

POLE PLACEMENT
IM S.S. DOMAIN

STATE OBSERVER
DISTURBANCE ESTIMATION

Advanced and Multivariable Control

Pole placement control → $\left\{ \begin{array}{l} \text{method for synthesis} \\ \text{of stabilizing regulator} \\ \text{on MIMO case} \end{array} \right\}$

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BASIC

Problem statement

From a linear system

Consider the system

Assumption: \Rightarrow system such that the state $x(t)$ is measurable (available)

$$\dot{x}(t) = Ax(t) + Bu(t), \quad x \in R^n, \quad u \in R^m$$

if $x(t)$ available you don't need output transform
because the state info is greater than output info

and the control law
state feedback

$$u(t) = -Kx(t) + \gamma(t), \quad K \in R^{m,n}, \quad \gamma \in R^m$$

$m \times n$ MATRIX

additional terms

↓ GOAL

Problem: we want to design a feedback matrix K such that the closed loop system

$$\dot{x}(t) = (A - BK)x(t) + B\gamma(t)$$

has prescribed eigenvalues (of the matrix $A - BK$)

stated in this way $\gamma(t)$ seems useless,
↑ the specification depends on $A - BK$

↑ has prescribed eig values

additional input used for
other reasons

L,
(2nd step
of design)

Remark: we are assuming that the state is measurable (not realistic in many cases). This hypothesis will be removed later on

NOT unique,
it can take
different
meanings

but sometimes this additional input can
be useful for pole placement and to consider
additional requirements

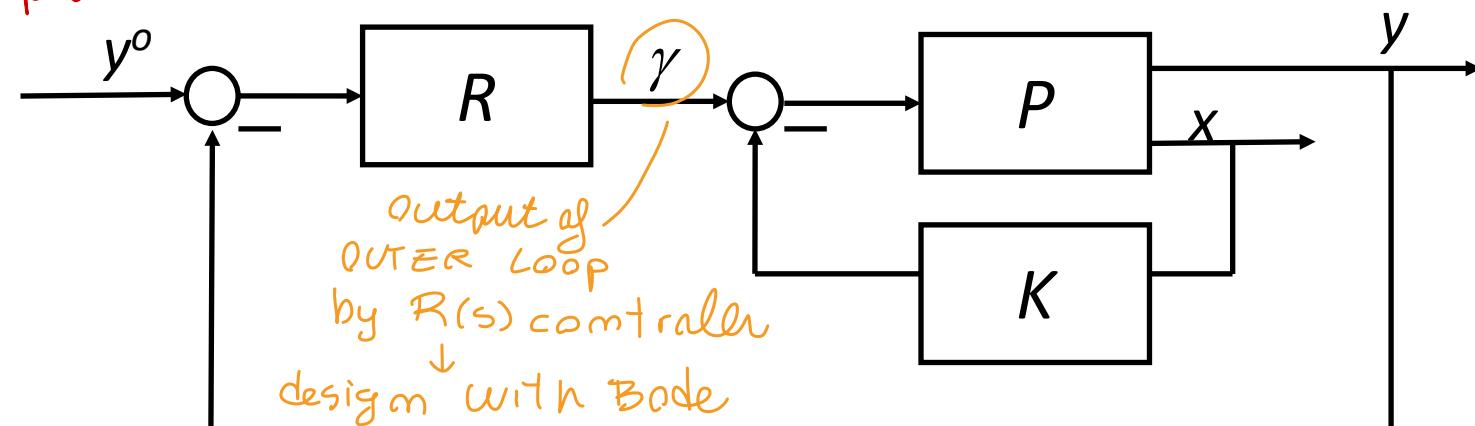
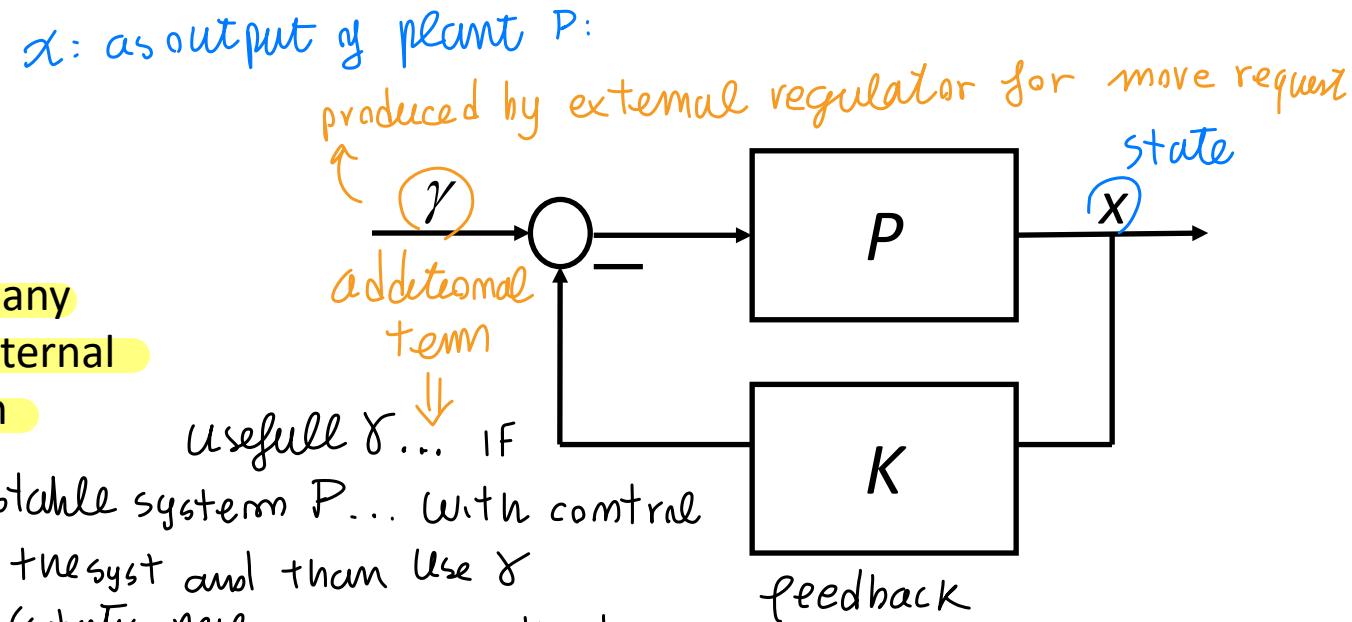
Remarks → Overall Schéma

The additional input γ does not play any role, but it can be useful to design external loops, while K is used for stabilization

for example we have an unstable system P ... with control law stabilizing we stabilize the syst and then use γ to close an external loop (static performance or other)

IF P unstable \rightarrow NO Bode \rightarrow Nyquist
 \rightarrow pole placement \Rightarrow by K MATRIX

Inner feedback used to stabilize, while $R(s)$ used for performance (for example, one can use the Bode criterion)

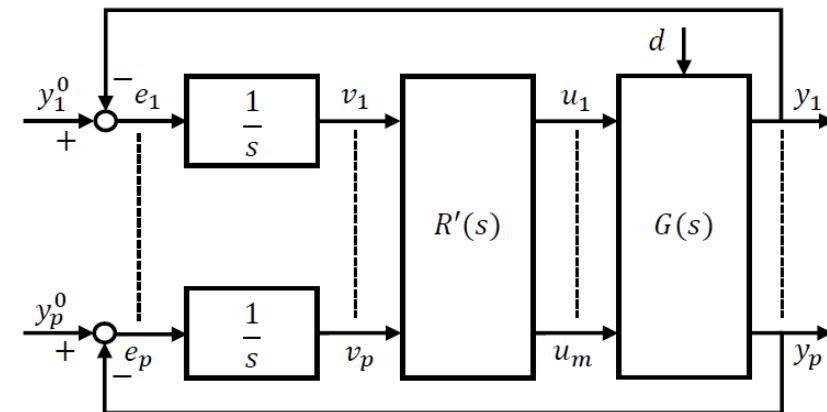


Criterions, standard techniques
IM stable syst.

Scheme with integrators

If we want integrators, first we add integral action... then we design the pole placement controller for enlarged system made by plant state $x(t)$ and integrator state $v(t)$

$$\begin{bmatrix} \dot{x}(t) \\ \dot{v}(t) \end{bmatrix} = \begin{bmatrix} A & 0 \\ -C & 0 \end{bmatrix} \begin{bmatrix} x(t) \\ v(t) \end{bmatrix} + \begin{bmatrix} B \\ 0 \end{bmatrix} u(t) + \begin{bmatrix} 0 \\ I \end{bmatrix} y^0$$

**Control law**

$$u(t) = -K \begin{bmatrix} x(t) \\ v(t) \end{bmatrix} = -\begin{bmatrix} K_x & K_v \end{bmatrix} \begin{bmatrix} x(t) \\ v(t) \end{bmatrix}$$

PARTITION of $[K]$ in 2 block

K designed for the new pair \bar{A}, \bar{B} such that $\bar{A} - \bar{B}K$ has prescribed eigenvalues

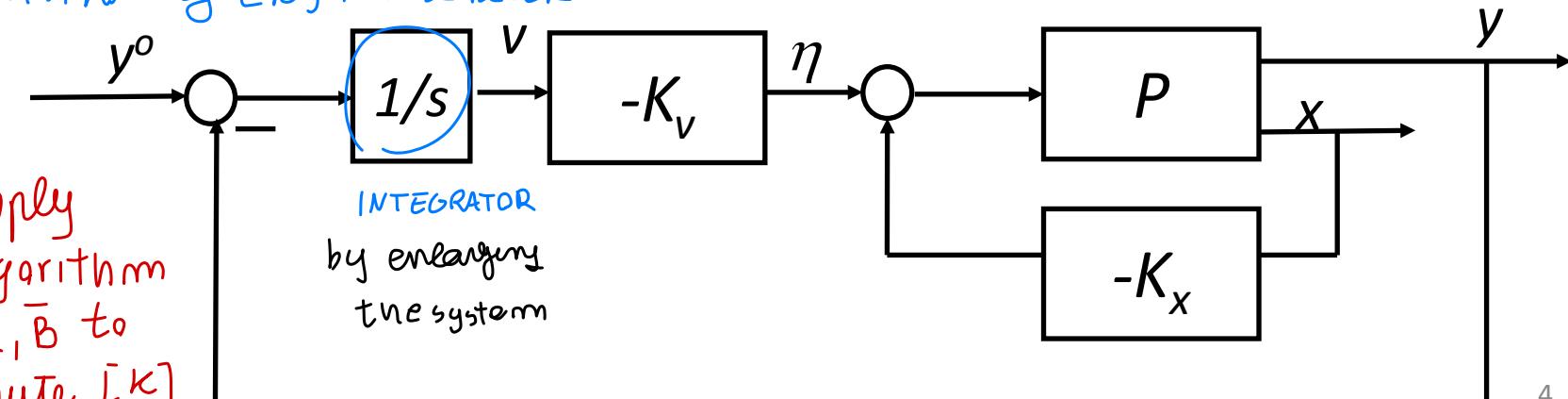
$$u(t) = -Kx(t) + K_v v(t)$$

new scheme!

{ to use structure with integral action... new system enlarged described by \bar{A}, \bar{B}

apply algorithm to \bar{A}, \bar{B} to compute $[K]$

INTEGRATOR by enlarging the system



Algorithm

Necessary and sufficient condition for the solution of the pole placement problem (with measurable state) is that the pair (A, B) is reachable

↑ for the control problem

↳ or of (\hat{A}, \hat{B}) of integrators

Sketch of the proof

↓ given a generic lm. system you can partition A as triang. matrix

By means of a suitable state transformation, it is possible to write the system as

$$\begin{bmatrix} \dot{\hat{x}}_r(t) \\ \dot{\hat{x}}_{nr}(t) \end{bmatrix} = \begin{bmatrix} \hat{A}_r & \hat{A}_x \\ 0 & \hat{A}_{nr} \end{bmatrix} \begin{bmatrix} \hat{x}_r(t) \\ \hat{x}_{nr}(t) \end{bmatrix} + \begin{bmatrix} \hat{B}_r \\ 0 \end{bmatrix} u(t)$$

Reachable ↓
subset of state Unreachable

the pole assignment
problem can be
solved only if all
reachable ...
↓

If we take

$$u(t) = -\hat{K}\hat{x}(t) = -[\hat{K}_r \quad \hat{K}_{nr}] \begin{bmatrix} \hat{x}_r(t) \\ \hat{x}_{nr}(t) \end{bmatrix}, \quad \hat{K}_r \in R^{m, \nu_r}$$

eig of $\hat{A}_r - \hat{B}_r \hat{K}_r$:

we obtain

Close loop matrix \dashrightarrow can be modified by a proper choice of \hat{K}_r

$$\hat{A} - \hat{B}\hat{K} = \begin{bmatrix} \hat{A}_r - \hat{B}_r \hat{K}_r & \hat{A}_x - \hat{B}_r \hat{K}_{nr} \\ 0 & \hat{A}_{nr} \end{bmatrix}$$

↑
eig of $\hat{A}_{nr} \rightarrow$ unreachable part

In case of **stabilizable System** \rightsquigarrow eig of unreach part are
asympt. stable so their

in general we assume
reachable system
and assign eig values of $A - BK$ (**Pole placement**)

PROBLEM \rightarrow at the end we have to find K such that
 $A - BK$ has prescribed eig values

Single input systems ($m=1$) - The Ackermann's formula

ALGORITHM

Define the reachability matrix (square for $m=1$)



$$M_r = [B \ AB \ \dots \ A^{n-1}B]$$

for single input system is a square $m \times m$ system

reachable system $\Rightarrow M_r$ NON SINGULAR

Assume you want to obtain with a closed-loop system with characteristic polynomial



$$P(s) = (s + \bar{p}_1)(s + \bar{p}_2)\dots(s + \bar{p}_n) = s^n + p_{n-1}s^{n-1} + \dots + p_1s + p_0$$

You specify the closed loop eigenvalues \bar{p}_i : desired \rightarrow (obtaining)

Define

evaluate $P(s=A)$:

$$P(A) = A^n + p_{n-1}A^{n-1} + \dots + p_1A + p_0I$$



Then, the solution to the pole placement problem is given by the Ackermann's formula

on practical P.O.V

there are other computation approaches to obtain K

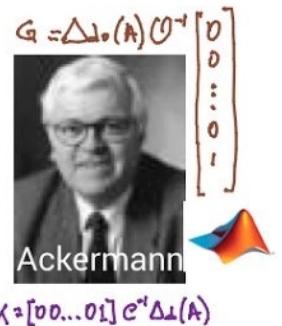


$$K = [0 \ 0 \ \dots \ 0 \ 1] (M_r^{-1}) P(A)$$

unique matrix

M_r^{-1} inverse, which exist

desired polynomial @ $A=s$
In the textbook proofs and many more considerations



[Acker.m] in Matlab!

Example Ackermann's formula

↓ [mechanical approach]

DIFFICULT task:

define specification

Where to place poles!?

3) compute reach. matrix

1) WRITE

2) compute $P(A)$

$$M_r = [B \ AB] = \begin{bmatrix} 1 & 5 \\ 2 & 4 \end{bmatrix}$$

$$K = [0 \ 1] M_r^{-1} P(A) = [2 \ 3]$$

Ackermann's formula ↗

II ORP system
 $A = \begin{bmatrix} -1 & 3 \\ 0 & 2 \end{bmatrix}, \quad B = \begin{pmatrix} 1 \\ 2 \end{pmatrix}$

desired eig position in $(-3), (-a)$ in closed loop

$$P(s) = (s + 3)(s + 4) = s^2 + 7s + 12$$

$$P(A) = A^2 + 7A + 12I = \begin{bmatrix} 6 & 24 \\ 0 & 30 \end{bmatrix}$$

compute inverse

$$M_r^{-1} = \begin{bmatrix} -4/6 & 5/6 \\ 2/6 & -1/6 \end{bmatrix}$$

exist of
Reachable syst

$$\boxed{A - BK} = \begin{bmatrix} -3 & 0 \\ -4 & -4 \end{bmatrix}$$

overall
closed loop

new eig values,
positioned
where desired!

Where to place the poles? This is the difficult task

(NNT specific rules)

(Define specification thinking of I / II ORD systems!)

An idea is try to impose that the closed loop system (approximately) behaves as a first order or second order system
(specify that by phase margin is a good indicator of closed loop system characteristic for SISO)

$\varphi_m \text{ small} \rightarrow \text{II ORD with } \xi \approx \varphi_m/100 / \varphi_m \text{ LARGE} \rightarrow \text{I ORD}$

In the case of first order systems, the «desired» transfer function and step response are

$$G(s) = \frac{\mu}{1+sT} \longleftrightarrow y(t) = \mu(1 - e^{-t/T})$$

↓

step input response

you could find the required poles looking to step response

respect our requirements you choose T to have desired behaviour

So, you can choose the time constant T , or (equivalently) the pole $-1/T$ and fix one of the closed-loop poles at that value
(I imagine closed loop \simeq I ORD, select ideal pole position, the others @ higher freq)

Then, the remaining closed-loop poles can be chosen to be «much faster» than this pole, which is the «dominant one»
@ higher frequency!

Always remember that it is not advisable to choose a closed-loop system must faster than the open-loop one.
This could lead to the saturation of the actuators, or to an excessive use of them

\uparrow too fast closed loop \rightarrow SATURATION \rightarrow No more linear model
TECHNOLOGICAL problem of fast feedback!

Two dominant poles

I may request for little oscillation \rightarrow so FASTER response

≈ II ORD systems similar reasoning

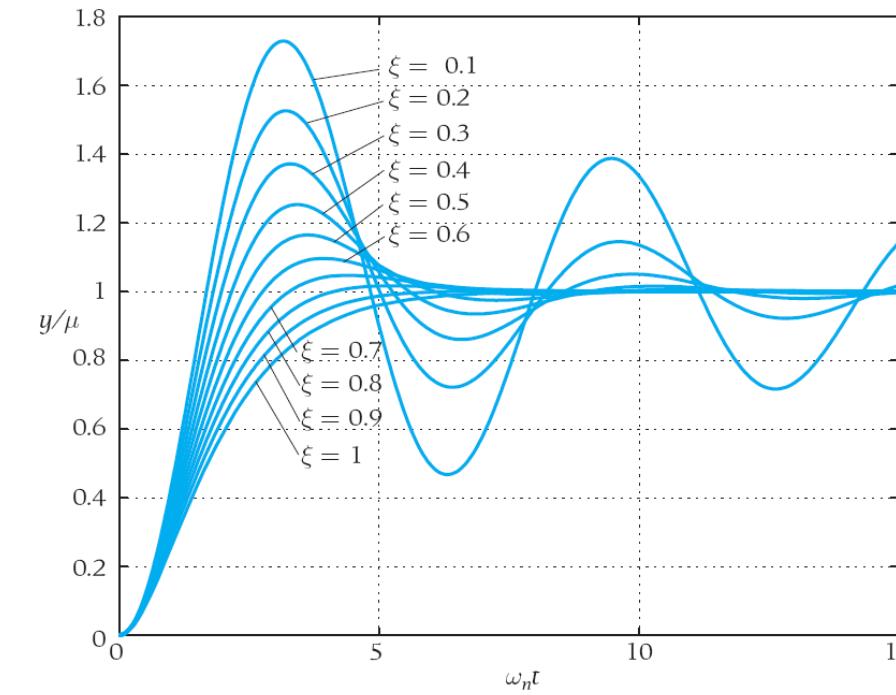
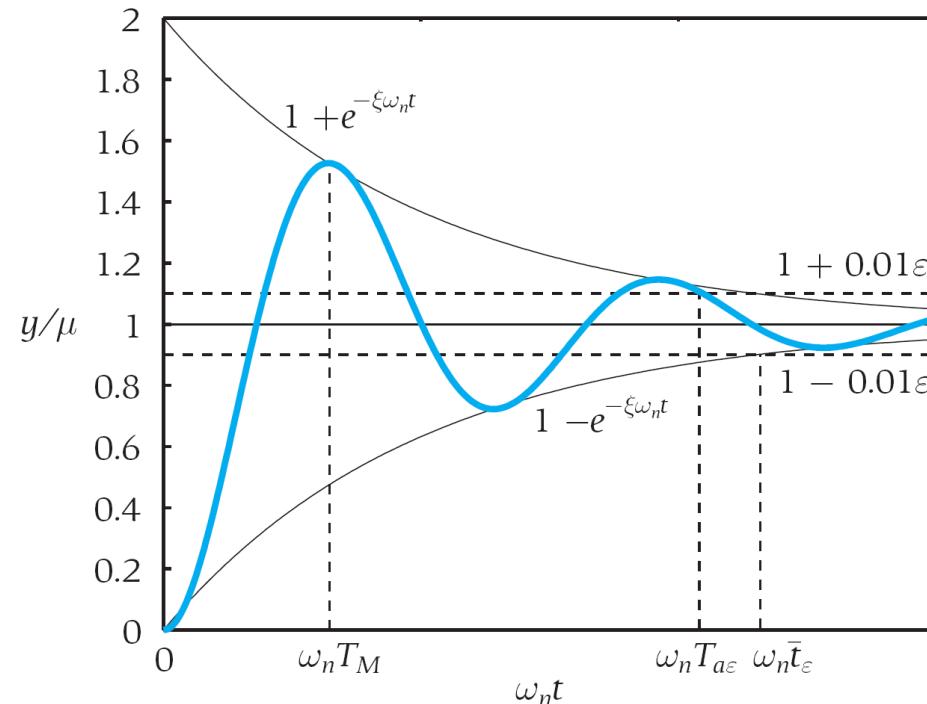
$$G(s) = \frac{\mu \omega_n^2}{s^2 + 2\xi \omega_n s + \omega_n^2}$$

MAX overshoot

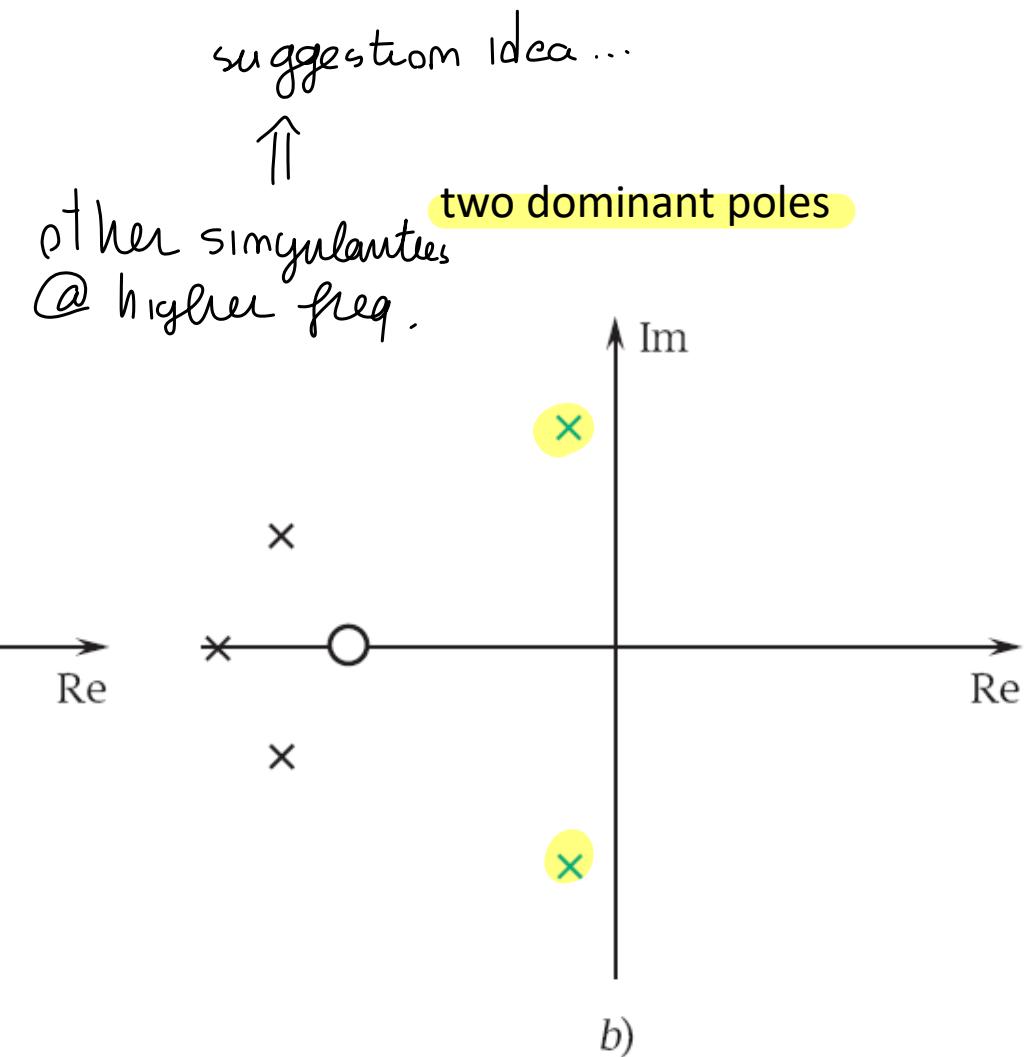
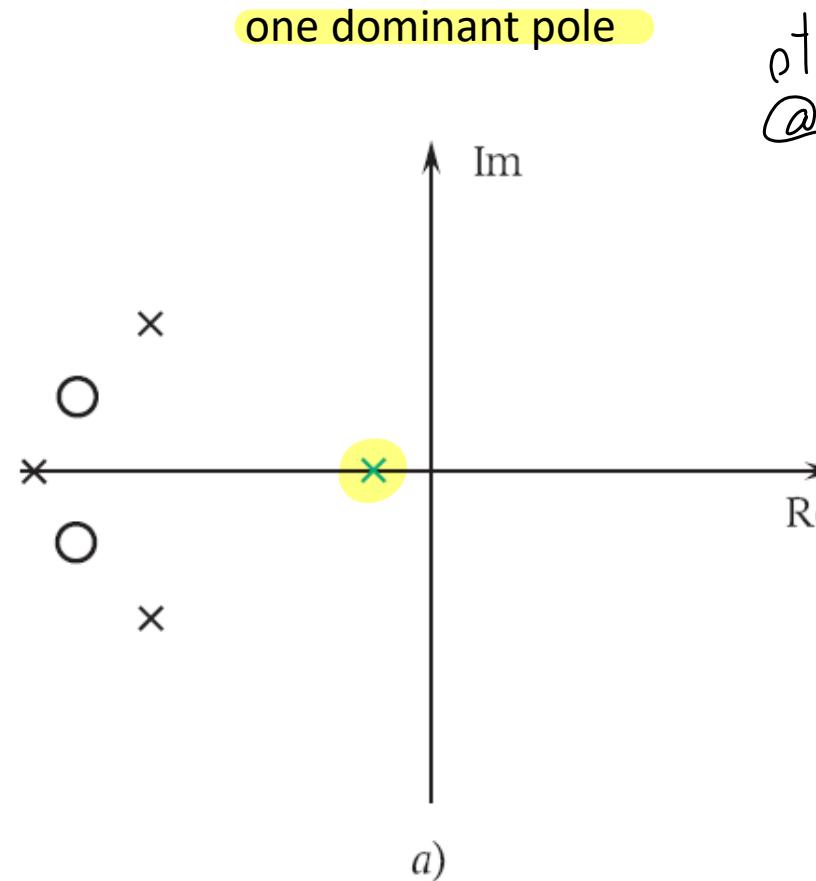
settling time
 \downarrow what you want close loop

(II ORD resp.)

Characteristics of the step response



select 2 poles and the others @ higher freq



Where to place the poles!

Systems with many inputs ($m > 1$)

HIMO case! for more inputs more DOF on the choice \Rightarrow in SISO case you have just one choice! Solution (Ackermann)

The problem can have many solutions. In any case, the assumption of reachability is mandatory
(difficult to understand BEST)

↑
necessary condition

Example

$$\dot{x}(t) = Ax(t) + Bu(t) = Ax(t) + b_1 u_1(t) + b_2 u_2(t) \quad \left\{ \begin{array}{l} 3 \text{ STATES} \\ 2 \text{ INPUT} \end{array} \right.$$

$\left\{ \begin{array}{l} \text{canonical} \\ \text{form} \end{array} \right\}$

$$A = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 1 & 0 \end{bmatrix}, \quad B = [b_1 \ b_2] = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \end{bmatrix}$$

2 INPUTS..
by 2 rows of B

Both (A, b_1) and (A, b_2) are reachable, so that we can use just one input to assign the closed-loop poles

because both reachable
↓
reduced reachability test!

↓
Use one input for Ackermann

formula to assign poles, while don't use the other input!

continued...

• Solution 1

$$\left\{ \begin{array}{l} u_1(t) = -K_1 x(t) \\ u_2(t) = 0 \end{array} \right. \longleftrightarrow \dot{x}(t) = (A - b_1 K_1)x(t)$$

, use only $u_1(t)$ with Ackermann !

impose this eeg

don't use $u_2(t)$!

2 SOLUTIONS! NOT unique!

• Solution 2

$$\left\{ \begin{array}{l} u_1(t) = 0 \\ u_2(t) = -K_2 x(t) \end{array} \right. \longleftrightarrow \dot{x}(t) = (A - b_2 K_2)x(t)$$

Use only $u_2(t)$
Same type of control law

K_2 can be computed with the Ackermann's formula (only one input!) to impose the eigenvalues of $(A - b_2 K_2)$

In both cases $u(t) = -Kx(t)$ where $K = \begin{bmatrix} K_1 \\ 0 \end{bmatrix}$ (first solution) or $K = \begin{bmatrix} 0 \\ K_2 \end{bmatrix}$ (second solution)

don't seems clever solution

using just one input BUT both usable ..

on MIMO case \rightarrow we can have many different solutions

What to do in practice?

(STANDARD SOFTWARE) ← NOT unique solution ↑

(to compute $[K]$ according to different criteria)

↪ algorithm such that from A, B nominal even for variation
good solution

Use the function `place.m` of Matlab, it uses all the available inputs and guarantees some robustness of the solution with respect to small variations of the parameters of the matrices A, B .

The only warning is that you must specify closed-loop poles each one different from the others (not a big issue, there is no difference in placing two poles in -1 or one pole in -1 and the other one in -1.001)

$[K, \text{PREC}] = \text{place}(A, B, P)$

take as input A, B matrix and
location of eigenvalues desired, by P

returns PREC, an estimate of how closely the eigenvalues of $A - B^*K$ match the specified locations P (PREC measures the number of accurate decimal digits in the actual closed-loop poles). A warning is issued if some nonzero closed-loop pole is more than 10% off from the desired location.

{ multiple solutions!
can compute $[K]$
with more SWs }

← { ^{LIMITATION} avoid to place the poles on same position of complex plane ⇒ use different pole position }
(im - 1, -1.01 ... NOT 2im - 1)

IF you don't have the state of syst..

Nonmeasurable state

↓ you need to use an element (meas dyn system) which given the in/out of syst compute an estimation \hat{x} of the state at any time used for control.

Procedure

(OBSERVER based on pole placement control)
(Kalman Filter) ↓ here simpler, using duality of control / observer design

1. first a **state observer** is built. The **observer** is a system which, based on the **input and output** measurements, computes an **estimate \hat{x}** of the system's state x . The **observer introduced** in the following is a **linear time-invariant system** with prescribed eigenvalues;
2. the **estimated state \hat{x}** is **used in the pole assignment control law** in place of the real state x .



State observers

System specified also by output transform

$$\begin{cases} \dot{x}(t) = Ax(t) + Bu(t), & x \in R^n, u \in R^m \\ y(t) = Cx(t) + Du(t), & y \in R^p \end{cases}$$

Observer ↓ only measurable signal
from $u(t), y(t)$ we have to estimate $x(t)$

$$\dot{\hat{x}}(t) = A\hat{x}(t) + Bu(t) + L[y(t) - C\hat{x}(t) - Du(t)]$$

Same system dyn. (open loop)

$L \in R^{n,p}$ ← $L \in R^{n,p}$ is the gain of the observer (and its design parameter) **and estimation**

We must consider also the output transformation now

we also use a feedback observer
improving performance
by a Gain (L)
by difference of output

$e_y(t) = [y(t) - C\hat{x}(t) - Du(t)]$ is the output estimation error

$$\text{if } \hat{x} = x : y(t) - (\hat{x} - Du(t)) = 0$$

$e(t) = x(t) - \hat{x}(t)$ is the state estimation error

we want to bring $e(t) \rightarrow 0$

|| We have to compute || → L

from OBSERVER formula

$$\dot{\hat{x}} = (A - LC) \hat{x} + (B - LD) u(t) + Ly(t)$$

↓
on a PRACTICAL P.O.V

{ on different
matrices }

to implement it on MATLAB

↳ to implement an observer

to understand how the system works...

Dynamics of the state estimation error



$$\begin{aligned}\dot{e}(t) &= \dot{x}(t) - \dot{\hat{x}}(t) \\ &= Ax(t) + Bu(t) - A\hat{x}(t) - Bu(t) - L[y(t) - C\hat{x}(t) - Du(t)] \\ &= A(x(t) - \hat{x}(t)) - LC(x(t) - \hat{x}(t)) = (A - LC)e(t)\end{aligned}$$

We have

to choose
($A - LC$) asymptotically stable region

No forcing input

evolution as the one of a linear system with only free motion

If the eigenvalues of ($A - LC$) are asymptotically stable, the state estimation error tends to zero. The idea is to use pole assignment to select the observer gain L to assign these eigenvalues

{ to design }
that regulator

FIND L such that $(A - LC)$ has eig as desired

The eigenvalues of $(A - LC)$ are the eigenvalues of $(A' - C'L')$

TRANSPOSED

equivalent problem

The problem of assigning the eigenvalues of $(A' - C'L')$ is equivalent to assigning the eigenvalues of $(A - BK)$

solve equivalent problem as described before!

$A \rightarrow A'$, $B \rightarrow C'$, $m \rightarrow p$, $K \rightarrow L'$

on pole placement control problem

as reachability means I'm able to have an input which modify system dyn. observability means that looking output I understand the state !!

A necessary and sufficient condition for the design of an asymptotic observer with arbitrarily specified eigenvalues is that the pair (A, C) is observable | (CNS)

↓
(detectability! \Rightarrow nom obs part has stable eig values,
they cannot be well estimated but they don't affect the syst)

► For single output systems we can use the Ackermann's formula with the proper substitutions

{ You can use and redefine
Ackermann formula from A' , B' obtain L' } ↓ polymomials of desired
eig on A'

$$\hookrightarrow L' = [0 \ 0 \ \dots \ 0 \ 1] M_o^{-1} P(A') \text{ or } \| L = P(A) M_o^{-1}$$

{ don't make syst
too much faster or you will } ↗ inverse of
observability, transposed
have problems physical
PREVIOUSLY

$$\begin{bmatrix} 0 \\ 0 \\ \dots \\ 0 \\ 1 \end{bmatrix} \|$$

► For multi output systems the suggestion is to use place.m with the proper substitutions ($A \rightarrow A'$, $B \rightarrow C'$)

for OBSERVER

a faster than state feedback control law!



How to choose the eigenvalues of $(A - LC)$? Typically **much faster** than the ones of $(A - BK)$.

The observer is an algorithm, it does not suffer from actuators' saturations

{ no speed variations }

→ don't impact actuator
(only problem for unbalanced problem) ↑ force the observer
reconstruct state faster

Matlab – Simulink implementation

OBSERVER

$$C = [I]$$

because $y = \hat{x}$

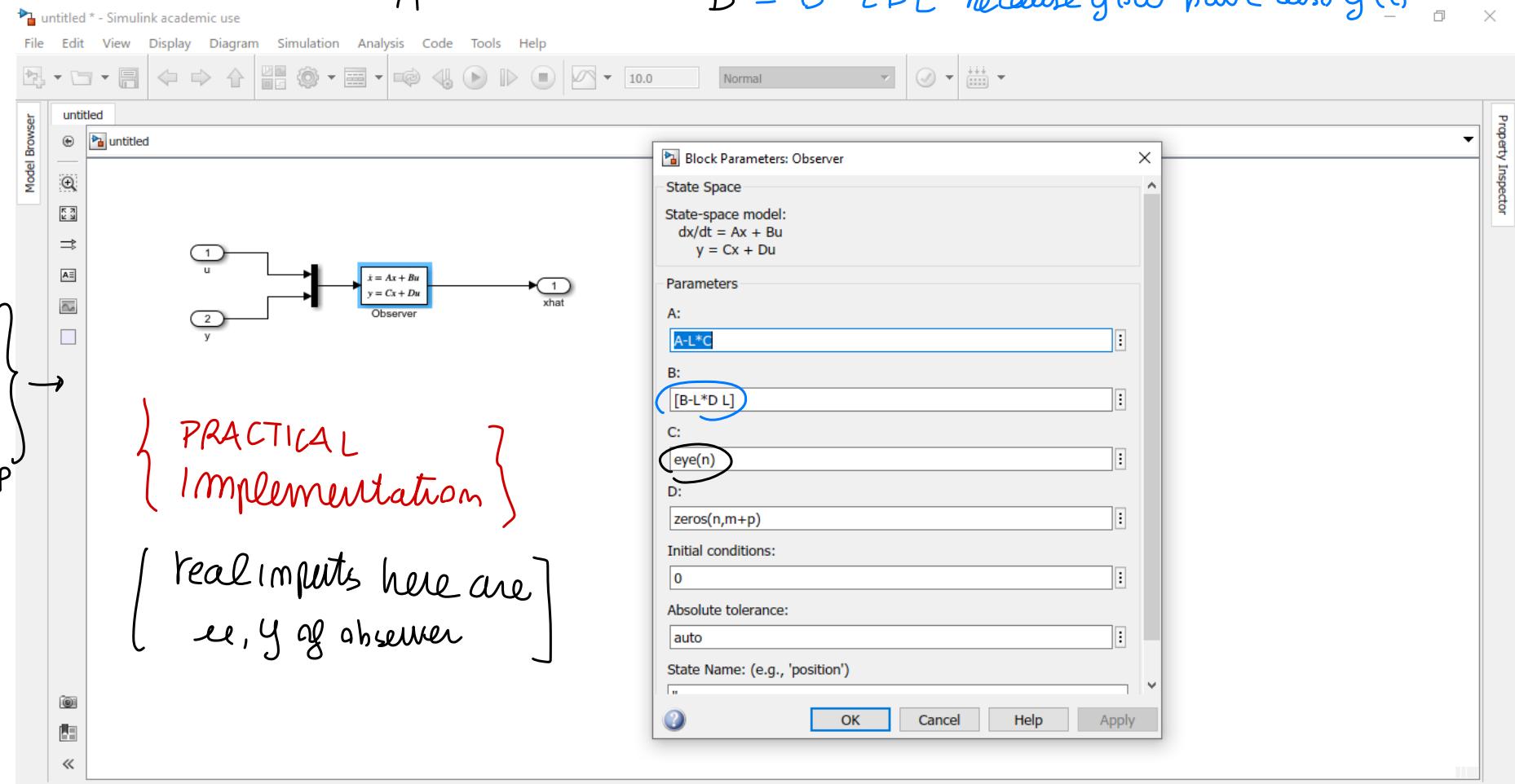
and $D = [0]_{m, m+p}$

dy m system
with proper I/O

$$\dot{\hat{x}}(t) = A\hat{x}(t) + Bu(t) + L[y(t) - C\hat{x}(t) - Du(t)]$$

$$\dot{\hat{x}}(t) = \underbrace{(A - LC)}_A \hat{x}(t) + \underbrace{(B - LD)}_B u(t) + Ly(t)$$

$\tilde{B} = B - LDL$ because you have also $y(t)$



Systems with disturbances

(2 CASES)

- 1) **measurable disturbance $d(t)$ KNOWN** \rightarrow SIMPLE Case, consider also $d(t)$, include disturbance exactly as an INPUT $u(t)$

$$\dot{\hat{x}}(t) = A\hat{x}(t) + Bu(t) + Md(t) + L[y(t) - C\hat{x}(t) - Du(t) - Nd(t)]$$

(end cancel from
this difference)

STATE influence

from computation

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

influence output
error!

same as before!
same PROBLEM of state estimation

- 2) **non measurable disturbance**

{ more difficult! }

\hookrightarrow IF unknown $d(t)$ you cannot use it for observer definition!

(OBSERVER theory : to estimate unmeasurable quantity)

$$\dot{\hat{x}}(t) = A\hat{x}(t) + Bu(t) + L[y(t) - C\hat{x}(t) - Du(t)]$$

computing $\dot{\hat{x}} - \dot{x}$...

$$\dot{e}(t) = (A - LC)e(t) + (M - LN)d(t)$$

as before
Free motion
 $e_0 \rightarrow 0$

for unk d, NOT possible to have $e(t) \rightarrow 0$!
forcing terms

also for constant d

additional term! \Rightarrow cannot be cancelled, UNKNOWN terms!

OBSERVER theory important in many application, also for monitoring process in safety application \Rightarrow OBSERVER to consider VIRTUAL sensor

$\hat{x} \rightarrow x$

TRUBLE

where you estimate variable

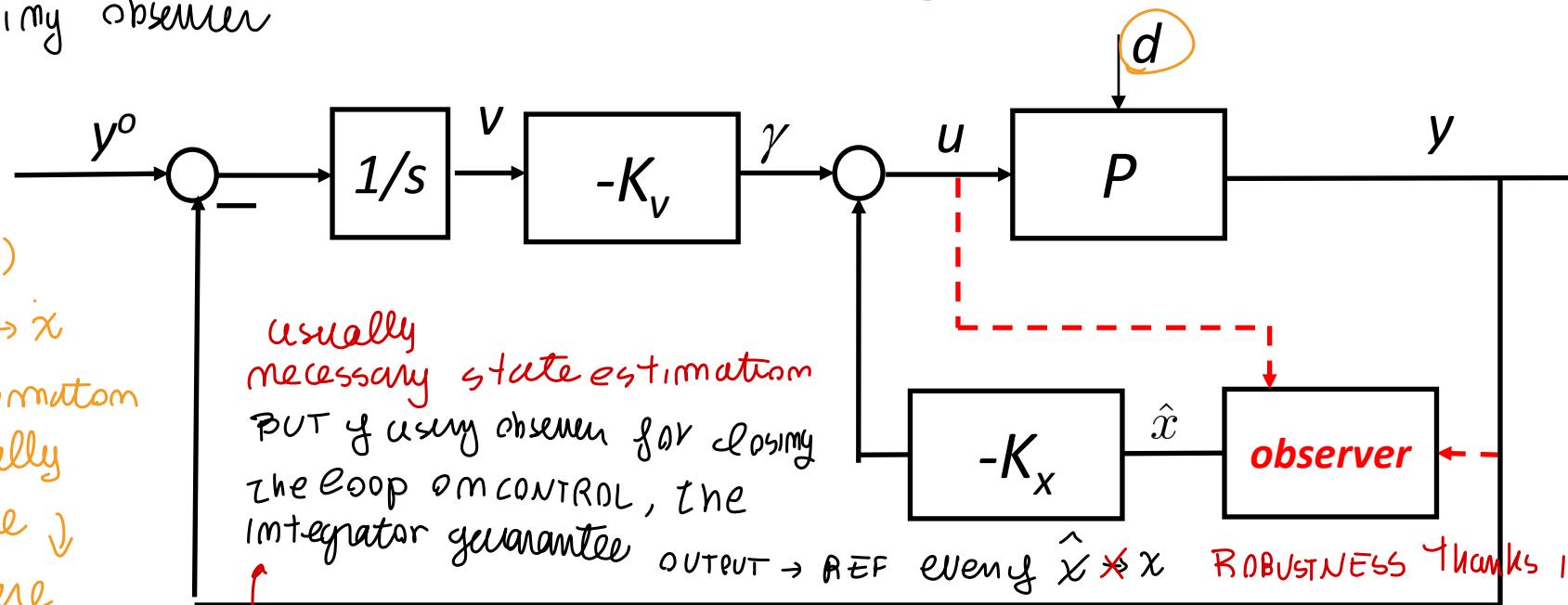
↳ other
consideration ...

Is it really necessary to have the correct state estimation?

Consider the system with integrators and observer (Q error for constant reference)

If the closed-loop system remains asymptotically stable, even when the state estimate is not correct, for constant references and disturbances the input of the integrators must be asymptotically zero

→ solve state optimiz by integrator \Rightarrow design (K_v, K_x) with pole placement, than stabilizing observer



IF act $d(t)$
unlikely $\hat{x} \rightarrow x$
WRONG estimation
BUT y overall
asympt. stable
if y^o, d are
constant, all signal tends to constant value, so far same u goes to error $e \rightarrow 0$
otherwise ramp!

Sometimes we need a state estimate to close the loop → instead of usual control law, we use \hat{x} , not x
(Q error for constant reference)

IF $d(t)$ unknown impossible proper $x(t)$ reconstruct \downarrow on a PARTICULAR CASE
Estimation of constant disturbances

Assume to know that d is constant, or at least constant for long periods of time.

It is useful to estimate it for two reasons: $\downarrow d(t)$ can be estimated to have $e(t) \rightarrow 0$

- { 1) It is possible to correctly estimate the state of the system
- 2) The estimate of the disturbance can be used in a control scheme with direct compensation
 \downarrow IF you have $d(t)$ estimation \Rightarrow compensator in control scheme

Procedure

Assign a dynamics to the disturbance IF $d(t) = \text{constant} \Rightarrow$ (fictitious dynamic term)

$$\dot{d}(t) = 0, \quad d(0) = \bar{d} \in \mathbb{R}^r \quad r := \text{disturbances number}$$

Enlarge the system

↳ well working system

$$\left\{ \begin{array}{l} \begin{bmatrix} \dot{x}(t) \\ \dot{d}(t) \end{bmatrix} = \begin{bmatrix} A & M \\ 0 & 0 \end{bmatrix} \begin{bmatrix} x(t) \\ d(t) \end{bmatrix} + \begin{bmatrix} B \\ 0 \end{bmatrix} u(t) \\ y(t) = [C \ N] \begin{bmatrix} x(t) \\ d(t) \end{bmatrix} + D u(t) \end{array} \right\}$$

from Enlarged system ...
 Basically provided some condition holds!

\hookrightarrow $d(t)$ estimation properly
 Algorithm to estimate
 $\hat{x}, \hat{d} \rightarrow \hat{x}, \hat{d}$!

Use an observer for the enlarged system. When is it possible?

apply observer theory on this new system,
Conditions I need conditions to guarantee
 OBSERVER existence

$$\begin{cases} \begin{bmatrix} \dot{x}(t) \\ \dot{d}(t) \end{bmatrix} = \begin{bmatrix} A & M \\ 0 & 0 \end{bmatrix} \begin{bmatrix} x(t) \\ d(t) \end{bmatrix} + \begin{bmatrix} B \\ 0 \end{bmatrix} u(t) \\ y(t) = \begin{bmatrix} C & N \end{bmatrix} \begin{bmatrix} x(t) \\ d(t) \end{bmatrix} + Du(t) \end{cases}$$

$\bar{A} \in R^{n+r, n+r}$
 $\bar{C} \in R^{p, n+r}$

The pair (\bar{A}, \bar{C}) must be observable !



Observable iff (CNS) for enlarged syst to be obs

- 1. (A, C) observable (ORIGINAL syst must be obs)
- 2. rank $\begin{bmatrix} A & M \\ C & N \end{bmatrix} = n + r$

you don't have invariant zeros...

you cannot reconstruct more disturbance than outputs!

OBSERVABILITY TEST!

Exercise: use the PBH test to prove this result

disturbances estimation
 IF constant dynamic!
 \Downarrow
 we can use this approach
 (used also for $d(t)$)
 ramp / sinusoid with
 proper generalization)

trivial! if original m obs!
 impossible...

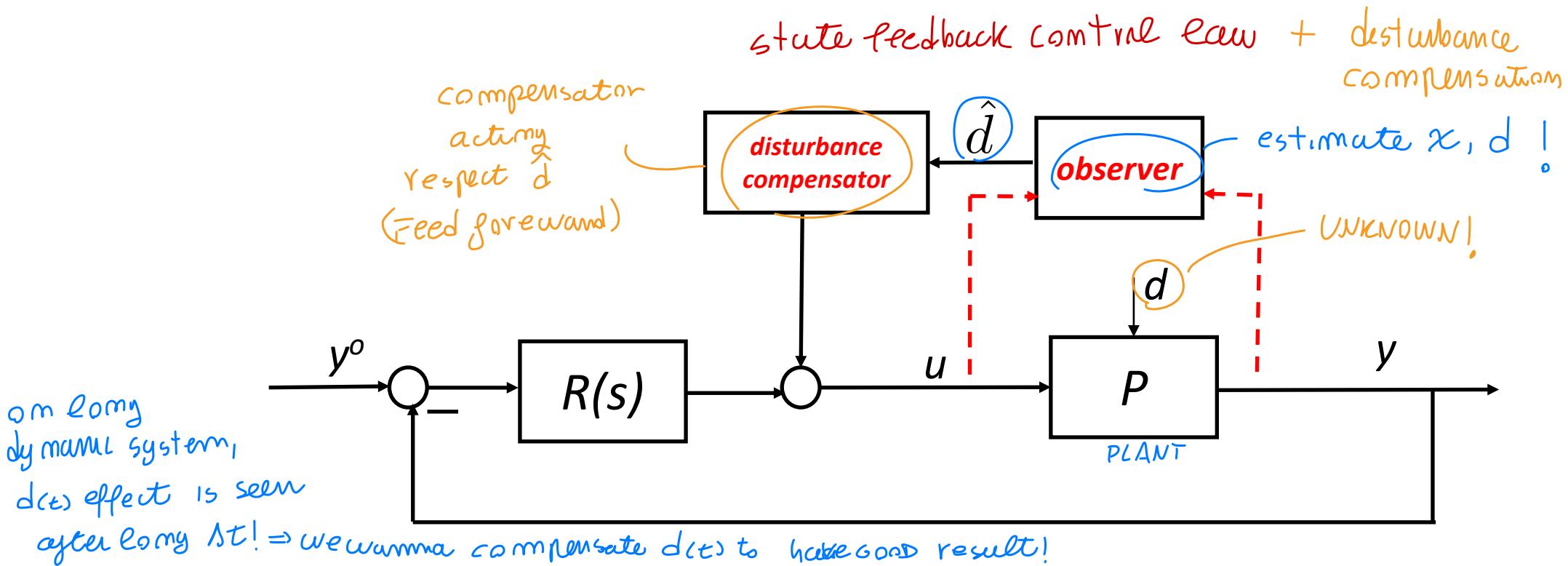
maximum number of
 disturbances that can be
 estimated

P as upper limit
 of d estimated

Scheme with compensator

usefull $d(t)$ estimation also to design control scheme with compensation!

The idea is to design the compensator in such a way to force to be null the transfer function from the (estimated) disturbance to the output



↑ { disturbance compensator: designed such that $H_{yd} = \frac{y}{d} \approx \infty$ overall \rightarrow causal dynamic!
and on practice \rightarrow usage of compensator is fundamental to compensate external effect \rightarrow BETTER PERFORMANCE !!

What are the eigenvalues of the closed-loop system?

Linear

system
(S.S)

$$\begin{cases} \dot{x}(t) = Ax(t) + Bu(t) \\ y(t) = Cx(t) + Du(t) \end{cases}$$

OBSERVER

+

$$\dot{\hat{x}}(t) = A\hat{x}(t) + Bu(t) + L[y(t) - C\hat{x}(t) - Du(t)]$$

CONTROL

$$\text{LAW} \quad u(t) = -K\hat{x}(t) + \gamma(t)$$

OVERALL

$$\begin{cases} \dot{x}(t) = Ax(t) - BK\hat{x}(t) + B\gamma(t) \end{cases}$$

$$\begin{cases} \dot{\hat{x}}(t) = A\hat{x}(t) - BK\hat{x}(t) + B\gamma(t) + L[Cx(t) - C\hat{x}(t)] \end{cases}$$

STATE ESTIMATION error

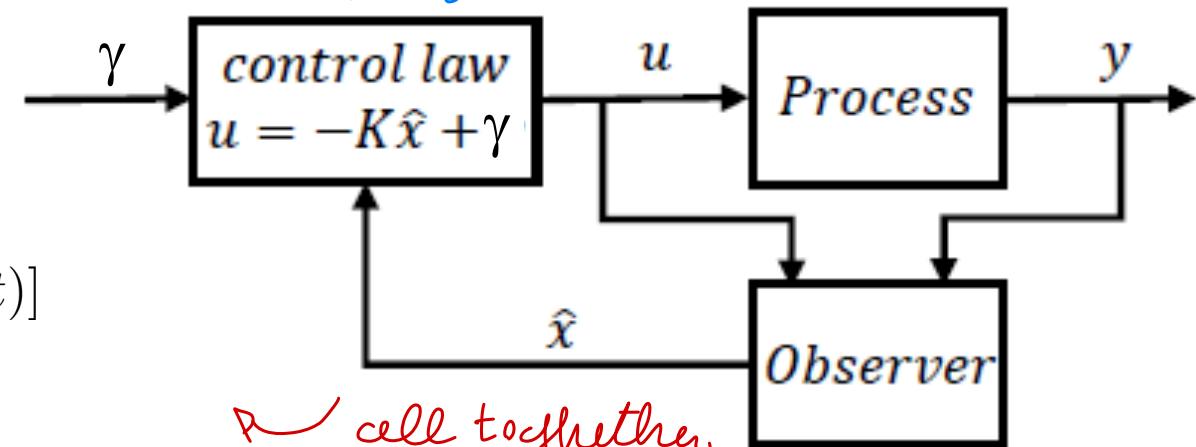
e = x - \hat{x} [No $u(t)$ exogenous variable effect!]

$$\begin{bmatrix} \dot{x}(t) \\ \dot{e}(t) \end{bmatrix} = \begin{bmatrix} A - BK & BK \\ 0 & A - LC \end{bmatrix} \begin{bmatrix} x(t) \\ e(t) \end{bmatrix} + \begin{bmatrix} B \\ 0 \end{bmatrix} \gamma(t)$$

$(x - \hat{x})$

L-dynam matrix of closed loop syst \Rightarrow block triangular

OPEN issues... after we saw the control law
to assign feedback system poles! \Rightarrow BUT you can design
observer A.s imposing A-LC



cell together
observer placement \Rightarrow all works well??!

The eigenvalues are those of $A-BK$ and $A-LC$ **Separation principle**

You can independently design the state feedback control law and the observer, the eigenvalues are maintained when putting everything together

really important RESULT!

eig values overall are the ones of
 $(A - BK)$ and $(A - LC)$
OBSERVER

control law

+

observer

} even overall mixing the two
you can design K, L independently
without issue, maintaining eig. of design

↓

{ SEPARATION PRINCIPLE }

maintained eig values

{ ↑ don't hold for non-ctr systems! \Rightarrow more difficult study }

Until now all studied referred to system described

in time domain \rightarrow overall control scheme, controller
made by OBSERVER (L) + control law (K)

in time domain

↳ how to go in
T.F forms \Rightarrow

Regulator transfer function T.F

Implementation in blocks
(in variable s)

$$\dot{\hat{x}}(t) = A\hat{x}(t) + Bu(t) + L[y(t) - C\hat{x}(t) - Du(t)]$$

+

$$u(t) = -K\hat{x}(t) + \gamma(t)$$

put controller on obs! (1)

$$\dot{\hat{x}}(t) = (A - LC)\hat{x}(t) - (B - LD)K\hat{x}(t) + (B - LD)\gamma(t) + Ly(t)$$

$$u(t) = -K\hat{x}(t) + \gamma(t)$$

↑ no more freedom of $u(t)$

Obtain Regulator on time domain, but you can elaborate the equations from y , δ input

redefine the matrix!

$$\bar{A} = A - LC - BK + LDK, \quad \bar{B} = (B - LD)$$

to write Regulator T.F

$$\dot{\hat{x}}(t) = \bar{A}\hat{x}(t) + \bar{B}\gamma(t) + Ly(t)$$

$$u(t) = -K\hat{x}(t) + \gamma(t)$$

Retransform on Laplace domain

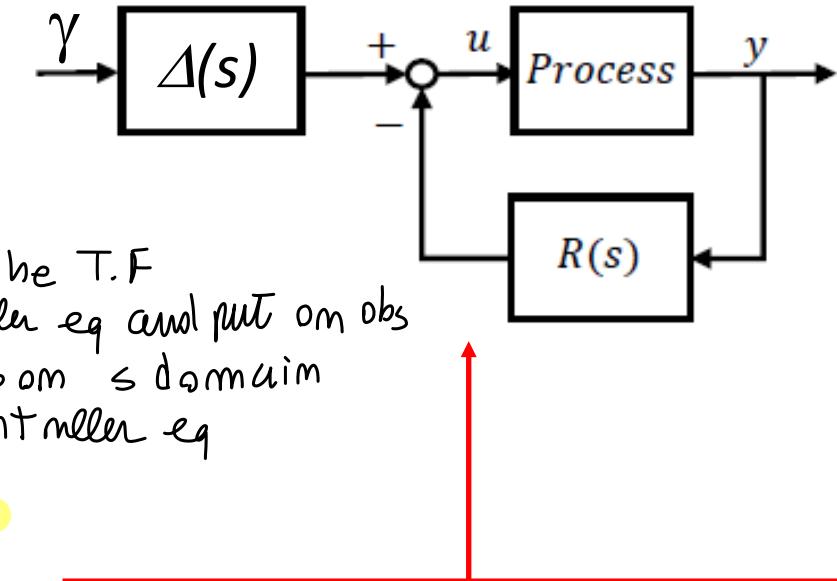
Y contribute

$$U(s) = -K(sI - \bar{A})^{-1}LY(s) + [-K(sI - \bar{A})^{-1}\bar{B} + I]\Gamma(s)$$

OBSERVER

to obtain the T.F

- 1) take controller eq and put on obs
- 2) transform obs on s domain
- 3) go back to controller eq



$$R(s) = K(sI - \bar{A})^{-1}L$$

$$\Delta(s) = -K(sI - \bar{A})^{-1}\bar{B} + I$$

The overall regulator can be expressed in transfer function form

 $\delta(t)$ contribution

From time or Laplace representation of overall Regulation
is a dynamic system \downarrow

$\left\{ \begin{array}{l} \text{made by } m \text{ state eq of observer (MORD)} \\ + \text{ set of algebraic eq (AORD)} \end{array} \right\}$

$(R(s) + A(s))$ are of order m

\Rightarrow if $x(t)$ measurable you have only ALGEBRAIC
equation of $R(s)$ (ORDER \neq)

all Reg dynamics comes from (state eqs) of order m dyn syst

\downarrow

When you have syst with $x(t)$ unmeasurable and a known output

transform $y(t) = Cx(t)$

\downarrow

We don't have $x(t)$ but we know $y(t)$ is a lin
combim. of $x(t)$ \Rightarrow p variable (output) as combination
of the states

IDEA

$\left\{ \begin{array}{l} \text{(Basically) make a STATE transformation (NOT unique)...} \\ \text{such that the output of this transformed variables} \\ \text{will coincide with some states!} \end{array} \right.$

\Downarrow

\rightarrow on transformed system \rightarrow p outputs coincide with the states
which are already available coinciding with output

\hookrightarrow so p states known, I only have to estimate remaining
 $m-p$ states!

\uparrow
(OBSERVER) estimating $x(t) \rightarrow$ because $y(t) = Cx(t)$ combination
of states by $C \rightarrow$ by proper state transform $x(t) = T\tilde{x}(t)$
corresponding to transformed state correspondingly
to the outputs $y(t) \rightarrow$ I obtain p state
REDUCED ORDER OBSERVER \Rightarrow

sequence of operations to apply! (Focus on main steps)

Reduced order observers

OVERALL System

$$\begin{cases} \dot{x}(t) = Ax(t) + Bu(t) \\ y(t) = Cx(t) \end{cases}$$

transform the states by
a suitable state trans Tx

T_1 any matrix such that T
is non singular

(To find T)

Transformed system

$$\begin{cases} \dot{\tilde{x}}(t) = \tilde{A}\tilde{x}(t) + \tilde{B}u(t) \\ y(t) = \tilde{C}\tilde{x}(t) \end{cases}$$

overall syst transform!

$$\tilde{A} = TAT^{-1}, \tilde{B} = TB, \tilde{C} = CT^{-1}$$

PARTITION

Idea: the output is a linear combination of the states

1. Apply a state transformation to the system so that the p outputs coincide with p new states
2. Estimate the remaining $n-p$ states

T_1 easy to build such that
 T is not singular

$$\tilde{x} = Tx = \begin{bmatrix} C \\ T_1 \end{bmatrix} x = \begin{bmatrix} y \\ \tilde{x}_r \end{bmatrix}, \quad \tilde{x}_r \in R^{n-p}$$

(such that \tilde{x} has first p values corresponding to y)

decomposing matrix. (PARTITIONING of \tilde{A})

$$\begin{aligned} \dot{\tilde{x}} &= \begin{cases} \dot{y}(t) \\ \dot{\tilde{x}}_r(t) \end{cases} = \tilde{A}_{11}y(t) + \tilde{A}_{12}\tilde{x}_r(t) + \tilde{B}_1u(t) \\ &= \tilde{A}_{21}y(t) + \tilde{A}_{22}\tilde{x}_r(t) + \tilde{B}_2u(t) \end{aligned}$$

L TO RECONSTRUCT

because
 $y = \tilde{C}\tilde{x}$ only
 $y = Cx$

$$\tilde{C} = \begin{bmatrix} I_p & 0 \end{bmatrix}$$

Observer

equation
on new formulation *KNOWN! measurable*

$$\begin{cases} \dot{y}(t) = \tilde{A}_{11}y(t) + \tilde{A}_{12}\tilde{x}_r(t) + \tilde{B}_1u(t) \\ \dot{\tilde{x}}_r(t) = \tilde{A}_{21}y(t) + \tilde{A}_{22}\tilde{x}_r(t) + \tilde{B}_2u(t) \end{cases}$$

(3)

define

redefine new variables!

$$\begin{cases} \eta(t) = \dot{y}(t) - \tilde{A}_{11}y(t) - \tilde{B}_1u(t) \\ \zeta(t) = \tilde{A}_{21}y(t) + \tilde{B}_2u(t) \end{cases}$$

to have on the right
only KNOWN quantities!Reduced order
system

Now, design an observer for this system

S.S. system form \rightarrow
 { STANDARD }
 OBS FORM OBSERVER

$$\begin{cases} \dot{\tilde{x}}_r(t) = \tilde{A}_{22}\tilde{x}_r(t) + \zeta(t) \\ \eta(t) = \tilde{A}_{12}\tilde{x}_r(t) \end{cases}$$

dyn eq. of STATE
↓
OUTPUT TRANSF.

if (A, C) is observable,
 $(\tilde{A}_{22}, \tilde{A}_{12})$ is observable

original syst obs..

same for new one

I can find L if I can guarantee
 $(\tilde{A}_{22}, \tilde{A}_{12})$ is OBS, say e.g. of
 ↑ OBS well selected

or (equivalent)

III equivalent

$$\dot{\hat{x}}_r(t) = (\tilde{A}_{22} - L\tilde{A}_{12})\hat{x}_r(t) + \tilde{A}_{21}y(t) + \tilde{B}_2u(t) + L\eta(t)$$

Problem: η contains the derivative of y , not wise from a numerical point of view

I have an OBS of order $\hat{x}_r \rightarrow (m-p)$ obtained by my obs design

If this is
 asympt stable you
 will be able to
 solve your problem

Issue ...

on $\dot{x}_r(t)$ we treat $y(t)$ as output
BUT it is measurable depending on $y(t)$
BAD dependency on numerical P.O.V



We wanna remove the derivative
dependency (Bad numerically)

remove y !

$$\dot{\hat{x}}_r(t) = \left(\tilde{A}_{22} - L\tilde{A}_{12} \right) \hat{x}_r(t) + \tilde{A}_{21}y(t) + \tilde{B}_2u(t) + L\eta(t)$$

put \dot{y} on the left hand side of the equation

$$\eta(t) = \dot{y}(t) - \tilde{A}_{11}y(t) - \tilde{B}_1u(t)$$

$$\dot{\hat{x}}_r(t) - L\dot{y}(t) = (\tilde{A}_{22} - L\tilde{A}_{12})\hat{x}_r(t) + (\tilde{A}_{21} - L\tilde{A}_{11})y(t) + (\tilde{B}_2 - L\tilde{B}_1)u(t)$$

sum and subtract $(\tilde{A}_{22} - L\tilde{A}_{12})Ly(t)$

define $\xi(t) = \hat{x}_r(t) - Ly(t)$, $\xi \in R^{n-p}$

Final reduced order observer

$$\dot{\xi}(t) = (\tilde{A}_{22} - L\tilde{A}_{12})\xi(t) + (\tilde{A}_{21} - L\tilde{A}_{11} + \tilde{A}_{22}L - L\tilde{A}_{12}L)y(t) + (\tilde{B}_2 - L\tilde{B}_1)u(t)$$

on state

$\xi(t)$ and input $y(t), u(t)$
(STANDARD observer) to
estimate ξ than from

ξ estimation I go
back on \hat{x} estimation!

$$\hat{x}_r(t) = \xi(t) + Ly(t)$$

from ξ definition!

$$\hat{x}(t) = T^{-1} \begin{bmatrix} y(t) \\ \hat{x}_r(t) \end{bmatrix}$$

(EXAM question about reduced observer and simple examples)

Example

Given the system

Design a reduced order observer for the system

$$\begin{cases} \dot{x}_1 = x_2 \\ \dot{x}_2 = u \\ y = x_1 \end{cases}$$

1) we should transform the state so that substitute
of output are states

equation of state dynamic, $A = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}$, $B = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$, $C = [1 \ 0]$

we have only to estimate x_2

One state already coincides with the output. There is no need of state transformation ($\tilde{x} = Tx$)

According to the general Theory $\boxed{y = \dot{\tilde{y}} = x_2}$

$$\underbrace{T = I_2}_{\text{we don't need to modify A,B,C}}$$

$$\begin{array}{l} \dot{y} = x_2 = y \\ \text{output transform} \end{array}$$

The observer can be written as \Rightarrow

The observer can be written as

$$\dot{\hat{x}}_2 = u + L(\dot{y} - \hat{x}_2) \quad \text{estimation equation for STATE ESTIMATION}$$

\downarrow

$$\dot{\hat{x}}_2 - L\dot{y} = -L(x_2 - Ly) - L^2y + u$$

$\underbrace{\dot{\hat{x}}_2}_{\xi} \quad \underbrace{- L\dot{y}}_{\xi}$

\downarrow

re-define, new variable ξ

$$\dot{\xi} = -L\xi - L^2y + u \quad \text{dyn. equation of the OBSERVER}$$

following my theoretical steps

inputs u, y

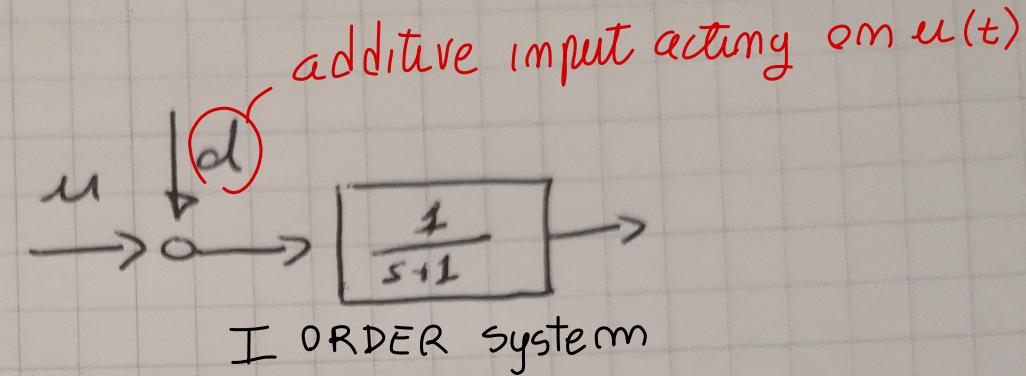
$\dot{\hat{x}}_2 = \xi + Ly$

to estimate \hat{x}_2
to come back

use of an OBSERVER to estimate a disturbance $d(t)$

Exercise

Given the system

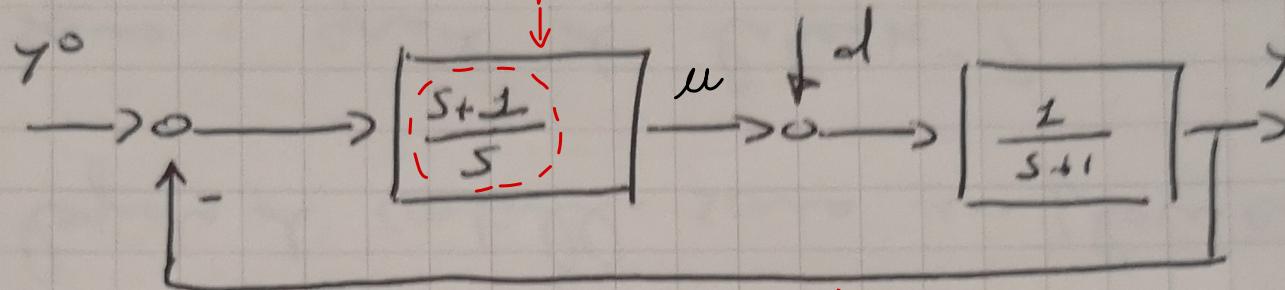


A) design a PI regulator

$$d(t) \approx \bar{d}$$

B) Design an estimator of the disturbance, assumed to be constant
and a compensator acting on the estimated value of d

Solution A) simple PI regulator design zeros to cancel the T.F pole



$$y(s) = \frac{1}{s+1} y^o + \frac{s}{(s+1)^2} d$$

hidden pole in (-1)

$$L(s) = \frac{1}{s} \text{ so } \varphi_m = 90^\circ$$

↑ both
poles
visible

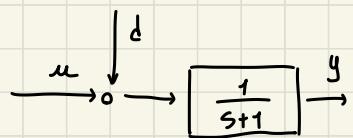
With this regulator... nice T.F. y/y^o

While y/d has two poles in (-1)

+ Derivative action from integrator \Rightarrow steady state $d(t) \rightarrow 0$ effect (derivative) error goes to 0 asympt

(note that due to the integrator in $R(s)$)
constant disturbances

are asymptotically
rejected)



$$sY + Y = u + d$$

$$\dot{y} + y = u + d$$

Solution B

$d(t)$ can be estimated? and can we build a COMPENSATOR for $\hat{d}(t)$ (integrator already remove d , but we want estimate $d(t)$)

The system is described in the state space by

$$\left\{ \begin{array}{l} \overset{\circ}{x} = -x + u + d \\ \dot{d} = 0 \\ y = x \end{array} \right. \begin{array}{l} \text{model equation} \\ \text{disturbance} \\ \text{OUTPUT transform} \end{array}$$

← fictitious dynamics
equation → to estimate it, we assume a
fictitious dyn. $\dot{d} = 0$ (do unkwn)

Output = state : We don't need
 $x(t)$ estimation ↓

We can use a reduced order observer to estimate d

According to the theory, write the system as

$$\left\{ \begin{array}{l} \dot{d} = 0 \\ \underbrace{\dot{x} + x - u}_{y} = d \end{array} \right. \Rightarrow \boxed{\dot{\hat{d}} = L \underbrace{[\dot{y} + y - u - \hat{d}]}_{\gamma} \text{ ESTIMATOR}}$$

equation
re-written

$$\dot{\hat{d}} = L \dot{y} + Ly - Lu - L\hat{d}$$

derivative of y om right. → undesired!

\Downarrow new variable ξ

$$\underbrace{\dot{\hat{d}} - Ly}_{\dot{\xi}} = -L\hat{d} + Ly - Lu$$

$$\dot{\xi} \quad \downarrow \quad \textcircled{+ L^2 y}$$

$$\dot{\xi} = -L \underbrace{[\hat{d} - Ly]}_{\tilde{\xi}} - L^2 y + Ly - Lu$$

new OBSERV.
of order I,
to build
for $\xi = 0$

$$\underbrace{\ddot{d} - Ly}_{\dot{\xi}} = -L\hat{d} + Ly - Lu$$

 $\dot{\xi}$

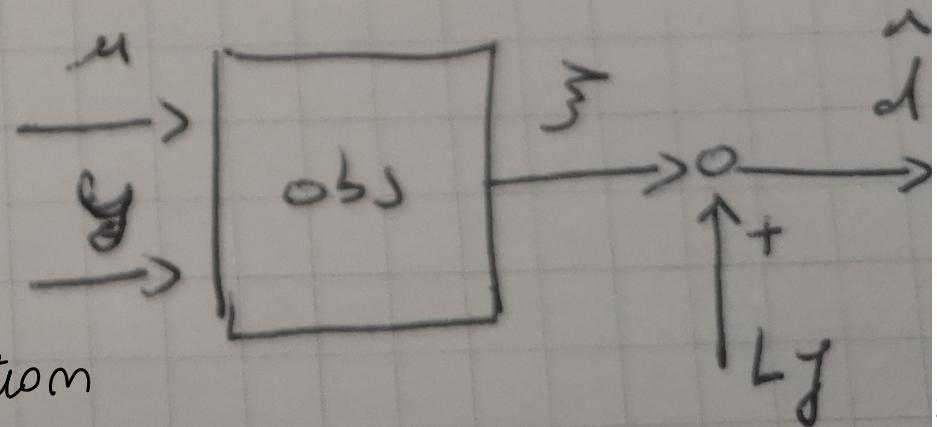
$$\downarrow \neq L^2y$$

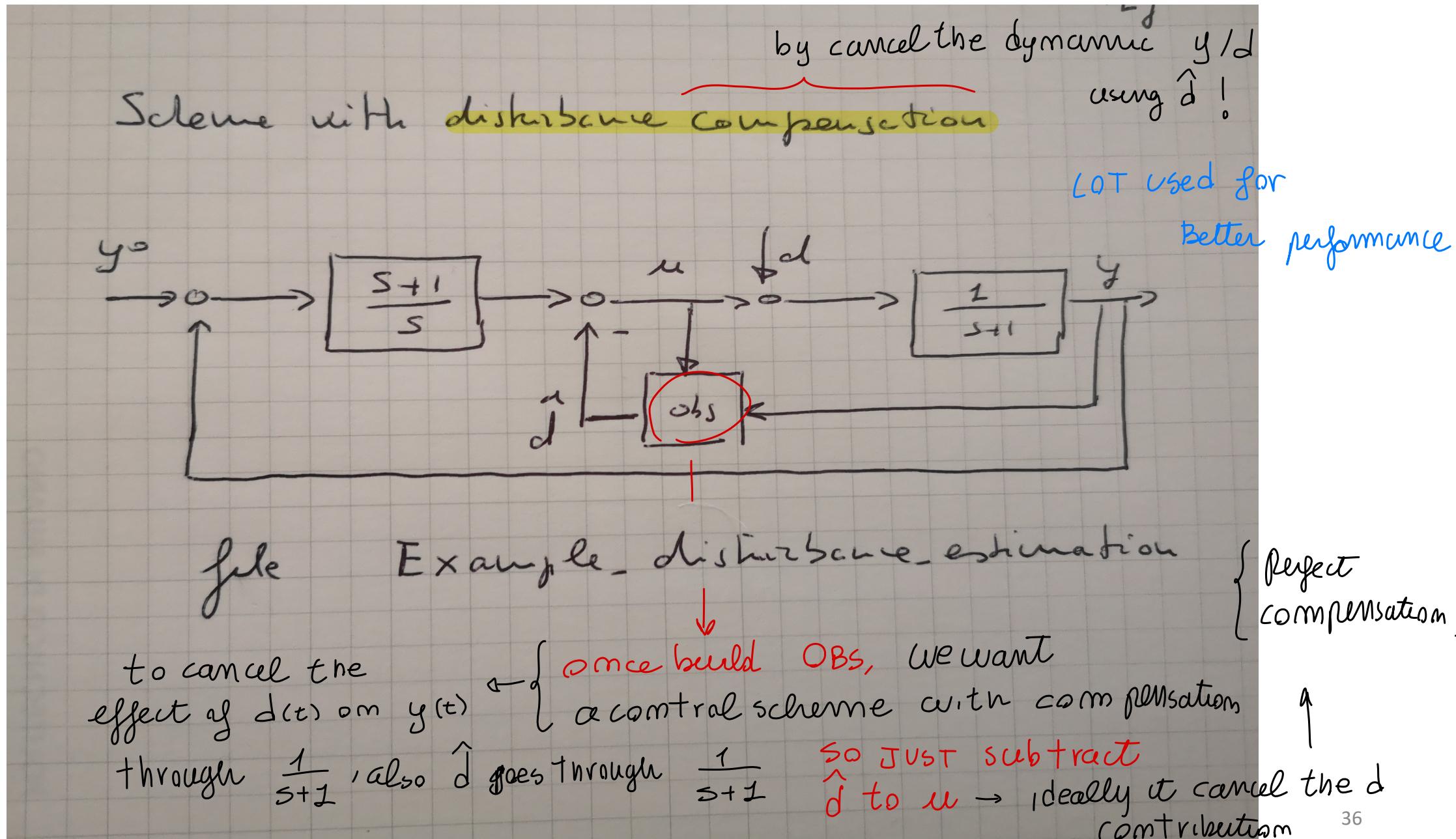
$$\dot{\xi} = -L \underbrace{[\ddot{d} - Ly]}_{\tilde{\xi}} - L^2y + Ly - Lu$$

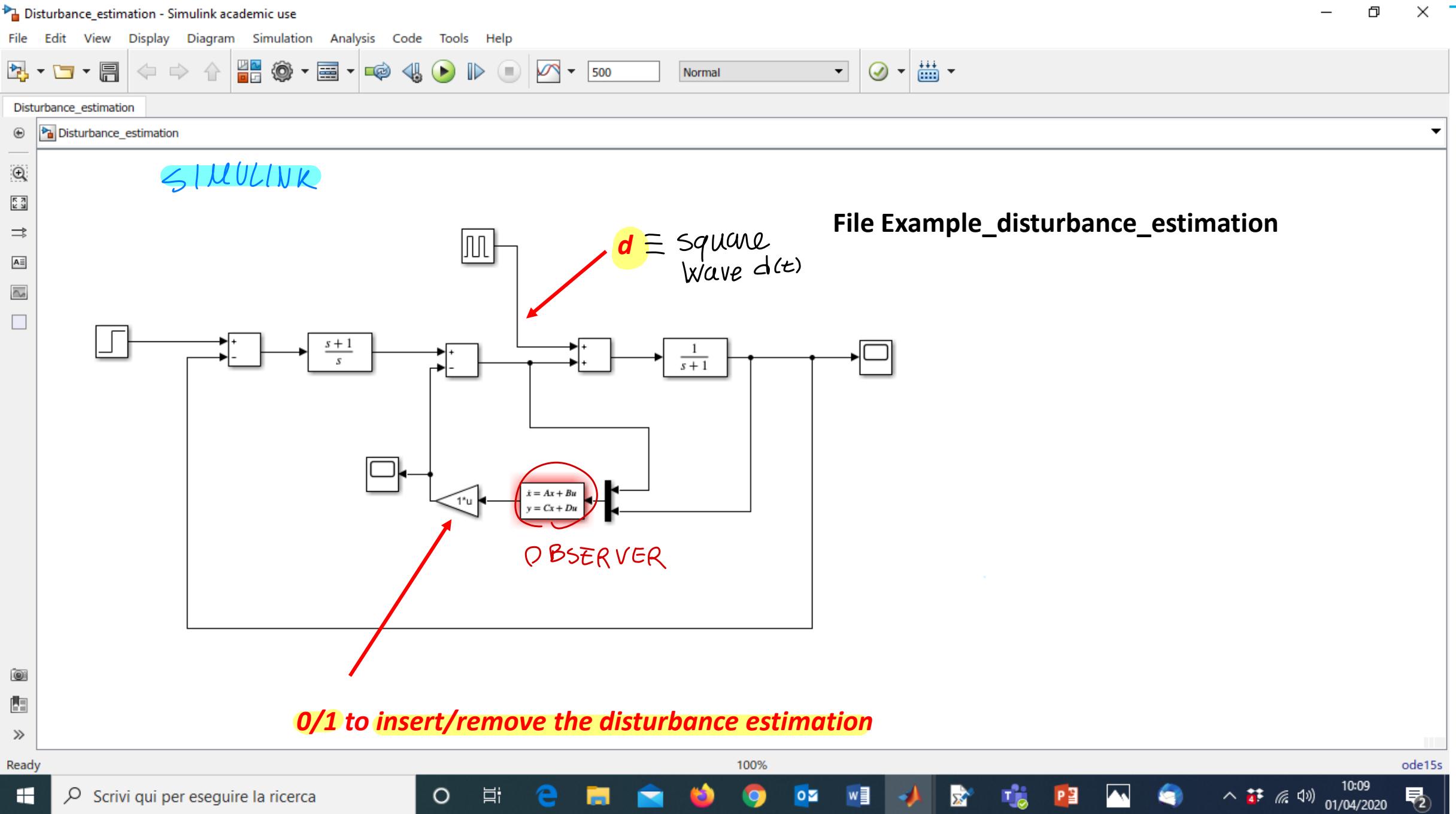
$$\boxed{\dot{\xi} = -L\tilde{\xi} - L^2y + Ly - Lu}$$

OBSERVER: given u, y compute $\tilde{\xi}$

by adding $Ly + \tilde{\xi} \rightarrow \hat{d}$ estimation

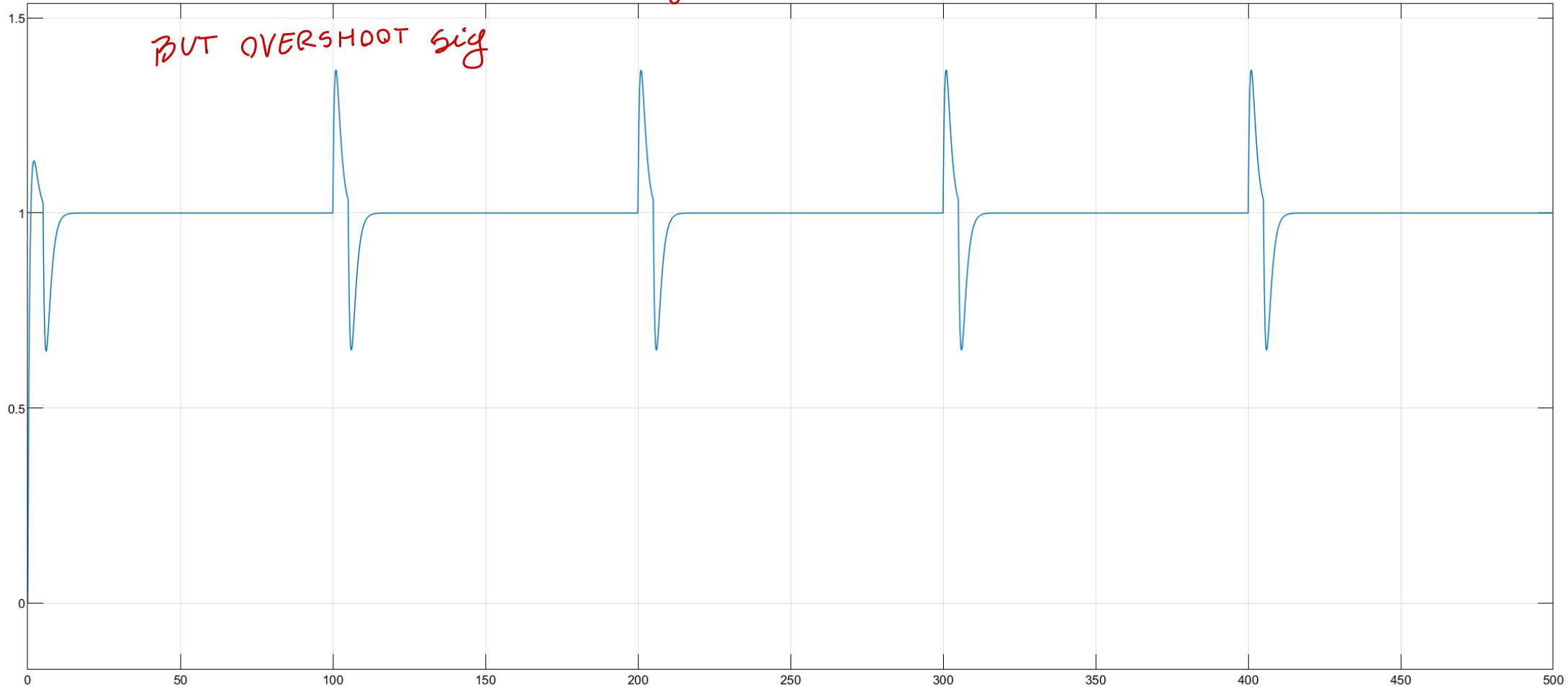






No disturbance compensationeffect of $d(t)$ during t ! \Rightarrow perturbation

you reject asymp disturbance effect for each period



Disturbance estimation and compensation

with obs+compensation

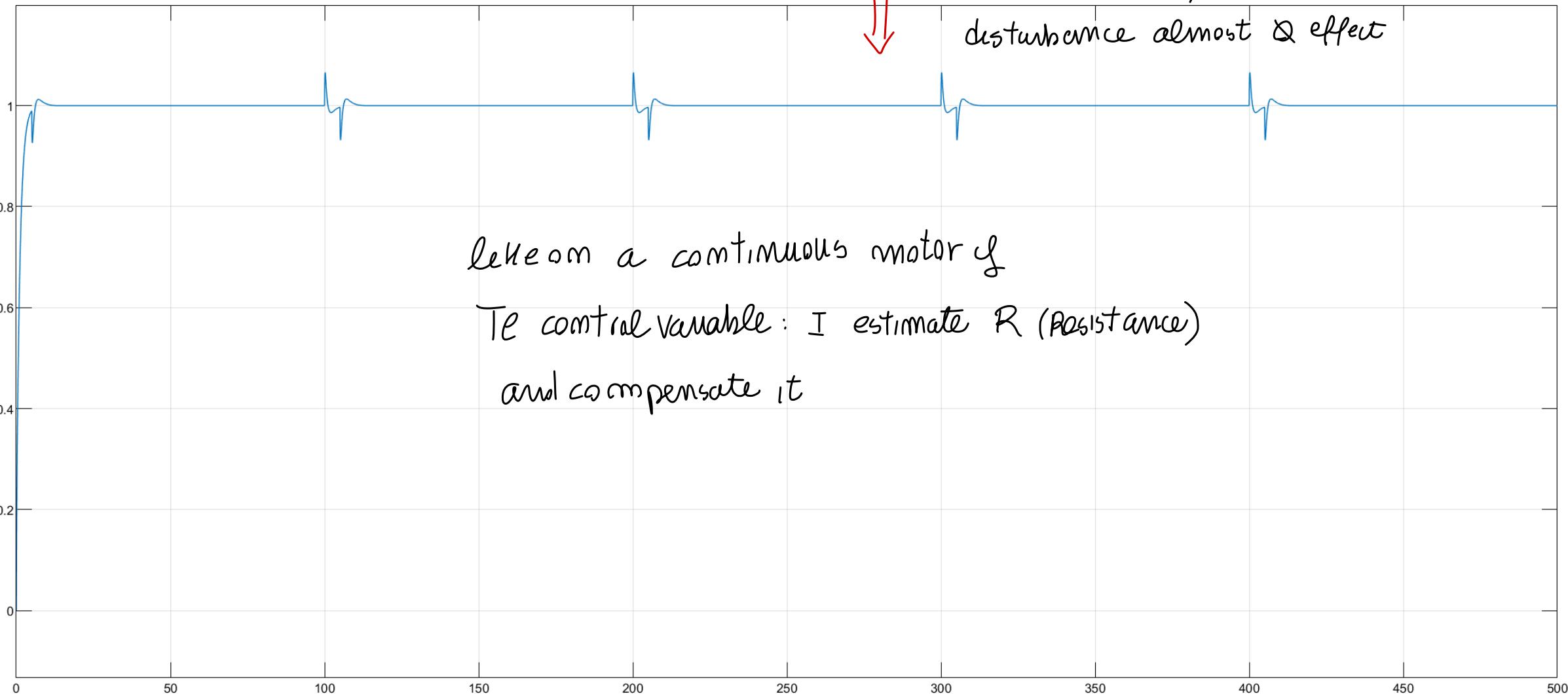
↓ smaller overshoot,

disturbance almost Q effect

let's assume a continuous motor of

Te control variable: I estimate R (Resistance)

and compensate it

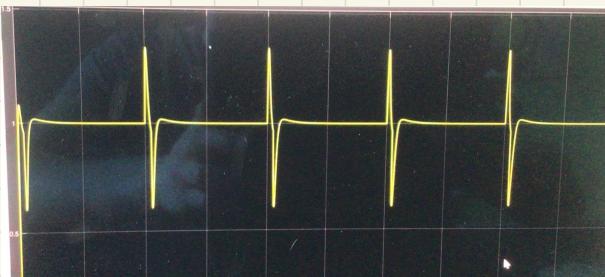


NOTICE on the estimator the eig are $(-L)$ values!

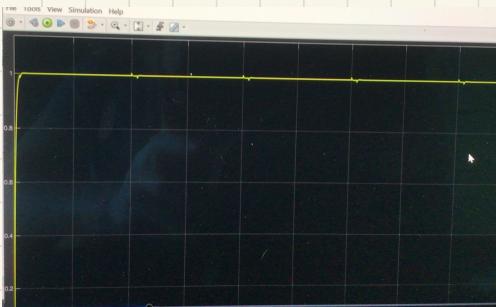
on simulation: for $L=10$: OBS has eig on $-10\dots$

- If reducing $L \downarrow \Rightarrow$ eig becomes slow, so $\hat{d} \xrightarrow{\uparrow}$ in much longer time
 \Downarrow

for $L=0.1$ on estimator by SIMULINK
we got a bad Result! too slow



for **FASTER OBS.** $L = 100$



for FAST OBS you have $+d - d \rightarrow$ remove properly disturb

to select eig value \Rightarrow OBS can be FAST,
but if you have **output noise**,
the OBS make filtering action...

Without noise on $y(t)$ is better a fast OBS.

Extend the results for Discrete Time
Discrete time systems – measurable state ↗ easy extension,
 But take care of obs.-form!

System (in S.S. Form)

$$\begin{cases} x(k+1) = Ax(k) + Bu(k), & x \in R^n, u \in R^m \\ y(k) = Cx(k) + Du(k), & y \in R^p \end{cases}$$

Control law

↗ state feedback control law to guarantee asymp. stable eig value
 $u(k) = -Kx(k) + \gamma(k)$, $K \in R^{m,n}$, $\gamma \in R^m$ of closed loop syst

ALGEBRAIC control law (we know $x(k)$)

Closed-loop

↳ $\gamma(k)$ additional signal

$$x(k+1) = (\boxed{A - BK})x(k) + B\gamma(k)$$

(same as continuous time) $\rightarrow (A - BK)$

must have prescribed
eig value!

The problem is exactly the same, the same algorithms can be used (Ackermann's formula, place.m,...)

same Algorithm

single input
systems

more
inputs

small difference with cont. time

Observers (already seen on MIDA)



Two possibilities: (of design)

- { 1. **State predictor:** the estimated state at time k depends on the values of u and y up to time $k-1$
@ $t = K$: you estimate $x(K)$ depending to u, y up to $K-1$
- 2. **State filter:** the estimated state at time k depends on the values of u and y up to time k
@ $t = K$: you estimate $x(K)$ using even the value of u, y up to k
 \Rightarrow most updated state estimation... But
you can have implementation issue

1

State predictor

(same as continuous time)

state estimate @ $k+1$, given data up to k

(same as c. t. observer!)

$$\hat{x}(k+1|k) = A\hat{x}(k|k-1) + Bu(k) + L[y(k) - C\hat{x}(k|k-1) - Du(k)]$$

By simulating system dyn. | Gain * (difference of estimation)

state

Estimation error $\hat{e}(k|k-1) = x(k) - \hat{x}(k|k-1)$

↓ dynamics of state estim. error

$$\hat{e}(k+1|k) = (A - LC)\hat{e}(k|k-1)$$

you need L such that $\text{eig}(A-LC)$ inside unit circle!

Also in this case the problem is to compute the gain L which assigns the eigenvalues of $A-LC$ in prescribed positions (so $\hat{e} \rightarrow 0$, so $\hat{x} \rightarrow x$ @ regime)

observer such that

Deadbeat observers: all the eigenvalues of $A-LC$ are at the origin. The state estimation error goes to zero in at most n steps (FASTEST possible observer in discrete time)

(sometimes NOT too fast or noise issue)

so you can design observer using pole placement algorithm to have (L)

2

State filter

→ care about filter structure

standard part: syst dym

$$\tilde{x}(k+1|k+1) = \underbrace{A\tilde{x}(k|k) + Bu(k)}_{\text{state evolution of } \tilde{x}(k+1)} + \underbrace{L[y(k+1) - C(\tilde{x}(k|k) + Bu(k))]}_{\substack{\text{GAIN} \\ \text{to use } y(k+1)}} - Du(k+1)$$

--- green available
measurable!
you estimate $\tilde{x}(k+1)$

Estimation error $\tilde{e}(k|k) = x(k) - \tilde{x}(k|k)$
from system

dym eq.
I obtain... → $\tilde{e}(k+1|k+1) = (A - LCA)\tilde{e}(k|k)$ linear system
without forcing

The gain L must be designed to assign the eigenvalues of $(A-LCA)$. To this end, the pair (A, CA) must be observable
(so $\tilde{e} \rightarrow 0$) different condition!

{ to assign for free[†] the eig of that system you have
to guarantee obs of the pair!

- The pair (A, CA) is observable iff the pair (A, C) is observable and A is \leftarrow (CNS)
nonsingular

{ $y \tilde{C} = CA \rightarrow (A - L \tilde{C})$ same problem! so using standard
Algorithm of pole placement }

Note however that if A is singular, the system is detectable, since the nonobservable eigenvalues are at the origin,
i.e. the fastest possible ones. (In other words, leave them where they are (algorithms in case of detectability are
only slightly more complex))

↑
↑
ERROR to apply Ackerman...
you should partition the system

you cannot move the eig values of A which are @ ORIGIN
{ NON OBS eig values! → NOT a real problem, you cannot move them but they
have an evolution → 0 } 43

Estimation of constant disturbances

Assume to know that d is constant, or at least constant for long periods of time.
It is useful to estimate it for two reasons:

(otherwise $\tilde{e}(t) \neq 0$
↑
ERROR estimation)

- 1) It is possible to correctly estimate the state of the system (If you have $d(t)$, you can estimate them for a proper overall state estimation)
you need for good estimate
- 2) The estimate of the disturbance can be used in a control scheme with direct compensation
estimate \hat{d} so you build feed forward compensator

Procedure (Similar to continuous time)

- ① Assign a dynamics to the disturbance (fictitious dyn.) on discrete time $d(k+1) = d(k)$ (constant)

$$d(k+1) = d(k), \quad d(0) = \bar{d} \in R^r$$

- ② Enlarge the system

provided that the pair (\bar{A}, \bar{C}) is obs.
you can estimate
 (x) enlarged state!

$$\begin{cases} \begin{bmatrix} x(k+1) \\ d(k+1) \end{bmatrix} = \begin{bmatrix} \bar{A} & M \\ 0 & I \end{bmatrix} \begin{bmatrix} x(k) \\ d(k) \end{bmatrix} + \begin{bmatrix} B \\ 0 \end{bmatrix} u(k) \\ y(k) = \begin{bmatrix} \bar{C} & N \end{bmatrix} \begin{bmatrix} x(k) \\ d(k) \end{bmatrix} + Du(k) \end{cases}$$

state eq.
+ Disturb
output eq.

→ by PB H
test.. condition
request (A, C) obs,
and $y \in R^r := \#$ disturb
 $d \in R^r \Rightarrow r \leq p$
 $y \in R^p$
and invariant zero
if y/d are mat
Im point (1)

Use an observer for the enlarged system. When is it possible? Try to find the condition

also, in this case we can design reduced order for FILTER, PREDICTOR, they are of order $m \rightarrow$ we wanna find an OBS of order $(m-p) \rightarrow$ useful to have a FASTER OBSERV

Reduced order observer

$$\begin{cases} x(k+1) = Ax(k) + Bu(k) \\ y(k) = Cx(k) \end{cases}$$

same trick as cont. time

Idea: the output is a linear combination of the states

1. Apply a state transformation to the system so that the p outputs coincide with p new states
2. Estimate the remaining $n-p$ states

①

T_1 any matrix such that T is non singular

$$\tilde{x} = Tx = \begin{bmatrix} C \\ T_1 \end{bmatrix} x = \begin{bmatrix} y \\ \tilde{x}_r \end{bmatrix}, \quad \tilde{x}_r \in R^{n-p}$$

available (measurable)

so that outputs coincide with some x to estimate

Transformed system

②

$$\begin{cases} y(k+1) = \tilde{A}_{11}y(k) + \tilde{A}_{12}\tilde{x}_r(k) + \tilde{B}_1u(k) & \text{output dyn} \\ \tilde{x}_r(k+1) = \tilde{A}_{21}y(k) + \tilde{A}_{22}\tilde{x}_r(k) + \tilde{B}_2u(k) & \text{to estimate dyn} \end{cases}$$

NO derivative here!

$$\tilde{A} = TAT^{-1}, \quad \tilde{B} = TB, \quad \tilde{C} = CT^{-1}$$

$$\tilde{A} = \begin{bmatrix} \tilde{A}_{11} & \tilde{A}_{12} \\ \tilde{A}_{21} & \tilde{A}_{22} \end{bmatrix}, \quad \tilde{A}_{11} \in R^{p,p}, \quad \tilde{A}_{22} \in R^{n-p,n-p}$$

$$\tilde{C} = [I_p \quad 0]$$

$$\tilde{B} = \begin{bmatrix} \tilde{B}_1 \\ \tilde{B}_2 \end{bmatrix}, \quad \tilde{B}_1 \in R^{p,m}, \quad \tilde{B}_2 \in R^{n-p,m}$$

3

Define

$$\begin{cases} \eta(k+1) = y(k+1) - \tilde{A}_{11}y(k) - \tilde{B}_1u(k) \\ \zeta(k) = \tilde{A}_{21}y(k) + \tilde{B}_2u(k) \end{cases}$$

on right what we know, I'm left the unknown



$$\begin{cases} y(k+1) = \tilde{A}_{11}y(k) + \tilde{A}_{12}\tilde{x}_r(k) + \tilde{B}_1u(k) \\ \tilde{x}_r(k+1) = \tilde{A}_{21}y(k) + \tilde{A}_{22}\tilde{x}_r(k) + \tilde{B}_2u(k) \end{cases}$$



$$\begin{cases} \tilde{x}_r(k+1) = \tilde{A}_{22}\tilde{x}_r(k) + \zeta(k) \\ \eta(k+1) = \tilde{A}_{12}\tilde{x}_r(k) \end{cases}$$

STATE EQ

OUTPUT transform

New system form

observer

4

$$\hat{x}_r(k+1|k+1) = \tilde{A}_{22}\hat{x}_r(k|k) + \zeta(k) + L[\eta(k+1) - \tilde{A}_{12}\hat{x}_r(k|k)]$$

or

↓ by substituting η, ζ

$$\hat{x}_r(k+1|k+1) = (\tilde{A}_{22} - L\tilde{A}_{12})\hat{x}_r(k|k) + \tilde{A}_{21}y(k) + \tilde{B}_2u(k)$$

{ simple to implement }

$$+ [Ly(k+1) - \tilde{A}_{11}y(k) - \tilde{B}_1u(k)]$$

this act like a sort of FILTER, the estimated @ $k+1$ uses data up to $k+1$ (NOT PREDICTOR!)

← OBS eig values

assigned by L ,

depends on request

← L assign eig value of OBS by fixing L

|| No problems with derivatives || |

Regulator transfer function with state predictor

which one to use? T.F with PREDICTOR or FILTER?
IMPLEMENTATION aspect to choose between

System

$$\begin{cases} x(k+1) = Ax(k) + Bu(k) \\ y(k) = Cx(k) \end{cases}$$

Control law

$$u(k) = -K\hat{x}(k|k-1)$$

using the PREDICTOR

Predictor

$$\hat{x}(k+1|k) = Ax(k|k-1) + Bu(k) + L[y(k) - C\hat{x}(k|k-1)]$$

Regulator

T.F

control law
+ Predictor
equation

$$U(z) = -K(zI - A + BK + LC)^{-1}LY(z)$$

#ROWS = #INPUT

same form of cont time

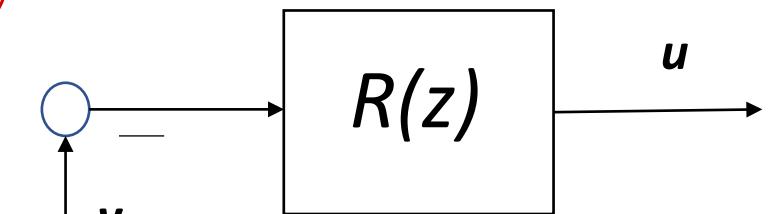
strictly proper transfer function $R(z)$

manipulation... (more poles than zeros)

take $u(k)$ put on $\hat{x}(k+1|k) \rightarrow$ transform

on z domain, then reevaluate $U(z)$

considered as SISO system!



$R(z)$ classical siso T.F

Regulator transfer function with state filter

System

$$\begin{cases} x(k+1) = Ax(k) + Bu(k) \\ y(k) = Cx(k) \end{cases}$$

Control law

$$u(k) = -K\tilde{x}(k|k)$$

Filter

$$\begin{aligned} \tilde{x}(k+1|k+1) &= A\tilde{x}(k|k) + Bu(k) \\ &+ L[y(k+1) - C(A\tilde{x}(k|k) + Bu(k))] \end{aligned}$$

Regulator

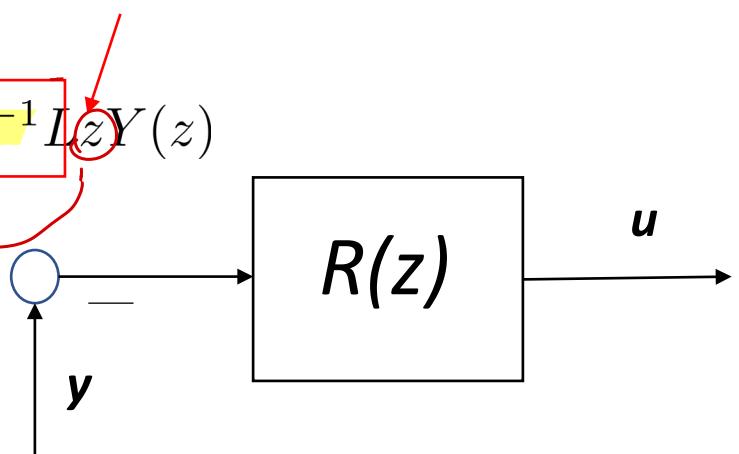
T.F computed
as before... you

Obtain a T.F matrix given by a proper T.F

{ NOT strictly proper due to ∞ ! same poles-zeros number ! }

$$U(z) = -K(zI - A + BK + LCA + LCBK)^{-1}Ly(z)$$

proper transfer function $R(z)$



WHAT HAPPENS?

Strictly proper regulator transfer function

↓
on SISO case (for simplicity)

$$R(z) = \frac{b_{n-1}z^{n-1} + \dots + b_1z + b_0}{z^n + a_{n-1}z^{n-1} + \dots + a_1z + a_0}$$

↓
to implement it by a suitable algorithm

$$\boxed{U(z) = R(z)E(z)}$$

↓

$$(z^n + a_{n-1}z^{n-1} + \dots + a_1z + a_0)U(z) = (b_{n-1}z^{n-1} + \dots + b_1z + b_0)E(z)$$

↓
 \mathcal{Z} time shift transform

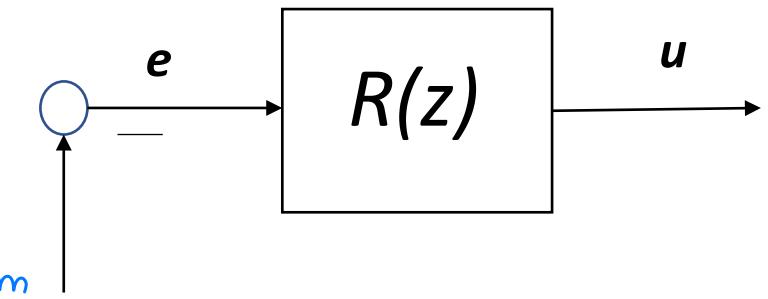
$$u(k) = -a_{n-1}u(k-1) - \dots - a_1u(k-n+1) - a_0u(k-n) + b_{n-1}e(k-1) +$$

$$\dots + b_1e(k-n+1) + b_0e(k-n)$$

you can make comput. in one sampling time ≤ 1 to estimate $x(k)$ update

and use **One sampling time to read the value of the error (A/D), make computations, write the input u (D/A)**

it on feedback control law if you use an observer!



↓ to compute $u(k)$ by strictly Proper T.F.. $u(k)$
depends on u, e up to $(k-1)$! \Rightarrow you have one sampling time to compute $u(k)$

... but less reactive

While!

Proper regulator transfer function

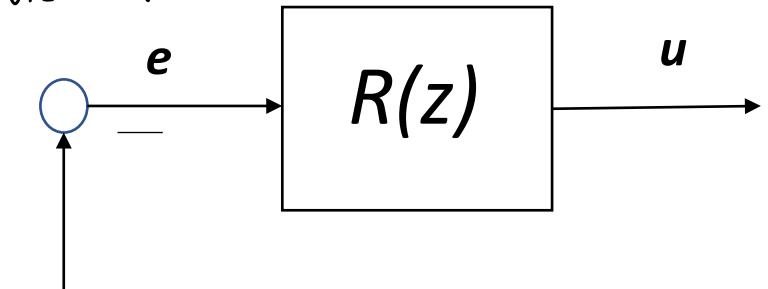
FILTER is more accurate, use updated value... BUT if introducing too much delay \rightarrow CRITICAL! } same number of poles / zeros } NOT relevant of high T system, but very problematic for FAST systems!

$$R(z) = \frac{b_n z^n + b_{n-1} z^{n-1} + \dots + b_1 z + b_0}{z^n + a_{n-1} z^{n-1} + \dots + a_1 z + a_0}$$



same computation $U(z) = R(z)E(z)$

$$(z^n + a_{n-1} z^{n-1} + \dots + a_1 z + a_0)U(z) = (b_n z^n + b_{n-1} z^{n-1} + \dots + b_1 z + b_0)E(z)$$



$$u(k) = -a_{n-1}u(k-1) - \dots - a_1u(k-n+1) - a_0u(k-n) + b_n e(k) + b_{n-1}e(k-1) + \dots + b_1e(k-n+1) + b_0e(k-n)$$

depends on $e(k)$, so at $k=0$,
at time I should read the error and compute $u(k) \rightarrow$ apply it!

More reactive, but introduces a delay due to A/D+computations+D/A

↓
 { NOT possible, you need computation time... (issue) }
 { delay \rightarrow phase lag critical for implementation }