\ x_n \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \\ 1 \end{bmatrix} u \\
 y &= [b_n - a_n b_0 \mid b_{n-1} - a_{n-1} b_0 \mid \cdots \mid b_1 - a_1 b_0] \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ \vdots \\ x_n \end{bmatrix} + b_0 u \\
 &\quad \text{CCF}
 \end{aligned}
 \qquad
 \begin{aligned}
 \begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \\ \vdots \\ \dot{x}_n \end{bmatrix} &= \begin{bmatrix} 0 & 0 & \cdots & 0 & -a_n \\ 1 & 0 & \cdots & 0 & -a_{n-1} \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & \cdots & 1 & -a_1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ \vdots \\ x_n \end{bmatrix} + \begin{bmatrix} b_n - a_n b_0 \\ b_{n-1} - a_{n-1} b_0 \\ \vdots \\ \vdots \\ b_1 - a_1 b_0 \end{bmatrix} u \\
 y &= [0 \quad 0 \quad \cdots \quad 0 \quad 1] \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ \vdots \\ x_n \end{bmatrix} + b_0 u \\
 &\quad \text{OCF}
 \end{aligned}$$

- Recall the expression of CCF allows state-feedback gains to be computed directly during Pole-Placement
- Same approach in OCF allows observer gains to be computed directly during Pole-Placement

Design By Pole-Placement for Observers

- To determine the Observer Gain Matrix, \mathbf{K}_e , the error dynamics is re-expressed into a form that is used extensively in Pole-Placement design for state feedback controllers

$$\dot{\mathbf{e}} = (\mathbf{A} - \mathbf{K}_e \mathbf{C}) \mathbf{e} \Rightarrow \dot{\mathbf{z}} = (\mathbf{A}^* - \mathbf{C}^* \mathbf{K}_e^*) \mathbf{z} \quad \dot{\mathbf{x}} = (\mathbf{A} - \mathbf{B} \mathbf{K}) \mathbf{x}$$

Eigenvalues of $(\mathbf{A} - \mathbf{K}_e \mathbf{C})$ are the same of $(\mathbf{A}^* - \mathbf{C}^* \mathbf{K}_e^*)$ Pole-placement design 'standard' form

- When designing the observer, chose desired eigenvalues such that it responds at least 2 to 5 times faster than the closed-loop system
- There are 3 ways to approach Pole-Placement design for Observer Design:
 - M1: Using the Observable Canonical Form as a start
 - M2: Using direct substitution & comparing with desired c.e. (30.101 method)
 - Because of the direct approach, operating in the original expression is possible: $\dot{\mathbf{e}} = (\mathbf{A} - \mathbf{K}_e \mathbf{C}) \mathbf{e}$
 - Not recommended for high order systems since it is not easily implemented/scalable on a computer program
 - M3: Using Ackermann's Formula

Method 1: Direct Computation from OCF

- Consider the system is in OCF:
- We can choose a set of desired eigenvalues (observer dynamics) at $-\mu_1, -\mu_2, \dots, -\mu_n$.

$$\begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \\ \vdots \\ \dot{x}_n \end{bmatrix} = \begin{bmatrix} 0 & 0 & \cdots & 0 & -a_n \\ 1 & 0 & \cdots & 0 & -a_{n-1} \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & \cdots & 1 & -a_1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} + \begin{bmatrix} b_n - a_n b_0 \\ b_{n-1} - a_{n-1} b_0 \\ \vdots \\ b_1 - a_1 b_0 \end{bmatrix} u$$

- Desired c.e:

$$(s + \mu_1)(s + \mu_2) \cdots (s + \mu_n) = s^n + \alpha_1 s^{n-1} + \cdots + \alpha_{n-1} s + \alpha_n = 0$$

- Define \mathbf{K}_e : $\mathbf{K}_e = \begin{bmatrix} \beta_n \\ \beta_{n-1} \\ \vdots \\ \beta_1 \end{bmatrix}$ $\mathbf{K}_e^* = [\beta_n \quad \beta_{n-1} \quad \cdots \quad \beta_1]$

$$y = [0 \quad 0 \quad \cdots \quad 0 \quad 1] \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_{n-1} \\ x_n \end{bmatrix} + b_0 u$$

- c.e. of system: $\dot{\mathbf{e}} = (\mathbf{A} - \mathbf{K}_e \mathbf{C}) \mathbf{e} \Rightarrow \dot{\mathbf{e}}^* = \mathbf{e}^* (\mathbf{A}^* - \mathbf{C}^* \mathbf{K}_e^*)$

$$|s\mathbf{I} - \mathbf{A}^* + \mathbf{C}^* \mathbf{K}_e^*| = \left| s\mathbf{I} - \begin{bmatrix} 0 & 1 & \cdots & 0 \\ 0 & 0 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ -a_n & -a_{n-1} & \cdots & -a_1 \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix} [\beta_n \quad \beta_{n-1} \quad \cdots \quad \beta_1] \right| = \left| \begin{bmatrix} s & -1 & \cdots & 0 \\ 0 & s & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ a_n + \beta_n & a_{n-1} + \beta_{n-1} & \cdots & s + a_1 + \beta_1 \end{bmatrix} \right|$$

$$|s\mathbf{I} - \mathbf{A}^* + \mathbf{C}^* \mathbf{K}_e^*| = s^n + (a_1 + \beta_1)s^{n-1} + \cdots + (a_{n-1} + \beta_{n-1})s + (a_n + \beta_n) = 0$$

Equating coefficients: $(a_1 + \beta_1) = \alpha_1, \dots, (a_n + \beta_n) = \alpha_n$

$$\mathbf{K}_e^* = [\alpha_n - a_n \quad \alpha_{n-1} - a_{n-1} \quad \cdots \quad \alpha_1 - a_1]$$

Method 2: Direct Substitution Method (30.101)

- If the system is low order ($n \leq 3$), direct substitution of the gain matrix \mathbf{K} into the desired characteristic equation may be simpler
- For a $n=3$ system, the gain may be expressed as:
$$\mathbf{K}_e = \begin{bmatrix} \beta_3 \\ \beta_2 \\ \beta_1 \end{bmatrix}$$
- Substituting directly into desired c.e: (Desired eigenvalues are $-\mu_1, -\mu_2, -\mu_3$)
$$|s\mathbf{I} - \mathbf{A} + \mathbf{K}_e \mathbf{C}| = (s + \mu_1)(s + \mu_2)(s + \mu_3) = 0$$
- Both sides of the c.e. are polynomials in s and the unknown k_1, k_2 and k_3 can be computed by comparing coefficients
- This method is great for $n=2,3$ but is very tedious for higher n
- If system is not completely controllable, the gain matrix cannot be determined (no solution exists).

Method 3: Using Ackermann's Formula

- Adapting Ackermann's approach for the observer dynamics:

$$\dot{\mathbf{e}} = (\mathbf{A} - \mathbf{K}_e \mathbf{C}) \mathbf{e} \Rightarrow \dot{\mathbf{e}}^* = \mathbf{e}^* (\mathbf{A}^* - \mathbf{C}^* \mathbf{K}_e^*) \Rightarrow \dot{\mathbf{z}} = (\mathbf{A}^* - \mathbf{C}^* \mathbf{K}_e^*) \mathbf{z}$$

$$\mathbf{K}_e^* = [0 \quad 0 \quad \dots \quad 1] \left[\mathbf{C}^* \quad \mathbf{A}^* \mathbf{C}^* \quad \dots \quad (\mathbf{A}^*)^{n-1} \mathbf{C}^* \right]^{-1} \phi(\mathbf{A}^*)$$

$$\begin{aligned} \mathbf{K}_e &= \phi(\mathbf{A}^*)^* \begin{bmatrix} \mathbf{C} \\ \mathbf{CA} \\ \vdots \\ \mathbf{CA}^{n-1} \end{bmatrix}^{-1} \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix} \\ &= \phi(\mathbf{A}) \begin{bmatrix} \mathbf{C} \\ \mathbf{CA} \\ \vdots \\ \mathbf{CA}^{n-1} \end{bmatrix}^{-1} \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix} \end{aligned}$$

Exercise

- For the following system, design a full-order state observer where the desired eigenvalues of the observer matrix are at -10 and -10.

$$\begin{aligned}\dot{\mathbf{x}} &= \mathbf{A}\mathbf{x} + \mathbf{B}u \\ y &= \mathbf{C}\mathbf{x}\end{aligned}\quad \mathbf{A} = \begin{bmatrix} 0 & 20 \\ 1 & 0 \end{bmatrix}, \quad \mathbf{B} = \begin{bmatrix} 0 \\ 1 \end{bmatrix} \quad \mathbf{C} = [0 \quad 1]$$

Let's check observability of the system before continuing!

$$\begin{aligned}\mathbf{O}_B &= [\mathbf{C}^* \quad \mathbf{A}^* \mathbf{C}^*] \\ &= \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 20 & 0 \end{bmatrix} \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}\end{aligned}$$

Observability matrix is rank 2. Hence system is completely observable

Desired C.E:

$$(s + \mu_1)(s + \mu_2) = (s + 10)(s + 10) = s^2 + 20s + 100$$

$$\begin{aligned}\alpha_1 &= 20 \\ \alpha_2 &= 100\end{aligned} \quad = s^2 + \alpha_1 s + \alpha_2 = 0$$

$$\mathbf{K}_e = \begin{bmatrix} \beta_2 \\ \beta_1 \end{bmatrix} \quad \mathbf{K}_e^* = [\beta_2 \quad \beta_1]$$

$$\begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \\ \vdots \\ \dot{x}_n \end{bmatrix} = \begin{bmatrix} 0 & 0 & \cdots & 0 & -a_n \\ 1 & 0 & \cdots & 0 & -a_{n-1} \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & \cdots & 1 & -a_1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} + \begin{bmatrix} b_n - a_n b_0 \\ b_{n-1} - a_{n-1} b_0 \\ \vdots \\ b_1 - a_1 b_0 \end{bmatrix} u$$

System already in OCF!

$$\begin{aligned}a_2 &= -20 \\ a_1 &= 0\end{aligned}$$

$$y = [0 \quad 0 \quad \cdots \quad 0 \quad 1] \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_{n-1} \\ x_n \end{bmatrix} + b_0 u$$

$$\dot{\mathbf{e}} = (\mathbf{A} - \mathbf{K}_e \mathbf{C}) \mathbf{e} \Rightarrow \dot{\mathbf{e}}^* = \mathbf{e}^* (\mathbf{A}^* - \mathbf{C}^* \mathbf{K}_e^*)$$

$$\begin{aligned} |s\mathbf{I} - \mathbf{A}^* + \mathbf{C}^* \mathbf{K}_e^*| &= \left| s\mathbf{I} - \begin{bmatrix} 0 & 1 \\ -a_2 & -a_1 \end{bmatrix} + \begin{bmatrix} 0 \\ 1 \end{bmatrix} [\beta_2 \quad \beta_1] \right| = \left| \begin{bmatrix} s & -1 \\ a_2 + \beta_2 & s + a_1 + \beta_1 \end{bmatrix} \right| \\ &= s^2 + (a_1 + \beta_1)s + (a_2 + \beta_2)\end{aligned}$$

$$\begin{aligned}\mathbf{K}_e^* &= [\alpha_2 - a_2 \quad \alpha_1 - a_1] \\ &= [100 + 20 \quad 20 - 0] = [120 \quad 20]\end{aligned}$$

Exercise (Use Direct Substitution)

$$\mathbf{K}_e = \begin{bmatrix} \beta_2 \\ \beta_1 \end{bmatrix}$$

$$|s\mathbf{I} - \mathbf{A} + \mathbf{K}_e\mathbf{C}| = (s+10)(s+10) = s^2 + 20s + 100 = 0$$

$$|s\mathbf{I} - \mathbf{A} + \mathbf{K}_e\mathbf{C}| = \left| s \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} - \begin{bmatrix} 0 & 20 \\ 1 & 0 \end{bmatrix} + \begin{bmatrix} \beta_2 \\ \beta_1 \end{bmatrix} \begin{bmatrix} 0 & 1 \end{bmatrix} \right|$$

$$= \left| \begin{bmatrix} s & 0 \\ 0 & s \end{bmatrix} - \begin{bmatrix} 0 & 20 \\ 1 & 0 \end{bmatrix} + \begin{bmatrix} 0 & \beta_2 \\ 0 & \beta_1 \end{bmatrix} \right|$$

$$= \left| \begin{bmatrix} s & -20 + \beta_2 \\ -1 & s + \beta_1 \end{bmatrix} \right|$$

$$= s^2 + \beta_1 s + (-20 + \beta_2)$$

$$\beta_1 = 20$$

$$\beta_2 = 120$$

$$\boxed{\mathbf{K}_e = \begin{bmatrix} 120 \\ 20 \end{bmatrix}}$$

Exercise (Use Ackermann's Formula)

$$\dot{\mathbf{e}} = (\mathbf{A} - \mathbf{K}_e \mathbf{C})\mathbf{e} \Rightarrow \dot{\mathbf{e}}^* = \mathbf{e}^* (\mathbf{A}^* - \mathbf{C}^* \mathbf{K}_e^*) \Rightarrow \dot{\mathbf{z}} = (\mathbf{A}^* - \mathbf{C}^* \mathbf{K}_e^*) \mathbf{z}$$

$$\mathbf{K}_e = \phi(\mathbf{A}) \begin{bmatrix} \mathbf{C} \\ \mathbf{CA} \end{bmatrix}^{-1} \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

Desired C.E: $(s + \mu_1)(s + \mu_2) = (s + 10)(s + 10) = s^2 + 20s + 100$ $\phi(s) = s^2 + 20s + 100$

$$= s^2 + \alpha_1 s + \alpha_2 = 0$$

$$\begin{aligned} \phi(\mathbf{A}) = \mathbf{A}^2 + 20\mathbf{A} + 100\mathbf{I} &= \begin{bmatrix} 0 & 20 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 0 & 20 \\ 1 & 0 \end{bmatrix} + 20 \begin{bmatrix} 0 & 20 \\ 1 & 0 \end{bmatrix} + 100 \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} 20 & 0 \\ 0 & 20 \end{bmatrix} + \begin{bmatrix} 0 & 400 \\ 20 & 0 \end{bmatrix} + \begin{bmatrix} 100 & 0 \\ 0 & 100 \end{bmatrix} \\ &= \begin{bmatrix} 120 & 400 \\ 20 & 120 \end{bmatrix} \end{aligned}$$

$$\begin{bmatrix} \mathbf{C} \\ \mathbf{CA} \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 0 & 20 \\ 1 & 0 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \quad \begin{bmatrix} \mathbf{C} \\ \mathbf{CA} \end{bmatrix}^{-1} = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}^{-1} = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$$

$$\begin{aligned} \mathbf{K}_e &= \begin{bmatrix} 120 & 400 \\ 20 & 120 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 0 \\ 1 \end{bmatrix} \\ &= \begin{bmatrix} 120 & 400 \\ 20 & 120 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 120 \\ 20 \end{bmatrix} \end{aligned}$$

Visualizing the State Observer

- Let's start with our system:
- $$\dot{\mathbf{x}} = \mathbf{A}\mathbf{x} + \mathbf{B}u$$
- $$y = \mathbf{C}\mathbf{x}$$
- $$\mathbf{A} = \begin{bmatrix} 0 & -5 \\ 1 & -4 \end{bmatrix}, \quad \mathbf{B} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}, \quad \mathbf{C} = \begin{bmatrix} 0 & 1 \end{bmatrix}$$
- We'll like to see how the observer poles affect the state observer performance
 - Try to design two observers: One with poles at -10, -10 and other at -1, -1.
 - This system has stable eigenvalues $(-2 \pm j)$

Let's check observability of the system before continuing!

$$\mathbf{O}_B = [\mathbf{C}^* \quad \mathbf{A}^* \mathbf{C}^*]$$

$$= \begin{bmatrix} 0 & 1 \\ 1 & -4 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 1 & -4 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 1 & -4 \end{bmatrix}$$

System already in OCF! $a_2 = 5$
 $a_1 = 4$

Observability matrix is rank 2. Hence system is completely observable

Desired C.E (poles @ -10,-10:

$$(s + \mu_1)(s + \mu_2) = (s + 10)(s + 10) = s^2 + 20s + 100$$

$$= s^2 + \alpha_1 s + \alpha_2 = 0$$

$$\alpha_1 = 20$$

$$\alpha_2 = 100$$

$$\mathbf{K}_e^* = \begin{bmatrix} \alpha_2 - a_2 & \alpha_1 - a_1 \end{bmatrix}$$

$$= \begin{bmatrix} 100 - 5 & 20 - 4 \end{bmatrix} = \begin{bmatrix} 95 & 16 \end{bmatrix}$$

Desired C.E (poles @ -1,-1:

$$(s + \mu_1)(s + \mu_2) = (s + 1)(s + 1) = s^2 + 2s + 1$$

$$= s^2 + \alpha_1 s + \alpha_2 = 0$$

$$\alpha_1 = 2$$

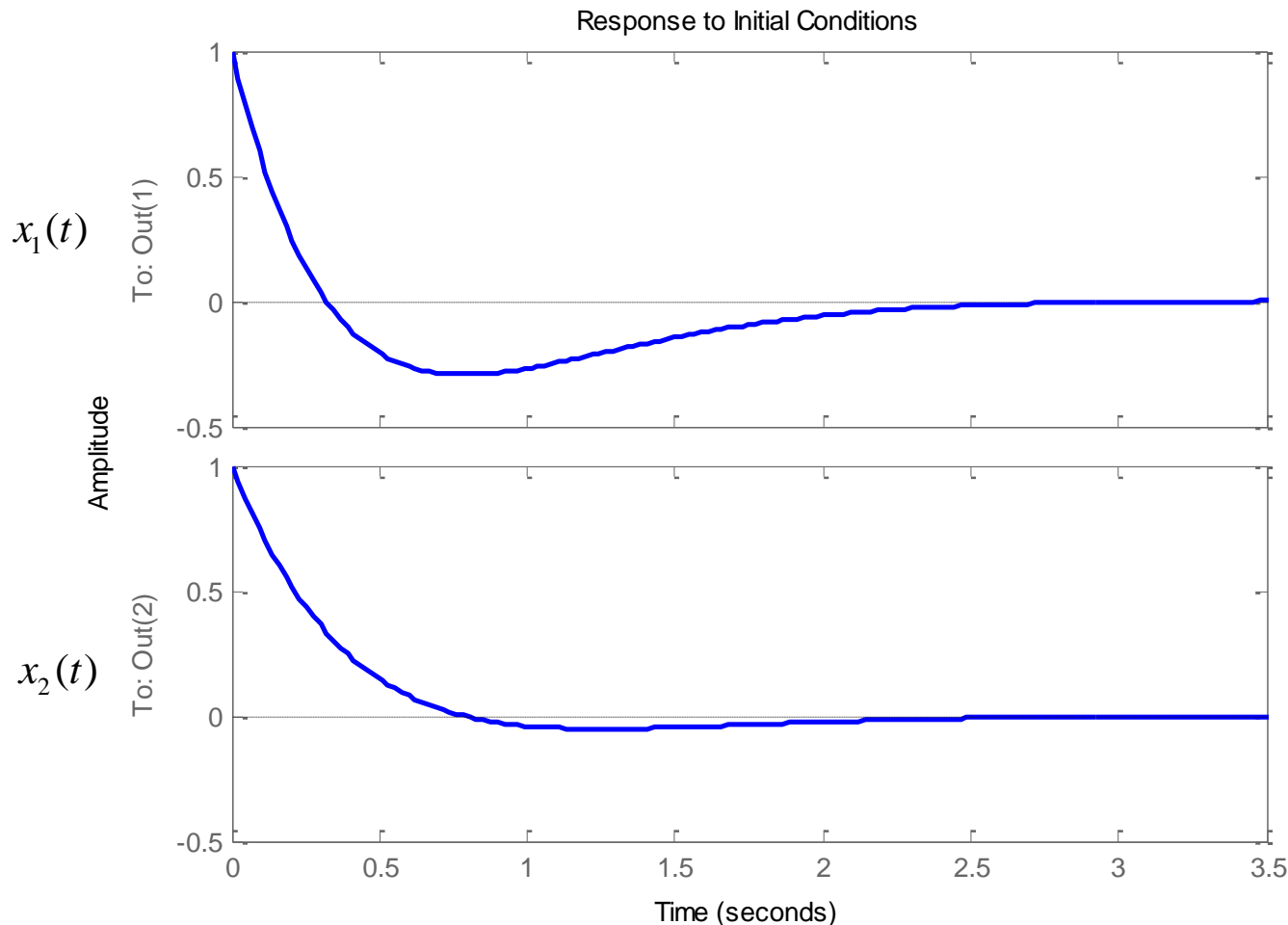
$$\alpha_2 = 1$$

$$\mathbf{K}_e^* = \begin{bmatrix} \alpha_2 - a_2 & \alpha_1 - a_1 \end{bmatrix}$$

$$= \begin{bmatrix} 1 - 5 & 2 - 4 \end{bmatrix} = \begin{bmatrix} -4 & -2 \end{bmatrix}$$

Visualizing the State Observer

- First let's see how the system is responding due a specified initial condition: $\mathbf{x}(0)=[1 \ 1]^T$



$$\dot{\mathbf{x}} = \mathbf{A}\mathbf{x} + \mathbf{B}u$$
$$y = \mathbf{C}\mathbf{x}$$
$$\mathbf{A} = \begin{bmatrix} 0 & -5 \\ 1 & -4 \end{bmatrix}, \mathbf{B} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}, \mathbf{C} = \begin{bmatrix} 0 & 1 \end{bmatrix}$$

Visualizing the State Observer

- Now, let's see how the state observer is performing
 - Note that the state observer has no idea on the value of $\mathbf{x}(0)$. So as an initial guess, it will assume its initial observation value is zero. $\tilde{\mathbf{x}}(0) = 0$

