

1. Example

Recall previous example. Eigenvalues of A are 1, 3:

$$A = \begin{bmatrix} 1 & 2 \\ 0 & 3 \end{bmatrix}, \quad b = \begin{bmatrix} 0 \\ 1 \end{bmatrix}.$$
$$A - bK = \begin{bmatrix} 1 & 2 \\ 0 & 3 \end{bmatrix} - \begin{bmatrix} 0 \\ 1 \end{bmatrix} [K_1 \quad K_2] = \begin{bmatrix} 1 & 2 \\ 0 & 3 \end{bmatrix} - \begin{bmatrix} 0 & 0 \\ K_1 & K_2 \end{bmatrix}$$
$$= \begin{bmatrix} 1 & 2 \\ -K_1 & 3 - K_2 \end{bmatrix}$$

If $K_1 = 0.5$, $K_2 = 3.5$,

$$A - bK = \begin{bmatrix} 1 & 2 \\ -0.5 & -0.5 \end{bmatrix}$$

- Eigenvalues are $0.25 \pm 0.6614j$. Stable.

Now use $K = e_n^T \mathcal{C}^{-1} \alpha_c(A)$

2. Example

$$\alpha_c = (z - (0.25 + j0.6614))(z - (0.25 - j0.6614))$$
$$= z^2 - 0.5z + 0.5$$

$$\alpha_c(A) = \begin{bmatrix} 1 & 2 \\ 0 & 3 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ 0 & 3 \end{bmatrix} - 0.5 \begin{bmatrix} 1 & 2 \\ 0 & 3 \end{bmatrix} + 0.5 \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$
$$= \begin{bmatrix} 1 & 8 \\ 0 & 9 \end{bmatrix} - \begin{bmatrix} 0.5 & 1 \\ 0 & 1.5 \end{bmatrix} + \begin{bmatrix} 0.5 & 0 \\ 0 & 0.5 \end{bmatrix} = \begin{bmatrix} 1 & 7 \\ 0 & 8 \end{bmatrix}$$

$$\mathcal{C} = [b \quad Ab] = \begin{bmatrix} 0 & 2 \\ 1 & 3 \end{bmatrix}, \quad \mathcal{C}^{-1} = -\frac{1}{2} \begin{bmatrix} 3 & -2 \\ -1 & 0 \end{bmatrix}$$

$$K = e_n^T \mathcal{C}^{-1} \alpha_c(A) = -\frac{1}{2} [0 \quad 1] \begin{bmatrix} 3 & -2 \\ -1 & 0 \end{bmatrix} \begin{bmatrix} 1 & 7 \\ 0 & 8 \end{bmatrix}$$
$$= -\frac{1}{2} [-1 \quad 0] \begin{bmatrix} 1 & 7 \\ 0 & 8 \end{bmatrix} = \frac{1}{2} [1 \quad 7] = [0.5 \quad 3.5]$$

Agree with the previous values

3. Controllability

- Can place poles at any place, provided $\mathcal{C} = [b \quad Ab \quad \dots \quad A^{n-1}b]$ is nonsingular
- Equivalently, the system can be made arbitrarily fast, by placing poles at suitable locations
- If \mathcal{C} is nonsingular, the system is said to be **Controllable**
- Example of a controllable system:

$$A = \begin{bmatrix} 1 & 2 \\ 0 & 3 \end{bmatrix}, \quad b = \begin{bmatrix} 0 \\ 1 \end{bmatrix}, \quad \mathcal{C} = \begin{bmatrix} 0 & 2 \\ 1 & 3 \end{bmatrix}.$$

\mathcal{C} is nonsingular and hence system is controllable

4. Controllability

- Example of an uncontrollable system:

$$A = \begin{bmatrix} 1 & 2 \\ 0 & 3 \end{bmatrix}, \quad b = \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \quad \mathcal{C} = \begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix}.$$

- \mathcal{C} is singular and hence the system is **uncontrollable**.
- Physical interpretation:

$$\begin{aligned} x(m+1) &= Ax(m) + bu(m) \\ \begin{bmatrix} x_1(m+1) \\ x_2(m+1) \end{bmatrix} &= \begin{bmatrix} 1 & 2 \\ 0 & 3 \end{bmatrix} \begin{bmatrix} x_1(m) \\ x_2(m) \end{bmatrix} + \begin{bmatrix} 1 \\ 0 \end{bmatrix} u \end{aligned}$$

- The control action does not affect x_2 at all!
- Unfortunately, x_2 is also **unstable**.

5. What Happens to Controller if States are Not Measured?

- State feedback that we used earlier:

$$u = -Kx + v.$$

Example

$$A = \begin{bmatrix} 1 & 2 \\ 0 & 3 \end{bmatrix}, \quad b = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$
$$u = -Kx = - \begin{bmatrix} K_1 & K_2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

- Suppose that all states are **not** measured.
- Suppose only one measurement is made.

6. What Happens to Controller if States are Not Measured?

- If only x_1 is measured

$$y = x_1 = \begin{bmatrix} 1 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = cx,$$
$$c = \begin{bmatrix} 1 & 0 \end{bmatrix}$$

- If only x_2 is measured

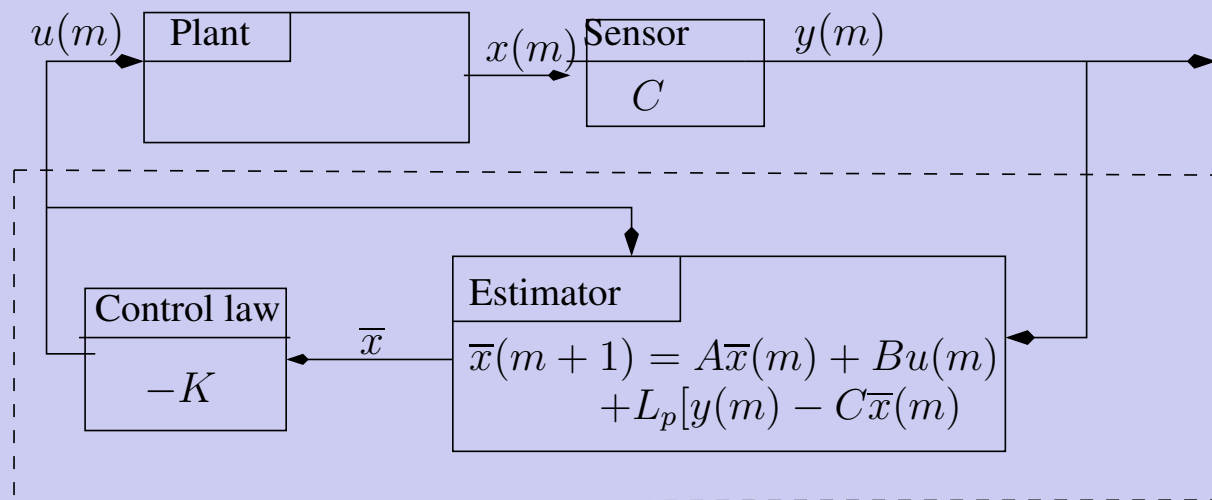
$$y = x_2 = \begin{bmatrix} 0 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = cx,$$
$$c = \begin{bmatrix} 0 & 1 \end{bmatrix}$$

- If a sum is measured

$$y = x_1 + x_2 = \begin{bmatrix} 1 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = cx,$$
$$c = \begin{bmatrix} 1 & 1 \end{bmatrix}$$

- In general, fewer than n states are measured.

7. Use Estimator to Determine States



The structure of the estimator will be explained shortly.

8. States are Not Available - Estimate Them

- If in $x(m+1) = Ax(m) + bu(m) + x_0\delta(m+1)$, not all states are not measured, it is equivalent to not knowing initial state x_0 .

- Because, if the initial state is known,

$$x(1) = Ax(0) + bu(0)$$

$x(1)$ can be calculated.

$$x(2) = Ax(1) + bu(1)$$

$x(2)$ can be calculated. All future states can be calculated recursively.

- All states are not measured = initial state unknown.

9. An Unsuccessful Approach to Estimator Design

- Plant:

$$\begin{aligned}x(m+1) &= Ax(m) + bu(m) + x_0\delta(m+1) \\ y(m) &= cx(m)\end{aligned}$$

- Will the following work as the estimator?

$$\bar{x}(m+1) = A\bar{x}(m) + bu(m) + \bar{x}_0\delta(m+1)$$

where, \bar{x} is a **calculated estimate** of x .

- Subtract and let $e = \bar{x} - x$:

$$e(m+1) = Ae(m) + e_0\delta(m+1)$$

How does this error dynamics behave?

10. Open Loop Analysis: Controller yet to be Designed

Recall \bar{x} , x , $e = \bar{x} - x$:

$$\begin{aligned}\bar{x}(m+1) &= A\bar{x}(m) + bu(m) + \bar{x}_0\delta(m+1) \\ x(m+1) &= Ax(m) + bu(m) + x_0\delta(m+1) \\ e(m+1) &= Ae(m) + e_0\delta(m+1)\end{aligned}$$

- Error dynamics governed by eigenvalues of A only
 - If inside unit circle, e will go to zero
 - e is zero, because it is of the form $e = \bar{x} - x = 0 - 0$
 - If outside, $e(m+1) = \infty - \infty = \infty$, as $k \rightarrow \infty$.
- In fact, we would like $e(m)$ to behave in a nice way: fast rise time, small overshoot, etc.
- Cannot use the above as the estimator.

11. A Potential Estimator?

- Plant:

$$\begin{aligned}x(m+1) &= Ax(m) + bu(m) + x_0\delta(m+1) \\ y(m) &= cx(m)\end{aligned}$$

- Will the following work as the estimator?

$$\begin{aligned}\bar{x}(m+1) &= A\bar{x}(m) + bu(m) + \bar{x}_0\delta(m+1) \\ &\quad + L_p[y(m) - c\bar{x}(m)]\end{aligned}$$

where, L_p is to be calculated.

- Subtract and let $e = \bar{x} - x$:

$$e(m+1) = Ae(m) + e_0\delta(m+1) + L_p[y(m) - c\bar{x}(m)]$$

Can we make the error dynamics behave the way we want?

12. A Potential Estimator?

Recall equations for \bar{x} and x , $e = \bar{x} - x$, $y = cx$

$$\bar{x}(m+1) = A\bar{x}(m) + bu(m) + \bar{x}_0\delta(m+1) + L_p[y(m) - c\bar{x}(m)]$$

$$x(m+1) = Ax(m) + bu(m) + x_0\delta(m+1)$$

$$\begin{aligned}e(m+1) &= Ae(m) + e_0\delta(m+1) + L_p[y(m) - c\bar{x}(m)] \\ &= Ae(m) + e_0\delta(m+1) + L_p[cx(m) - c\bar{x}(m)] \\ &= Ae(m) + e_0\delta(m+1) + L_p c[x(m) - \bar{x}(m)] \\ &= Ae(m) + e_0\delta(m+1) - L_p c e(m) \\ &= (A - L_p c)e(m) + e_0\delta(m+1)\end{aligned}$$

- A is unstable
- Can I choose L_p so as to make $(A - L_p c)$ stable and well behaved?

13. Regulator - Estimator Duality

- Pole placement problem:

$$x(m+1) = Ax(m) + bu(m) + x_0\delta(m+1)$$

We chose K so that $(A - bK)$ is well behaved, that is,

$$|zI - (A - bK)| = \alpha_c(z) = z^n + \alpha_1 z^{n-1} + \dots + \alpha_{n-1} z + \alpha_n =$$

- Solution: $K = e_n^T C^{-1} \alpha_c(A)$.
- Estimator problem:

$$e(m+1) = (A - L_p c)e(m) + e_0\delta(m+1)$$

A not well behaved. Can we choose L_p so that $(A - L_p c)$ is well behaved? i.e.

$$|zI - (A - L_p c)| = \alpha_e(z)$$

Want eigenvalues of $A - L_p c$ to be at desirable locations

14. Regulator - Estimator Duality

- Pole placement problem:

$$x(m+1) = Ax(m) + bu(m) + x_0\delta(m+1)$$

We chose K so that $(A - bK)$ is well behaved.

- Estimator problem:

$$e(m+1) = (A - L_p c)e(m) + e_0\delta(m+1)$$

Can we choose L_p so that $(A - L_p c)$ is well behaved?

- K and L_p are at different locations. Can't use directly.
- But, eigenvalues of $(A - L_p c) =$ eigenvalues of $(A - L_p c)^T$
= eigenvalues of $A^T - c^T L_p^T$.
- Now the unknowns are in the same location!

15. Ackermann's Formula for Estimator Design

To place eigenvalues of $A - bK$ at required locations,

$$\begin{aligned}|zI - (A - bK)| &= \alpha_c(z) \\ &= z^n + \alpha_1 z^{n-1} + \dots + \alpha_{n-1} z + \alpha_n \\ &= (z - \beta_1)(z - \beta_2) \dots (z - \beta_n) \\ K &= e_n^T \mathcal{C}^{-1} \alpha_c(A) \\ \mathcal{C} &= [b \quad Ab \quad \dots \quad A^{n-1}b].\end{aligned}$$

To place eigenvalues of $A^T - c^T L_p^T$ at desirable locations,

$$|zI - (A^T - c^T L_p^T)| = \alpha_e(z),$$

Use the formula,

$$L_p^T = e_n^T [\mathcal{O}^T]^{-1} \alpha_e(A^T)$$

16. Ackermann's Formula for Estimator Design

Estimator design:

$$L_p^T = e_n^T [\mathcal{O}^T]^{-1} \alpha_e(A^T)$$

Assume the following matrix to be nonsingular:

$$\mathcal{O}^T = [c^T \quad A^T c^T \quad \dots \quad (A^T)^{n-1} c^T]$$

Taking transpose of first equation,

$$\begin{aligned}L_p &= \alpha_e(A) \mathcal{O}^{-1} e_n \\ &= \alpha_e(A) \begin{bmatrix} c \\ cA \\ \vdots \\ cA^{n-1} \end{bmatrix}^{-1} e_n\end{aligned}$$