

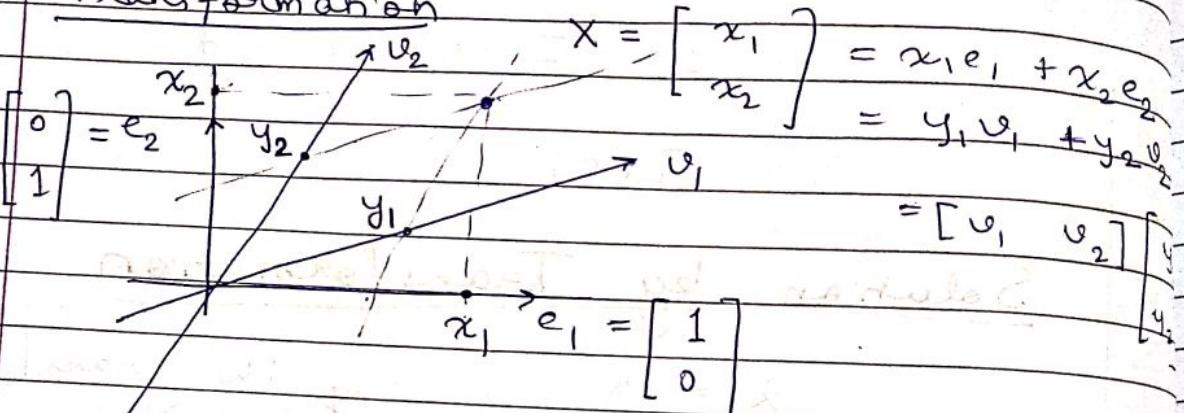
Solution of State Space Equations

by Coordinate transforms

$$\dot{x} = Ax + bu$$

$$w = Cx + du$$

Transformation



$$x = (V y)$$

$$\rightarrow \dot{x} = Ax + bu \quad \left. \begin{array}{l} w = Cx + du \end{array} \right\} \text{original } \underline{\Phi}_x(t)$$

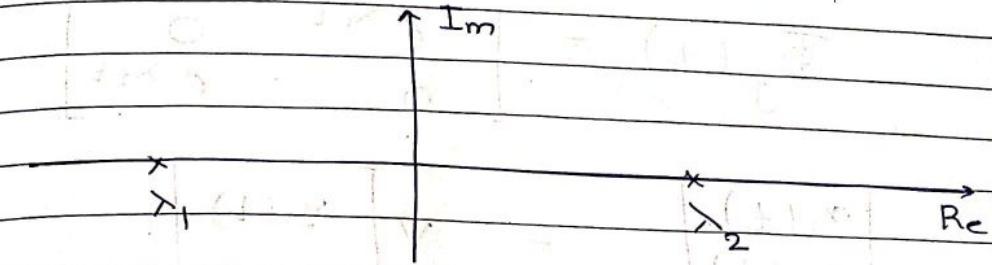
$$\rightarrow \dot{V}y = AVy + bu \quad \left. \begin{array}{l} w = CVy + du \end{array} \right\} \underline{\Phi}_y(t) = V^{-1} \underline{\Phi}_x(t) V$$

$$\Rightarrow \dot{y} = (V^{-1}AV)y + V^{-1}bu$$

$$w = (CV)y + du$$

\rightarrow Can the transformation give us a simpler State transition Matrix

v_1, v_2 are eigenvectors
with distinct eigen values λ_1, λ_2
(Since distinct $\lambda \Rightarrow LI$)



λ_1, λ_2 are eigen values of A
(Must be atleast complex conj. of each other)

$$Av_1 = \lambda_1 v_1, \quad Av_2 = \lambda_2 v_2$$

$$x = y_1 v_1 + y_2 v_2$$

$$\dot{x} = \dot{y}_1 v_1 + \dot{y}_2 v_2 = A(y_1 v_1 + y_2 v_2) + bu$$

$$\text{Say, } b = \beta_1 v_1 + \beta_2 v_2$$

$$= y_1 \lambda_1 v_1 + y_2 \lambda_2 v_2 + \beta_1 u v_1 + \beta_2 u v_2$$

(components)

from: $\dot{y}_1 = \lambda_1 y_1 + \beta_1 u$

$$\dot{y}_2 = \lambda_2 y_2 + \beta_2 u$$

$$y_1(t) = \int e^{\lambda_1 t} y_1(0) + \int \beta_1 e^{\lambda_1(t-\tau)} d\tau$$

$$y_2(t) = -e^{\lambda_2 t} y_2(0) + \int \beta_2 e^{\lambda_2(t-\tau)} d\tau$$

→ We know the initial conditions on the x variables, not necessarily the y variables.

$$\Phi_y(t) = \begin{bmatrix} e^{\lambda_1 t} & 0 \\ 0 & e^{\lambda_2 t} \end{bmatrix}$$

$$\begin{bmatrix} x_1(t) \\ x_2(t) \end{bmatrix} = V \begin{bmatrix} y_1(t) \\ y_2(t) \end{bmatrix}$$

$$\begin{aligned} x &= Vy \\ x(0) &= Vy(0) \end{aligned}$$

$$\dot{y} = \begin{bmatrix} e^{\lambda_1 t} & 0 \\ 0 & e^{\lambda_2 t} \end{bmatrix} V^{-1} \begin{bmatrix} x_1(0) \\ x_2(0) \end{bmatrix} + "$$

$$\dot{x} = (V[\phi_y] V^{-1}) x + V[\int"]$$

ex

$$A = \begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix} \quad \lambda^2 + 3\lambda + 2 = 0$$

$$\lambda_1, \lambda_2 = -1, -2$$

$$v_1 = \begin{bmatrix} \alpha \\ \beta \end{bmatrix}$$

$$\begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix} \begin{bmatrix} \alpha \\ \beta \end{bmatrix} = (-2) \begin{bmatrix} \alpha \\ \beta \end{bmatrix}$$

$$\beta = -\alpha$$

$$v_1 = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

Q30

-2β

$$\begin{matrix} \beta \\ -2\alpha - 3\beta \end{matrix} = (-1) \begin{bmatrix} \alpha \\ \beta \end{bmatrix}$$

$$v_2 = \begin{bmatrix} 1 \\ -2 \end{bmatrix} \quad \text{using } (\beta = -2\alpha)$$

$$V = \begin{bmatrix} 1 & 1 \\ -1 & -2 \end{bmatrix} \quad V^{-1} = \frac{1}{(-1)} \begin{bmatrix} -2 & -1 \\ 1 & 1 \end{bmatrix} = \begin{bmatrix} 2 & 1 \\ -1 & -1 \end{bmatrix}$$

$$V^T A V^{-1} = [\lambda]$$

$$\begin{bmatrix} 1 & 1 \\ -1 & 2 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ -3 & -2 \end{bmatrix} \begin{bmatrix} 2 & 1 \\ -1 & -1 \end{bmatrix} = \begin{bmatrix} -1 & 0 \\ 0 & 2 \end{bmatrix}$$

For $u = 0$,

$$y_1 = e^{-t} \quad y_2 = e^{-2t}$$

$$x(t) = \Phi_x(t) \times (0) + \int_0^t \Phi_x(t-\tau) b u(\tau) d\tau$$

→ For zero State Solution ($x_{(0)} = 0$)

$$w(t) = c \int_0^t \Phi_x(t-\tau) b u(\tau) d\tau$$

→ Step Response

$$c(t) = \mathcal{L}^{-1} \frac{1}{s(s^2 + 3s + 2)} \quad \text{L.T solution}$$

$$\omega(t) = C \int_0^t \Phi_x(t-\tau) b d\tau \quad \text{State space solution}$$

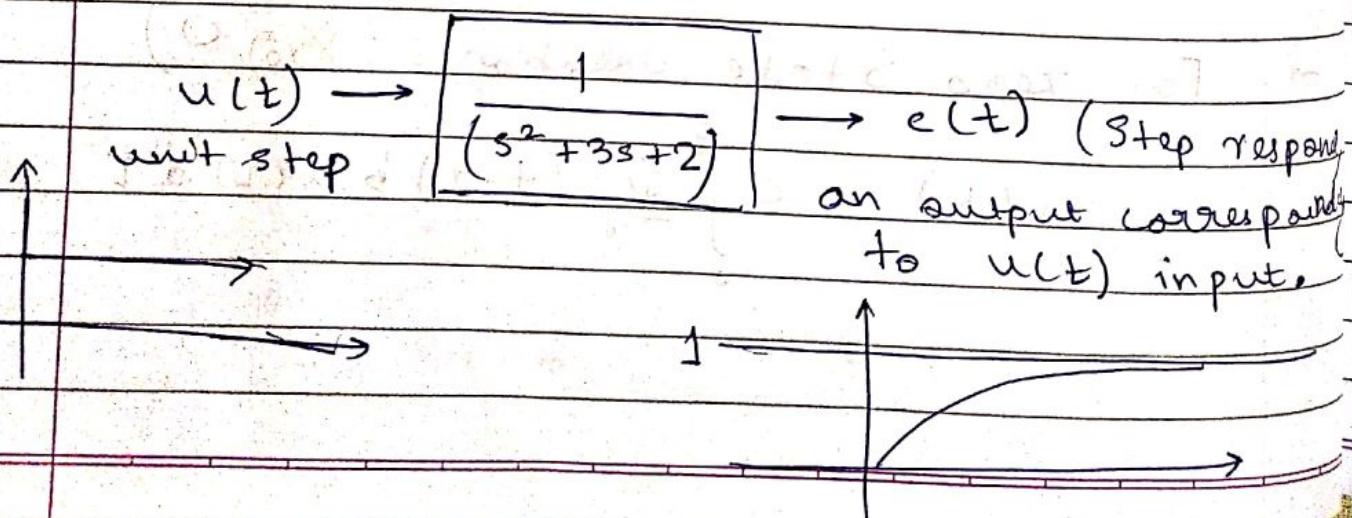
$$\dot{x} = Ax \quad x = Vy$$

$$V\dot{y} = AVy \quad \Phi_x(t) = \mathcal{L}^{-1}$$
$$\dot{y} = (V^{-1}AV)y$$

Re-derive

$$\begin{aligned} \Phi_x(t) &= \mathcal{L}^{-1}(sI - A)^{-1} \\ &= \mathcal{L}^{-1}(sI - VBV^{-1})^{-1} \\ &= V \mathcal{L}^{-1}(sI - B)^{-1} V^{-1} \end{aligned}$$

Diagram



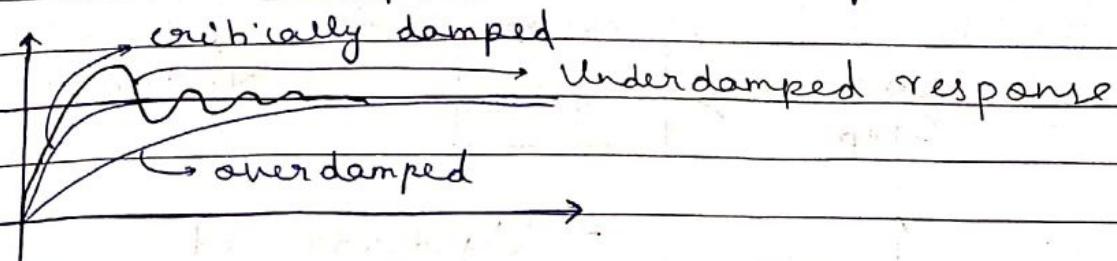
Big picture idea of C.S

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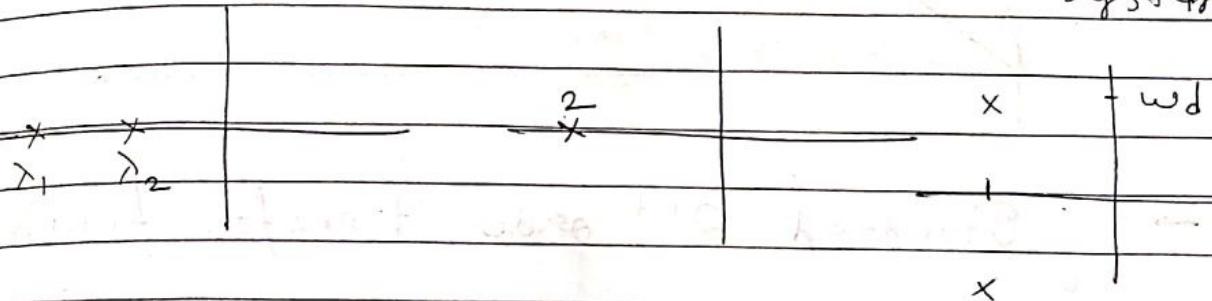
→ Second order response to unit step



$$c(t) = K + c_1 e^{-t} + c_2 e^{-2t}$$

(Artificial)

~~2nd order~~ Location of Poles (Vs Real System)



→ Standard parameters of a 2nd order System

$$\frac{\omega_n^2}{s^2 + 2\zeta\omega_n s + \omega_n^2}$$

$$\lambda_1, \lambda_2 = -\zeta\omega_n \pm \omega_n \sqrt{1 - \zeta^2}$$

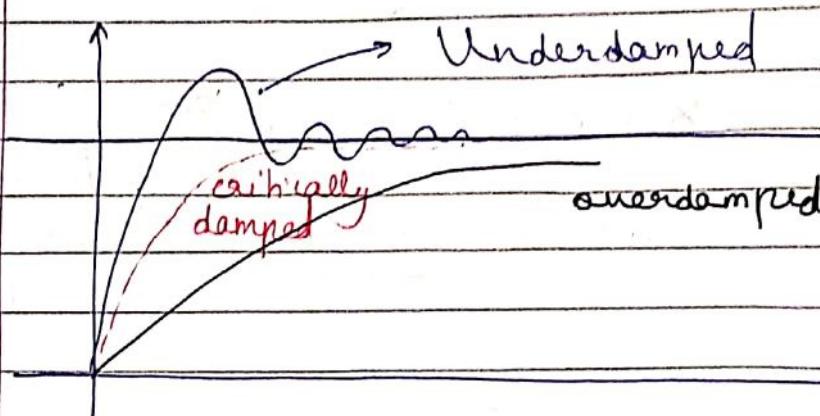
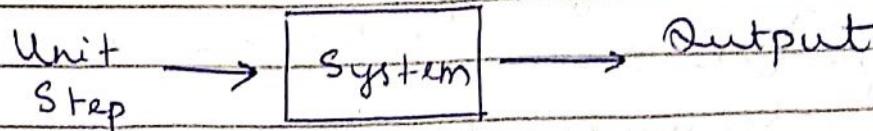
$\zeta \rightarrow$ Damping factor

$\omega_n \rightarrow$ Natural frequency

$\omega_d \rightarrow$ damped frequency

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Second Order Response Model



→ Standard 2nd order transfer function

$$\frac{\omega_n^2}{s^2 + 2\zeta\omega_n s + \omega_n^2} = \frac{\omega_n^2}{(s-\lambda)(s-\bar{\lambda})}$$

$$\lambda_1, \lambda_2 = -\zeta\omega_n \pm j\omega_n \sqrt{1-\zeta^2} \quad \zeta < 1$$

$\sigma = -\zeta\omega_n$, ω_d : damped natural frequency

$$\omega_d = \omega_n \sqrt{1-\zeta^2}$$

$$\lambda_1, \lambda_2 = -\sigma \pm j\omega_d$$

Exercise, Show that General form of the solution is,

$$c(t) = 1 - \frac{1}{\sqrt{1-\zeta^2}} e^{-\sigma t} \cos(\omega_d t + \phi)$$

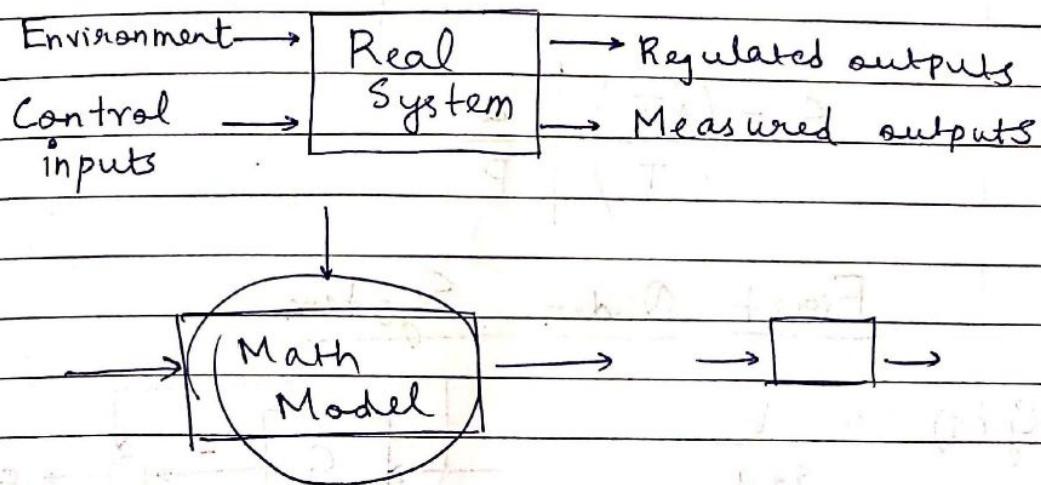
$\phi = \tan^{-1} \zeta$

Qty's of Interest

- Time to peak T_p
- Settling time T_s to settle in $\pm 2\%$ Band
- Rise time
- Steady State Error $e_{ss} = 1 - c(\infty)$

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The Big Picture



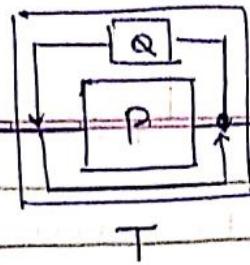
Sensitivity

→ In the real world, parameters are not constant, this behaviour must be modelled.

$$\tilde{P}(s) = P(s) + \Delta(s)$$

% Change ad for the model $\frac{\Delta}{P(s)}$

→ The perturbations that we need to model here are inside the system.



- After embedding P inside a network of controllers

T - Nominal System

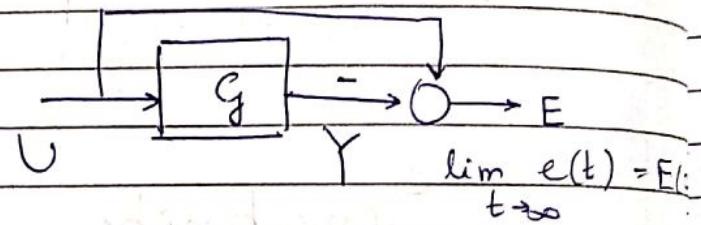
\tilde{T} - Perturbed "

$$\Delta T = \tilde{T} - T, \frac{\Delta T}{T}$$

$$S = \left(\frac{\Delta T}{T} \right) \Bigg| \frac{\Delta}{P}$$

First Order System

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



$$E = \frac{1}{s} - Y = \frac{1}{s} - \frac{b}{s(s+a)}$$

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

$$e_{ss} = \frac{a-b}{a} = \frac{1-b/a}{1}$$

$$\frac{\tilde{b}}{\tilde{a}} = \frac{b}{a} + \Delta$$

$$\Delta e_{ss} = \tilde{e}_{ss} - e_{ss}$$

$$= \left(1 - \frac{\tilde{a}}{a}\right) - 1 + \frac{b}{a} = -\Delta$$

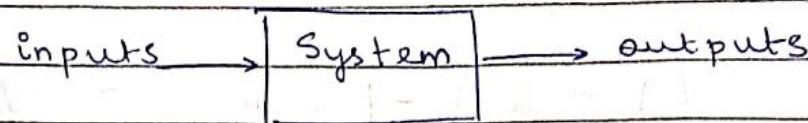
% var in e_{ss} without control

$$= \frac{\Delta e_{ss}}{e_{ss}} = -\Delta \cdot \left(1 - \frac{b}{a}\right)$$

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Bode Sensitivity

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→ Given System Model

$$P(s)$$

P - model
transfer function
 \tilde{P} - transfer
function for actual
systems

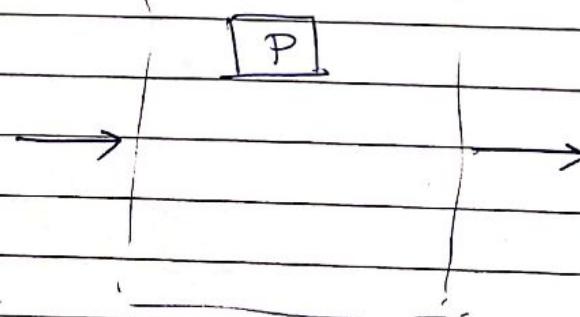
→ $\Delta P \equiv$ Perturbation / uncertainty

p.u or % variation

$$\frac{\Delta P}{P}$$

another T.F

→ Modify P by encapsulating in T



(T) → Also has T associated

p.u variation after control = $\frac{\Delta T}{T}$

→ Sensitivity of T configuration,

Sensitivity $(S) = \left(\frac{\Delta T / \tilde{T}}{\Delta P / \tilde{P}} \right)$

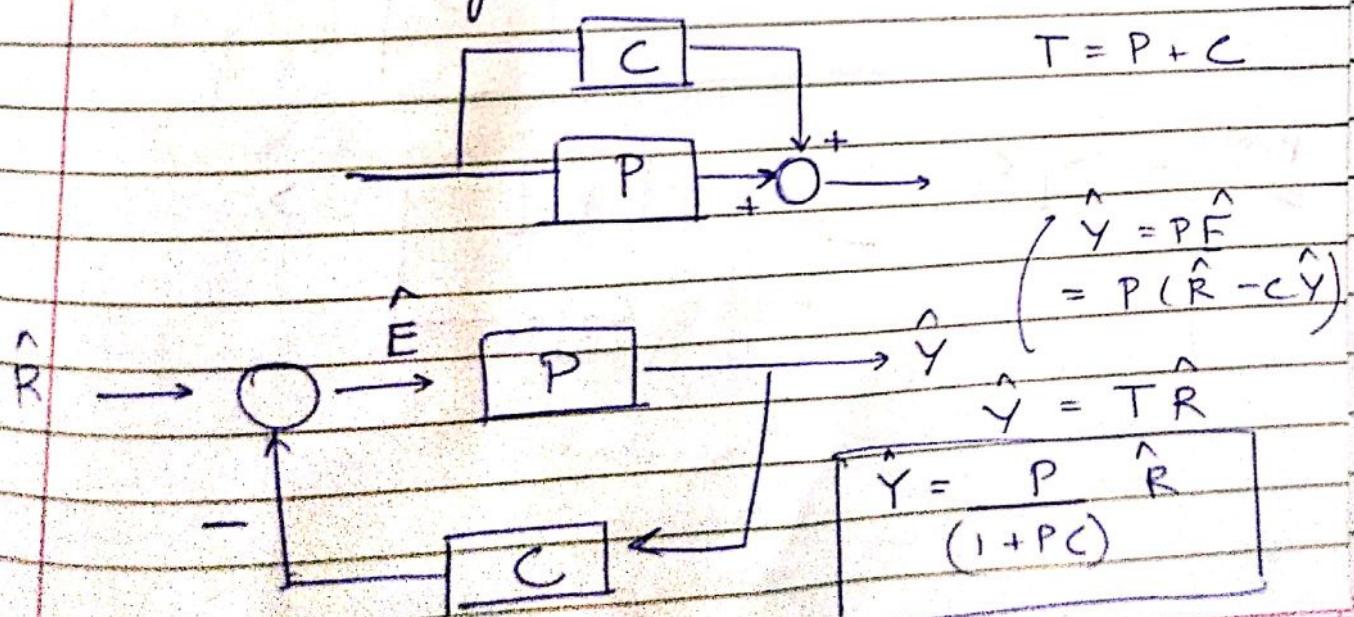
\rightarrow No control $\Rightarrow (S = 1)$

$\rightarrow S < 1$ can be done only over a small b/w.
 \rightarrow Ideal perfect system / known exactly
 $\rightarrow C \rightarrow P$

ex $C \rightarrow$ Electronics $T = CP$
 $P \rightarrow$ Plant $\tilde{T} = C\tilde{P}$
 $\underline{\Delta T = C\Delta P}$

$$S = \frac{C\Delta P / C\tilde{P}}{\Delta P / \tilde{P}} = 1$$

\rightarrow Simple Cascade Control has no effect on Sensitivity



12.2

$$\hat{Y} = P \hat{R} \rightarrow T.F. \text{ of} \\ \text{closed low} \\ \text{system}$$

$$\Delta T = \tilde{T} - T$$

$$\tilde{P} - P$$

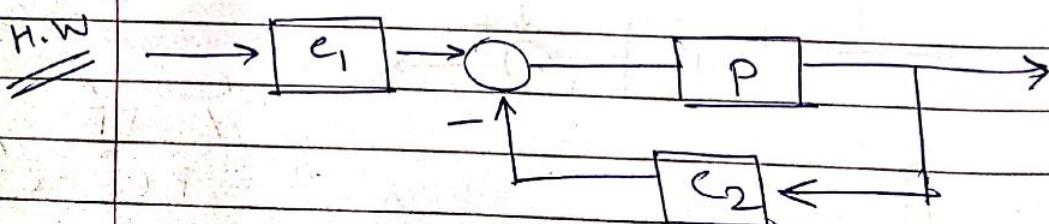
$$\tilde{P} + P \tilde{C} - P - P \tilde{C} = \Delta P$$

$$(1+P C)$$

~~10~~

$$S = \frac{\frac{\Delta P}{(1+P C)} / \frac{\tilde{P}}{(1+\tilde{P} C)}}{D \Delta T / \tilde{P}}$$

$$S = \frac{\tilde{P} / (1+P C)}{1}$$



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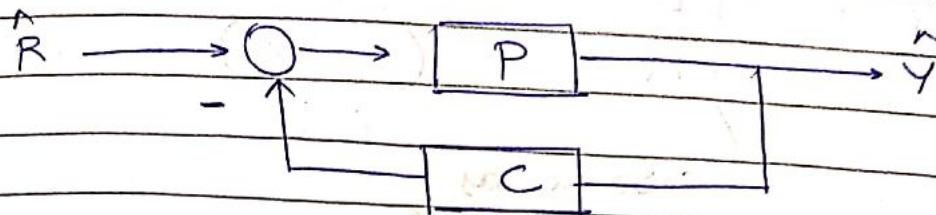
Sensitivity Function in feedback Control

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1)



2)

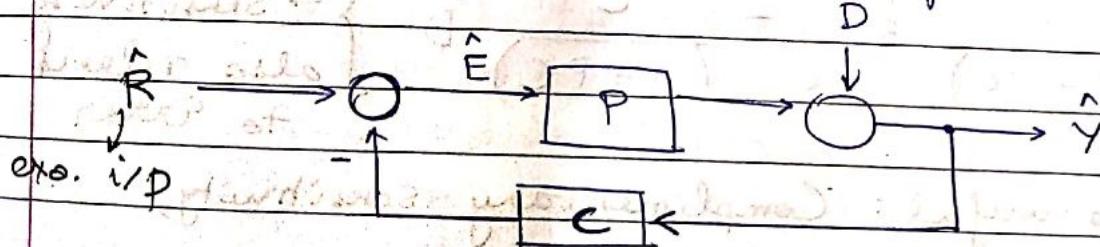


$$\hat{Y} = \left(\frac{P}{1 + PC} \right) \cdot \hat{R}$$

$\underbrace{T}_{\text{Time constant}}$

$$S = \frac{\Delta T / \tilde{T}}{\Delta P / \tilde{P}} = \frac{1}{1 + PC}$$

→ Properties of S in FB System



→ The qty measured is always a disturbed

version

$$\hat{Y} \rightarrow \hat{Y} + \hat{D}$$

$\hat{P} - \tilde{P} \rightarrow$ Uncertainty in System

$\hat{D} \rightarrow$ external Disturbances

- outside the Plant . . .

exogenous input

Equations,

$$\hat{Y} = P\hat{E} + \hat{D}$$

$$= P(\hat{R} - C\hat{Y}) + \hat{D}$$

$$\frac{\hat{Y}}{1+PC} = \frac{P\hat{R}}{1+PC} + \frac{\hat{D}}{(1+PC)}$$

Same as
effect of \hat{T} from
 $\hat{R} \rightarrow \hat{Y}$

T is only from $\hat{R} \rightarrow \hat{Y}$

$$\hat{E} = \hat{R} - C\hat{Y}$$

$$= \hat{R} - C \left(\frac{P}{1+PC} \hat{R} + \frac{1}{1+PC} \hat{D} \right)$$

$$\frac{1}{1+PC} \hat{R} - \frac{C}{1+PC} \hat{D}$$

3

$\left. \begin{array}{l} \text{→ Sensitivity} \\ \text{also related} \\ \text{to error} \end{array} \right\}$

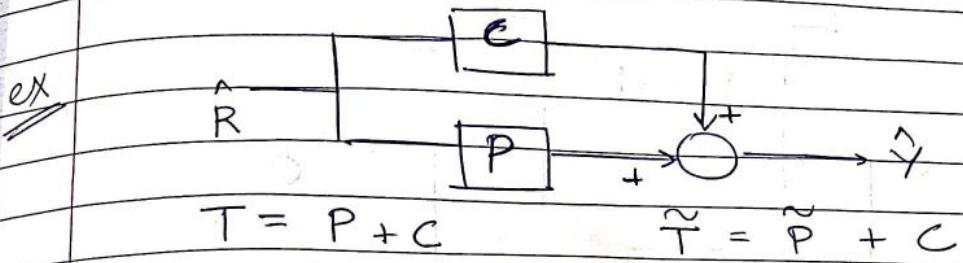
→ Also useful: Complementary sensitivity,

$$S_c = \frac{PC}{1+PC} = 1 - S, S = \frac{1}{1+PC}$$

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1. Making sensitivity "small" is important in CSD

2. Feedback control scheme has scope for making Sensitivity small



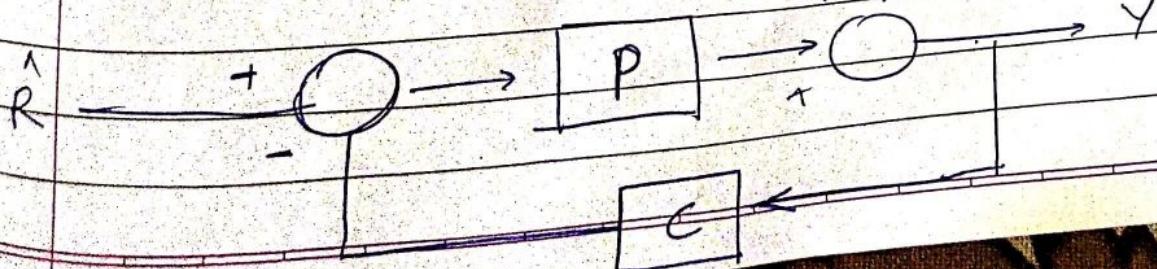
$$\begin{aligned} \Delta T &= \Delta P \\ S &= \frac{\Delta T / \tilde{T}}{\Delta P / \tilde{P}} = \frac{\Delta P}{\tilde{P}} \cdot \frac{\tilde{P}}{\tilde{P} + C} = \frac{\Delta P}{\tilde{P} + C} \end{aligned}$$

3. Mathematical framework for handling S

* Ideas: Nbd (neighbourhood), distance, layers, smallness of transfer functions

4. What are possible limits to changing sensitivity

5. Under what constraints should we design a control system?



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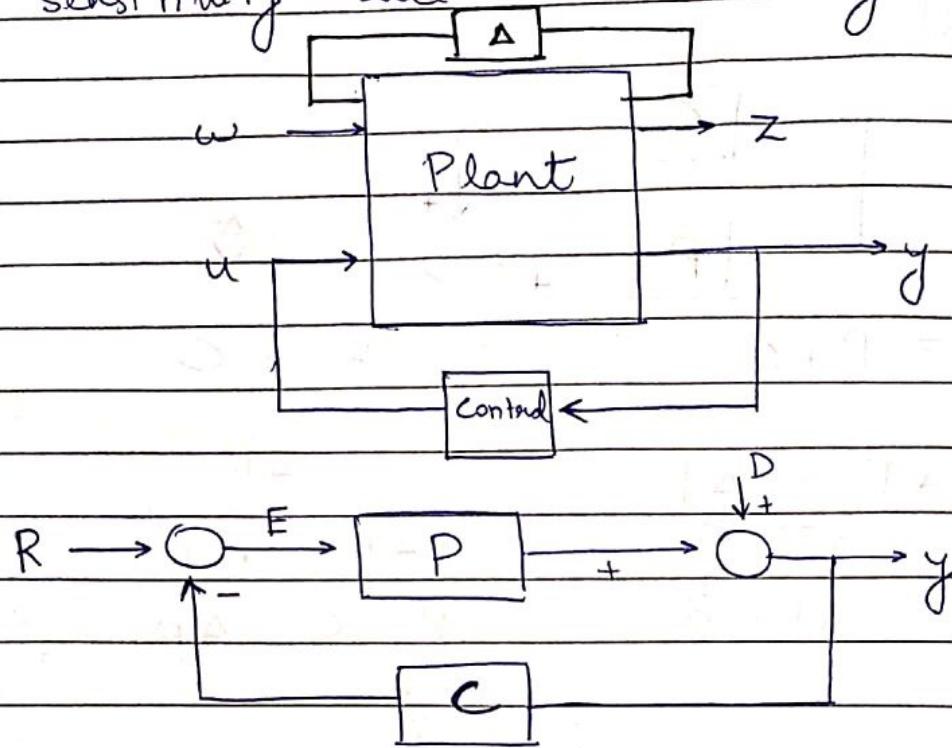
Control System Objectives

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Young

- Desired performance against reference inputs, regulation etc.
- Feedback control is necessary to reduce sensitivity due to uncertainty in the model



$$\hat{Y} = \left(\frac{1}{1+PC} \right) \hat{D} + \left(\frac{P}{1+PC} \right) \hat{R}$$

$$\hat{E} = \left(\frac{1}{1+PC} \right) \hat{R} + \left(\frac{C}{1+PC} \right) \hat{D}$$

Mathematical Stability Theory

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YUVRAJ

1. Constraints of feedback control?
2. Limits to achievable performance?
3. What is a Stable System?
4. How to stabilize a given system?
5. Is there a characterization/classification of stable linear feedback systems?

Constraints

- (i) Consider only Laplace transformable signals

$$f(t) \in [0 \infty)$$

$$\hat{F}(s) = \int_0^\infty f(t) e^{-st} dt, \quad \hat{F}(s) \text{ exists}$$

when $|\hat{F}(s)| < \infty$

over

$$s = \sigma + j\omega$$

→ Also, $f(t)$ should be bounded,

$$|f(t)| \leq c e^{\sigma t}$$

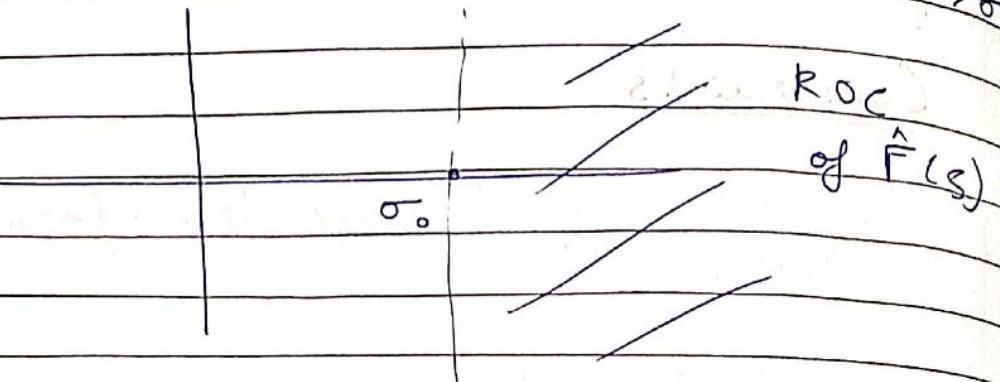
$$|\hat{F}(s)| = \left| \int_0^\infty f(t) e^{-\sigma t} e^{-j\omega t} dt \right|$$

$$\leq \int_0^\infty |f(t)| e^{-\sigma t} dt \leq \int_0^\infty c e^{(\sigma - \sigma)t} dt$$

$$\text{if } |f(t)| \leq ce^{\sigma_0 t} < \infty \quad \text{if } \sigma > \sigma_0$$

$$\rightarrow \text{if } |f(t)| \leq ce^{\sigma_0 t}$$

then $|\hat{F}(s)| < \infty$ for all $\sigma = \text{Re } s > \sigma_0$



$\rightarrow B$ is the set of all functions $f(t)$
s.t

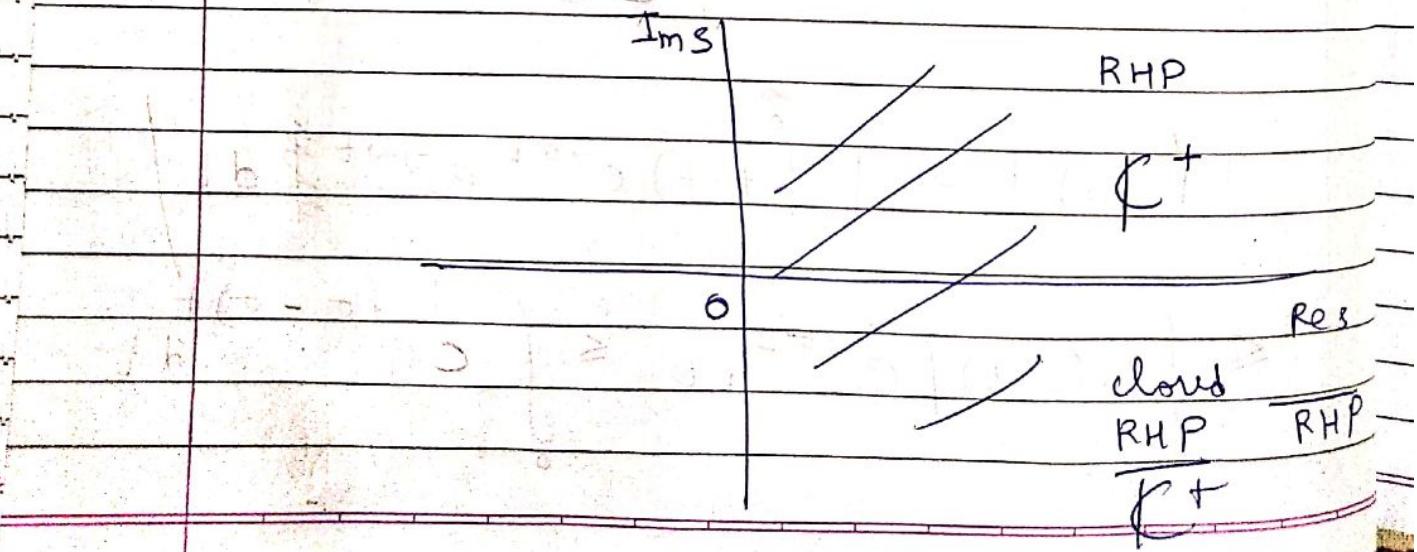
$$|f(t)| \leq M < \infty$$

hence,

ROC of any function

$f \in B$ in the R.H.P

$$\text{R.H.P} = \{s \in \mathbb{C} \mid \text{Re } s > 0\}$$



2.19

Stable Systems

- Bounded functions: (Recap)

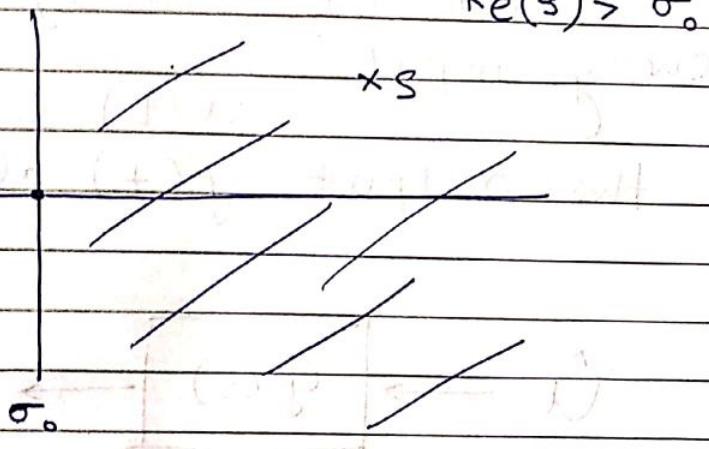
Uniformly bounded functions $f(t)$, $t \in [0, \infty)$ satisfy,

$$|f(t)| < M < \infty$$

$f(t)$ is Laplace Transformable if $\exists \sigma_0$ s.t

$$|f(t)| \leq ce^{\sigma_0 t}$$

When $f(t)$ is uniformly bounded ($< M$) then $\hat{F}(s)$ has no poles in $\Re(s) > \sigma_0$



- If $\hat{F}(s)$ has no poles in $\Re(s) > \sigma_0$ and those poles on jw Axis are simple then $f(t)$ is uniformly bounded

(ex $\frac{1}{s} \rightarrow u(t)$: uniformly bounded
Simple poles)

$\frac{1}{s^2} \rightarrow \text{ramp}(t)$: Not uniformly bounded

ex $\frac{1}{s^2} / \frac{1}{(j\omega)^2} = \frac{s}{(s^2 + \omega^2)}$

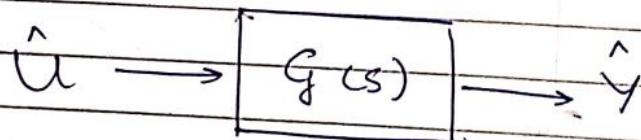
Not uniformly bounded

→ Let B be the set,
 B : the set of bounded functions of
 $t \in [0, \infty)$
which have Laplace Transforms satisfying,

1. No poles in the \mathcal{P}^+ (RHP)
2. Poles on jw Axis are simple

→ Then, an input - output linear system represented by a transfer function $G(s)$ is said to be "stable" (BIBO stable), if for every input

$u(t) \in B$,
the output $y(t)$ also $\in B$



BIBO stable system

$$u(t) \in B \Rightarrow y(t) \in B$$

$$\mathcal{L}^{-1}(u) \in B$$

$$\mathcal{L}^{-1}(u \cdot G(s)) \in B$$

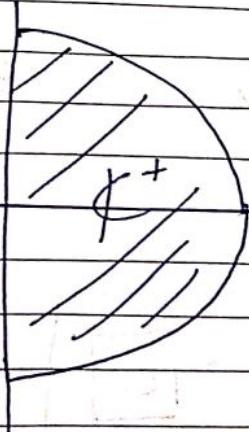
Thm If a system $G(s)$ is represented by a transfer function $G(s)$ which has no pole-zeros cancellation in the closed RHP (\mathcal{P}^+ or $\overline{\text{RHP}}$) then the system is BIBO stable iff $G(s)$ has no poles in \mathcal{P}^+

$$G(s) = \frac{q(s)}{p(s)}$$

\Rightarrow q. p(s) has no roots in \mathbb{C}^+

BIBO Stable Systems are denoted as \mathcal{S}

\mathcal{S} has no poles in $\overline{\text{RHP}}$ (or \mathbb{C}^+)

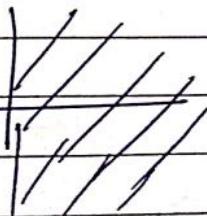


Properties

$G(s), H(s) \in \mathcal{S}$
then $G+H \in \mathcal{S}$
 $GH \in \mathcal{S}$
all constants

$iR \in \mathcal{S}$

Same as \therefore If $G \in \mathcal{S}$, when does



$$\frac{1}{G} \in \mathcal{S} ?$$

- $q(s)$ also has no roots in RHP
- $\deg(p(s)) = \deg(q(s))$

Examples

$$\frac{100(s-2)}{(s+1)(s+3)} \in \mathcal{S}$$

$$\frac{s^2 + 3}{(s+1)^3} \in \mathcal{S}$$

$$\frac{(s+1)^3}{(s+1)^3}$$

$$\frac{(s-1)(s+2)}{(s+1)(s-2)}$$

$\neq s$

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Hurwitz Polynomials

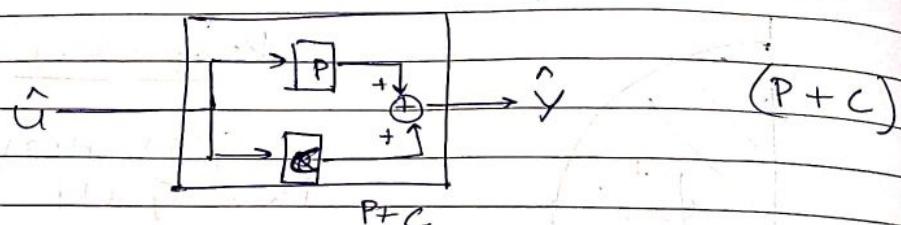
→ Polynomials (pc's) with no roots in the left half-plane.
(Note relation with $s \rightarrow$ denominator)

Alexa

Anat

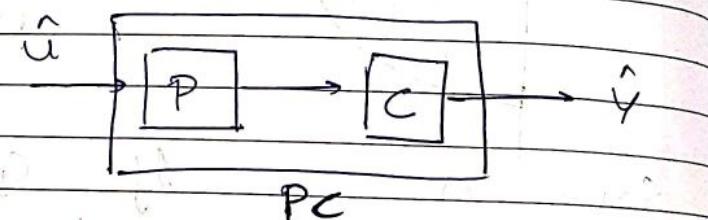
Connections of Systems

1) Sum.



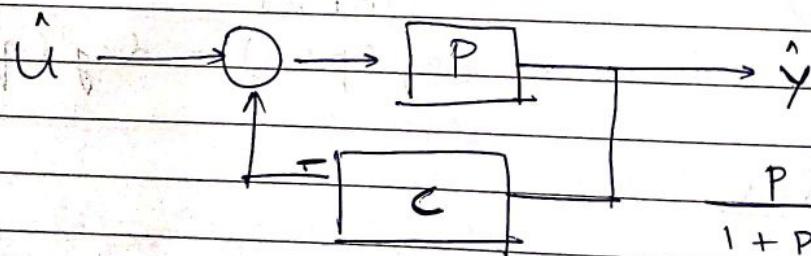
2)

Cascade



3)

Feedback Connection



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P
 $1 + PC$ → transfer function

Thm

Open Loop TF

PC

Sensitivity

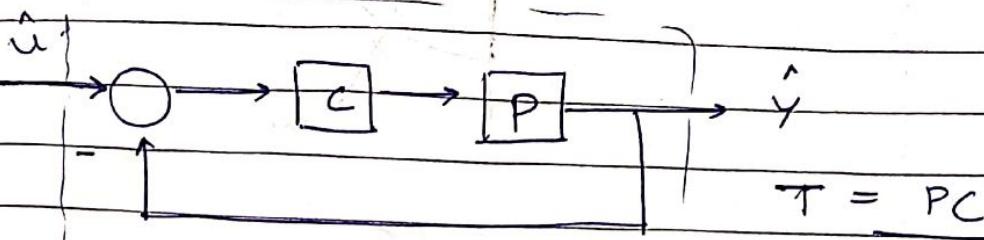
$$S = \frac{1}{1 + PC}$$

Complementary
Sensitivity

$$S_C = \frac{PC}{1 + PC}$$

Alternately

→ Another form of feedback System

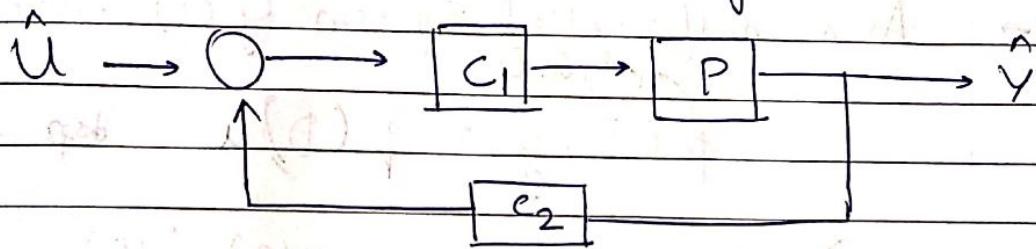


$$OL \quad TF \quad PC$$

$$\begin{aligned} &= 1 + PC \\ &= 1 - S \end{aligned}$$

Sensitivity

→ 2 - DOF feedback System,



$$\text{Open Loop TF } (c_1 c_2 P)$$

~~1.2.19~~ Class of Stable Systems
The set of BIBO stable systems

Thm If $G(s) = \frac{q(s)}{p(s)}$

has no pole-zero cancellation in \mathbb{C}^+

then $G(s)$ represents a BIBO

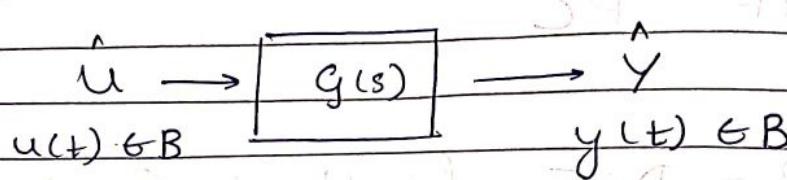
stable system iff

$G(s) \in \mathcal{S}$ iff

$p(s)$ is a Hurwitz polynomial

No poles in
closed RHP

$$y \quad g(s) \in \mathcal{S}$$



→ As a differential system $G(s)$ corresponds

$$p(D)y = q(D)u \quad \text{deg } p = \text{deg } q$$

$$D = \frac{d}{dt} \quad u = 0, \quad y(0), y'(0), \dots, y^{(n-1)}(0)$$

Explanation
Details

$$G(s) = \frac{(s+1)}{(s^2 + 3s + 5)} \quad \begin{array}{l} \text{Ratio of } L(\text{out}) \\ \text{under zero init. condns.} \end{array}$$

$$(s^2 + 3s + 5)\hat{Y} = (s+1)\hat{u}$$

$$(D^2 + 3D + 5)y = (D+1)u \rightarrow \text{zero state hom. soln.}$$

where $D = \frac{d}{dt}$

→ Zero input response of $y(t)$ is a soln. of

$$(D^2 + 3D + 5) y(t) = 0$$

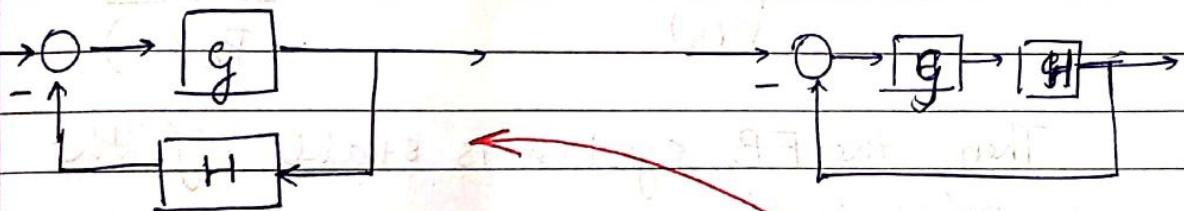
when $y(0) = y_0$, $y'(0) = y_1$

Internal Stability

$G(s)$ represents internally stable system
if $y(t) \rightarrow 0$ as $t \rightarrow \infty$ under
0 input.

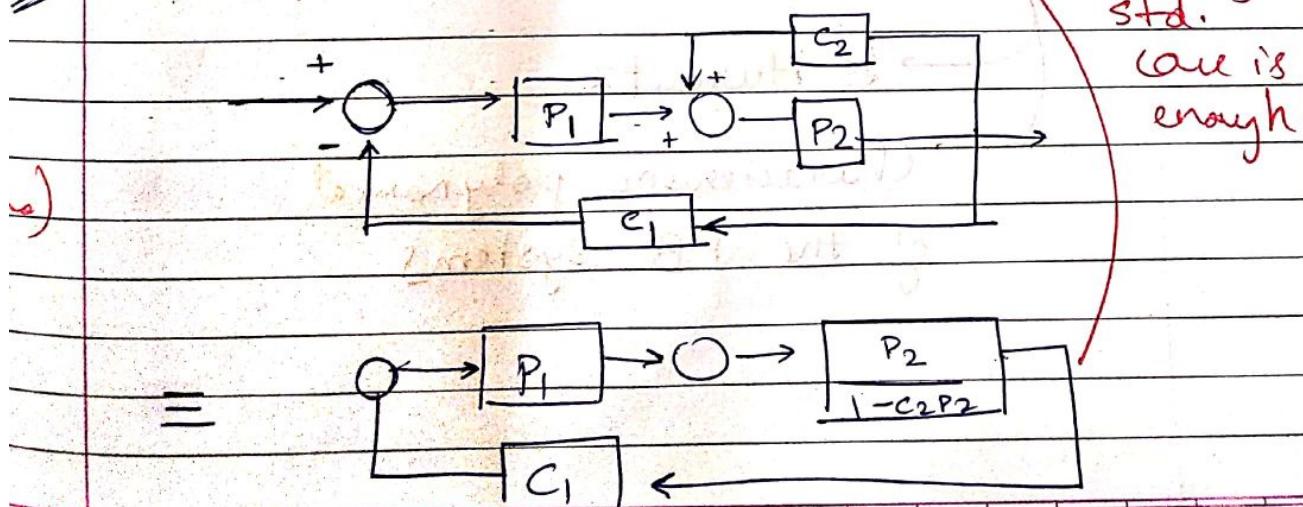
Thm $G(s)$ is internally stable iff $P(s)$
has no roots in \bar{P}^+ (Same as BIBO)

Feedback Configurations



ex Combinations of F.B Systems

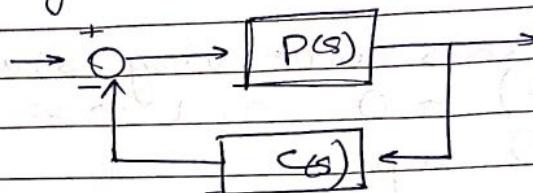
Analyzing std.
one is enough



Defn

- Stability of the closed loop system (BIBO)
 - 1. Input - Output Stability
 - 2. Internal Stability for a single block
- The same only when no pole = zero cancellation
 How it is defined for F.B system? ... later
- Condition for stability of a closed loop system:

Thm



Let $P(s)$ and $C(s)$ be transfer functions, each of which have no pole zeros cancellation in \mathbb{R}^+

$$P(s) = \frac{q(s)}{P(s)} \quad C(s) = \frac{q_c(s)}{P_c(s)}$$

Then the FB system is stable iff the polynomial

$$\psi(s) = P(s)P_c(s) + q(s)q_c(s)$$

is Hurwitz

Characteristic polynomial
 of the FB system