

# Sufficiently Rich Inputs and Parameter Identification

Recall:  $z = \theta^T \phi$  (linear parametrization)

$\theta$ : system parameters (freq. domain description)

$z = y^{(n)}$ ,  $\phi = (u^{(n-1)}, \dots, u, u, -y^{(n-1)}, \dots, -y, -y)$   
(filtered versions,  $n$ : model order)

Observe  $(u(t), y(t))$ ,  $t \geq 0$

$\hat{\theta}(t)$ : estimates of  $\theta$  based on  $(z(t), \phi(t))$ ,  $0 \leq t \leq T$   
(e.g. use gradient flow or online LS)

Output prediction error:  $\hat{z}(t) - z(t)$   
 $= (\hat{\theta}(t) - \theta)^T \phi(t)$

Param. identification error:  $\hat{\theta}(t) - \theta$

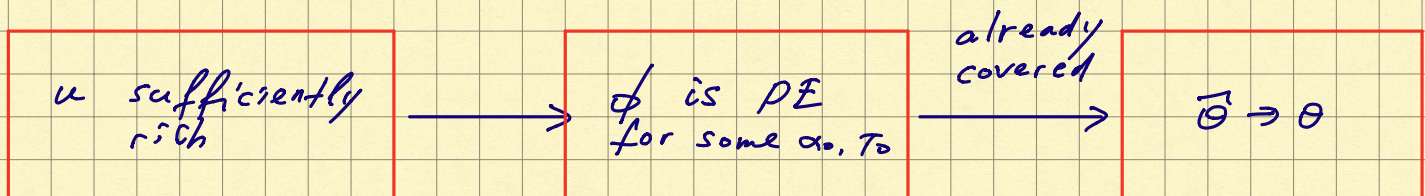
$\hat{\theta}(t) \rightarrow \theta$  under a **Persistence of Excitation** assumption  
on  $\phi$ :  
 $t_0 + T_0$

$$\int_{t_0}^{t_0 + T_0} \phi(t) \phi(t)^T dt \geq \alpha_0 T_0 I \quad \forall t_0 \geq 0$$

for some  $\alpha_0, T_0 > 0$

Note:  $\phi$  consists of filtered versions of  $u$  (input) and  $y$  (the output) but the only signal we have direct control over is  $u$ .

**Need conditions on  $u$  alone to guarantee PE of  $\phi$ .**





# Example ① 1st order plant

$$\dot{y} = -ay + bu$$

$a > 0$  (stable system)

$b \in \mathbb{R}$  (arbitrary)

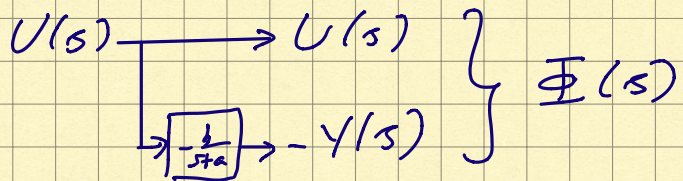
$u \in L_\infty$  (bdd input)

What makes  $u$  "sufficiently rich"?

Ignore filtering:  $z = y$   $\phi = \begin{pmatrix} u \\ -y \end{pmatrix}$ ,  $\theta = \begin{pmatrix} b \\ a \end{pmatrix}$

$$H(s) = \frac{b}{s+a}$$

$$\phi = \begin{pmatrix} 1 \\ -\frac{b}{s+a} \end{pmatrix} u$$



$$\underline{PE}: \int_{t_0}^{t_0+T} \phi(t) \phi(t)^T dt = \int_{t_0}^{t_0+T} \begin{pmatrix} u(t) \\ -y(t) \end{pmatrix} \begin{pmatrix} u(t) & -y(t) \end{pmatrix} dt$$

What is a good choice of  $u(t)$ ?

- try a constant input,  $u(t) = c$ ,  $t \geq 0$

$$\begin{aligned} \dot{y} &= -ay + bc \\ &= -a\left(y - \frac{bc}{a}\right) \end{aligned}$$

$$\text{Let } \tilde{y}(t) := y(t) - \frac{bc}{a}$$

$$\dot{\tilde{y}} = \dot{y} = -a\left(y - \frac{bc}{a}\right) = -a\tilde{y} \quad \Leftrightarrow \quad \tilde{y}(t) = \tilde{y}(0) e^{-at}$$

$$y(t) = \underbrace{\frac{bc}{a}}_{\substack{\text{depends} \\ \text{only on} \\ b/a}} + \underbrace{e^{-at} \left(y(0) - \frac{bc}{a}\right)}_{\text{decays exp. fast}}$$

- can't identify both  $a$  and  $b$  if we inject constant input.



More directly:

$$\phi(t) \phi(t)^T = \begin{pmatrix} u(t) \\ -y(t) \end{pmatrix} \begin{pmatrix} u(t) & -y(t) \end{pmatrix}$$

$$u(t) = c \quad (\text{const})$$

$$y(t) = \frac{bc}{a} + \tilde{y}(t)$$

$\tilde{y}(t)$  decays exponentially fast

$$\phi \phi^T = \begin{pmatrix} c & -\frac{bc}{a} - \tilde{y} \end{pmatrix} \begin{pmatrix} c & -\frac{bc}{a} - \tilde{y} \end{pmatrix}$$

$$= \begin{pmatrix} c^2 & -\frac{bc^2}{a} - c\tilde{y} \\ -\frac{bc^2}{a} - c\tilde{y} & \left(\frac{bc}{a} + \tilde{y}\right)^2 \end{pmatrix}$$

$$= \begin{pmatrix} c^2 & -\frac{bc^2}{a} - c\tilde{y} \\ -\frac{bc^2}{a} - c\tilde{y} & \frac{b^2c^2}{a^2} + \frac{2bc}{a}\tilde{y} + \tilde{y}^2 \end{pmatrix} \quad \tilde{y}(t) = e^{-at} \tilde{y}(0)$$

$$= \underbrace{\begin{pmatrix} c^2 & -\frac{bc^2}{a} \\ -\frac{bc^2}{a} & \frac{b^2c^2}{a^2} \end{pmatrix}}_{\det \equiv 0} + \underbrace{\begin{pmatrix} 0 & -c\tilde{y}(t) \\ -c\tilde{y}(t) & \frac{2bc}{a}\tilde{y}(t) + \tilde{y}^2(t) \end{pmatrix}}_{\det = -c^2\tilde{y}^2(t) \rightarrow 0 \text{ exp. fast}}$$

Try some other inputs (keep them bdd)

Sinusoids!

$$u(t) = \sin \omega t$$

$$\omega \neq 0$$

- constant freq.

$$y(t) = e^{-at} y(0) + \int_0^t e^{-a(t-s)} \sin(\omega s) ds$$

$$\approx A \sin(\omega t + \alpha) \quad (\text{in steady state})$$

where  $A$  and  $\alpha$  are constants that depend on  $a, b$



$$\phi(t) = \begin{pmatrix} \sin \omega t \\ A \sin(\omega t + \alpha) \end{pmatrix} \quad \text{is PE} \\ \text{(homework)}$$

Ex. (2): 2nd-order plant

$$Y(s) = \frac{b_1 s + b_0}{s^2 + a_1 s + a_0} U(s)$$

$(a_1, a_0 > 0$  : stable, by Routh-Hurwitz)

$$\theta = (b_1, b_0, a_1, a_0)^T \in \mathbb{R}^4$$

$$u \rightarrow \boxed{G(s)} \rightarrow y$$

$$u(t) = \sin(\omega t) \rightarrow y(t) = A \sin(\omega t + \alpha)$$

Linearity:  $u(t) = \sin(\omega_1 t) + \sin(\omega_2 t)$   $\omega_1 \neq \omega_2$

$$y(t) = A_1 \sin(\omega_1 t + \alpha_1) + A_2 \sin(\omega_2 t + \alpha_2)$$

Here:  $z(t) = \dot{y}(t)$ ,  $\phi(t) = (u'(t), u(t), -y'(t), -y(t))^T$

General idea: for an  $n$ th-order SISO LTI system, an input  $u$  is sufficiently rich if it contains at least  $n$  distinct frequencies (assuming no pole-zero cancellations)

$$\int_0^T \phi(t) \phi(t)^T dt \quad - \text{ "autocovariance" of } \phi$$

- examine in Fourier domain (spectral measures)

Caveats:

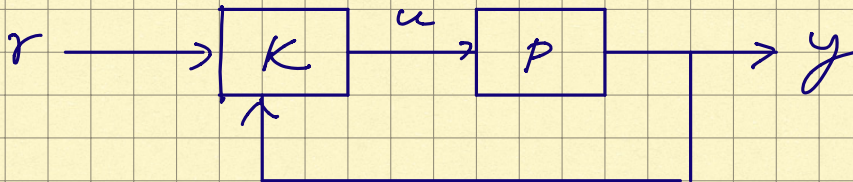
- parameter identification ( $\hat{\theta} \rightarrow \theta$ ) is not necessary for control: PE may conflict w/ control goals
- parameter estimation may still be useful even w/o PE
- PE may be of use in settings like reference tracking



# Model Reference Adaptive Control (MRAC)

Goal: get the closed-loop system (w/ unknown plant) to reproduce i/o behavior according to a given reference model.

Recall:



Ex.:

plant  $\dot{y} = ay + bu$

$a \in \mathbb{R}$  (unknown, no assumption of stability)

$b > 0$  (sign of  $b$  known)

reference:  $\dot{y}_m = -a_m y_m + b_m r$

$a_m > 0$  (stable reference)

$r \in L_\infty$

$a_m, b_m$ :  
given desired  
closed-loop  
system specs

control:  $u = -ky + lr$

feedback gain  $\downarrow$   
feedforward gain  $\rightarrow$

If  $a, b$  are known,  $k$  and  $l$  can be specified:

$$\dot{y} = ay + bu$$

$$u = -ky + lr$$

$$= (a - bk)y + blr$$

Want:

$$a - bk = -a_m$$

$$bl = b_m$$

$\iff$

$$k = \frac{a + a_m}{b}$$

( $b > 0$ )

$$l = \frac{b_m}{b}$$

Since  $a, b$  are unknown, let's use:

$$u = -\hat{k}y + \hat{l}r$$

$\hat{k}, \hat{l}$  - dynamic estimates



MRAC: design update laws  $\hat{k}, \hat{l}$

- direct: update  $\hat{k}, \hat{l}$  (control params) directly
- indirect: estimate  $a, b$  via  $\hat{a}, \hat{b}$ , use these to generate  $\hat{k}, \hat{l}$

## Direct MRAC

• reparametrize

$$\dot{y} = ay + bu$$

$$= -a_m y + b_m r + (a + a_m)y + bu - b_m r$$

$$= -a_m y + b_m r + bk y + bu - bl r$$

$$\dot{y} = -a_m y + b_m r + b(u + ky - lr)$$
$$= -\tilde{k}y + \tilde{l}r$$

$$\dot{y} = -a_m y + b_m r + b(-\tilde{k}y + \tilde{l}r)$$

$$\tilde{k} := \hat{k} - k$$

$$\tilde{l} := \hat{l} - l$$

$$e := y_m - y$$

Preview: candidate Lyapunov fcn

$$V(e, \tilde{k}, \tilde{l}) := \frac{1}{2} \left( \frac{e^2}{b} + \frac{\tilde{k}^2}{\gamma} + \frac{\tilde{l}^2}{\gamma} \right),$$

where  $\gamma > 0$  is a tunable constant