

EECS 16B: Lec. 15:
[Module 2 (Lecture 3)]

Today :
• System ID (Finish up)
• Stability

Reminders

- ① MT next
Mon. 4/13
8-10 p.m.
- ② Review
SYSTEM
tomorrow (7-9pm)
- ③ No class next
Tues.

RECAP:

SYSTEM ID: (SCALAR CASE)

$$x_d[k+1] = \lambda_d x_d[k] + b_d u_d[k] + e[k]$$

MODEL
PARAMETERS
NEED TO BE
ESTIMATED

need
to be
estimated

error or
disturbance
due to
model mismatch,
noise,
perturbation,
etc.

Goal: "Learn" λ_d, b_d from data

Unknowns: $\lambda_d, b_d \Rightarrow 2$ unknowns

$\begin{bmatrix} \lambda_d \\ b_d \end{bmatrix}$ vector of unknowns

Start system @ time $k=0 \Rightarrow x_d[0]$: observation

$$x_d[k+1] = \lambda_d x_d[k] + b_d u_d[k] + e[k]$$

$$x_d[1] = \lambda_d x_d[0] + b_d u_d[0] + e[0]$$

$$x_d[2] = \lambda_d x_d[1] + b_d u_d[1] + e[1]$$

$$x_d[3] = \lambda_d x_d[2] + b_d u_d[2] + e[2]$$

Least Squares!

Scalar Diff. Eq. : Least-Squares

Unknowns: λ_d, b_d

Goal: "Learn" λ_d, b_d from data

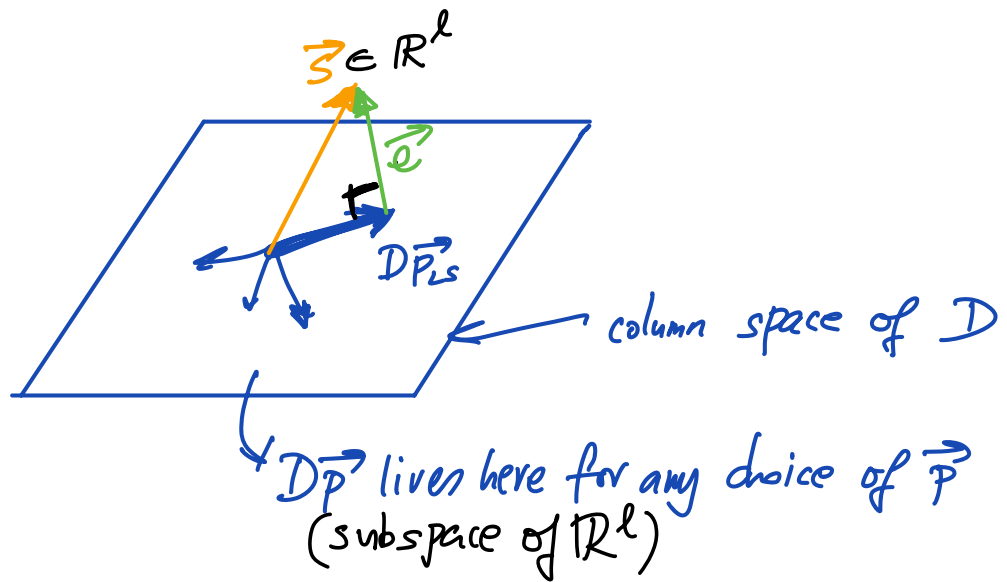
$$x_d[0] = \lambda_d x_d[0] + b_d u_d[0] + e[0]$$

$$x_d[1] = \lambda_d x_d[0] + b_d u_d[1] + e[1]$$

⋮

$$\underbrace{\begin{bmatrix} x_d[0] & u_d[0] \\ x_d[1] & u_d[1] \\ \vdots & \vdots \\ x_d[n-1] & u_d[n-1] \end{bmatrix}}_{\mathcal{D}} \underbrace{\begin{bmatrix} \lambda_d \\ b_d \end{bmatrix}}_{\vec{p}} = \underbrace{\begin{bmatrix} x_d[1] \\ x_d[2] \\ \vdots \\ x_d[n] \end{bmatrix}}_{\vec{s}}$$

Solve for $\hat{\vec{p}} = \begin{bmatrix} \hat{\lambda}_d \\ \hat{b}_d \end{bmatrix}$ using
Least-Squares.



$$\vec{P}_{LS} = (D^T D)^{-1} D^T \vec{s}$$

(if $[D^T D]$ is invertible)

General vector setting for Least Squares:

$$\vec{x}[i+1] = A \vec{x}[i] + B \vec{u}[i] + \vec{e}[i]$$

$(n \times 1)$ $(n \times n)$ $(n \times 1)$ $(n \times m)$ $(m \times 1)$ $(n \times 1)$

Example: $n=2$; $m=1$ (2 state vars., 1 control input)

$$\begin{bmatrix} x_1[i+1] \\ x_2[i+1] \end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \begin{bmatrix} x_1[i] \\ x_2[i] \end{bmatrix} + \begin{bmatrix} b_{11} \\ b_{12} \end{bmatrix} u[i] + \begin{bmatrix} e_1[i] \\ e_2[i] \end{bmatrix}$$

How many unknowns? $A: n \times n = n^2$ } Total of
 $B: n \times m = nm$ } $n(n+m)$

In our ex, $n=2, m=1 \Rightarrow 6$ unknowns

1st row: $x_1[i+1] = a_{11} x_1[i] + a_{12} x_2[i] + b_{11} u[i] + e_1[i]$

$i=0$: $x_1[1] = a_{11} x_1[0] + a_{12} x_2[0] + b_{11} u[0] + e_1[0]$

$i=1$: $x_1[2] = a_{11} x_1[1] + a_{12} x_2[1] + b_{11} u[1] + e_1[1]$

⋮

$i=l-1$: $x_1[l] = a_{11} x_1[l-1] + a_{12} x_2[l-1] + b_{11} u[l-1] + e_1[l-1]$

Q) How do we set up a L-S formulation? ($D \vec{p} + \vec{e} = \vec{s}$)
→ unknown

$$\begin{bmatrix} x_1[0] & x_2[0] & u[0] \\ x_1[1] & x_2[1] & u[1] \\ \vdots & \vdots & \vdots \\ x_1[l-1] & x_2[l-1] & u[l-1] \end{bmatrix} \underbrace{\begin{bmatrix} a_{11} \\ a_{12} \\ b_{11} \end{bmatrix}}_{\vec{p}_1} + \begin{bmatrix} e_1[0] \\ \vdots \\ e_1[l-1] \end{bmatrix} = \begin{bmatrix} x_1[1] \\ x_1[2] \\ \vdots \\ x_1[l] \end{bmatrix}$$

$\underbrace{\hspace{10em}}_{\vec{e}}$
 $\underbrace{\hspace{10em}}_{\vec{s}_1}$

D

$$\begin{bmatrix} \hat{a}_{11} \\ \hat{a}_{12} \\ \hat{b}_{11} \end{bmatrix}$$

$$\vec{\hat{p}}_1 = (D^T D)^{-1} D^T \vec{s}_1$$

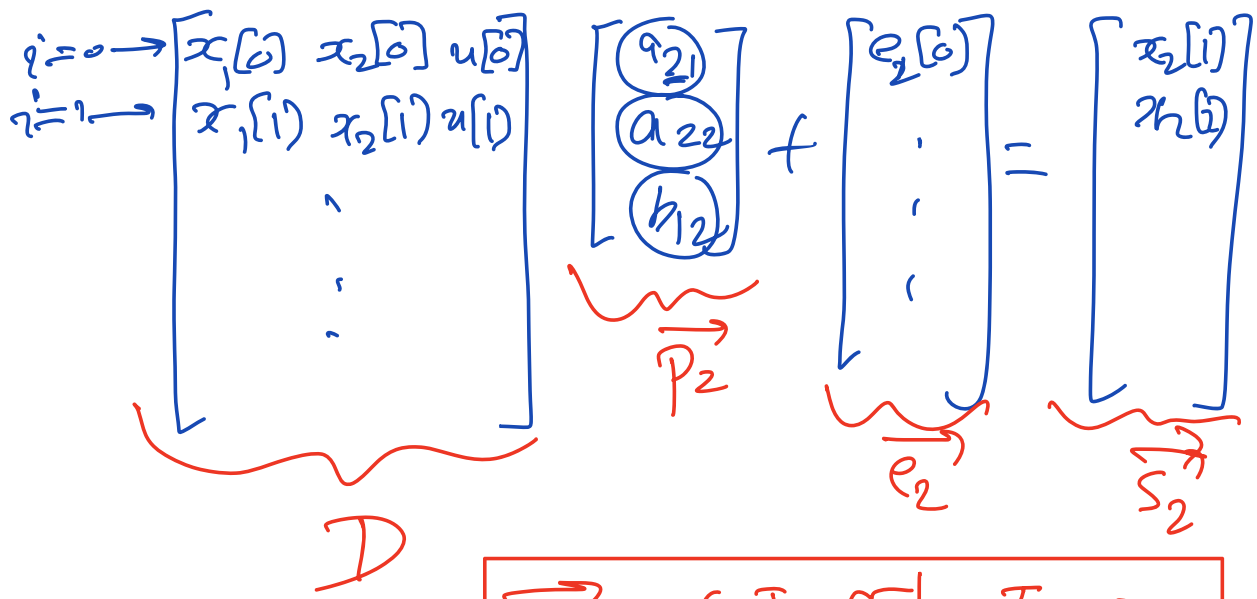
(if $D^T D$ is invertible)

$$\begin{bmatrix} x_1[i+1] \\ x_2[i+1] \end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \begin{bmatrix} x_1[i] \\ x_2[i] \end{bmatrix} + \begin{bmatrix} b_{11} \\ b_{12} \end{bmatrix} u[i] + \begin{bmatrix} e_1[i] \\ e_2[i] \end{bmatrix}$$

Let's est. the second row of parameters

$$\hat{a}_{21}, \hat{a}_{22}, \hat{b}_{12} \text{ in the same way:}$$

2nd row: $x_2[i+1] = a_{21}x_1[i] + a_{22}x_2[i] + b_{12}u[i] + e_2[i]$



Least-squares!

$$\vec{\hat{p}}_2 = (D^T D)^{-1} D^T \vec{s}_2$$

$$\begin{bmatrix} \hat{a}_{21} \\ \hat{a}_{22} \\ \hat{b}_{12} \end{bmatrix}$$

Note: We need $l \geq 3$ obs.

of $\begin{cases} x_1[i] \\ x_2[i] \end{cases}_{i=1}^3$ & input controls
 $\begin{cases} u[i] \end{cases}_{i=0}^2$ (plus init. state $x_1[0], x_2[0]$)

Q Do we need to solve separate L-S eqns. for each row?

$$\vec{\hat{p}}_1 = (D^T D)^{-1} D^T \vec{s}_1 \longrightarrow (1)$$

$$\vec{\hat{p}}_2 = (D^T D)^{-1} D^T \vec{s}_2 \longrightarrow (2)$$

$$\underbrace{\begin{bmatrix} \vec{\hat{p}}_1 & \vec{\hat{p}}_2 \end{bmatrix}}_P = (D^T D)^{-1} D^T \underbrace{\begin{bmatrix} \vec{s}_1 & \vec{s}_2 \end{bmatrix}}_S$$

Exactly follows rules of matrix-multiplication!

$$\hat{P} = \begin{bmatrix} \hat{a}_{11} & \hat{a}_{21} \\ \hat{a}_{12} & \hat{a}_{22} \\ \hat{b}_{11} & \hat{b}_{12} \end{bmatrix} = (D^T D)^{-1} D^T S$$

General vector case:

$$\vec{x}(i+1) = A \vec{x}(i) + B \vec{u}(i) + \vec{e}(i)$$

$(n \times 1)$ $(n \times n)$ $(n \times 1)$ $(n \times m)$ $(m \times 1)$ $(n \times 1)$

$$\begin{bmatrix} x_1[i+1] \\ x_2[i+1] \\ \vdots \\ x_n[i+1] \end{bmatrix} = \underbrace{\begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{bmatrix}}_{\text{unknowns}} \begin{bmatrix} x_1[i] \\ x_2[i] \\ \vdots \\ x_n[i] \end{bmatrix} + \underbrace{\begin{bmatrix} b_{11} \dots b_{1m} \\ \vdots \\ b_{n1} \dots b_{nm} \end{bmatrix}}_{\text{unknowns}} \begin{bmatrix} u_1[i] \\ \vdots \\ u_m[i] \end{bmatrix} + \begin{bmatrix} e_1[i] \\ e_2[i] \\ \vdots \\ e_n[i] \end{bmatrix}$$

$$a_i^T = [a_{i1} \ a_{i2} \ \dots \ a_{im}]$$

$$b_i^T = [b_{i1} \ \dots \ b_{im}]$$

Goal: Identify (i.e. do L-S est. of) system parameters A, B.

How do we set it up as a L-S set-up?

$$D P = S \Rightarrow \hat{P} = (D^T D)^{-1} D^T S$$

↓
unknowns

$$x_{i+1} = a_i^T \vec{x}[i] + b_i^T \vec{u}[i] + e_i[i]$$

Let's drop this term from here on.

$$\begin{aligned} x_{i+1} &\cong \langle \vec{a}_i, \vec{x}[i] \rangle + \langle \vec{b}_i, \vec{u}[i] \rangle \\ &\cong \langle \vec{x}[i], \vec{a}_i \rangle + \langle \vec{u}[i], \vec{b}_i \rangle \end{aligned}$$

$$x_{i+1} \cong \vec{x}[i]^T \vec{a}_i + \vec{u}[i]^T \vec{b}_i$$

One equation for $i=0$.

Can get multiple observations by spanning
 $i=0, 1, \dots$ (recall we have to solve for
 $(\underbrace{n^2}_{A_{n \times n}} + \underbrace{mn}_{B_{n \times m}})$ unknown scalar parameters)

$$x_1[i+1] \cong \vec{x}^T[i] \vec{a}_1 + \vec{u}^T[i] b_1$$

One equation
for $i=0$.

$$\begin{array}{c}
 i=0 \\
 i=1 \\
 \vdots \\
 i=l-1
 \end{array}
 \begin{bmatrix}
 \vec{x}^T[0] & \vec{u}^T[0] \\
 \vec{x}^T[1] & \vec{u}^T[1] \\
 \vdots & \vdots \\
 \vec{x}^T[l-1] & \vec{u}^T[l-1]
 \end{bmatrix}
 \begin{bmatrix}
 \vec{a}_1 \\
 b_1
 \end{bmatrix}
 \cong
 \begin{bmatrix}
 x_1[1] \\
 x_1[2] \\
 \vdots \\
 x_1[l]
 \end{bmatrix}$$

$\underbrace{\hspace{15em}}_D \quad \underbrace{\begin{bmatrix} a_{11} \\ a_{12} \\ \vdots \\ a_{1n} \\ b_{11} \\ \vdots \\ b_{1m} \end{bmatrix}}_{\vec{p}_1} \quad \underbrace{\hspace{10em}}_{\vec{s}_1}$

LA's now estimate \vec{a}_2, \vec{b}_2 :

$$x_2[i+1] \cong \vec{x}^T[i] \vec{a}_2 + \vec{u}^T[i] \vec{b}_2$$

$i=0, 1, \dots, l-1$

$$\begin{array}{l}
 i=0 \rightarrow \\
 i=1 \rightarrow \\
 \vdots \\
 i=l-1 \rightarrow
 \end{array}
 \begin{bmatrix}
 \vec{x}^T[0] & \vec{u}^T[0] \\
 \vec{x}^T[1] & \vec{u}^T[1] \\
 \vdots & \vdots \\
 \vec{x}^T[l-1] & \vec{u}^T[l-1]
 \end{bmatrix}
 \begin{bmatrix}
 \vec{a}_2 \\
 \vec{b}_2
 \end{bmatrix}
 \approx
 \begin{bmatrix}
 x_1[2] \\
 x_1[3] \\
 \vdots \\
 x_1[l+1]
 \end{bmatrix}$$

$\underbrace{\hspace{15em}}_D \qquad \underbrace{\hspace{10em}}_{P_2} \qquad \underbrace{\hspace{10em}}_{S_2}$

$$\left. \begin{array}{l}
 D \vec{p}_1 \approx \vec{s}_1 \\
 D \vec{p}_2 \approx \vec{s}_2 \\
 \vdots \\
 D \vec{p}_n \approx \vec{s}_n
 \end{array} \right\} \Rightarrow D \begin{bmatrix} \vec{p}_1 & \vec{p}_2 & \dots & \vec{p}_n \end{bmatrix} \approx \begin{bmatrix} \vec{s}_1 & \vec{s}_2 & \dots & \vec{s}_n \end{bmatrix}$$

$\underbrace{\hspace{15em}}_P \qquad \underbrace{\hspace{15em}}_S$

$$\Rightarrow \hat{P} = (D^T D)^{-1} D^T S \quad (*)$$

LS est. \vec{p}_1 $\begin{bmatrix} \hat{a}_{11} \\ \hat{a}_{1n} \\ \hat{b}_{11} \\ \hat{b}_{1m} \end{bmatrix}$ $\begin{bmatrix} \hat{a}_{21} \\ \vdots \\ \hat{a}_{2n} \\ \hat{b}_{21} \\ \vdots \\ \hat{b}_{2m} \end{bmatrix}$ \dots $\begin{bmatrix} \hat{a}_{n1} \\ \vdots \\ \hat{a}_{nn} \\ \hat{b}_{n1} \\ \vdots \\ \hat{b}_{nm} \end{bmatrix} \vec{p}_n$

\vec{s}_1 $\begin{bmatrix} x_1[2] & x_2[2] & \dots & x_n[2] \\ \vdots & \vdots & \ddots & \vdots \\ x_1[l+1] & x_2[l+1] & \dots & x_n[l+1] \end{bmatrix} \vec{s}_n$

Q) How many (vector) observations $\{\vec{x}[i]\}_{i=1}^l$ and input controls $\{\vec{u}[i]\}_{i=0}^{l-1}$ do we need to solve for the unknown sys. ID parameters A, B ?

A) Need $l \geq (n+m)$.

Note: When doing system ID, we are learning the model param. by trying various input controls $\vec{u}[i]$ and observing the system states $\vec{x}[i]$.

(1) System-ID

$$\vec{x}[i+1] = A\vec{x}[i] + Bu[i] \quad *$$

(2) Use (*) to do control.

STABILITY:

DT: $x(t+1) = \lambda x(t) + u(t) + e(t)$

\swarrow
 ASSUMP known

\uparrow
 input (engineer's choice)

\uparrow
 noise (math's choice)

e.g. $x(t+1) = -2x(t)$

$x(0) = x(0)$

$x(1) = -2x(0)$

$x(2) = (-2)^2 x(0)$

\vdots

$x(t) = (-2)^t x(0) \xrightarrow{t \rightarrow \infty} \pm \infty$

"unstable"

e.g. $x(t+1) = \left(\frac{1}{2}\right)^t x(t)$

"stable"

State stability:

A system is stable iff \exists a K such that $|x(t)| < K$ for all time t . ($\forall t$)

if \exists only if there exists

\rightarrow for all

e.g. $x(t+1) = -2x(t) + u(t)$
 $x(0) = 1$
is unstable even for $u(t) = 0$!

e.g. $x(t+1) = \frac{1}{2}x(t) + u(t)$

New definition:

Bounded-input bounded-output stability
(BIBO)

BIBO stable: for every bounded input \vec{u} & init. state $\vec{x}[0]$, the resulting state trajectory $\vec{x}[t]$ is bounded.

BIBO stable iff for $E > 0$,
 $|u(t)| \leq E \forall t$, then \exists a $K (< \infty)$
s.t. $|x(t)| \leq K \forall t$

(must hold for any initial state $\vec{x}[0]$)

e.g. $x(t+1) = \lambda x(t) + u(t)$

$$\lambda > 1$$

$$|u(t)| \leq E$$

Guess: Not BIBO stable

Proof: Find some $u(t)$ s.t.
the system blows up.

Try: $u(0) = E, u(t) = 0, t > 0$

$$x(0) = x(0)$$

$$x(1) = \lambda x(0) + u(0) = \lambda x(0) + E$$

$$x(2) = \lambda(\lambda x(0) + E) + 0$$

$$\vdots$$
$$x(t) = \lambda^{t-1} (\lambda x(0) + E)$$

$$|x(t)| = \underbrace{|\lambda^{t-1}|}_{\infty} | \lambda x(0) + E |$$

BIBO
unstable

as $t \rightarrow \infty$
since $\lambda > 1$

Ex: $x(t+1) = \lambda x(t) + u(t)$
 $|\lambda| < 1$ $(u(t)) < E$

Guess: BIBO stable

Proof: Want to be bounded
for all $u(t)$!

$$x[1] = \lambda x(0) + u(0)$$

$$x(2) = \lambda x(1) + u(1)$$

$$= \lambda (\lambda x(0) + u(0)) + u(1)$$

$$= \lambda^2 x(0) + \lambda u(0) + u(1)$$

$$x(3) = \lambda^3 x(0) + (\lambda^2 u(0) + \lambda u(1) + u(2))$$

⊕

$$x(t) = \lambda^t x(0) + \sum_{i=0}^{t-1} \lambda^i u(t-i)$$

We want $|x(t)| < \infty \quad \forall t$
 Take abs. value of both sides of

$$|x(t)| = \left| \lambda^t x(0) + \sum_{i=0}^{t-1} \lambda^i u(t-1-i) \right|$$

(use: $|A+B| \leq |A|+|B|$)
 a.k.a. Triangle-ineq



$$|x(t)| \leq \underbrace{|\lambda^t x(0)|}_{\leq |x(0)|} + \underbrace{\left| \sum_{i=0}^{t-1} \lambda^i u(t-1-i) \right|}$$

$$\leq |x(0)| + \sum_{i=0}^{t-1} |\lambda^i| |u(t-1-i)|$$

(by repeated \leq
 app. of Δ ineq)

$$\leq |x(0)| + E \underbrace{\sum_{i=0}^{t-1} |\lambda^i|}$$

$$\sum_{i=0}^{t-1} |\lambda|^i = 1 + |\lambda| + |\lambda|^2 + \dots$$

(Geometric sum)

$$\leq \frac{1}{1-|\lambda|} \quad \text{if } |\lambda| < 1$$

$$\leq |x(0)| + E \frac{1}{1-|\lambda|} = K$$

Bounded!