

16.513 Control systems (Lecture #11)

Last time,

- Controllability and observability (Chapter 6)
- Two approaches to state feedback design (Chapter 8)
 - Using controllable canonical form
 - By solving matrix equations

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Today, we continue to work on feedback design (Chapter 8)

- Other state-feedback problems
 - Regulation and tracking
 - Robust tracking and disturbance rejection
 - Stabilization of uncontrollable systems
- Full dimensional estimator
 - SISO case via observable canonical form
 - MIMO case by solving matrix equation

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Project and Final Exam: Due 12pm, Dec 16, 2018

Final exam problems will be sent to your email box at uml, at 9am, Dec 15 (Saturday).

- The written part of the project should be complete with all results clearly presented. 3 points out of 25 will be given on presentation.
- All the Matlab and Simulink files for the project and the final exam should be contained in a zip file for possible verification.
- The project and final exam should be done independently.

Please send 4 files to me via email between 12 - 12:30pm, 12/16/18

- 1) Project; 2) Final exam; 3) Homework #12 (pdf file, or MS word)
- 4) Zip file for all Matlab/Simulink source files

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Regulation and Tracking (§8.3)

- Generally, regulation is about bringing the output or state to certain desired value asymptotically and keep it there. It can be transformed into a stabilization problem
- Tracking is a relationship between the output and reference signals. It describes the property of how the output $y(t)$ follows a desired reference $r(t)$.
 - The simplest tracking problem is to track a step signal.
 - A more complicated case is to track a sinusoidal signal, a polynomial signal, or periodic signals.
 - We need more advanced tools to address the second case.

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Tracking a step signal

Recall that we use $u = r - kx$ to stabilize a system and the resulting closed-loop system is

$$\dot{x} = (A - bk)x + br, \quad y = cx$$

By choosing k appropriately, the transfer function from r to y is

$$g(s) = c(sI - A + bk)^{-1}b = \frac{\beta_1 s^3 + \beta_2 s^2 + \beta_3 s + \beta_4}{s^4 + (\alpha_1 + k_1)s^3 + (\alpha_2 + k_2)s^2 + (\alpha_3 + k_3)s + \alpha_4 + k_4}$$

Suppose that $\beta_4 \neq 0$. Then the DC gain from r to y is

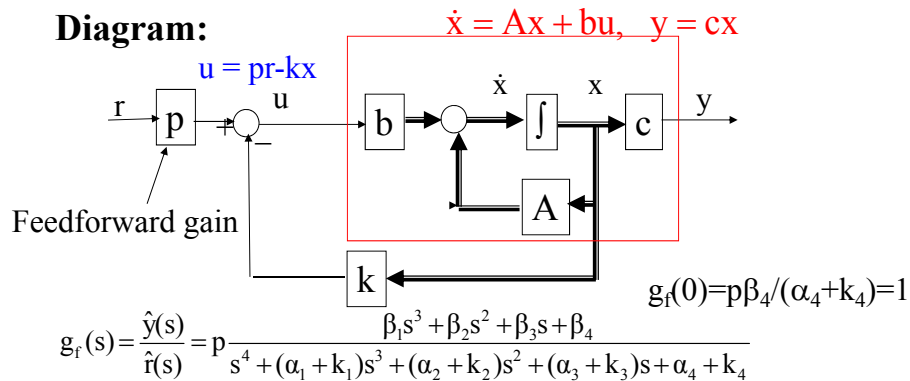
$$g(0) = \beta_4 / (\alpha_4 + k_4). \text{ If } g(0) \text{ is 1 and } r(t) \text{ is a step, then } y(t) - r(t) \rightarrow 0.$$

If $g(0) \neq 1$, we need to introduce a feedforward gain p , i.e., let

$$u = pr - kx, \text{ with } p = (\alpha_4 + k_4) / \beta_4. \text{ Then}$$

$$g_f(s) = \frac{\hat{y}(s)}{\hat{r}(s)} = p \frac{\beta_1 s^3 + \beta_2 s^2 + \beta_3 s + \beta_4}{s^4 + (\alpha_1 + k_1)s^3 + (\alpha_2 + k_2)s^2 + (\alpha_3 + k_3)s + \alpha_4 + k_4}$$

$$g_f(0) = 1 \Rightarrow \text{The output } y(t) \text{ can track any step signal } r(t). \quad 5$$



Comments:

- The feedforward strategy should work well when the parameters are accurate and there is no external disturbance.
- However, if the parameters have errors, $p \neq (\alpha_4 + k_4) / \beta_4$, then the final DC gain may not be exactly one. Also, some disturbance may cause steady-state tracking errors.
- Robust tracking problem is formulated to deal with these issues.

Robust Tracking and disturbance rejection

Consider an open-loop system

$$\dot{x} = Ax + bu + bw, \quad y = cx$$

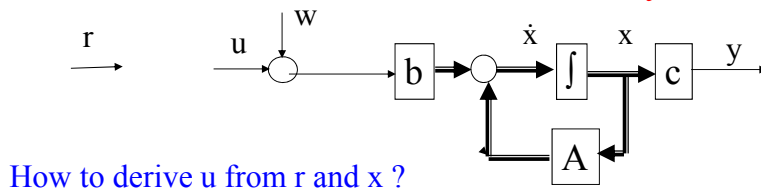
where w is the disturbance. Assume that (A,b) is controllable.

Suppose that there are uncertainties in A , b and c :

$$A \rightarrow A + \delta A, \quad b \rightarrow b + \delta b, \quad c \rightarrow c + \delta c.$$

Robust tracking requires $y(t)$ to follow a step $r(t)$ in the presence of uncertainties and disturbances.

$$\dot{x} = Ax + bu + bw, \quad y = cx$$

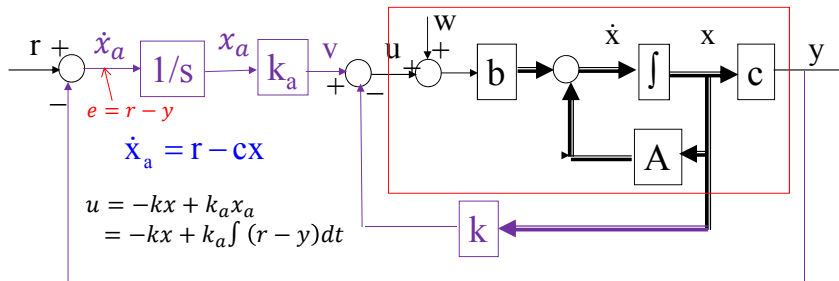


How to derive u from r and x ?

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Proposed configuration

$$\dot{x} = Ax + bu + bw, \quad y = cx$$



$$\dot{x} = Ax - bkx + bk_a x_a + bw = (A - bk)x + bk_a x_a + bw$$

$$\begin{bmatrix} \dot{x} \\ \dot{x}_a \end{bmatrix} = \begin{bmatrix} A - bk & bk_a \\ -c & 0 \end{bmatrix} \begin{bmatrix} x \\ x_a \end{bmatrix} + \begin{bmatrix} 0 \\ 1 \end{bmatrix} r + \begin{bmatrix} b \\ 0 \end{bmatrix} w, \quad y = \begin{bmatrix} c & 0 \end{bmatrix} \begin{bmatrix} x \\ x_a \end{bmatrix}$$

$$\text{Let } A_L = \begin{bmatrix} A & 0 \\ -c & 0 \end{bmatrix}, \quad b_L = \begin{bmatrix} b \\ 0 \end{bmatrix}. \Rightarrow \begin{bmatrix} A - bk & bk_a \\ -c & 0 \end{bmatrix} = A_L - b_L [k \quad -k_a]$$

- If $c(sI-A)^{-1}b$ has no zero at $s=0$, then (A_L, b_L) is controllable
- Then the eigenvalues of $A_L - b_L [k \quad -k_a]$ can be arbitrarily assigned, internal stability can be achieved.

- Now we discuss tracking and disturbance rejection:

$$\text{Let } \Delta_f(s) = \det(sI - A_L + b_L[k \quad -k_a]) = \det \begin{bmatrix} sI - A + bk & -bk_a \\ c & s \end{bmatrix}$$

Suppose $c(sI - A + bk)^{-1}b = N(s)/D(s)$ Then $\Delta_f(s) = sD(s) + k_a N(s)$

$$\text{Let } y(s) = y_r(s) + y_w(s) = \hat{g}_r(s)r(s) + \hat{g}_w(s)w(s)$$

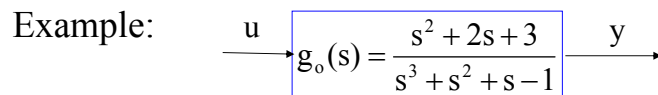
$$\hat{g}_r(s) = \frac{y_r(s)}{r(s)} = \frac{k_a N(s)}{\Delta_f(s)} = \frac{k_a N(s)}{sD(s) + k_a N(s)} \quad \rightarrow \quad \hat{g}_r(0) = 1$$

$$\hat{g}_w(s) = \frac{sN(s)}{\Delta_f(s)}, \quad \hat{g}_w(0) = 0 \quad \text{If } w(t) \text{ is a step signal, } y_w(t) \rightarrow 0$$

Conclusions:

- The DC gain from r to y is always 1. Tracking step signal asymptotically even if parameters A, b, c change.
- The DC gain from w to y is always 0. Step disturbance can be rejected.

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Design a robust tracking control strategy such that y tracks a step signal $r(t)$ asymptotically.

State space realization of $g_o(s)$:

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

$$A_L = \begin{bmatrix} A & 0 \\ -c & 0 \end{bmatrix} = \begin{bmatrix} -1 & -1 & 1 & 0 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ -1 & -2 & -3 & 0 \end{bmatrix}, \quad b_L = \begin{bmatrix} b \\ 0 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

The basic requirement is stability. The convergence rate depends on the eigenvalues of $A_L - b_L k_L$ ($k_L = [k \quad -k_a]$)

We first choose k_L that assign the eigenvalues.

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We use the second approach of pole assignment. Pick

$$F = \begin{bmatrix} -2 & -2 & 0 & 0 \\ 2 & -2 & 0 & 0 \\ 0 & 0 & -4 & -4 \\ 0 & 0 & 4 & -4 \end{bmatrix} \quad \begin{array}{l} \text{Pick } k_0 = [1 \ 1 \ 1 \ 1]; \\ T = \text{lyap}(AL, -F, -bL * k_0) \\ k_L = k_0 * \text{inv}(T) \end{array}$$

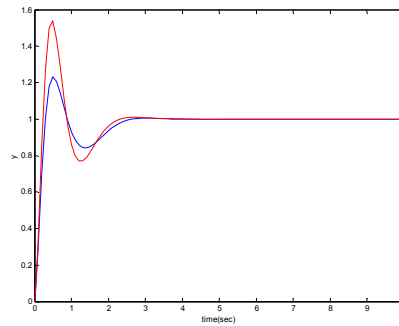
$$k_L = [11.0000 \ -14.3333 \ 22.3333 \ -85.3333] \\ k = [11.0000 \ -14.3333 \ 22.3333], \quad k_a = 85.3333$$

$y(t)$ is plotted in red curve.

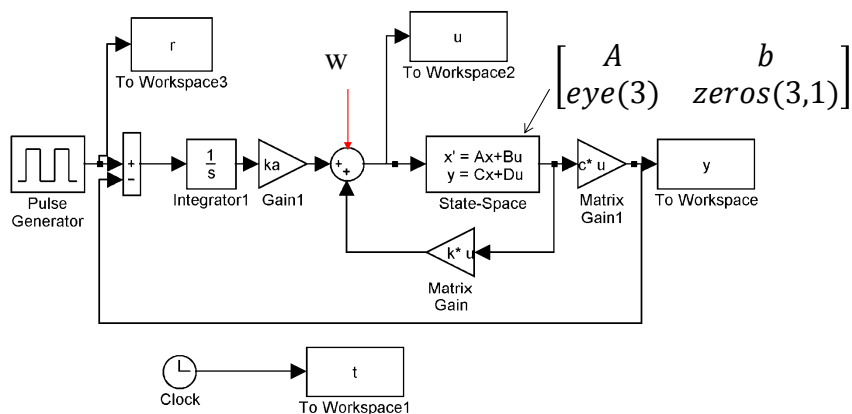
If we pick

$$F = \begin{bmatrix} -2 & -2 & 0 & 0 \\ 2 & -2 & 0 & 0 \\ 0 & 0 & -4 & 0 \\ 0 & 0 & 0 & -8 \end{bmatrix}$$

The tracking performance is improved. See $y(t)$ plotted in blue.

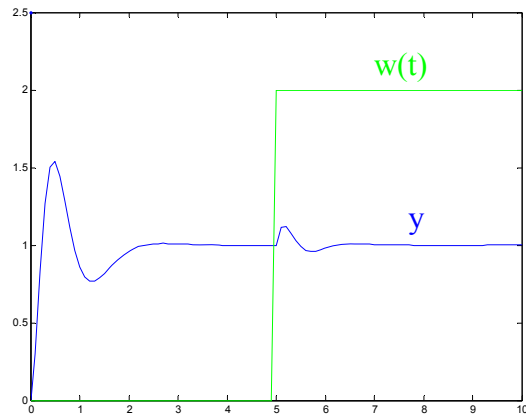


The simulink model



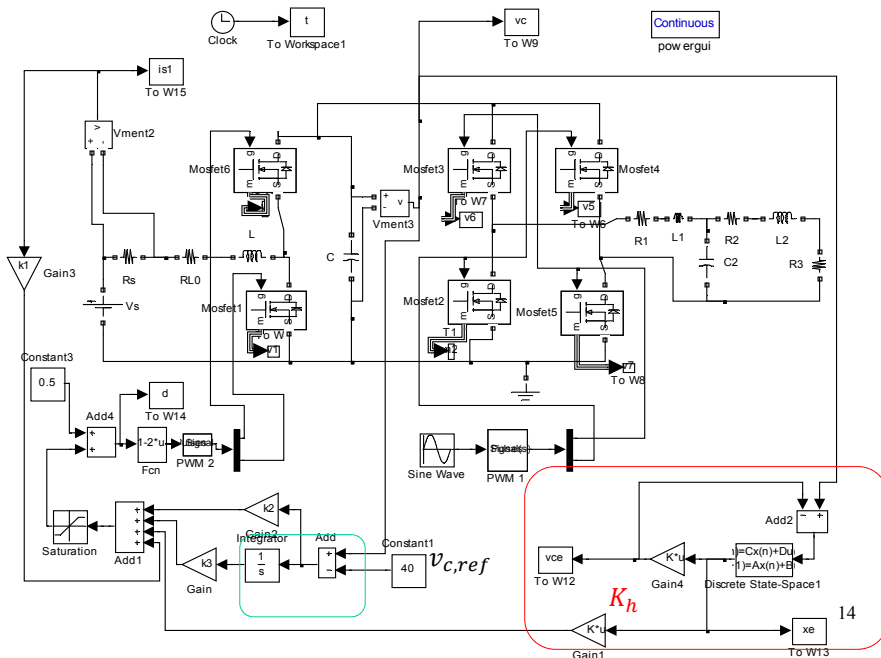
You can try to change the parameters A , or b and the tracking property is maintained.

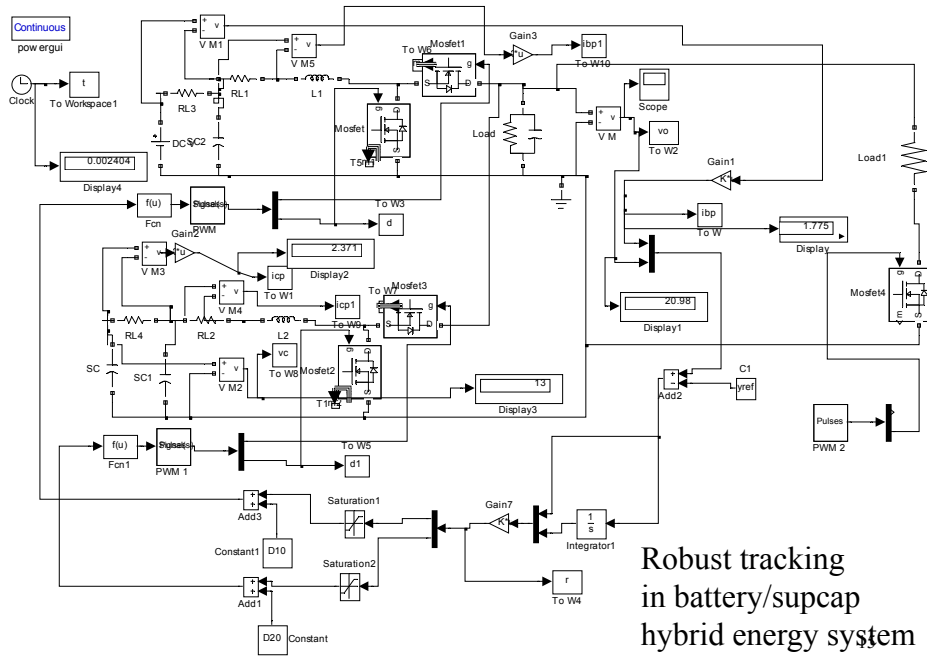
A response in the presence of step disturbance



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Robust tracking in power converter





Robust tracking in battery/supcap hybrid energy system

Problem set #11

1. Design a robust tracking strategy for the system

$$\mathbf{u} \rightarrow \mathbf{g}_0(s) = \frac{s^2 + s + 1}{s^3 - s^2 + 2s - 1} \rightarrow \mathbf{y}$$

so that the output y follows a step signal asymptotically. Choose design parameters so that the closed-loop poles are at $-2+j2$, $-2-j2$, -4 and -8 . Simulate the system from $t=0s$ to $t=20s$. (print the simulink model)

- 1) with the given $\mathbf{g}_0(s)$
- 2) Keep all the design parameters but replace $\mathbf{g}_0(s)$ with

$$\mathbf{g}_1(s) = \frac{s^2 + (1 + \delta)s + 1}{s^3 - s^2 + (2 - \delta)s - 1} \quad \text{for } \delta = 0, 0.5, 1, 2$$

Plot $y(t)$ for each of the cases with 0 initial condition for the state. You may plot all responses in the same figure and identify them with the value of δ .

What is the **minimal** δ to make the closed-loop system unstable?

(Note: use the same state-feedback for all cases.)

Stabilization of uncontrollable systems

Recall: If (A,B) is controllable, then the eigenvalues of $(A+BK)$ can be arbitrarily assigned.

- What if (A,B) is not controllable? Can the system be stabilized?

The original system: $\dot{x} = Ax + Bu$

Suppose that the system is transformed (by $z = Px$) into the following:

$$\dot{z} = \bar{A}z + \bar{B}u, \quad \bar{A} = PAP^{-1}, \quad \bar{B} = PB, \quad z = \begin{bmatrix} z_1 \\ z_2 \end{bmatrix}$$

$$\begin{bmatrix} \dot{z}_1 \\ \dot{z}_2 \end{bmatrix} = \begin{bmatrix} \bar{A}_c & \bar{A}_{12} \\ 0 & \bar{A}_{\bar{c}} \end{bmatrix} \begin{bmatrix} z_1 \\ z_2 \end{bmatrix} + \begin{bmatrix} \bar{B}_c \\ 0 \end{bmatrix} u, \quad (\bar{A}_c, \bar{B}_c) \text{ controllable}$$

Note: $\text{eig}(A) = \text{eig}(\bar{A}) = \text{eig}(\bar{A}_c) \cup \text{eig}(\bar{A}_{\bar{c}})$

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$$\begin{bmatrix} \dot{z}_1 \\ \dot{z}_2 \end{bmatrix} = \begin{bmatrix} \bar{A}_c & \bar{A}_{12} \\ 0 & \bar{A}_{\bar{c}} \end{bmatrix} \begin{bmatrix} z_1 \\ z_2 \end{bmatrix} + \begin{bmatrix} \bar{B}_c \\ 0 \end{bmatrix} u, \quad (\bar{A}_c, \bar{B}_c) \text{ controllable}$$

Note: $\text{eig}(A) = \text{eig}(\bar{A}) = \text{eig}(\bar{A}_c) \cup \text{eig}(\bar{A}_{\bar{c}})$

Consider a state feedback: $u = r - Kx = r - \bar{K}z = r - \begin{bmatrix} \bar{K}_1 & \bar{K}_2 \end{bmatrix} \begin{bmatrix} z_1 \\ z_2 \end{bmatrix}$, $K = \bar{K}P$

$$\rightarrow \begin{bmatrix} \dot{z}_1 \\ \dot{z}_2 \end{bmatrix} = \begin{bmatrix} \bar{A}_c - \bar{B}_c \bar{K}_1 & \bar{A}_{12} - \bar{B}_c \bar{K}_2 \\ 0 & \bar{A}_{\bar{c}} \end{bmatrix} \begin{bmatrix} z_1 \\ z_2 \end{bmatrix} + \begin{bmatrix} \bar{B}_c \\ 0 \end{bmatrix} r,$$

Then $\text{eig}(A - BK) = \text{eig}(\bar{A} - \bar{B}\bar{K}) = \text{eig}(\bar{A}_c - \bar{B}_c \bar{K}_1) \cup \text{eig}(\bar{A}_{\bar{c}})$

Conclusion:

- Since (\bar{A}_c, \bar{B}_c) is controllable, $\text{eig}(\bar{A}_c - \bar{B}_c \bar{K}_1)$ can be arbitrarily assigned
- $\text{eig}(\bar{A}_{\bar{c}})$ cannot be changed by any state feedback. They are called uncontrollable modes.
- For the system to be stabilizable, the uncontrollable modes have to be stable, i.e., $\text{Re}(\lambda_i(\bar{A}_{\bar{c}})) < 0$ for each i .

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Today's topics:

- Other state-feedback problems
 - Regulation and tracking
 - Robust tracking and disturbance rejection
 - Stabilization of uncontrollable systems
- Full dimensional estimator
 - SISO case via observable canonical form
 - MIMO case by solving matrix equation

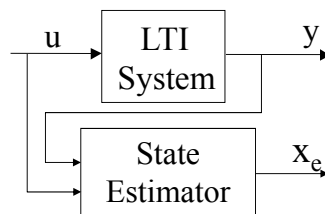
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State Estimators

- Previously, we assumed that x is available.
However, generally x is not available

Q. What to do?

- Construct a system to estimate $x \sim$ **State estimator** (or **observer**)



- Want $x_e(t)$ to be a "good" estimate of $x(t)$, i.e.,
 - $x(t) - x_e(t)$ should go to zero asymptotically with fast convergence rate.

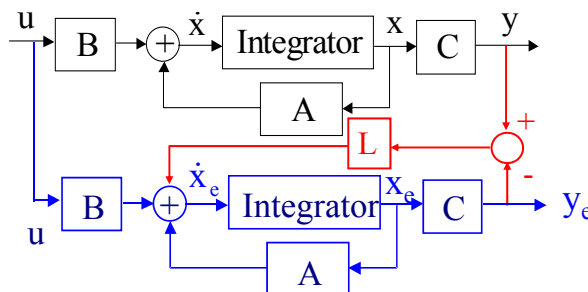
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- **Main result:** If the system is observable, then by properly designing the state estimator, the poles of the dynamics associated with $x(t) - x_e(t)$ can be arbitrarily assigned
 - Error goes to zero as quickly as possible
- We will discuss several types of estimators
 - Full-dimensional state estimator
 - Single output case
 - Multivariable output case
 - Reduced-dimensional state estimator
- Finally, connecting state estimator and state feedback

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Full-Dimensional State Estimators

- How to estimate x based on u and y ?
 $\dot{x} = Ax + Bu$; $y = Cx$ ~ Assuming that $D = 0$
- A proposed configuration:



- Duplicate the system dynamics
- Make correction on dx_e/dt when y and y_e are different
- Pick the correction as a linear function of $(y - y_e)$

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Will the proposed configuration work?

The original system

$$\dot{x} = Ax + Bu; \quad y = Cx$$

The duplicated system (estimator)

$$\dot{x}_e = Ax_e + Bu + L(y - y_e), \quad y_e = Cx_e$$

- The correction term $L(y - y_e)$ plays the essential role.
- Can we choose L appropriately to make x_e approach x ?

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- We now analyze the system

– First, consider the estimator dynamics

$$\begin{aligned} \dot{x}_e &= Ax_e + Bu + L(y - y_e), \quad y_e = Cx_e \\ &= Ax_e + Bu + L[Cx - Cx_e] \\ &= (A - LC)x_e + LCx + Bu \end{aligned}$$

– Next, consider the dynamics of $e := x - x_e$

$$\begin{aligned} \dot{e} &= \dot{x} - \dot{x}_e = (Ax + Bu) - [(A - LC)x_e + LCx + Bu] \\ &= (A - LC)(x - x_e) = (A - LC)e \end{aligned}$$

$$\dot{e} = (A - LC)e$$

- If the eigenvalues of $(A - LC)$ have negative real parts, any error will converge to 0, $x - x_e \rightarrow 0$.
- Can the eigenvalues of $A - LC$ be arbitrarily assigned? 24

Theorem. If the system (A,C) is observable, then all the eigenvalues of (A - LC) can be arbitrarily placed, provided that complex eigenvalues appear in pairs

Proof : The result follows from duality

- The eigenvalues of (A-LC) and (A' - C'L') are the same.
- By Theorem 6.5: (A, C) is observable iff (A', C') is controllable (page 155)
- Now, since (A, C) is observable, (A', C') is controllable
- Eigenvalues of A' - C'L' can be arbitrarily placed

- Note: All the state feedback design procedure can be used to design state estimator, and some of them will be highlighted next

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Full Dimensional State Estimator, SISO Case

Theorem. If a SISO system is **observable**, then it can be transformed, by an **equivalent transformation**, to an **observable canonical form**

- Observable Canonical Form:

$$\dot{x} = \begin{bmatrix} 0 & 0 & 0 & \dots & -\alpha_n \\ 1 & 0 & 0 & \dots & -\alpha_{n-1} \\ 0 & 1 & 0 & \dots & -\alpha_{n-2} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & \dots & 1 & -\alpha_1 \end{bmatrix} x + \begin{bmatrix} \beta_n \\ \beta_{n-1} \\ \beta_{n-2} \\ \vdots \\ \beta_1 \end{bmatrix} u; \quad y = [0 \quad 0 \quad 0 \quad \dots \quad 1] x + du$$

$$c(sI - A)^{-1}b + d = \frac{\beta_1 s^{n-1} + \beta_2 s^{n-2} + \dots + \beta_n}{s^n + \alpha_1 s^{n-1} + \dots + \alpha_{n-1} s + \alpha_n} + d$$

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The transformation matrix Q is formed as follows:

$$q_n \equiv c'$$

$$q_{n-1} \equiv A'q_n + \alpha_1 c' = A'c' + \alpha_1 c'$$

$$q_{n-i} \equiv A'q_{n-i+1} + \alpha_i c' = (A')^{i-1} + \alpha_1 (A')^{i-2} + \dots + \alpha_{i-1} c'$$

$$q_1 \equiv A'q_2 + \alpha_{n-1} c' = (A')^{n-1} + \alpha_1 (A')^{n-2} + \dots + \alpha_{n-1} c'$$

For example, with n=4

$$Q = \begin{bmatrix} (A')^3 c' & (A')^2 c' & A' c' & c' \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ \alpha_1 & 1 & 0 & 0 \\ \alpha_2 & \alpha_1 & 1 & 0 \\ \alpha_3 & \alpha_2 & \alpha_1 & 1 \end{bmatrix}$$

– With $P \equiv Q'$, then (PAP^{-1}, Pb, cP^{-1}) is in the observable canonical form.

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Example $\dot{x} = \begin{bmatrix} 1 & 2 & 0 \\ 3 & -1 & 1 \\ 0 & 2 & 0 \end{bmatrix} x + \begin{bmatrix} 2 \\ 1 \\ 1 \end{bmatrix} u; \quad y = [0 \ 0 \ 1] x$

– Is the system observable?

$$G^o = \begin{bmatrix} C \\ CA \\ CA^2 \end{bmatrix} = \begin{bmatrix} 0 & 0 & 1 \\ 0 & 2 & 0 \\ 6 & -2 & 2 \end{bmatrix}$$

$$|G^o| = -12 \neq 0 \sim \text{Observable}$$

$$\Delta(\lambda) = \begin{vmatrix} \lambda-1 & -2 & 0 \\ -3 & \lambda+1 & -1 \\ 0 & -2 & \lambda \end{vmatrix} = \lambda(\lambda^2 - 1) - 2(\lambda - 1) - 6\lambda$$

$$= \lambda^3 + 0\lambda^2 - 9\lambda + 2 \quad \sim \alpha_1 = 0, \alpha_2 = -9, \alpha_3 = 2$$

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$$q_3 = c' = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}; \quad q_2 = A'q_3 + \alpha_1 c' = \begin{bmatrix} 0 \\ 2 \\ 0 \end{bmatrix}$$

$$q_1 = A'q_2 + \alpha_2 c' = \begin{bmatrix} 6 \\ -2 \\ 2 \end{bmatrix} - \begin{bmatrix} 0 \\ 0 \\ 9 \end{bmatrix} = \begin{bmatrix} -6 \\ -2 \\ -7 \end{bmatrix}$$

$$Q = \begin{bmatrix} 6 & 0 & 0 \\ -2 & 2 & 0 \\ -7 & 0 & 1 \end{bmatrix}; \quad P \equiv Q' = \begin{bmatrix} 6 & -2 & -7 \\ 0 & 2 & 0 \\ 0 & 0 & 1 \end{bmatrix}; \quad \bar{A} = \begin{bmatrix} 0 & 0 & -2 \\ 1 & 0 & 9 \\ 0 & 1 & 0 \end{bmatrix}$$

$$\bar{c} = [0 \ 0 \ 1]; \quad \bar{b} = Pb = \begin{bmatrix} 3 \\ 2 \\ 1 \end{bmatrix}$$

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- Now, how to select L?
 - Consider the equivalent system in Observable Canonical Form:

$$\dot{\bar{x}} = \begin{bmatrix} 0 & 0 & 0 & \dots & -\alpha_n \\ 1 & 0 & 0 & \dots & -\alpha_{n-1} \\ 0 & 1 & 0 & \dots & -\alpha_{n-2} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 1 & -\alpha_1 \end{bmatrix} \bar{x} + \begin{bmatrix} \beta_n \\ \beta_{n-1} \\ \beta_{n-2} \\ \vdots \\ \beta_1 \end{bmatrix} u; \quad y = [0 \ 0 \ 0 \ \dots \ 1] \bar{x}$$

$$\dot{\bar{e}} = (\bar{A} - \bar{L}\bar{c})\bar{e}$$

$$\bar{A} - \bar{L}\bar{c} = \begin{bmatrix} 0 & 0 & 0 & \dots & -\alpha_n \\ 1 & 0 & 0 & \dots & -\alpha_{n-1} \\ 0 & 1 & 0 & \dots & -\alpha_{n-2} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 1 & -\alpha_1 \end{bmatrix} - \begin{bmatrix} \bar{l}_1 \\ \bar{l}_2 \\ \bar{l}_3 \\ \vdots \\ \bar{l}_n \end{bmatrix} [0 \ 0 \ 0 \ \dots \ 1]$$

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$$\bar{A} - \bar{L}\bar{c} = \begin{bmatrix} 0 & 0 & 0 & \dots & -(\alpha_n + \bar{l}_1) \\ 1 & 0 & 0 & \dots & -(\alpha_{n-1} + \bar{l}_2) \\ 0 & 1 & 0 & \dots & -(\alpha_{n-2} + \bar{l}_3) \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & \dots & 1 & -(\alpha_1 + \bar{l}_n) \end{bmatrix}$$

– The characteristic polynomial:

$$\Delta_d(s) = s^n + (\alpha_1 + \bar{l}_n)s^{n-1} + \dots + (\alpha_{n-1} + \bar{l}_2)s + (\alpha_n + \bar{l}_1)$$

– Suppose that the desired estimator poles are $\{\hat{\lambda}_i\}_{i=1}^n$ and the desired characteristic function is

$$\Delta_d(s) = s^n + \hat{\alpha}_1 s^{n-1} + \dots + \hat{\alpha}_{n-1} s + \hat{\alpha}_n$$

Then it is clear that

$$\bar{l}_1 = \hat{\alpha}_n - \alpha_n; \bar{l}_2 = \hat{\alpha}_{n-1} - \alpha_{n-1}; \dots \bar{l}_{n-1} = \hat{\alpha}_2 - \alpha_2; \bar{l}_n = \hat{\alpha}_1 - \alpha_1$$

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$$\bar{L} = \begin{bmatrix} \hat{\alpha}_n - \alpha_n \\ \hat{\alpha}_{n-1} - \alpha_{n-1} \\ \vdots \\ \hat{\alpha}_1 - \alpha_1 \end{bmatrix}$$

– Finally, Let $L = P^{-1} \bar{L}$, then

$$A - LC = P^{-1} \bar{A} P - P^{-1} \bar{L} C P = P^{-1} (\bar{A} - \bar{L} C) P$$

A-LC also has the desired eigenvalues.

– L is uniquely determined by the desired eigenvalues of the estimator.

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Procedure for designing the estimator gain

Step 1. Choose the desired eigenvalue set $\{\lambda_i, i=1,2,\dots,n\}$ of $A-Lc$ and obtain the coefficients of

$$\Delta_d(s) = (s-\lambda_1)(s-\lambda_2)\cdots(s-\lambda_n) = s^n + \bar{\alpha}_1 s^{n-1} + \cdots + \bar{\alpha}_{n-1} s + \bar{\alpha}_n$$

Step 2. Compute the characteristic polynomial of A

$$\Delta(s) = s^n + \alpha_1 s^{n-1} + \cdots + \alpha_{n-1} s + \alpha_n$$

and the transformation matrix, e.g., for $n = 4$

$$Q = \begin{bmatrix} (A')^3 c' & (A')^2 c' & A' c' & c' \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ \alpha_1 & 1 & 0 & 0 \\ \alpha_2 & \alpha_1 & 1 & 0 \\ \alpha_3 & \alpha_2 & \alpha_1 & 1 \end{bmatrix}, \quad P = Q'$$

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$$\text{Then } \bar{A} = PAP^{-1} = \begin{bmatrix} 0 & 0 & 0 & -\alpha_4 \\ 1 & 0 & 0 & -\alpha_3 \\ 0 & 1 & 0 & -\alpha_2 \\ 0 & 0 & 1 & -\alpha_1 \end{bmatrix}, \quad \bar{c} = CP^{-1} = [0 \quad 0 \quad 0 \quad 1],$$

$$\bar{A} - \bar{b}\bar{k} = \begin{bmatrix} 0 & 0 & 0 & -\alpha_4 - \bar{l}_1 \\ 1 & 0 & 0 & -\alpha_3 - \bar{l}_2 \\ 0 & 1 & 0 & -\alpha_2 - \bar{l}_3 \\ 0 & 0 & 1 & -\alpha_1 - \bar{l}_4 \end{bmatrix}$$

Step 3: Choose $\bar{l}_i = \bar{\alpha}_i - \alpha_i$

$$\text{Then } \bar{A} - \bar{L}\bar{c} = \begin{bmatrix} 0 & 0 & 0 & -\bar{\alpha}_4 \\ 1 & 0 & 0 & -\bar{\alpha}_3 \\ 0 & 1 & 0 & -\bar{\alpha}_2 \\ 0 & 0 & 1 & -\bar{\alpha}_1 \end{bmatrix}$$

Step 4: Compute $L = P^{-1}\bar{L}$.

then $A - Lc = P^{-1}(\bar{A} - \bar{L}\bar{c})P$ has the desired eigenvalues $\{\lambda_i, i = 1, 2, \dots, n\}$

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Example (Continued)

$$\dot{x} = \begin{bmatrix} 1 & 2 & 0 \\ 3 & -1 & 1 \\ 0 & 2 & 0 \end{bmatrix} x + \begin{bmatrix} 2 \\ 1 \\ 1 \end{bmatrix} u; \quad y = [0 \quad 0 \quad 1]x$$

- Find L s.t. the estimator poles are all at -3, -3, -10
- As discussed earlier, the system is observable, and

$$\Delta(\lambda) = \lambda^3 + 0\lambda^2 - 9\lambda + 2 \sim \alpha_1 = 0, \alpha_2 = -9, \alpha_3 = 2$$

- The desired estimator characteristic function

$$\Delta_d(s) = (s+3)^2(s+10) = s^3 + 16s^2 + 69s + 90$$

$$\Rightarrow \bar{\alpha}_1 = 16, \bar{\alpha}_2 = 69, \bar{\alpha}_3 = 90; \quad \bar{L} = \begin{bmatrix} \bar{\alpha}_3 - \alpha_3 \\ \bar{\alpha}_2 - \alpha_2 \\ \bar{\alpha}_1 - \alpha_1 \end{bmatrix} = \begin{bmatrix} 88 \\ 78 \\ 16 \end{bmatrix}$$

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- From the previous example,

$$P = Q^T = \begin{bmatrix} 6 & -2 & -7 \\ 0 & 2 & 0 \\ 0 & 0 & 1 \end{bmatrix}; \quad P^{-1} = \begin{bmatrix} 1/6 & 1/6 & 7/6 \\ 0 & 1/2 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$L = P^{-1}\bar{L} = \begin{bmatrix} 1/6 & 1/6 & 7/6 \\ 0 & 1/2 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 88 \\ 78 \\ 16 \end{bmatrix} = \begin{bmatrix} 139/3 \\ 39 \\ 16 \end{bmatrix}$$

- Verification:

$$A - Lc = \begin{bmatrix} 1 & 2 & 0 \\ 3 & -1 & 1 \\ 0 & 2 & 0 \end{bmatrix} - \begin{bmatrix} 139/3 \\ 39 \\ 16 \end{bmatrix} [0 \quad 0 \quad 1] = \begin{bmatrix} 1 & 2 & -139/3 \\ 3 & -1 & -38 \\ 0 & 2 & -16 \end{bmatrix}$$

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$$\begin{aligned}
|\lambda - (A - Lc)| &= \begin{vmatrix} \lambda - 1 & -2 & 139/3 \\ -3 & \lambda + 1 & 38 \\ 0 & -2 & \lambda + 16 \end{vmatrix} \\
&= (\lambda^2 - 1)(\lambda + 16) + 278 + 76(\lambda - 1) - 6(\lambda + 16) \\
&= (\lambda^3 + 16\lambda^2 - \lambda - 16) + 278 + (76\lambda - 76) - (6\lambda + 96) \\
&= \lambda^3 + 16\lambda^2 + 69\lambda + 90 \sim \text{As desired}
\end{aligned}$$

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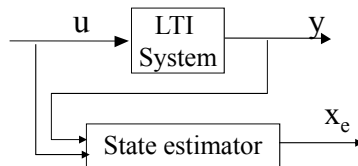
Today's topics:

- Other state-feedback problems
 - Regulation and tracking
 - Robust tracking and disturbance rejection
 - Stabilization of uncontrollable systems
- Full dimensional estimator
 - SISO case via observable canonical form
 - **MIMO case by solving matrix equation**

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General MIMO State Estimator

- Dual to MIMO state feedback ~ Methods discussed earlier has a counterpart here
 - We will discuss the method based on matrix equation, assuming that the system is **observable**



- A full-dimensional state estimator with $D = 0$:

$$\dot{x}_e = Ax_e + Bu + L(y - y_e), \quad y_e = Cx_e$$

Let $e = x - x_e$, the error dynamics:

$$\dot{e} = (A - LC)e$$

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Similar to the state feedback design, the objective is to pick an observer gain L such that $A - LC$ is equivalent to a certain F which has the desired eigenvalues, i.e.,

$$A - LC = T^{-1}FT \text{ for a nonsingular } T$$

$$\Leftrightarrow TA - FT = TLC, \quad \text{Let } L_0 = TL$$

$$\Leftrightarrow TA - FT = L_0C,$$

The procedure: Given A, C and F . Pick L_0 , solve

$$TA - FT = L_0C \text{ for } T.$$

If T is nonsingular, let $L = T^{-1}L_0$. Then $A - LC$ has the desired eigenvalues.

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About the solution to $TA - FT = L_0C$, we have:

- The matrix equation has a unique solution iff A and F have no common eigenvalues;
- The solution T is nonsingular only if (A,C) is observable and (F,L₀) is controllable;
- In case that C has one row, T is nonsingular if and only if (A,C) is observable and (F,L₀) is controllable.

The above result can be derived from the results for state feedback design from duality: taking transpose of the equation, we obtain

$$A'T' - T'F' = C'L_0' \text{ as compared with}$$

$$A T - T F = B K_0 \text{ for the state feedback design}_{41}$$

Example

$$\dot{x} = \begin{bmatrix} 1 & 2 & 0 \\ 3 & -1 & 1 \\ 0 & 2 & 0 \end{bmatrix} x + \begin{bmatrix} 2 \\ 1 \\ 1 \end{bmatrix} u; \quad y = \begin{bmatrix} 0 & 0 & 1 \\ 0 & 1 & 1 \end{bmatrix} x$$

Find T s.t. the estimator poles are at -5, -5, -10

– Check observability:

$$G^o = \begin{bmatrix} C \\ CA \\ CA^2 \end{bmatrix} = \begin{bmatrix} 0 & 0 & 1 \\ 0 & 1 & 1 \\ 0 & 2 & 0 \\ 3 & 1 & 1 \\ x & x & x \\ x & x & x \end{bmatrix}$$

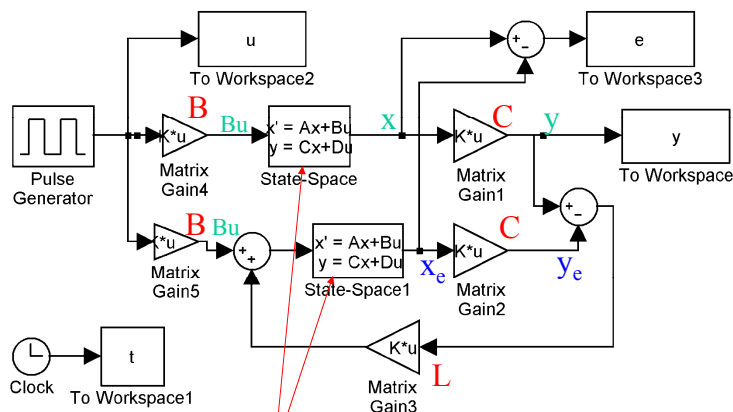
~ Observable

- Select F: $F = \begin{bmatrix} -5 & 0 & 0 \\ 0 & -5 & 0 \\ 0 & 0 & -10 \end{bmatrix}$
- Select L_0 : $L_0 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 2 \end{bmatrix} \Rightarrow \{F, L_0\}$ controllable
- Solve $TA - FT = L_0C$ with matlab: $T = \text{lyap}(-F, A, -L_0 * C)$

$$T = \begin{bmatrix} 0.1250 & -0.2500 & 0.2500 \\ -0.1875 & 0.3750 & 0.1250 \\ -0.0602 & 0.2206 & 0.1779 \end{bmatrix}, \quad L = T^{-1}L_0 = \begin{bmatrix} -3.5 & 24 \\ -3.75 & 15 \\ 3 & 2 \end{bmatrix}$$
- Verify: $\text{eig}(A - LC) = \{-5, -5, -10\}$
- If we pick: $L_0 = \begin{bmatrix} 1 & 0 \\ -1 & 1 \\ 1 & 0 \end{bmatrix} \Rightarrow L = \begin{bmatrix} -117 & 24 \\ -60.5 & 15 \\ 3 & 2 \end{bmatrix}$

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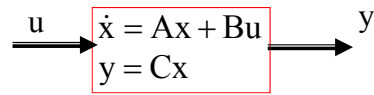
$$\dot{x} = Ax + Bu; \quad y = Cx; \quad \dot{x}_e = Ax_e + Bu + L(y - y_e), \quad y_e = Cx_e$$



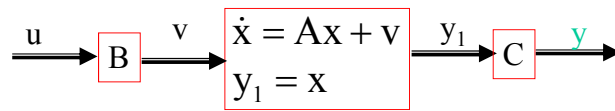
Take A to be the original A, B=I, C=I, D=0
 $B = \text{eye}(3), C = \text{eye}(3), D = \text{zeros}(3);$

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Explanation:

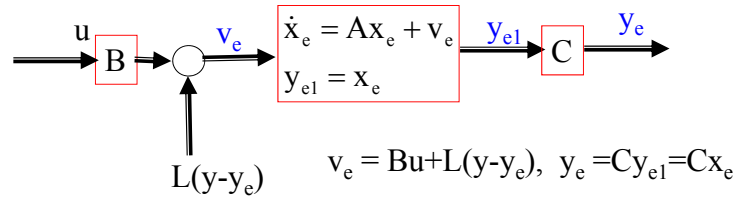


Can be equivalently realized with



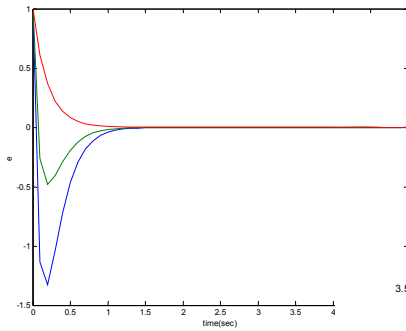
$$v = Bu, \quad y = Cy_1 = Cx$$

The purpose of doing this is to get x and x_e



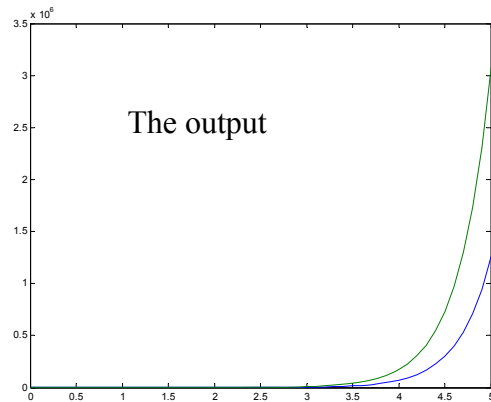
$$v_e = Bu + L(y - y_e), \quad y_e = Cy_{e1} = Cx_e$$

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The estimation error: $x - x_e$

$$x(0) = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}, \quad x_e(0) = 0$$



The output

Observer for discrete-time system

The original system

$$x(k+1) = Ax(k) + Bu(k); \quad y(k) = Cx(k)$$

The duplicated system (estimator)

$$x_e(k+1) = Ax_e(k) + Bu(k) + L(y(k) - y_e(k)), \quad y_e(k) = Cx_e(k)$$

Define the observer error, $e(k) := x(k) - x_e(k)$

$$\begin{aligned} e(k+1) &= (Ax(k) + Bu(k)) - [(A - LC)x_e(k) + LCx(k) + Bu(k)] \\ &= (A - LC)e(k) \end{aligned}$$

$$e(k+1) = (A - LC)e(k)$$

- If the eigenvalues of $(A - LC)$ are inside the unit disk, $|\lambda_i(A - LC)| < 1$, any error will converge to 0, $x(k) - x_e(k) \rightarrow 0$.

Use same algorithms to assign the eigenvalues of $A - LC$.

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Problem set #11

2. For the system

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

Design an observer to estimate the state x . The poles of the observer are $-4+j4$, $-4-j4$ and -8 . Simulate the system from $t=0$ to $t=10$ with $x(0)=[1 \ 1 \ 1]'$, $x_e(0)=[0 \ 0 \ 0]'$ and $u=0$. Plot the state $x(t)$ in one figure and $e = x(t) - x_e(t)$ in another figure. Also print the simulink model.

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Next Time:

- Reduced-order estimator
- Connection of state-feedback with state estimation
- LQR optimal control

Problem set #11: see slides 16, 48

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