
EECS 16B

Spring 2022

Lecture 15

3/8/2022 

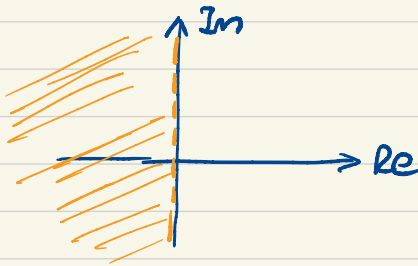
LECTURE 15

- complete feedback stabilization
- controllability

Definition: We say that a system is (bounded-input, bounded state) stable if state x is bounded for any initial condition and bounded disturbance pair. Unstable otherwise, i.e., if x grows unbounded for some initial condition, disturbance.

Stability Criteria:

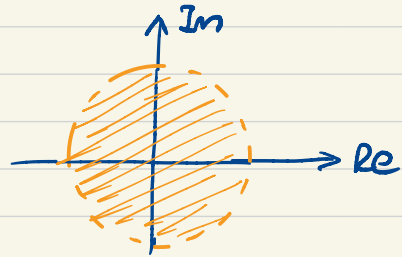
Continuous-time
 $\frac{d}{dt} \vec{x}(t) = A_c \vec{x}(t) + \vec{w}(t)$



$$\text{Re } \lambda_k < 0$$

for each eigenvalue of A_c

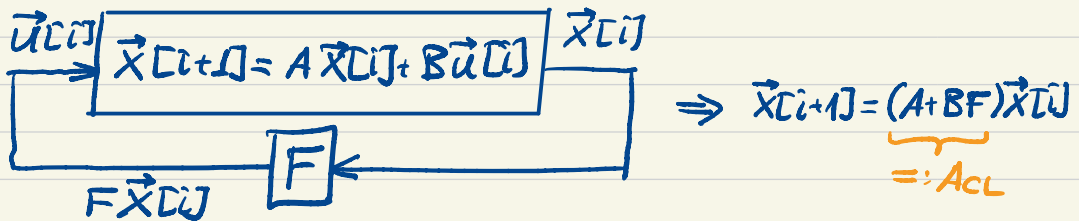
Discrete-time
 $\vec{x}[i+1] = A_d \vec{x}[i] + \vec{w}[i]$



$$|\lambda_k| < 1$$

for each eigenvalue of A_d

Stabilization by Feedback:



We can assign the eigenvalues of A_{CL} arbitrarily with choice of F if we can bring (A, B) to ...

Controller Canonical Form (for scalar u) :

$$TAT^{-1} = \begin{bmatrix} 0 & 1 & 0 & \dots & 0 \\ 0 & 0 & 1 & 0 & \dots \\ & & & \ddots & \\ * & * & & & 1 \\ & & & & * \end{bmatrix} \quad TB = \begin{bmatrix} 0 \\ \vdots \\ 0 \\ 1 \end{bmatrix}$$

Claim: An invertible $n \times n$ matrix T bringing TAT^{-1} and TB to the form above exists if

$$[A^{n-1}B \quad A^{n-2}B \quad \dots \quad AB \quad B]$$

is nonsingular, i.e. the columns $A^{n-1}B, \dots, AB, B$ are linearly independent.

Proof: Let q^T denote the top row of

$$[A^{n-1}B \quad A^{n-2}B \quad \dots \quad AB \quad B]^{-1}$$

$$\begin{bmatrix} q^T \\ \text{--- second row ---} \\ \vdots \\ \text{--- } n\text{th row ---} \end{bmatrix} [A^{n-1}B \quad A^{n-2}B \quad \dots \quad AB \quad B] = \begin{bmatrix} 1 & 0 & \dots \\ 0 & 1 & \dots \\ & & \ddots \\ & & & 1 \end{bmatrix}$$

$$\left. \begin{aligned} q^T A^{n-1}B &= 1 \\ q^T A^{n-2}B &= 0 \\ &\vdots \\ q^T AB &= 0 \\ q^T B &= 0 \end{aligned} \right\} \dots (1)$$

Take $T = \begin{bmatrix} q^T \\ q^T A \\ \vdots \\ q^T A^{n-1} \end{bmatrix}$

$TB = \begin{bmatrix} q^T B \\ q^T AB \\ \vdots \\ q^T A^{n-1} B \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix}$

$TA = \begin{bmatrix} q^T A \\ q^T A^2 \\ \vdots \\ q^T A^n \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 & \dots & 0 \\ 0 & 0 & 1 & 0 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & \dots & \dots & \dots & 1 & \dots \\ * & \dots & \dots & \dots & * & \dots \end{bmatrix} \begin{bmatrix} q^T \\ q^T A \\ \vdots \\ q^T A^{n-1} \end{bmatrix}$

$\underbrace{\hspace{10em}}_T$

TAT^{-1} has this form

How do we know this T is invertible? Here is how: note

$\begin{bmatrix} q^T \\ q^T A \\ \vdots \\ q^T A^{n-1} \end{bmatrix} [A^{n-1}B \dots AB B]$

$= \begin{bmatrix} q^T A^{n-1}B & q^T A^{n-2}B & \dots & q^T AB & q^T B \\ q^T A^n B & q^T A^{n-1}B & \dots & q^T AB & q^T B \\ q^T A^{n+1}B & q^T A^n B & \dots & q^T AB & q^T B \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ q^T A^{2n-1}B & q^T A^{2n-2}B & \dots & q^T AB & q^T B \end{bmatrix}$

by (1) $\begin{bmatrix} 1 & 0 & \dots & 0 \\ * & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ * & \dots & \dots & 1 \end{bmatrix}$

-- (2)

If rows of T were linearly dependent we could find $\vec{w} \neq 0$ such that $\vec{w}^T T = 0$. But then (2) implies $\vec{w}^T \begin{bmatrix} 1 & 0 & \dots & 0 \\ * & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ * & \dots & \dots & 1 \end{bmatrix} = 0$
 $\Rightarrow \vec{w} = 0$
 Contradiction!

Note: The condition of the Claim is a certificate for being able to assign values of $A+BF$ by choice of F . We don't need to bring the system to the canonical form to design F .

Example 1: $A = \begin{bmatrix} 1 & 1 \\ 0 & 2 \end{bmatrix}$ $B = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$ $n=2$

Are B and AB linearly independent? $AB = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$
 YES.

$$A + BF = \begin{bmatrix} 1 & 1 \\ 0 & 2 \end{bmatrix} + \underbrace{\begin{bmatrix} 0 \\ 1 \end{bmatrix} \begin{bmatrix} f_1 & f_2 \end{bmatrix}}_{\begin{bmatrix} 0 & 0 \\ f_1 & f_2 \end{bmatrix}} = \begin{bmatrix} 1 & 1 \\ f_1 & 2+f_2 \end{bmatrix} = A_{CL}$$

Note: evalues of A are $\{1, 2\}$.

Evalues of A_{CL} : $\det(\lambda I - A_{CL}) = \det \begin{bmatrix} \lambda - 1 & -1 \\ -f_1 & \lambda - 2 - f_2 \end{bmatrix}$
 $= \lambda^2 - (3 + f_2)\lambda + 2 + f_2 - f_1$

Let's say we want closed-loop evalues at $\{0, 0\}$.

That means characteristic polynomial = λ^2

Match coefficients:

$$3 + f_2 = 0, \quad 2 + f_2 - f_1 = 0$$

$$f_1 = -1, \quad f_2 = -3$$

Note: upper triangular matrix

$$A = \begin{bmatrix} \lambda_1 & * & \dots & * \\ 0 & \ddots & & \vdots \\ \vdots & & \ddots & * \\ 0 & \dots & 0 & \lambda_n \end{bmatrix}$$

has evalues $\lambda_1, \dots, \lambda_n$ (diagonal entries). Why?

Look at $\lambda I - A = \begin{bmatrix} \lambda - \lambda_1 & \dots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \dots & \lambda - \lambda_n \end{bmatrix}$ singular bc/ last row $= \vec{0}$

$\lambda I - A = \begin{bmatrix} \lambda - \lambda_1 & \dots & 0 & * \\ \vdots & \ddots & \vdots & \vdots \\ 0 & \dots & 0 & \lambda - \lambda_n \end{bmatrix}$ } linearly dependent rows $\Rightarrow$ singular

think about $\lambda_1, \lambda_2, \dots, \lambda_n$ and other λ_i 's.

Example 2 (Ex. 3 in last lecture):

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

$$AB = \begin{bmatrix} 1 \\ 0 \end{bmatrix} \text{ linearly dependent on } B$$

$$A + BF = \begin{bmatrix} 1+f_1 & 1+f_2 \\ 0 & 2 \end{bmatrix}$$

2 is an eigenvalue independent of f_1, f_2
can't overcome instability

Controllability:

Recall solution of discrete-time system:

$$\vec{x}[i+1] = A \vec{x}[i] + B u[i] \quad (\text{assume single input})$$

$$i=0: \quad \vec{x}[1] = \underbrace{A \vec{x}[0] + B u[0]}$$

$$\begin{aligned} i=1: \quad \vec{x}[2] &= A \vec{x}[1] + B u[1] \\ &= \underbrace{A^2 \vec{x}[0] + AB u[0] + B u[1]} \end{aligned}$$

$$\begin{aligned} \vec{x}[3] &= A \vec{x}[2] + B u[2] \\ &= A^3 \vec{x}[0] + A^2 B u[0] + AB u[1] + B u[2] \end{aligned}$$

$$\vec{x}[i] = A^i \vec{x}[0] + A^{i-1} B u[0] + \dots + AB u[i-2] + B u[i-1]$$

$$\vec{x}[L] - A^L \vec{x}[0] = [A^{L-1}B, A^{L-2}B, \dots, AB, B] \begin{bmatrix} u[0] \\ u[1] \\ \vdots \\ u[L-1] \end{bmatrix}$$

Can we find input sequence $u[0], \dots, u[L-1]$ that brings state $\vec{x}$ from $\vec{x}[0]$ to a target $\vec{x}_{\text{target}}$ at time L ?

Yes, if $\vec{x}_{\text{target}} - A^L \vec{x}[0]$ lies in the column space of

$$C_L := [A^{L-1}B, \dots, AB, B].$$

"Controllability": ability to reach any target state $\vec{x}_{\text{target}}$ from any $\vec{x}[0]$.

Def'n: A system is called controllable if, given any target state $\vec{x}_{\text{target}}$ and initial condition $\vec{x}[0]$, we can find a time L and input sequence $u[0], \dots, u[L-1]$ such that $\vec{x}[L] = \vec{x}_{\text{target}}$.

How do we check that? If C_L has n linearly independent columns for some L , then column space is $\mathbb{R}^n$, n : state dimension. This means we

can make $C_L \begin{bmatrix} u[0] \\ \vdots \\ u[L-1] \end{bmatrix}$ anything we want in $\mathbb{R}^n$ by choosing this vector $\uparrow$. In particular, we

can assign $C_L \begin{bmatrix} u[0] \\ \vdots \\ u[L-1] \end{bmatrix} = \vec{x}_{\text{target}} - A^L \vec{x}[0]$. Then,

$$\vec{x}[L] = A^L \vec{x}[0] + C_L \begin{bmatrix} u[0] \\ \vdots \\ u[L-1] \end{bmatrix} = A^L \vec{x}[0] + \vec{x}_{\text{target}} - A^L \vec{x}[0]$$

Example 1: $B = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$ $AB = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$

$$C_1 = B \quad \dim = 1$$

$$C_2 = [AB \ B] = \begin{bmatrix} 1 & 0 \\ 2 & 1 \end{bmatrix} \quad \dim = 2 = n$$

✓ controllable

Example 2: $A = \begin{bmatrix} 1 & 1 \\ 0 & 2 \end{bmatrix}$ $B = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$

$$C_1 = B \quad \dim = 1$$

$$C_2 = [AB \ B] = \begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix} \quad \dim = 1$$

$$C_3 = [A^2B \ AB \ B] = \begin{bmatrix} 1 & 1 & 1 \\ 0 & 0 & 0 \end{bmatrix}$$

$\dim = 1$

$\dim$ can't increase further

stuck at $1 < n = 2$. ✓ uncontrollable

What we observe about dimension getting stuck is a consequence of the following result...

Lemma: If $A^L B$ is linearly dependent on
 $\{A^{L-1}B, \dots, AB, B\}$
 then $A^{L+1}B$ is also linearly dependent on
 $\{A^{L-1}B, \dots, AB, B\}$.

Proof: $A^L B = \alpha_{L-1} A^{L-1} B + \dots + \alpha_1 AB + \alpha_0 B$
 for some $\alpha_{L-1}, \dots, \alpha_1, \alpha_0$ by linear dependence.

Then,

$$\begin{aligned} A^{L+1} B &= A \cdot A^L B \\ &= A(\alpha_{L-1} A^{L-1} B + \dots + \alpha_1 AB + \alpha_0 B) \\ &= \alpha_{L-1} A^L B + \alpha_{L-2} A^{L-1} B + \dots + \alpha_1 A^2 B + \alpha_0 AB \\ &= * A^{L-1} B + * A^{L-2} B + \dots + * B \end{aligned}$$

$$\parallel$$

$$\alpha_{L-1} \alpha_0$$

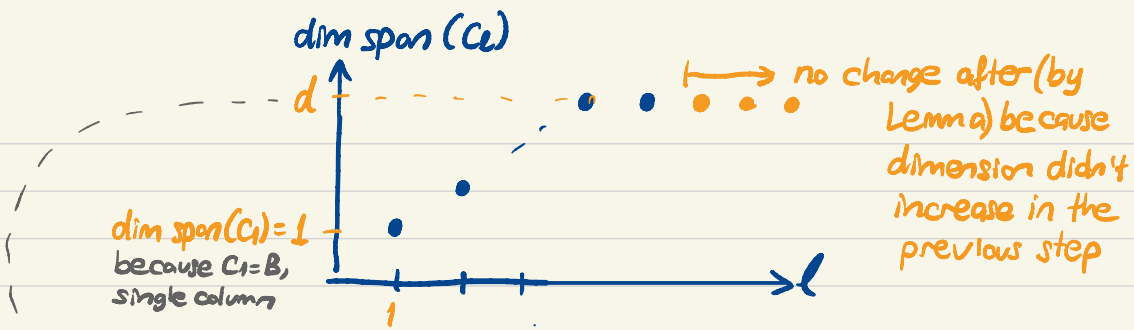
Note: $C_{L+1} = [A^L B \underbrace{A^{L-1} B \dots AB B}_{C_L}]$

Lemma implies:

if $\dim \text{span}(C_{L+1}) = \dim \text{span}(C_L) = d$

then $\dim \text{span}(C_{L+2}) = \dim \text{span}(C_{L+3}) = \dots = d$.

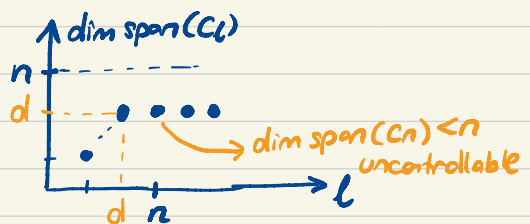
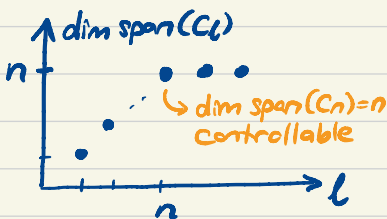
Once the dimension stops growing it stops for good.



Let d denote the dimension at which we get stuck.
 If $d = n$: CONTROLLABLE
 If $d < n$: UNCONTROLLABLE (will never reach n).

We could write a code that increments l by one as long $\dim \text{span}(C_l)$ keeps growing and terminates once growth stops (it has to stop because dimension can't exceed n). Then we can apply the test above with the dimension d reached to check controllability...

OR we can be smarter: dimension grows by one at each step as long as it grows (we are adding a single column when we increment l by 1). If we are able to reach n , we will reach it at $l=n$. Otherwise, growth will have stopped before n , so $\dim \text{span}(C_n)$ will be $d < n$. Thus, all we have to do is check if $\dim \text{span}(C_n) = n$. Controllable if so, uncontrollable if not. One-shot test!



Controllability condition: $\dim \text{span}(C_n) = n$
 i.e., $C_n = [A^n B \ \dots \ AB \ B]$ has n linearly indep. columns.