

11/4-18

Föreläsning 6

Recap to wrap-up the modeling part

Input-output models (ODE ↑ degree)

State-space models (matrix diff. eq., 1st order)



$$\tilde{x}(t) = T x(t)$$

Solved the state eq.

$$x(t) = e^{At} x_0 + \int_0^t e^{A(t-\tau)} B u(\tau) d\tau$$

for $\dot{x}(t) = Ax(t) + Bu(t)$, x_0

$$y(t) = Cx(t) + Du(t) = Ce^{At} x_0 + \int_0^t Ce^{A(t-\tau)} B u(\tau) d\tau$$

Stability

eigenvalues for A

$$\forall i=1 \dots n, \operatorname{Re}\{\lambda_i\} \leq 0$$

Lyapunov / bounded
stable

$$\operatorname{Re}\{\lambda_i\} < 0 \Rightarrow \lim_{t \rightarrow \infty} x(t) = 0 \quad \text{asymptotic stable}$$

ControllableCan we reach/control each and every state
with the input?

$$R(A, B) = [B \ ; \ AB \ \dots \ A^{n-1}B]$$

$$\uparrow \dim(x) = n$$

reachability matrix

rank $R = n \Rightarrow u$ can reach every x

Today:

1) Observability

————— || ————— modeling done

2) What does it stand for PID?

3) How to tune a PID compensator?

Time
Laplace] domain specs.

4) Nyquist stability I

1) Observability:

with $\{u, y\}$ can we observe the x ?

$$\text{e.g. } \begin{bmatrix} \dot{x}_1(t) \\ \dot{x}_2(t) \end{bmatrix} = \begin{bmatrix} -1 & 0 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} x_1(t) \\ x_2(t) \end{bmatrix} + \begin{bmatrix} 1 \\ 0 \end{bmatrix} u(t)$$

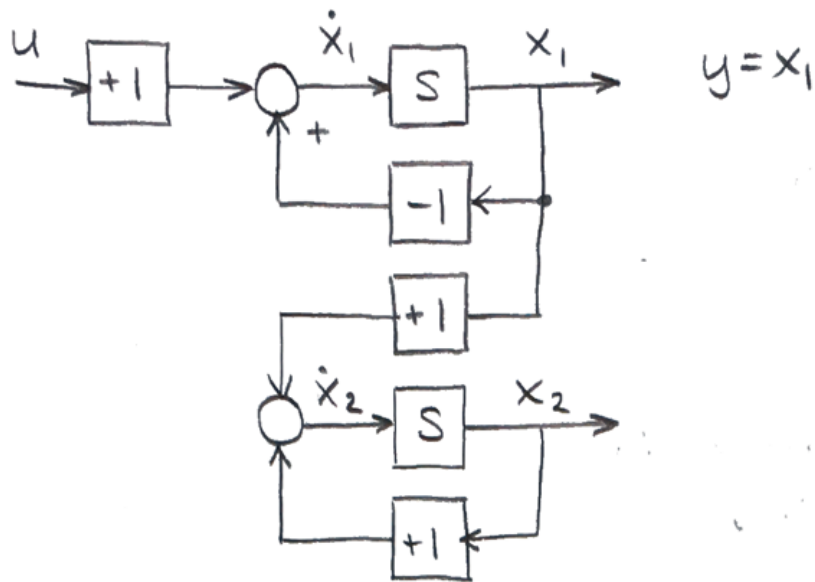
$$y(t) = \begin{bmatrix} 1 & 0 \end{bmatrix} \begin{bmatrix} x_1(t) \\ x_2(t) \end{bmatrix}$$

$$\dim(x) = 2$$

Draw its block diagram!

$$\dot{x}(t) = Ax(t) + Bu(t)$$

$$y(t) = Cx(t) + Du(t)$$



$x_1 \rightarrow y$ $x_2(t)$ will never be in $y(t)$

$x_2 \not\rightarrow y$

$x_2 \not\rightarrow x_1 \rightarrow y$

Def. The LTI state-space model is observable if for any T $x(T)$ can be observed by the $\{u, y\}$ sequence over $[0, T]$.

How to check?

Theorem (Kalman observability condition)

A continuous time LTI model is observable iff $\text{rank } \sigma = n$, where

$$\sigma(C, A) = \sigma = \begin{bmatrix} C \\ CA \\ \vdots \\ CA^{n-1} \end{bmatrix}, \quad \sigma \in \mathbb{R}^{n \times n} \quad (\text{for SISO})$$

e.g. $A = \begin{bmatrix} -1 & 0 \\ 1 & 1 \end{bmatrix}, \quad C = [1 \ 0]$

Observable?

$$n=2 \Rightarrow CA^{n-1} = CA$$

$$\sigma = \begin{bmatrix} C \\ CA \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ -1 & 0 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 0 \\ -1 & 0 \end{bmatrix} \begin{bmatrix} -1 & 0 \\ 1 & 1 \end{bmatrix} = \begin{bmatrix} -1 & 0 \\ 0 & 1 \end{bmatrix}$$

To check full rank property:

$$\det \sigma = 1 \cdot 0 - (-1) \cdot 0 = 0$$

↓

rank $\sigma < 2 \Rightarrow$ not observable

Note if $\det \sigma \neq 0 \Rightarrow \sigma$ has full rank.

Theorem

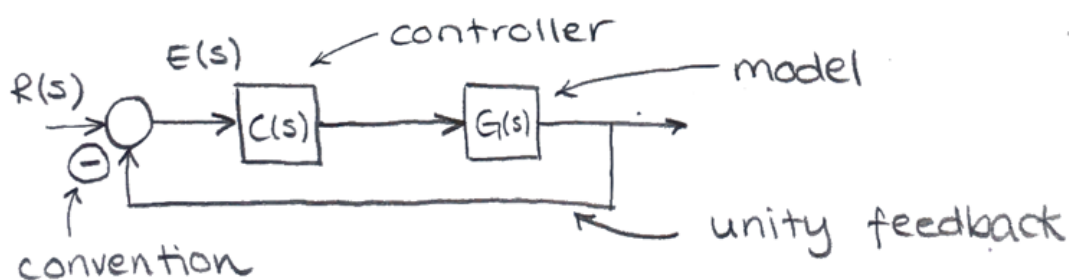
An LTI model is called minimal order both controllable and observable at the same time.

Minimality: $u \rightarrow y$ relation is given with the least element of x .

Input/output stability implies internal stability.

2) What does it stand for PID?

Def. PID is a structured, dynamic and output feedback controller



$$E(s) = R(s) - Y(s)$$

Basic form for PID

controller
regulator
compensator

$$C(s) = K_p + \frac{K_i}{s} + K_d s$$

proportional
(P)
constant

integrator
(I)
dynamic

differentiator
(D)
dynamic

$$u(t) = \mathcal{L}^{-1}\{u(s)\} = K_p e(t) + K_i \int_0^t e(\tau) d\tau + K_d \frac{de(t)}{dt}$$

output feedback because only $R(s)$ and $Y(s)$ are required for it!

Roles of:

P - constant / fast speed up closed loop

I - cumulate tracking error / slow / get rid of error $\lim_{t \rightarrow \infty} e = 0$

D - "look ahead" feat future.

e.g. virus epidemic model

$$\ddot{I}(t) + a_1 \dot{I}(t) + a_2 I(t) = b u(t)$$

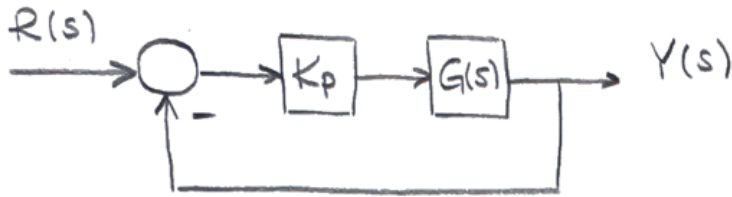
$I(t)$ - # of infected species

$u(t)$ - immunisation

Find a stable controlled epidemic behaviour!

$$y(t) = I(t) ; C(s) = K_p$$

$$G(s) = \frac{Y(s)}{U(s)} = \frac{b}{s^2 + a_1 s + a_2}$$



$$G_c(s) = \frac{Y(s)}{R(s)} = \frac{G(s) \cdot C(s)}{1 + G(s)C(s)} \quad (\text{see L4, L3})$$

after creating the closed-loop,

$$L(s) = C(s)G(s) = \frac{K_p \cdot b}{s^2 + a_1 s + a_2}$$

↑ loop transfer

$$f(s) = 1 + L(s) \leftarrow \text{loop ratio return}$$

$$G_c(s) = \frac{L(s)}{L(s) + 1} = \frac{K_p b}{s^2 + a_1 s + a_2 + K_p b}$$

K_p only influences the $a_2 \Rightarrow p_1, p_2$ can not be arbitrary.

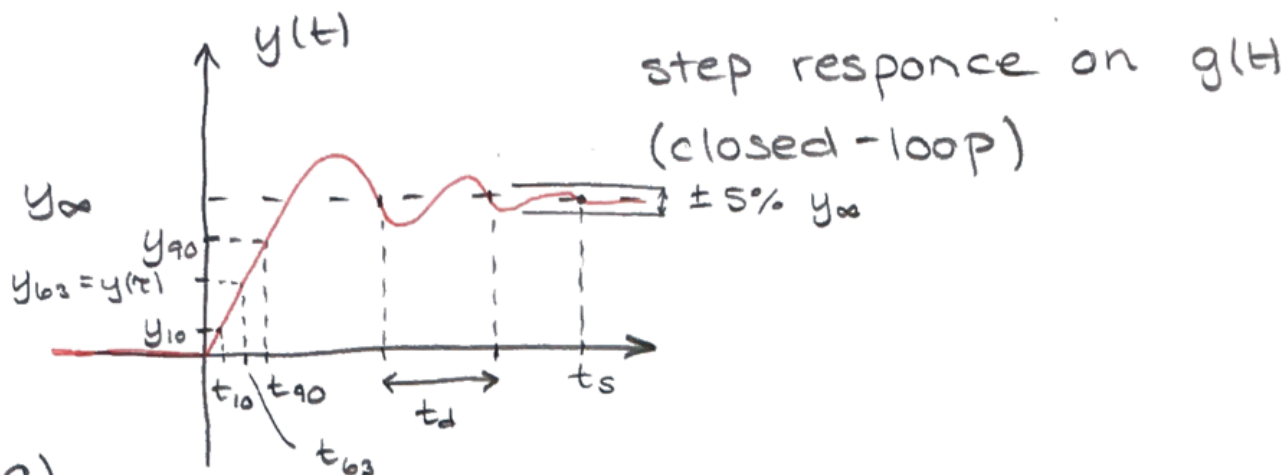
changed \Rightarrow It may remain unstable under

$$u(t) = K_p (r(t) - y(t))$$

$e(t)$ can be unbounded depending on (a_1, a_2)

3) How to tune a PID controller?

Time domain specifications



a) y_{∞} steady-state output.
 $\Rightarrow e_{\infty} = r_{\infty} - y_{\infty}$ stationary tracking error.

b) $y_{10} = y(t_{10})$ - t_{10} is when you reach 10% of y_{∞}
 $t_r = t_{90} - t_{10}$ rise time

c) Equivalent time constant τ
 $y_{63} = y(\tau)$ 63% of y_{∞} reached.

d) t_d = damping time

e) t_s = settling time

$$0,95 \leq y(t) \leq 1,05$$

enters into \uparrow and stay there \Rightarrow the $y(t) \approx y_{\infty}$

f) t_p = peak value for $y(t)$

\Rightarrow largest $y(t_p)$

g) overshoot $\sigma = \frac{y(t_p) - y_{\infty}}{y_{\infty}} \quad [\%]$

	t_r	σ	t_s	e_∞
K_p		↑		↓
K_i		↑		0
K_d		↓		-

homework: fill in table.

↑ increase, ↓ decrease, 0 removes, - no effort

Method for tuning PID

1) Heuristics $\left\{ \begin{array}{l} \text{Graph} \\ \text{Transfer function} \end{array} \right.$

2) Guaranteed stability

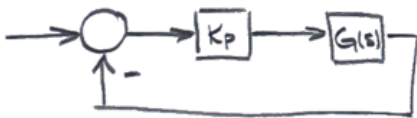
1.1 Graphical: Ziegler-Nichols

You aim finding K_p, K_i, K_d ?

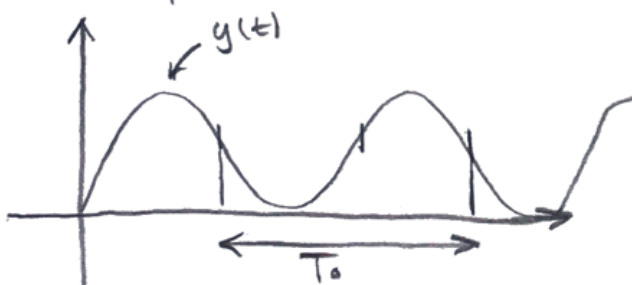
step 1)

closed-loop. Assumption, $G(s)$ can be stabilized by $C(s) = K_p$.

$K_i = 0; K_d = 0$ and use K_p



from $K_p \ll 1$ start increasing $K_p = K_0$



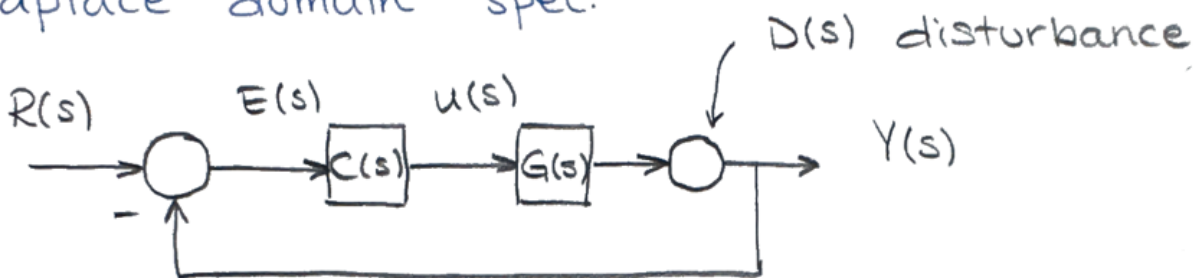
T_0 - period

with K_0 we are at the stability limit.

step 2)

Z/N	K_p	Λ/K_i	$1/K_d$
P	$0,5K_0$	-	-
P I	$0,45K_0$	$T_0/1,2$	-
PID	$0,6K_0$	$T_0/2$	$T_0/8$

Laplace domain spec.



$$D(s) \equiv 0$$

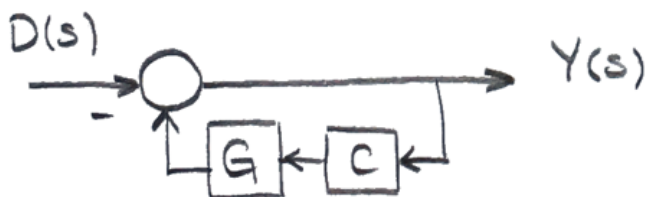
$$G_c(s) = \frac{Y(s)}{R(s)}$$

$$Y(s) = \frac{L(s)}{1 + L(s)} \cdot R(s) = T(s) \cdot R(s)$$

complementary sensitivity

how good is the tracking?

$$R(s) \equiv 0 \quad Y(s)/D(s)$$



$$Y(s) = \frac{L(s)}{1 + L(s)} D(s) \left[\approx \frac{\text{forward}}{\text{loop ratio}} \right]$$

$$Y(s) = S(s) \cdot D(s)$$

$S(s)$ - sensitivity

No matter what you select for $C(s)$

Theorem

$$T(s) + S(s) = 1$$

Guaranteed tracking

$$e_{\infty} = 0$$

$$E(s) = 1 \cdot R(s) - Y(s) = (\cancel{T(s)} + S(s)) R(s) - \cancel{T(s)} \cdot R(s)$$

↑
theorem

$$Y(s) = T(s) \cdot R(s)$$

$$E(s) = S(s) \cdot R(s) = \frac{1}{L(s) + 1} \cdot R(s)$$

$$\lim_{t \rightarrow \infty} e(t) \Big| = \lim_{t \rightarrow 0} s \cdot \frac{1}{L(s) + 1} \cdot \frac{1}{s}$$

$R(s) = \frac{1}{s}$

$$e_{\infty} = \lim_{s \rightarrow 0} \frac{1}{L(s) + 1}$$

Suppose

$$L(s) = \frac{1}{s} \cdot L_0(s)$$

remaining rational transfer fcn.

$$e_{\infty} = \lim_{s \rightarrow 0} \frac{s}{s + L_0(s)} = 0$$

If the loop has integrator it eliminates $e(t)$.

4) Nyquist theory I

$$L(s) = \frac{b(s)}{a(s)} \quad - \text{ loop behaviour}$$

$$T(s) = \frac{L(s)}{L(s) + 1} \quad - \text{ closed loop}$$

1) understand what is the diff. between $L(s) \sim T(s)$

2) Use conditions on $L(s)$ to get stability for closed-loop.