

TEKNILLINEN KORKEAKOULU
Systemiteknikan laboratorio

PID Controllers: Theory, Design and Tuning

Lecture content

- Introduction
- Basics of PID controllers
- Tuning of PID controllers
- Optimization in Matlab
- Auto tuning

PID-controllers: introduction

- By far the most popular controller
- In process control >95 controllers are of PI(D)-type
- Good for linear process control
- Relatively easy to understand (important reason for wide popularity)
- Still many of the PID-control loops are poorly tuned...

Typical paper mill

- Over 2000-500 control loops
- 97 % PI-controllers
- Only 20% of PI-controllers work well decreasing process variability
- Reason for poor performance:
 - 30% poor tuning
 - 30% valve problems
 - 20 % variety of problems (e.g sensor problems, bad choice of sampling rates...)

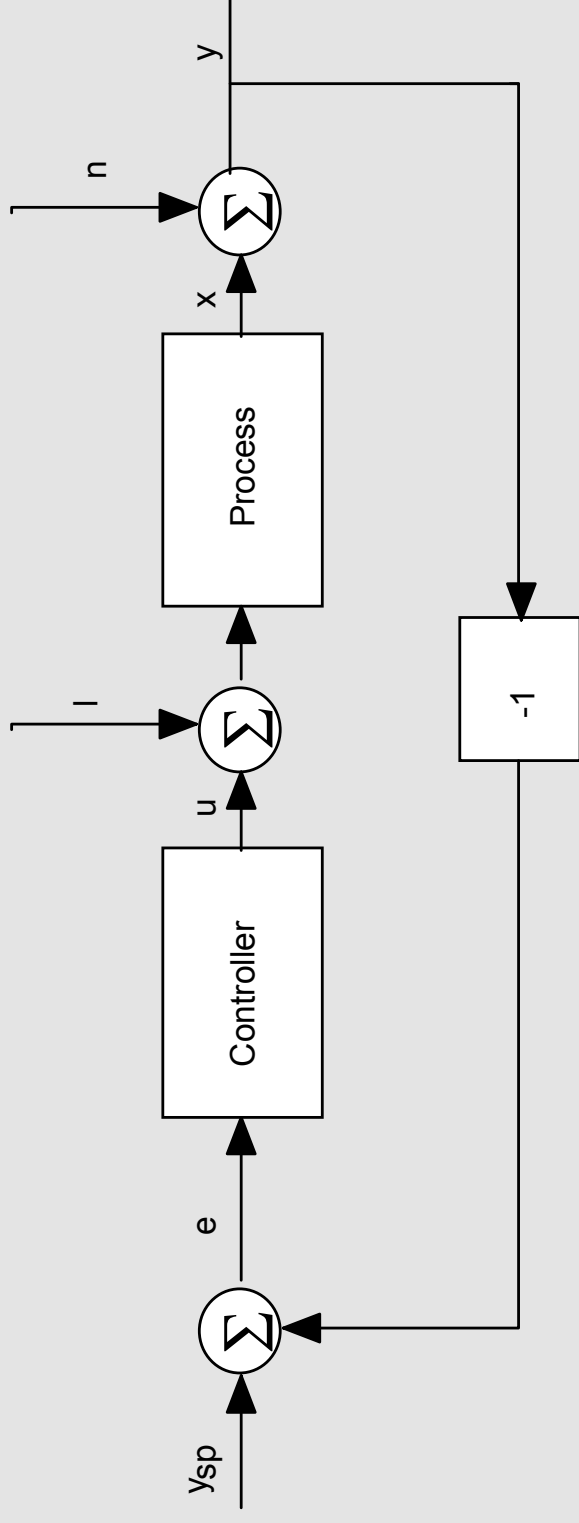
PID-controller

- Today most of the PID controllers are microprocessor based
- DAMATROL MC100: digital single-loop unit controller which is used, for example, as PID controller, ratio controller or manual control station.
- Often PID controllers are integrated directly into actuators (e.g valves, servos)



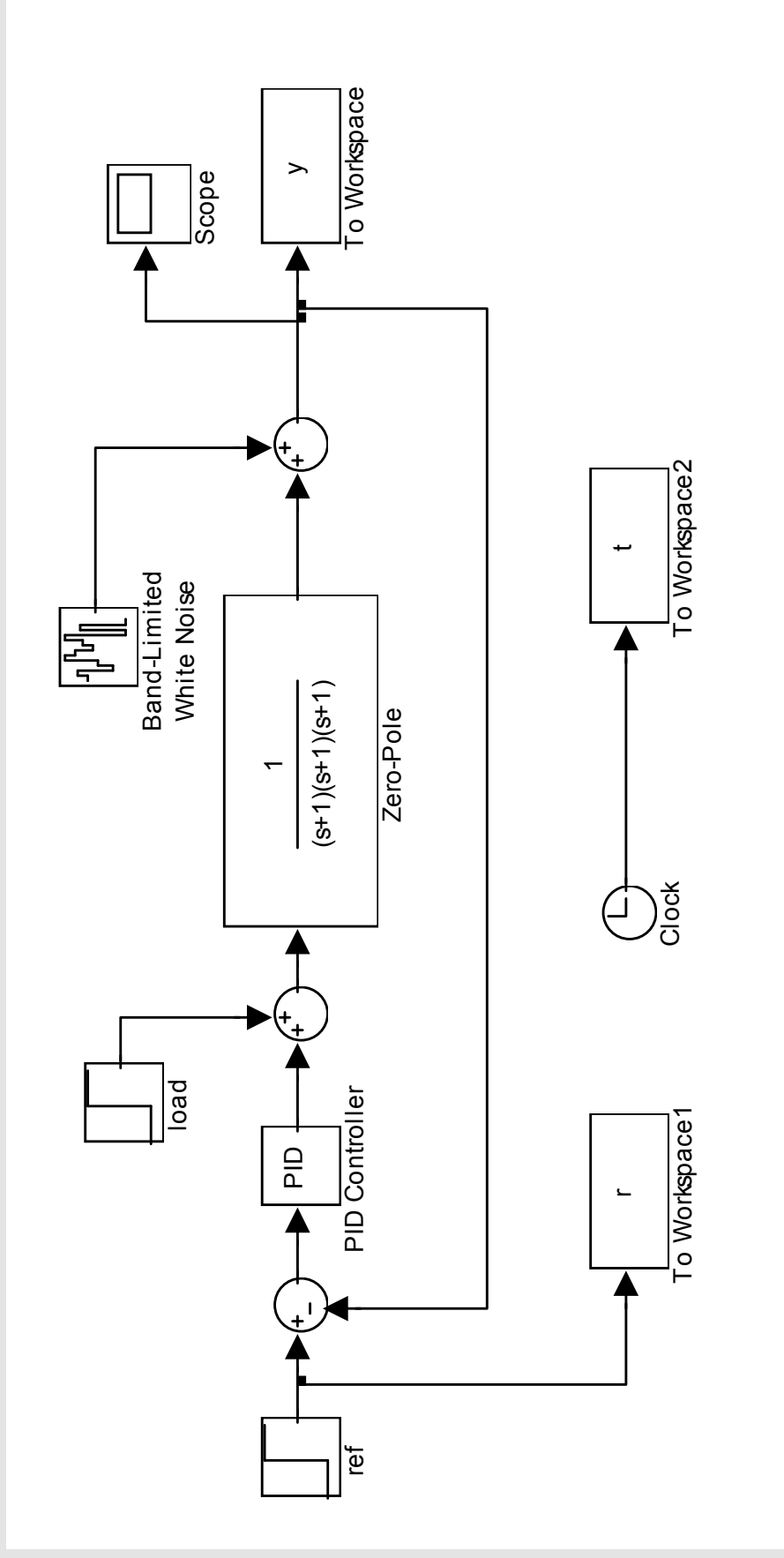
Simple idea of feedback...

- Principle of *negative* feedback:
 - ”Increase the manipulated variable when process variable is smaller than the setpoint and decrease the manipulated variable when the process variable is larger than the setpoint”



Simulink model

28.1.2009



PID control

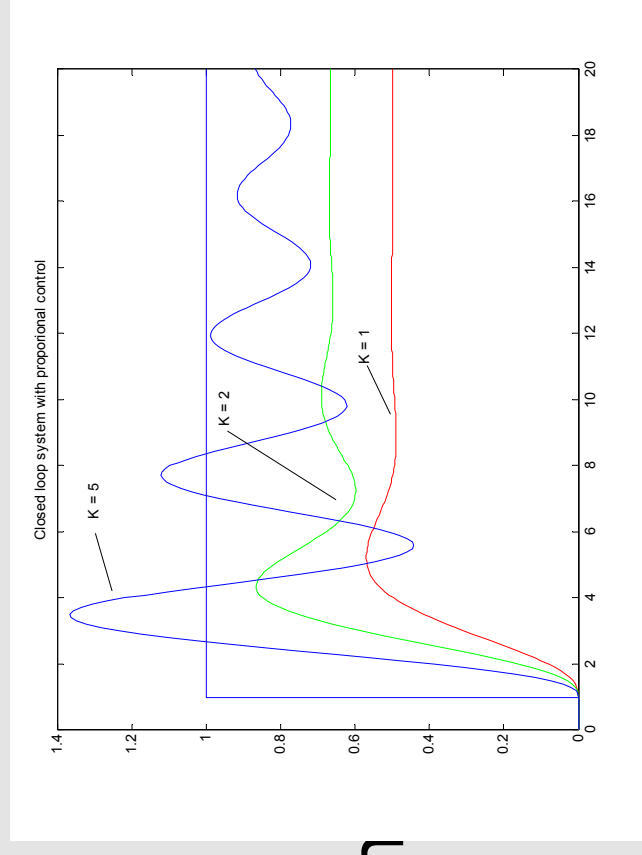
- "textbook" version of PID
- Control variable u is a sum of:
 - P -term (proportional to e)
 - I -term (proportional to integral of e)
 - D -term (proportional to derivative of e)
- Controller parameters:
 - K = proportional gain
 - T_i = integral time
 - T_d = derivative time

$$u(t) = K \left(e(t) + \frac{1}{T_i} \int_0^t e(\tau) d\tau + T_d \frac{de(t)}{dt} \right)$$

Proportional action

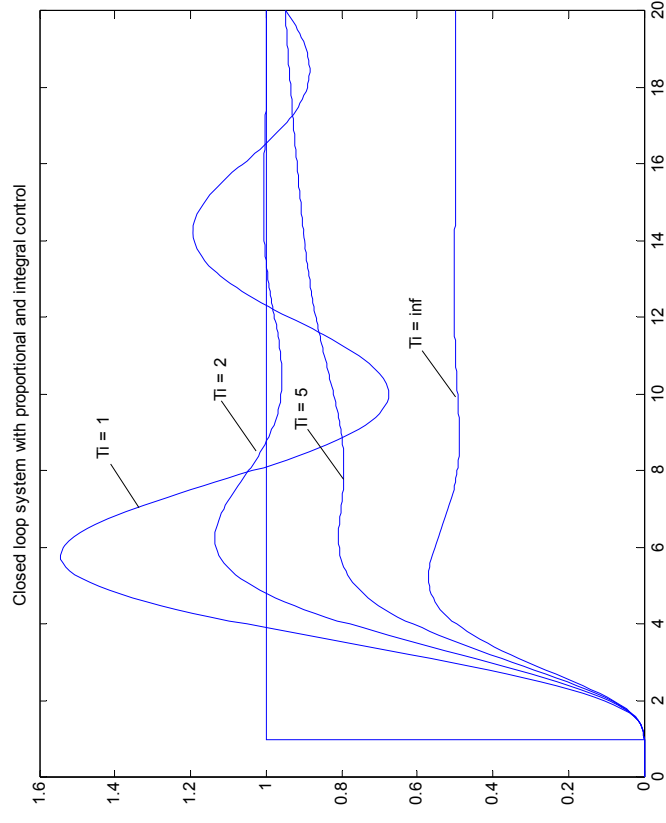
- High value of gain makes the system more insensitive to load disturbance
- Too large a gain makes the system more sensitive to measurement noise
- Steady-state error decreases when gain increases
- Oscillation however often increases

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



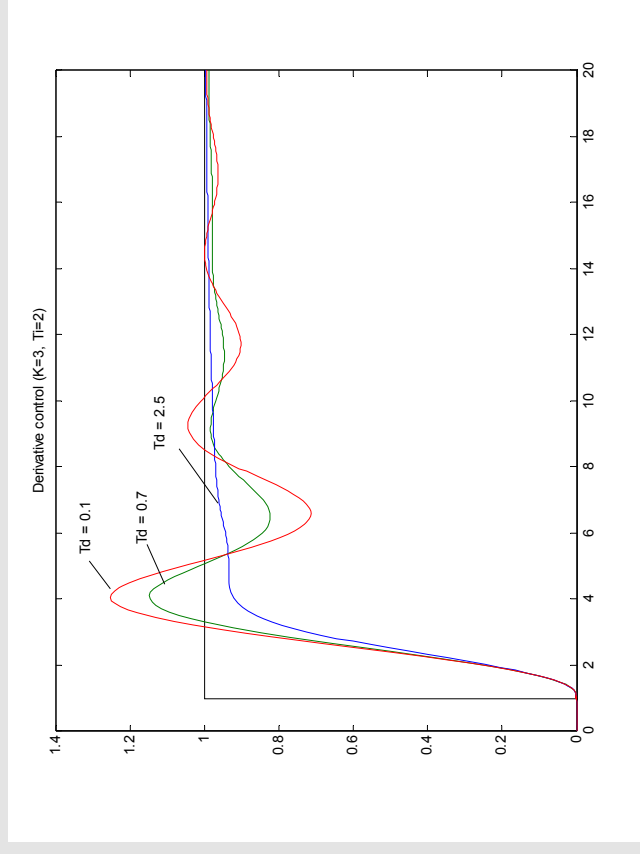
Integral action

- Integral term removes steady state error
- Short integration time often leads to oscillation
- Long integration time common in process control



Derivative action

- Derivative term can *predict* output
- Fast and stable response
- Noise can make derivative control problematic
- Also long delays are problematic when using derivative term



Derivative action

- Fast changes in reference signal result high derivatives → control signal saturates
- Fixes:
 - computing derivatives from process output
 - using filtered derivative term (this is used often in real applications)

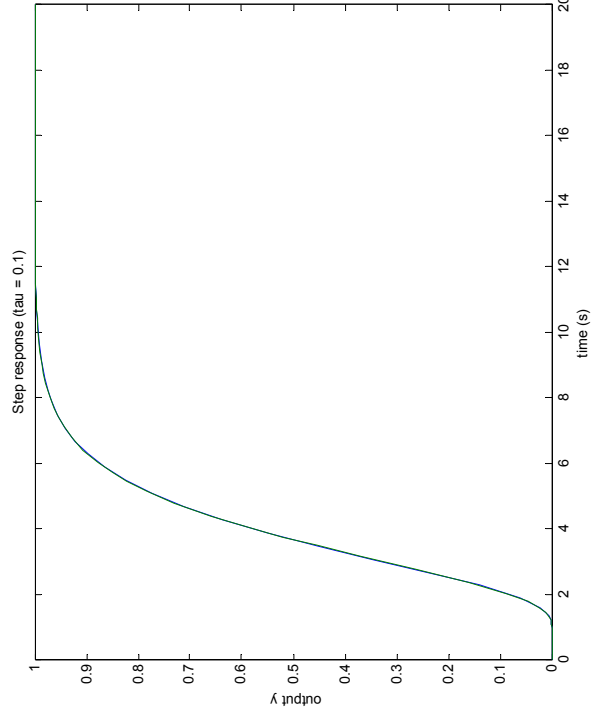
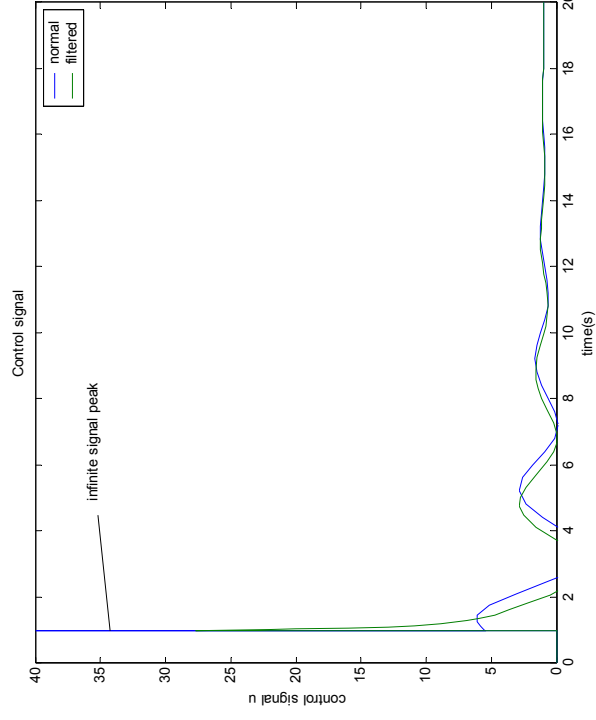
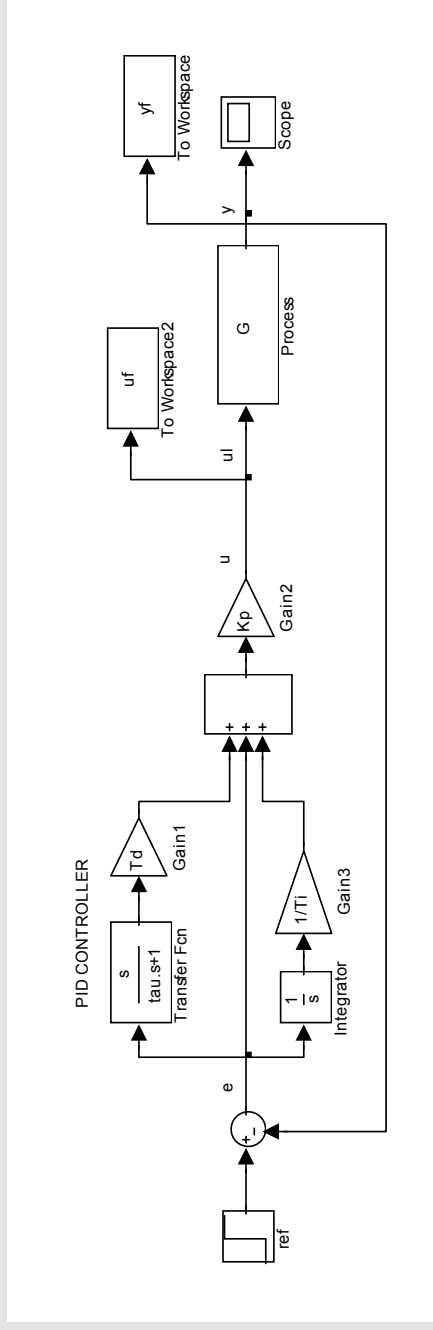
Filtered derivative term:

- Benefits:
 - easy to implement in practise
 - more insensitive to noise than normal derivative term
 - corresponds with derivation of low pass filtered signal
 - by choosing tau small
- system has same response as by using normal derivation

$$\left\{ \begin{array}{l} G_{s,1}(s) = s \\ G_{s,2}(s) = \frac{s}{\tau_s s + 1} \end{array} \right., \lim \{ G_{s,2}(s) \} = s = G_{s,1}(s)$$

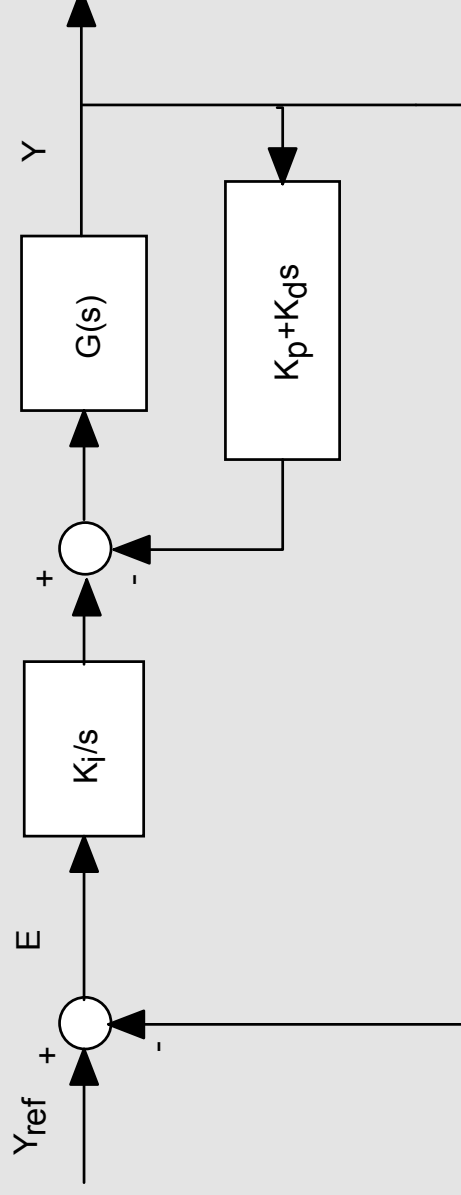
Filtered derivative

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Output derivation – Tachometer feedback

- Use process output for derivation and gain
- No zeros to controlled closed-loop system
(prevents overshoots)

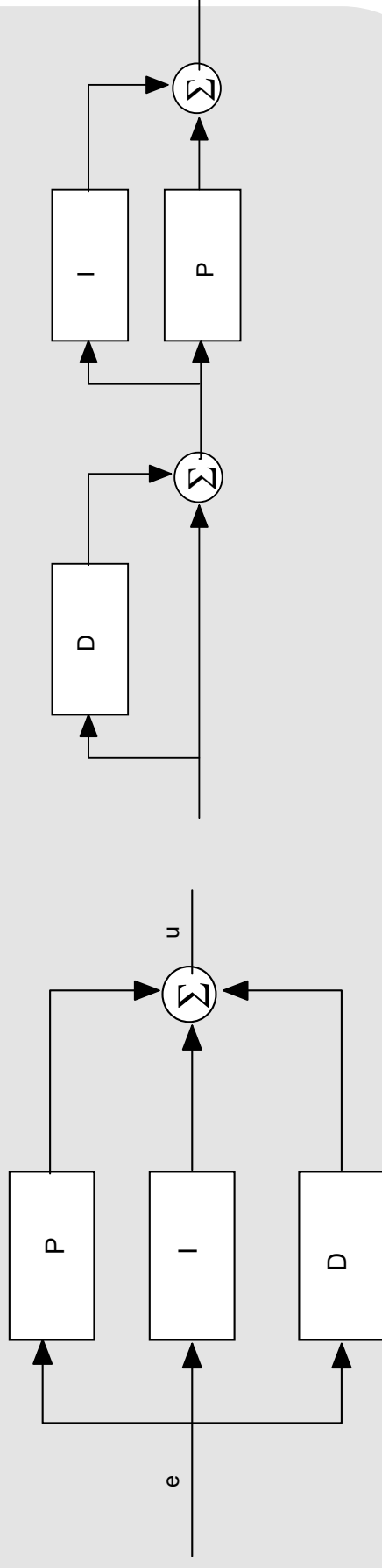


Alternative representations

- Non-interacting & interacting
- Non-interacting more general
- Interacting common in commercial controllers (said to be easier to tune manually)

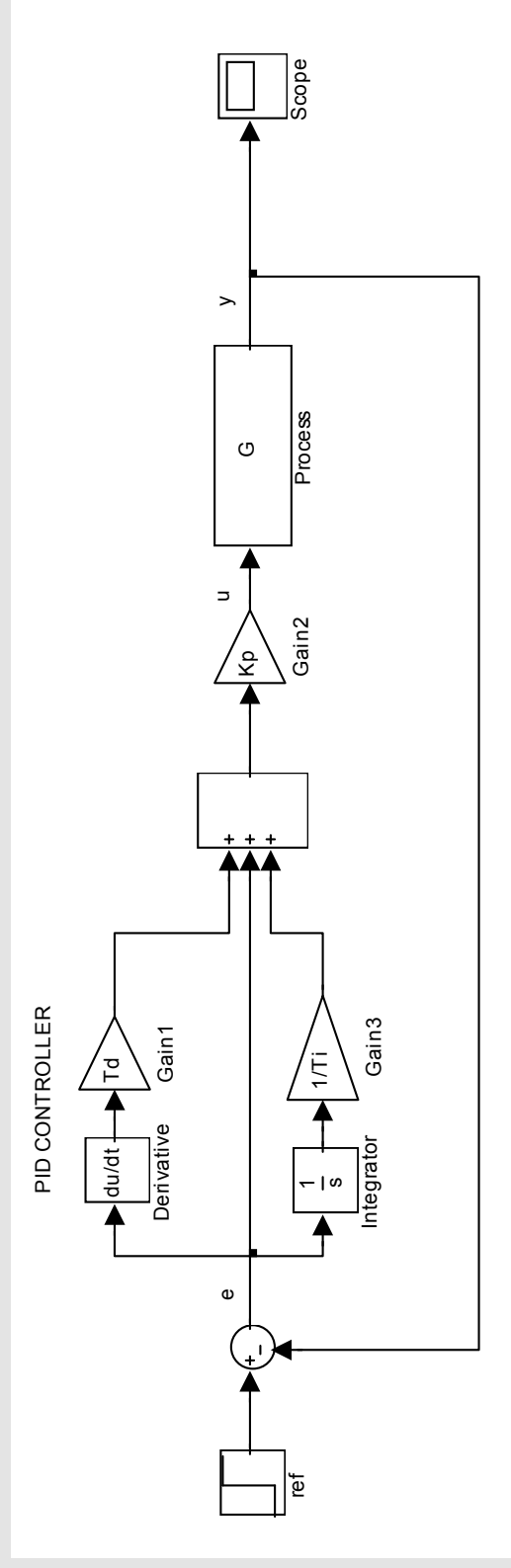
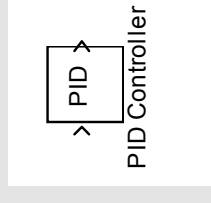
$$G(s) = K \left(1 + \frac{1}{sT_i} + sT_d \right)$$

$$G'(s) = K' \left(1 + \frac{1}{sT_i'} \right) (1 + sT_d')$$

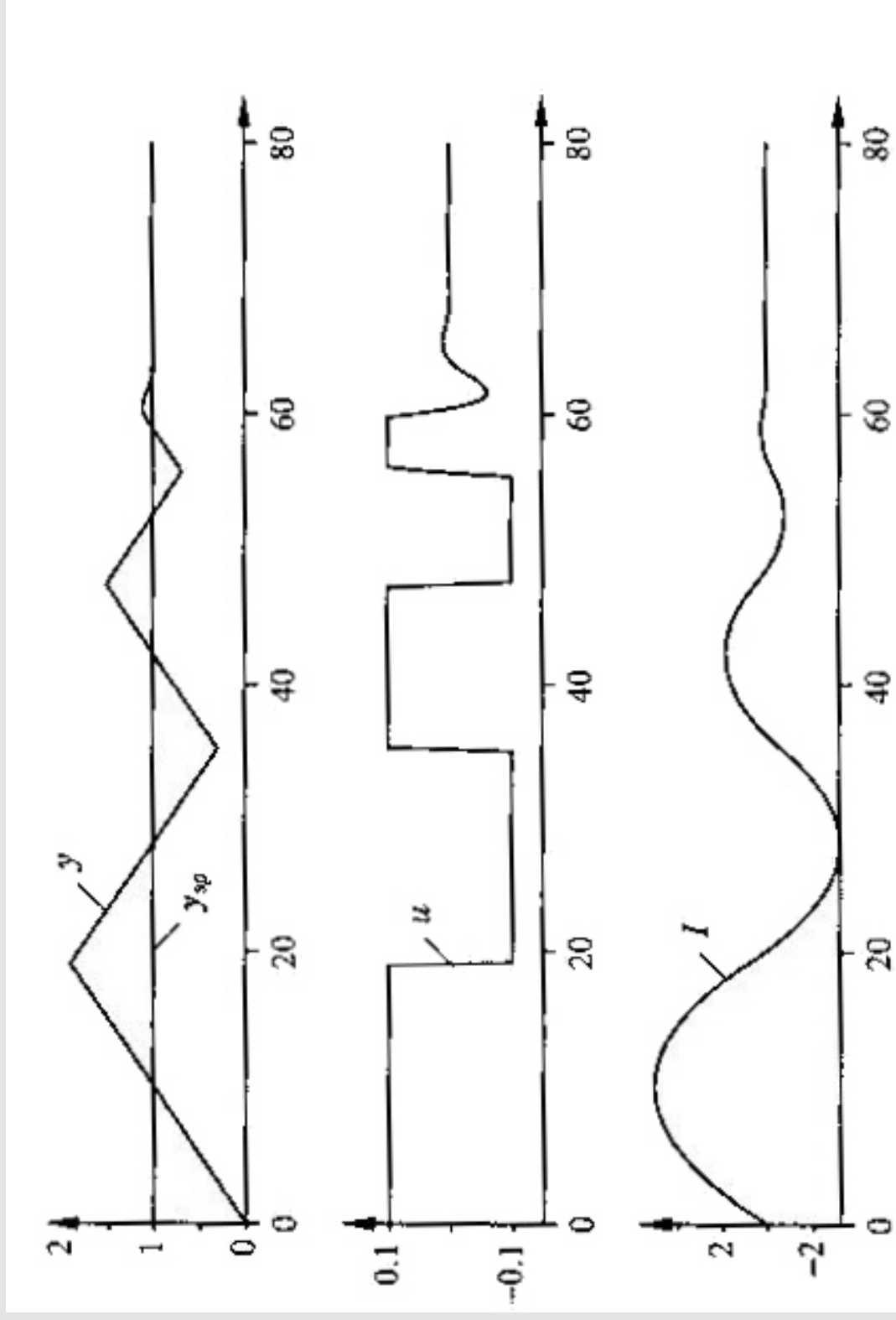


SIMULINK PID-controller

- Simulink PID-block is of form:
$$K_p + K_i \frac{1}{s} + T_d s$$
- ”text book” version is:



Integrator windup



Integrator windup

- All actuators have limitations:
 - limited speed
 - valve opening
- If the control variable reaches actuator limits → feedback loop is broken!
- Still error will continue to integrate → very large integral term (“wind up”)
- Large transients when the actuator saturates

Integrator windup

- Integrator action must be stopped when output saturates!
- Solutions:
 - setpoint limitation (limit performance, windup caused by disturbances?)
 - incremental algorithms
 - back calculation and tracking
 - conditional integration

When is PI control sufficient?

- Often derivative action switched off
 - Dominant dynamics are of the 1. order
 - For example:
 - level control in single tank
 - stirred tank with perfect mixing...
 - When tight control not needed
- PI-control adequate

When is PID control sufficient?

- Dominant dynamics are of the 2. order
- PID speeds up the response versus PI
 - damping improved
 - higher gain can be used to speed up transient response
- For example:
 - temperature control

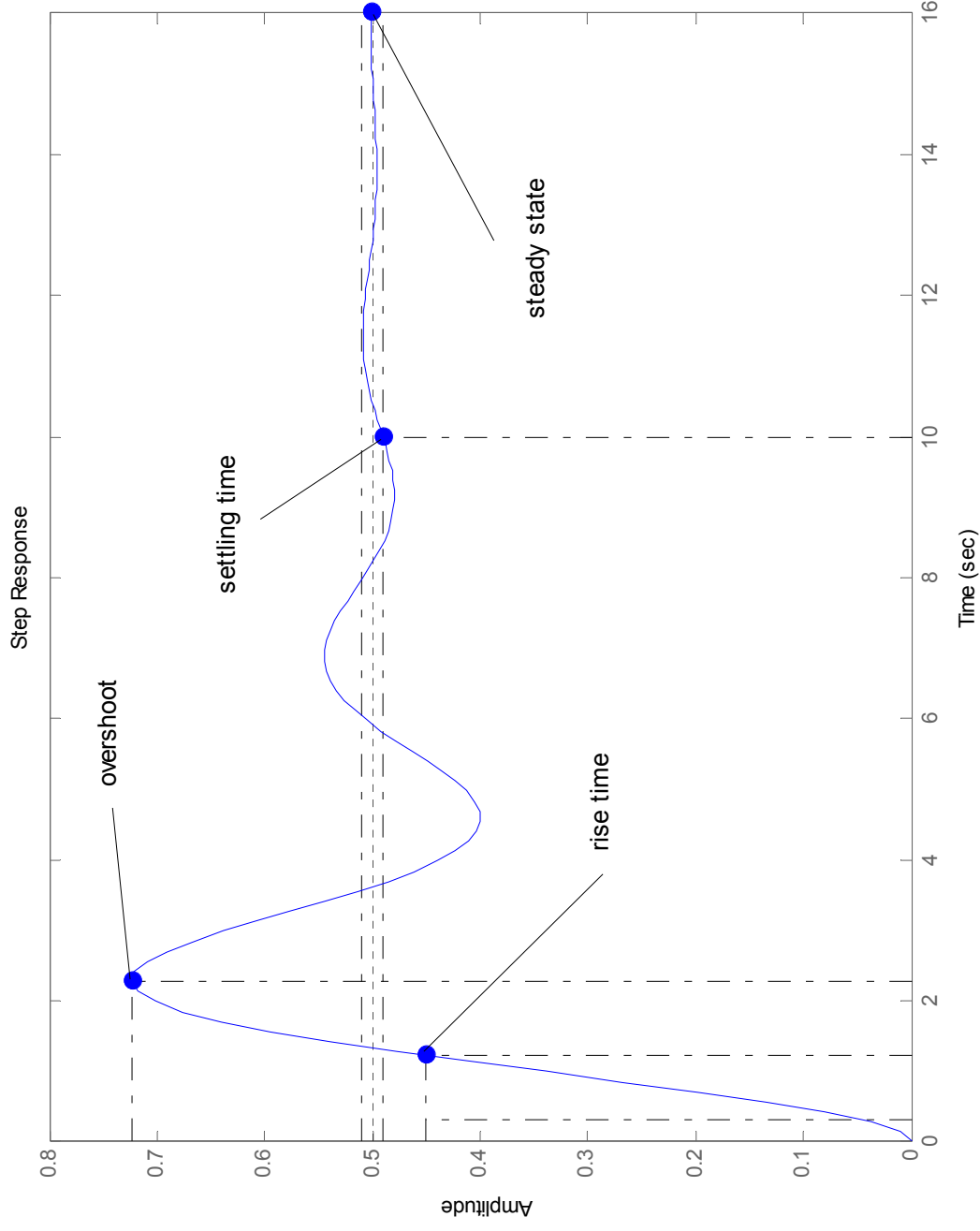
When PID control is insufficient?

- System has high order dynamics
- System is time variant
- Long delays
- Non-linear process
- MIMO/MISO system with strong cross dependencies

Controller design

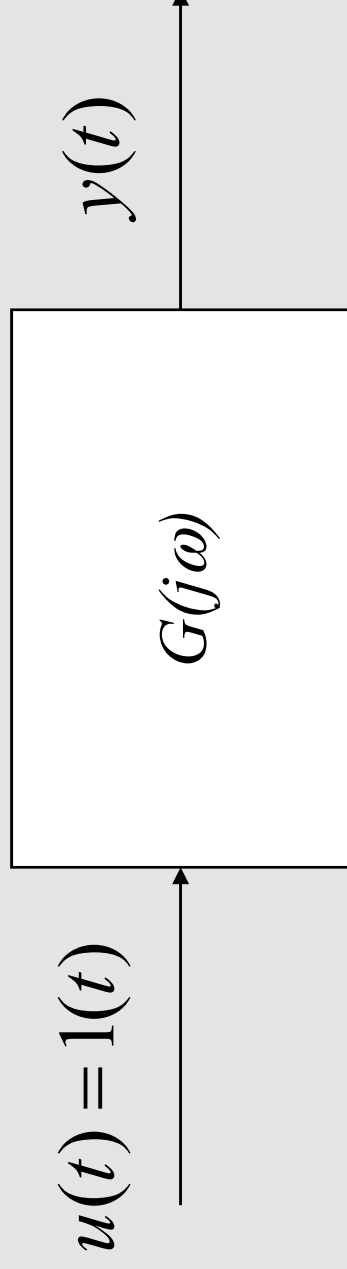
- Problem: how to determine the parameters?
- Tuning is a trade-offs between:
 - load disturbance attenuation
 - effects of measurement noise
 - robustness to process variations
 - response to setpoint change
 - model requirements
 - (computational requirements)

Performance criteria

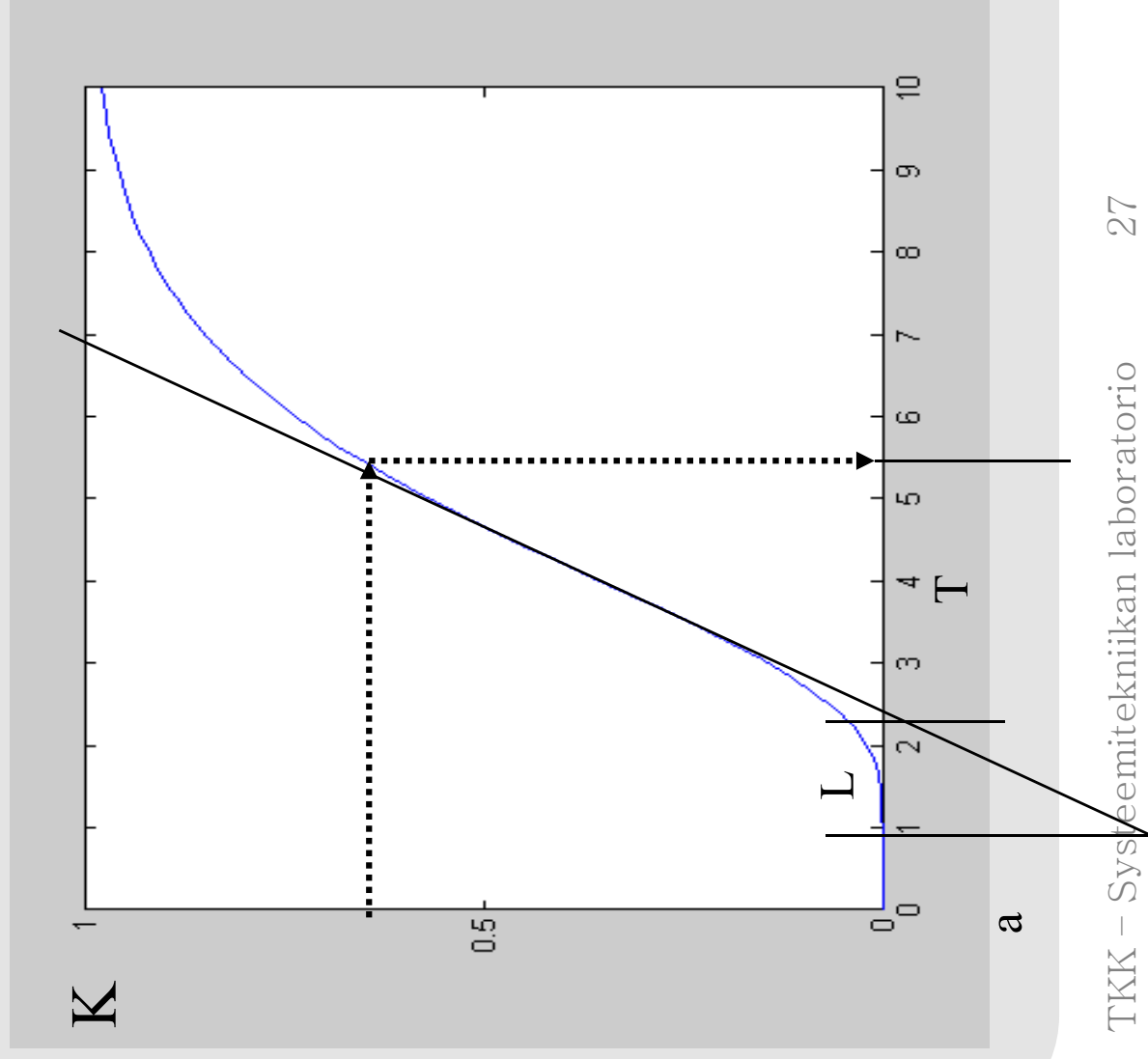


Open loop tuning

- Open loop
- Identify plant dynamics
- Let $u(t)$ be a unit step ('good' test signal)
- Measure output



Open loop tuning



$$G(s) = \frac{K}{1+sT} e^{-sL}$$

From figure

$$K=1$$

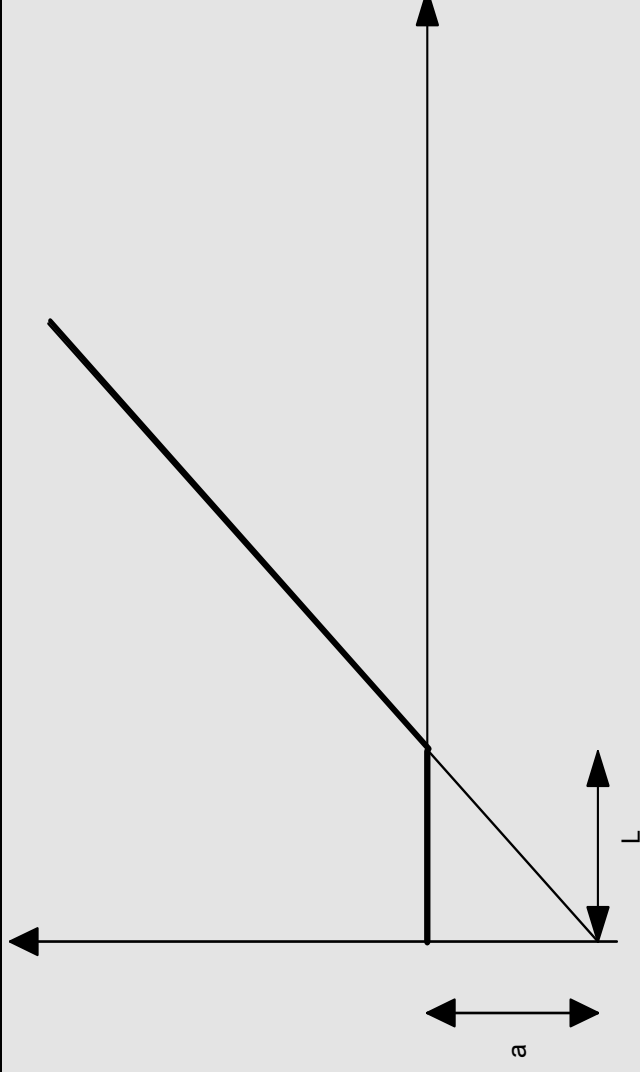
$$L=1.3 \text{ s}$$

$$T=4.4 \text{ s}$$

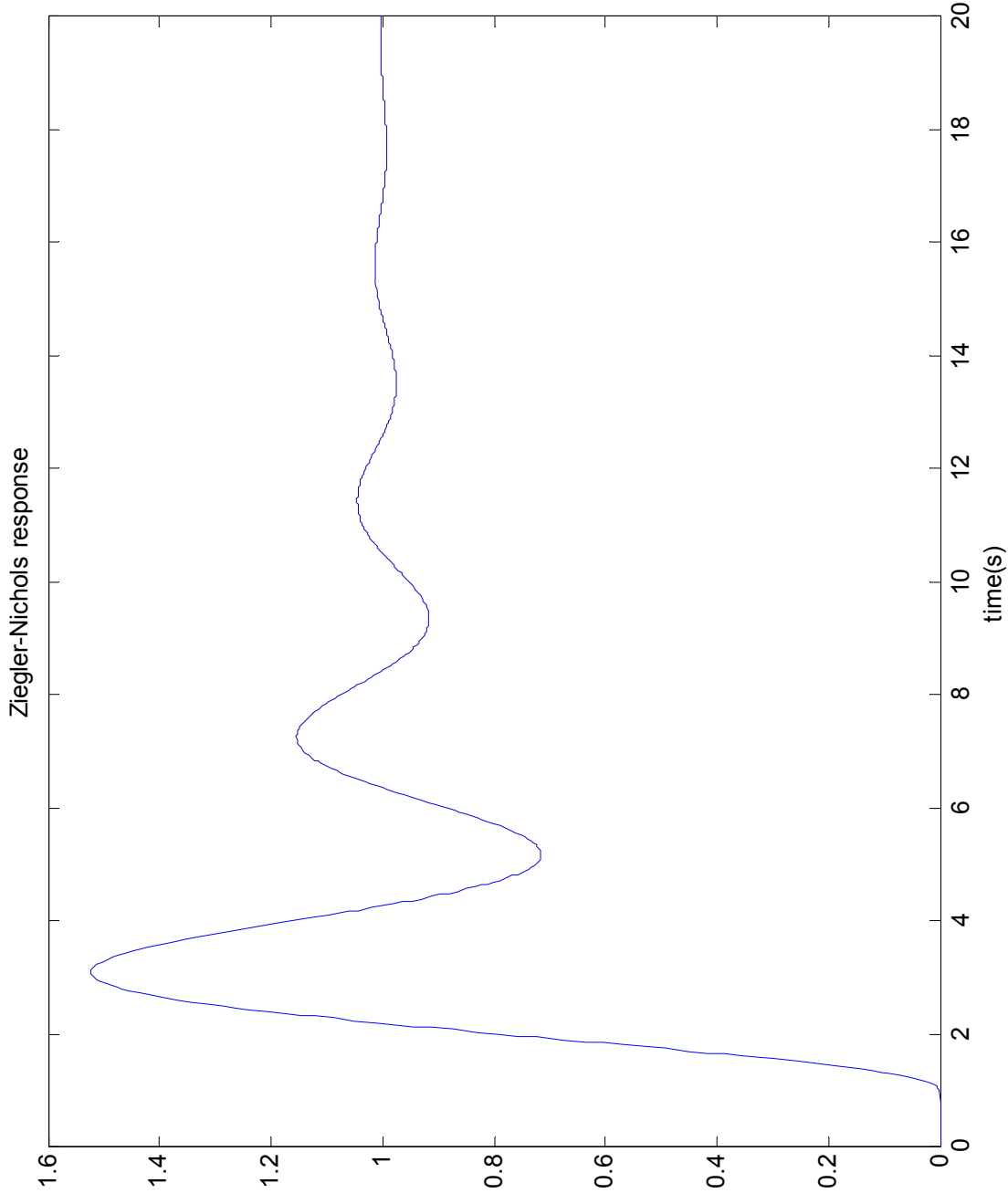
$$a=0.4$$

Ziegler-Nichols

Controller	K	Ti	Td	Tp
P	$1/a$			$4L$
PI	$0.9/a$	$3L$		$5.7L$
PID	$1.2/a$	$2L$	$L/2$	$3.4L$



Ziegler-Nichols: response



Ziegler-Nichols analysis

- Decay ratio is close to one quarter
- Overshoot is quite large
- Simple and widely used
- Often insufficient → necessary to have more data about process dynamics
- Gives a starting point for fine tuning
- Also frequency response method can be used

Ziegler-Nichols

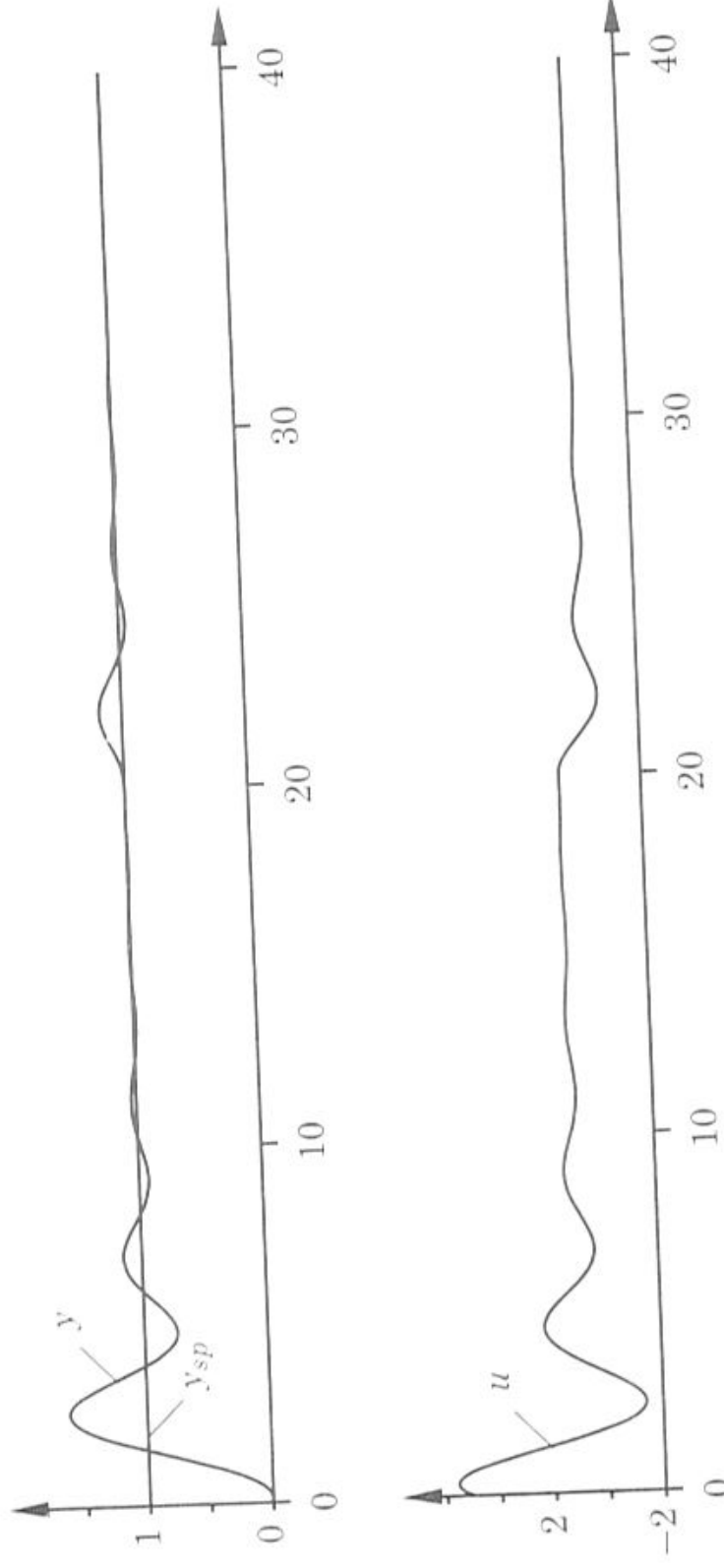


Figure 6.2 Set-point and load disturbance response of a process with transfer function $1/(s + 1)^3$ controlled by a PID controller tuned with the Ziegler-Nichols step response method. The diagrams show set point y_{sp} , process output y , and control signal u .

Chien, Hrones & Reswick Method

- Modification of Ziegler-Nichols method
- Better damped closed-loop step response
- Parameter tables for:
 - quickest response without overshoot
 - quickest response with 20% overshoot
- Different parameters for tuning:
 - setpoint response
 - load disturbance

CHR for setpoint response

- T = time constant

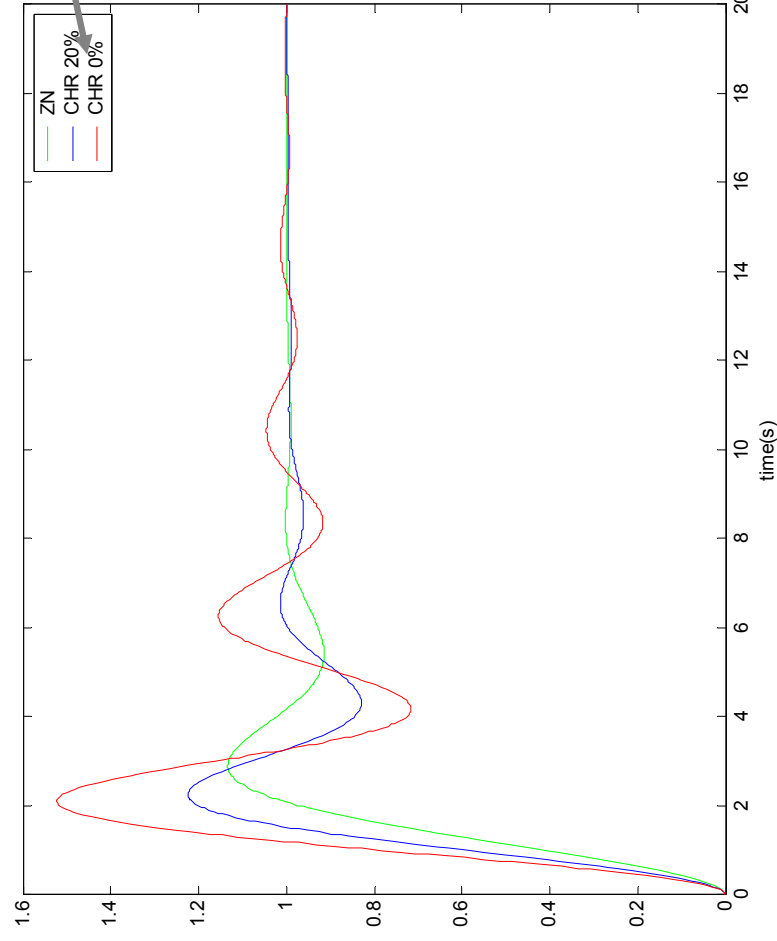
PID Controller parameters;					
Chien, Rhones Reswick setpoint response method		0 %	20 %	overshoot	
Control	K	Ti	Td	K	Ti
					Td
P	0.3/a			0.7/a	
PI	0.35 a	1.2T		0.6 a	T
PID	0.6/a	T	0.5L	0.95/a	1.4T 0.47L

CHR for load disturbance resp.

PID Controller parameters;					
Chien, Rhones Reswick load disturbance response method		0 %	20 % overshoot		
Control	K	Ti	Td	K	Ti
					Td
P	0.3/a			0.7/a	
PI	0.6 a	4L		0.7 a	2.3L
PID	0.95/a	2.4L	0.42L	1.2/a	2L
					0.42L

Z-N vs. CHR

- Overshoot exist but response is much better when CHR parameters are used



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

Analytical tuning methods

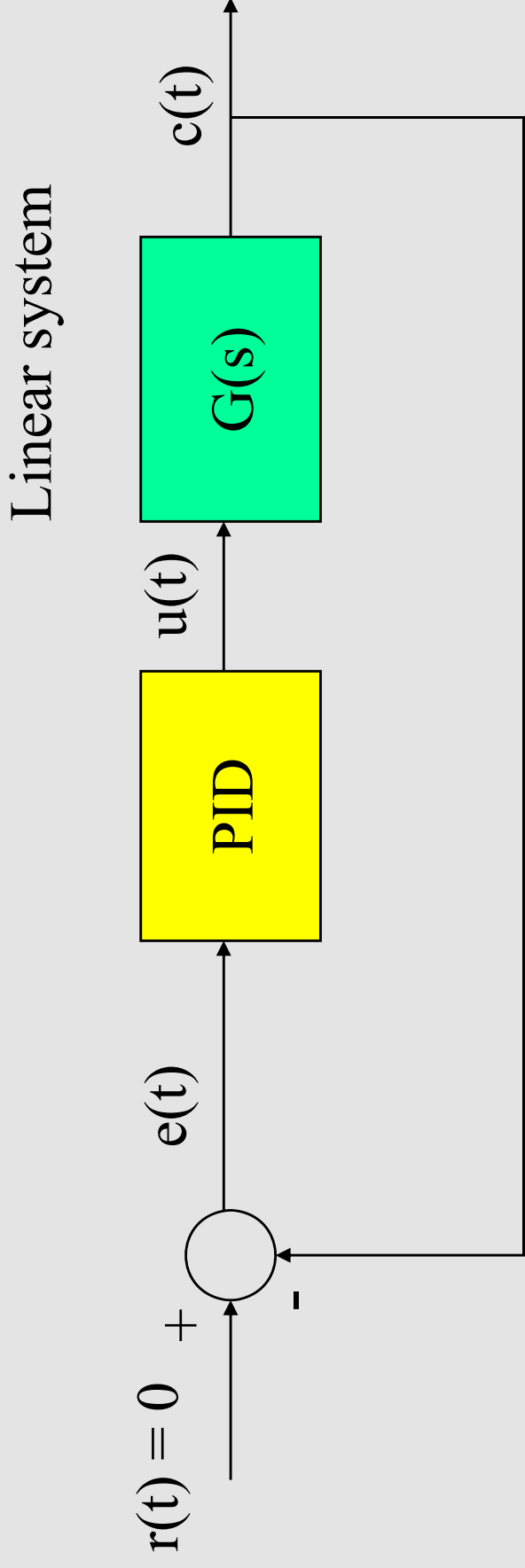
G_p (process transfer function)

G_c (controller transfer function)

$$\rightarrow G_0 = \frac{G_p G_c}{1 + G_p G_c} \quad (\text{closed loop transfer function})$$

$$\rightarrow G_c = \frac{1}{G_p} \cdot \frac{G_0}{1 - G_0} \quad (\text{closed loop transfer function})$$

Closed-loop tuning



1. Set $T_i = \infty$ and $T_d = 0$.
2. Increase K_p until the system oscillates to obtain critical gain K_{cr} and critical period T_{cr} (or frequency)

$$u(t) = K_p \left(e(t) + \frac{1}{T_i} \int_0^t e(\alpha) d\alpha + T_d \frac{de}{dt} \right)$$

Closed-loop tuning

PID Controller parameters;	
Ziegler-Nichols frequency response method	
Control K	Td
P	Ti
0.5Kcr=2.3	
PI	
0.4Kcr=1.9	0.8Tcr=1.8
PID	
0.6Kcr=2.8	0.5Tcr=1.1 0.125Tcr=0.3

$$u(t) = K_p \left(e(t) + \frac{1}{T_i} \int_0^t e(\alpha) d\alpha + T_d \frac{de}{dt} \right)$$

Frequency response method

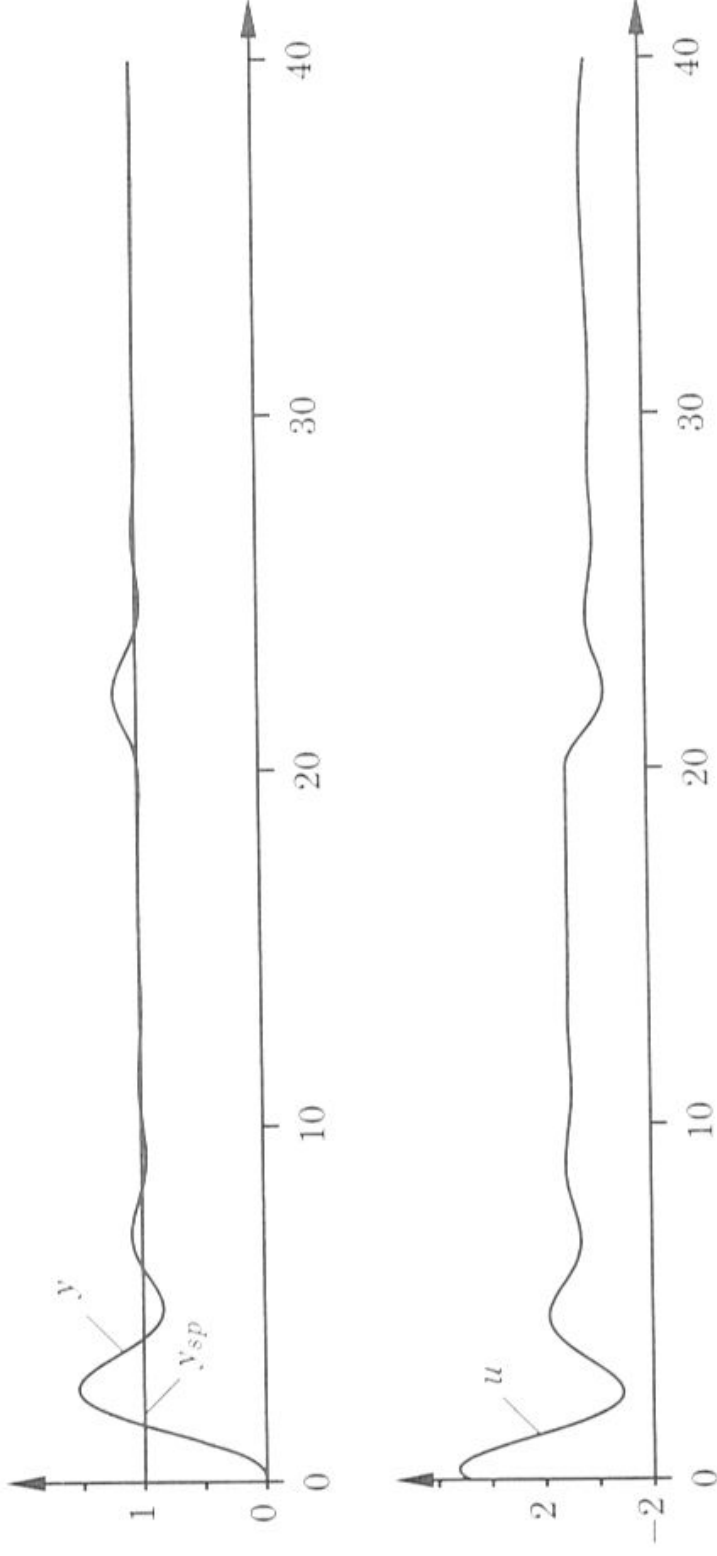


Figure 6.3 Set-point and load disturbance response of a process with the transfer function $1/(s+1)^3$ controlled by a PID controller that is tuned with the Ziegler-Nichols frequency response method. The diagrams show set point y_{sp} , process output y , and control signal u .

Frequency response interpretation

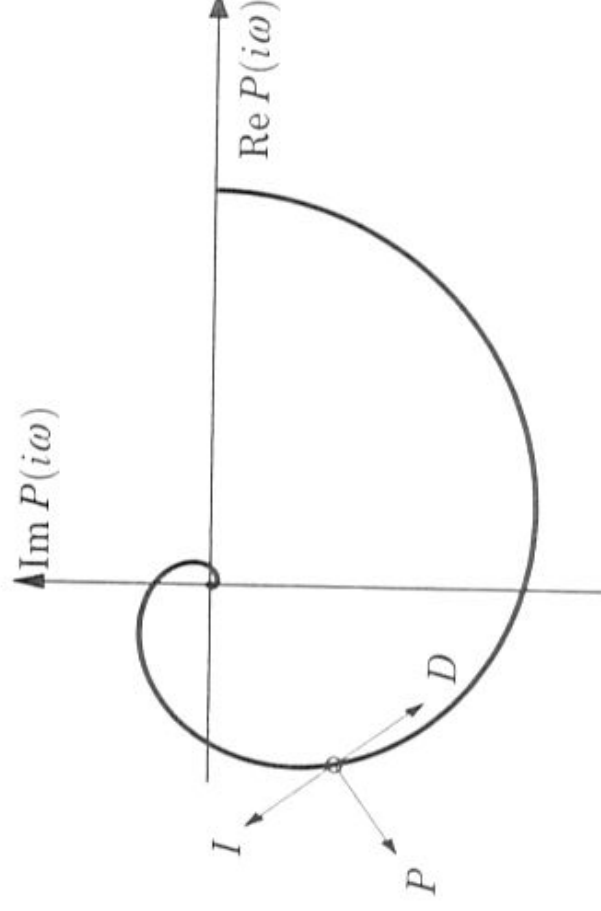


Figure 6.4 Illustrates that a point on the Nyquist curve of the process transfer function may be moved to another position by PID control. The point marked with a circle may be moved in the directions $P(i\omega)$, $-iP(i\omega)$, and $iP(i\omega)$ by changing the proportional, integral, and derivative gain, respectively.

Frequency response – phase margin

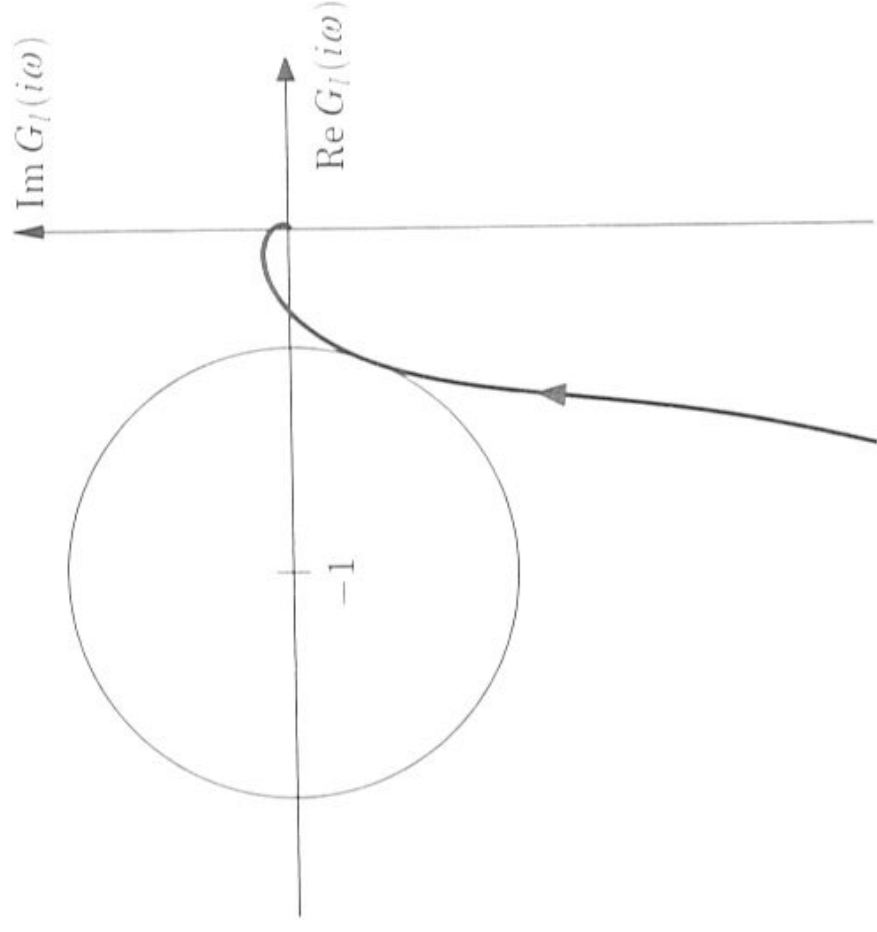


Figure 6.5 Nyquist plot for the loop transfer function G_I for PI control of the process $P(s) = e^{-\lambda s}$. The controller was designed to give the phase margin of 60° .

Comparison

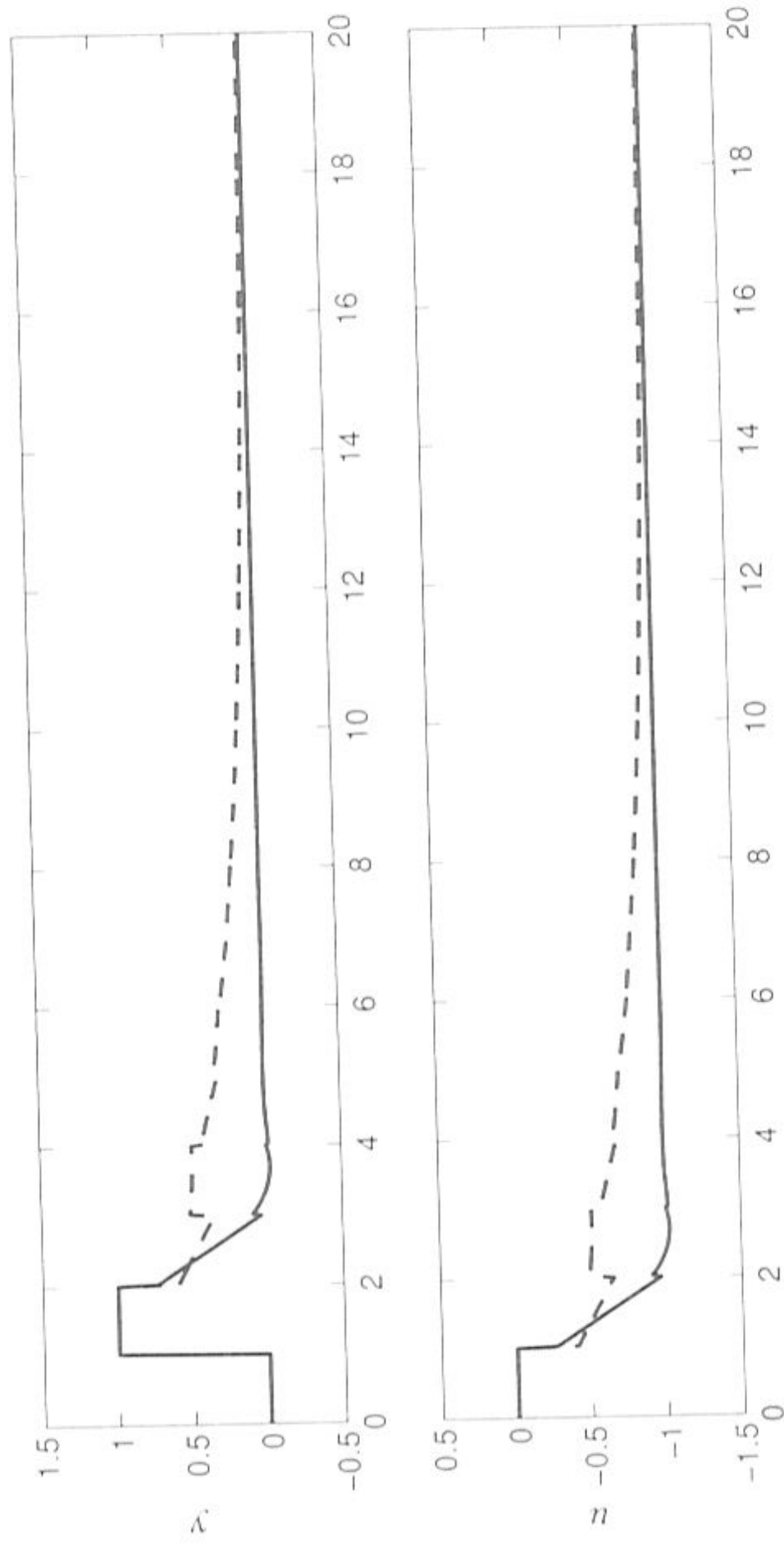
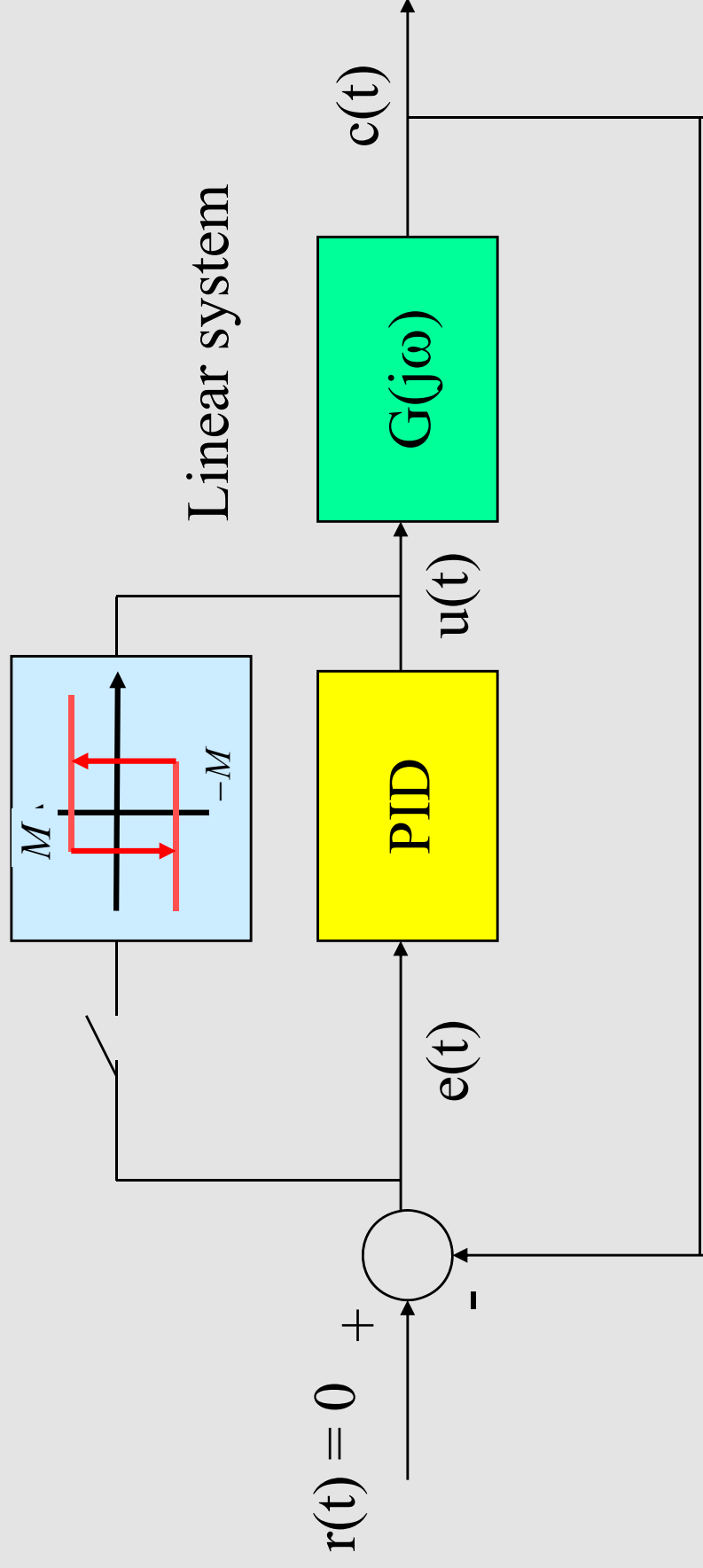


Figure 6.6 Responses to a load disturbance for a process with pure delay ($L = 1$) with PI controllers tuned by Ziegler-Nichols frequency response method (dashed) and a proper method (solid).

Relay tuning

$$u(t) = K_p \left(e(t) + \frac{1}{T_i} \int_0^t e(\alpha) d\alpha + T_d \frac{de}{dt} \right)$$



First relay on, to obtain critical gain K_{cr} and critical period T_{cr}

Rule-based empirical tuning

- Increasing proportional gain decreases stability
- Error decays more rapidly if integration time is decreased
- Decreasing integration time decreases stability
- Increasing derivative time improves stability

Tuning maps

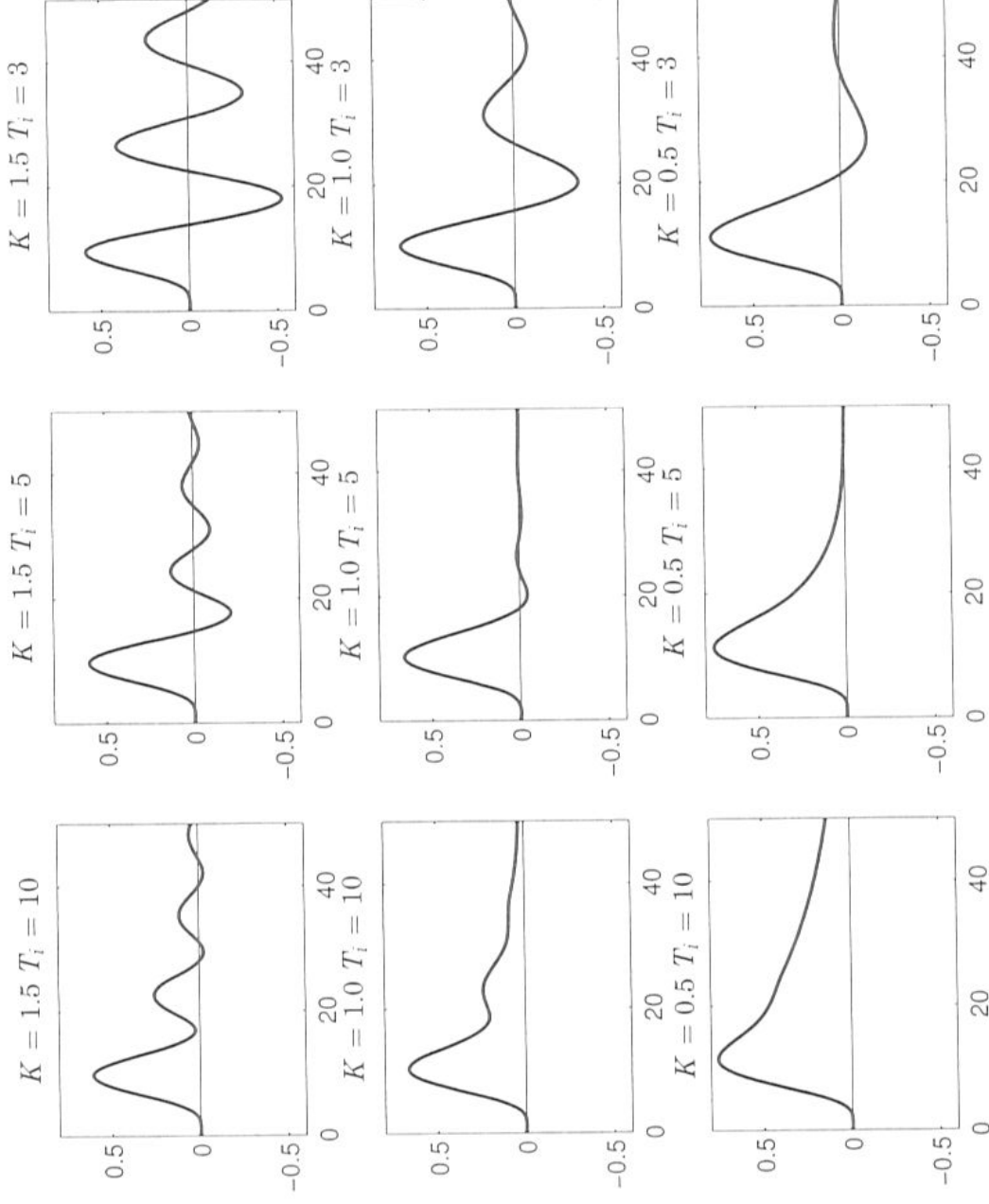


Figure 6.7 Tuning map for PID control of a process with the transfer function $P(s) = (s + 1)^{-8}$. The figure shows the responses to a unit step disturbance at the process input. Parameter T_d has the value 1.9.

Tuning map

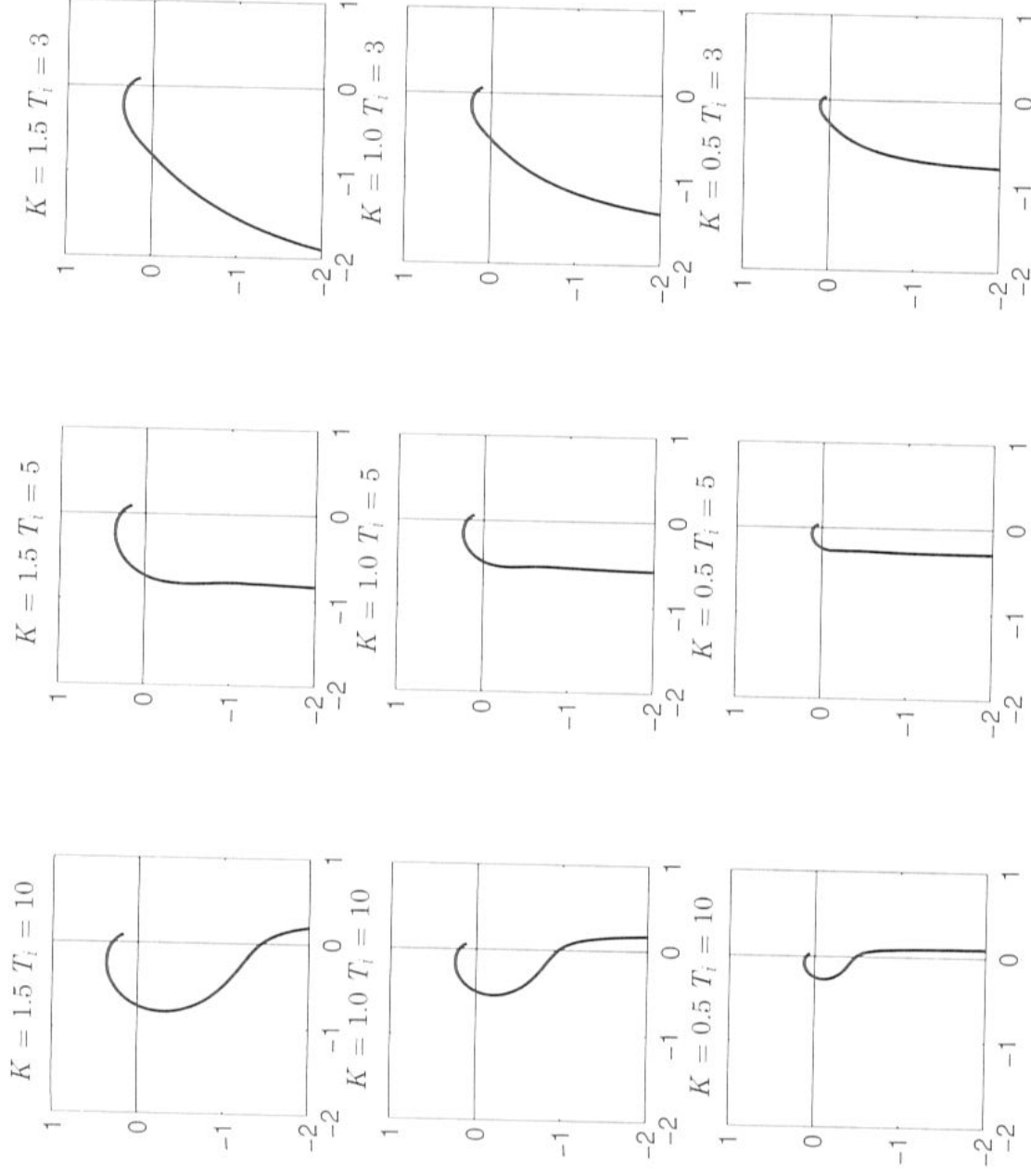


Figure 6.8 Tuning map for PID control of a process with the transfer function $P(s) = (s+1)^{-8}$. The figure shows the Nyquist plots of the loop transfer functions. Parameter T_d has the value 1.9.

Counterintuitive

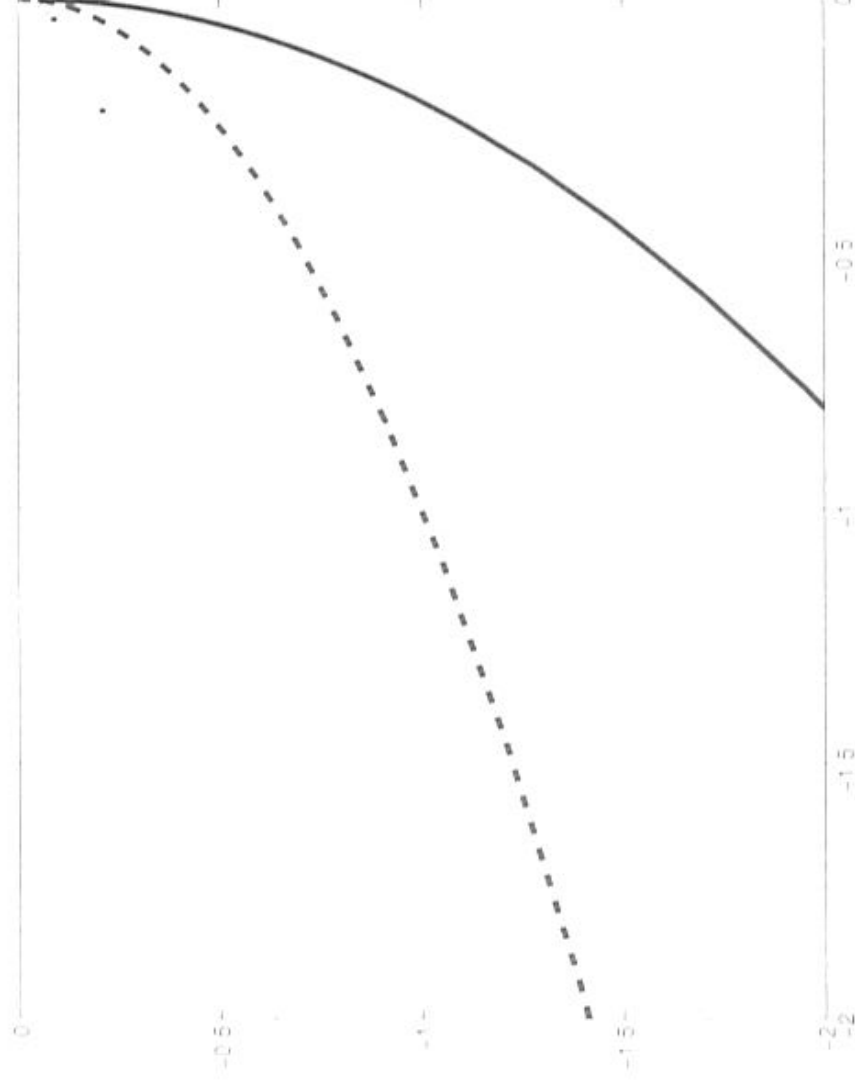


Figure 6.9 Nyquist curves for the loop transfer functions for an integrator with PI control. Integration time T_i is constant, and the gain has the values $K = 0.2$ (dotted), 1 (dashed), and 5 (solid). Notice the counterintuitive behavior that phase margin increases with increasing controller gain.

Pole Placement

- Process must be modelled
- Define the desired closed-loop poles
- PID-controller gives desired closed-loop poles
- Process parameters max. 3 → poles can be set freely using PID-controller

Pole Placement: example

- Process:
$$G_p = \frac{K_p}{(1+sT)}$$
 - Controller (PI):
$$G_c = K\left(1 + \frac{1}{sT_1}\right)$$
 - Closed loop system:
$$G(s) = \frac{G_p G_c}{1 + G_p G_c}$$
 - Characteristic eq:
$$1 + G_c G_p = 0$$
-
- $$s^2 + s \frac{1 + K_p K}{T} + \frac{K_p K}{TT_i} = 0$$

Pole placement: example

- Desired close-loop poles characterized by relative damping (ζ) and frequency (ω):

$$s^2 + 2\xi\omega_0s + \omega_0^2 = 0$$

- By making coefficient equal in characteristics equations and solving controller parameters gives:

$$K = \frac{2\xi\omega_0T - 1}{K_p}$$

$$T_i = \frac{2\xi\omega_0T - 1}{\omega_0^2T}$$

Gang of six

$$\frac{PC}{1+PC} = \frac{(2\zeta\omega_0 - 1/T)s + \omega_0^2}{s^2 + 2\zeta\omega_0s + \omega_0^2} \quad \frac{C}{1+PC} = \frac{K(s + 1/T_i)(s + 1/T)}{s^2 + 2\zeta\omega_0s + \omega_0^2}$$

$$\frac{P}{1+PC} = \frac{K_p s/T}{s^2 + 2\zeta\omega_0s + \omega_0^2} \quad \frac{1}{1+PC} = \frac{s(s + 1/T)}{s^2 + 2\zeta\omega_0s + \omega_0^2}$$

$$\frac{PC_{ff}}{1+PC} = \frac{b(2\zeta\omega_0 - 1/T)s + \omega_0^2}{s^2 + 2\zeta\omega_0s + \omega_0^2} \quad \frac{C_{ff}}{1+PC} = \frac{K(bs + 1/T_i)(s + 1/T)}{s^2 + 2\zeta\omega_0s + \omega_0^2}.$$

- Largest value of transfer function from load disturbance to process output

$$\max_{\omega} |G_{xd}(i\omega)| = \max_{\omega} \left| \frac{P(i\omega)}{1 + P(i\omega)C(i\omega)} \right| = \frac{K_p}{\omega_0 T \min(1, \zeta)}.$$

Choosing ω_0

$$\frac{1}{2\zeta} \leq \omega_0 T < \min\left(\frac{0.25}{T_e}, \frac{1 + K_p K_{max}}{2\zeta}\right).$$

- More sensitive when $\omega_0 T$ is small

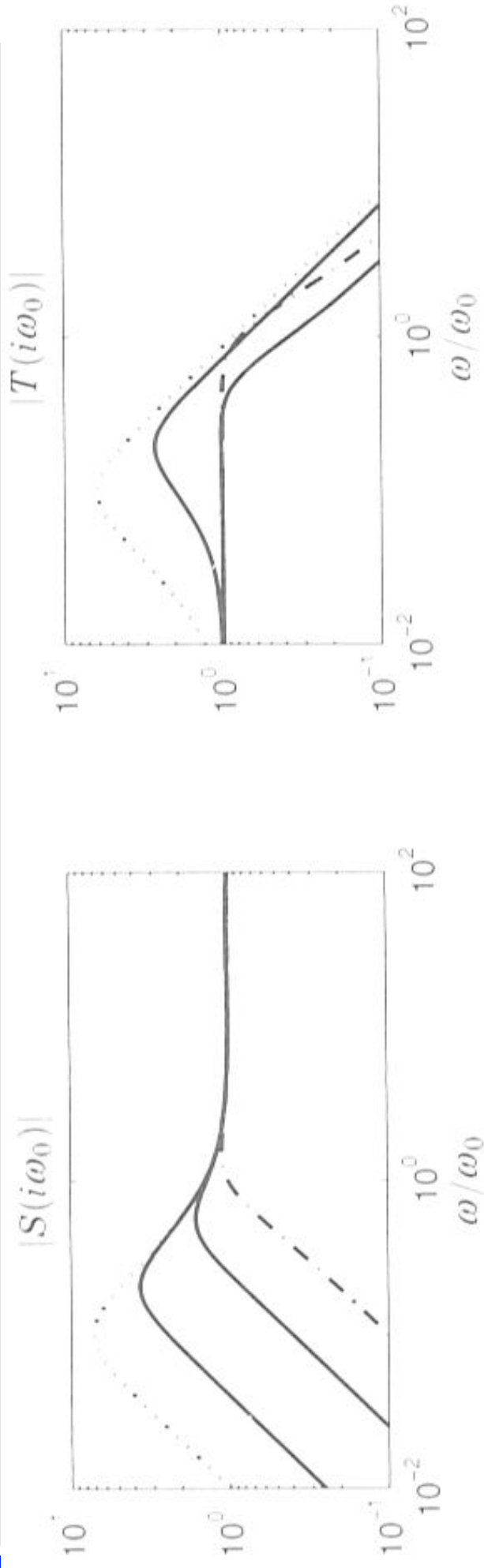


Figure 6.10 Gain curves of the sensitivity functions for $\zeta = 0.7$ and $\omega_0 T = 0.1, 0.2, 0.5,$ and 1. The dotted curve corresponds to $\omega_0 L = 0.1$ and the dash-dotted curve to $\omega_0 L = 1$.

Approximate models

- First order approx

$$P(s) = \frac{1}{1 + 1.26s}$$

$$P(s) = \frac{1}{(1 + s)(1 + 0.2s)(1 + 0.05s)(1 + 0.01s)}$$

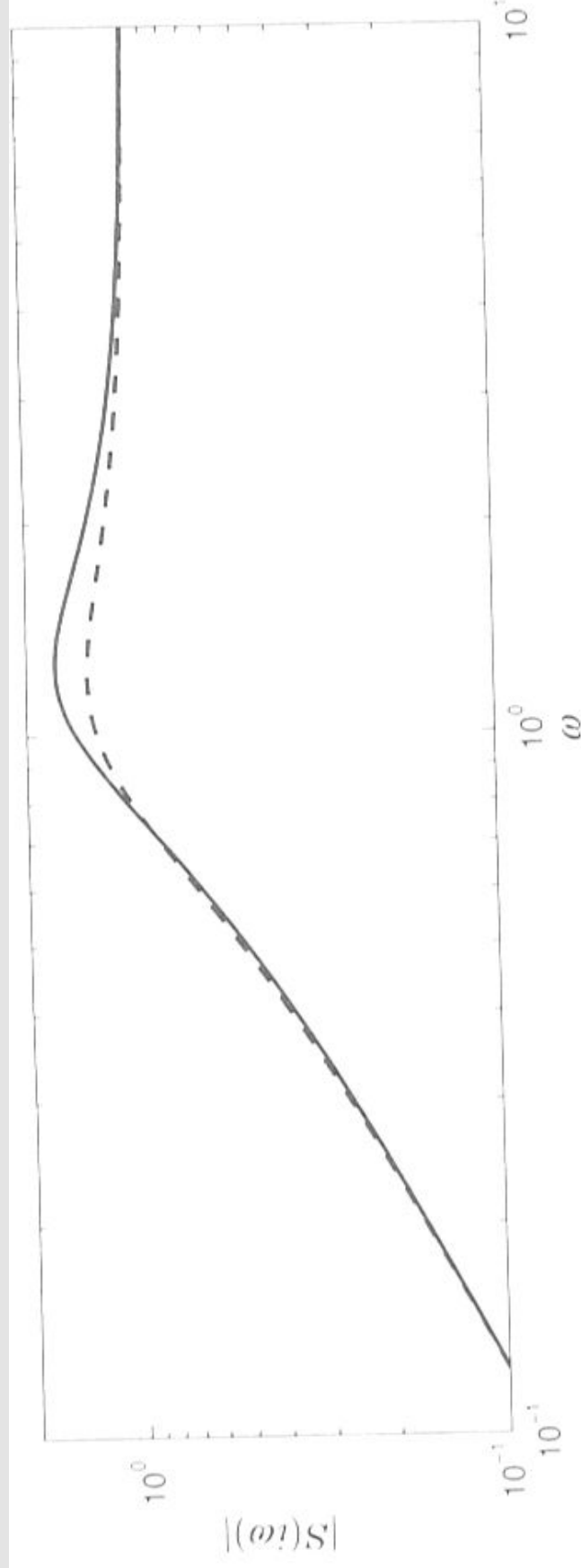


Figure 6.11 Sensitivity functions for the approximate system (dashed) and the true system in Example 6.11.

Approximate models

$$P(s) = \frac{1}{(1+s)(1+0.26s)}$$

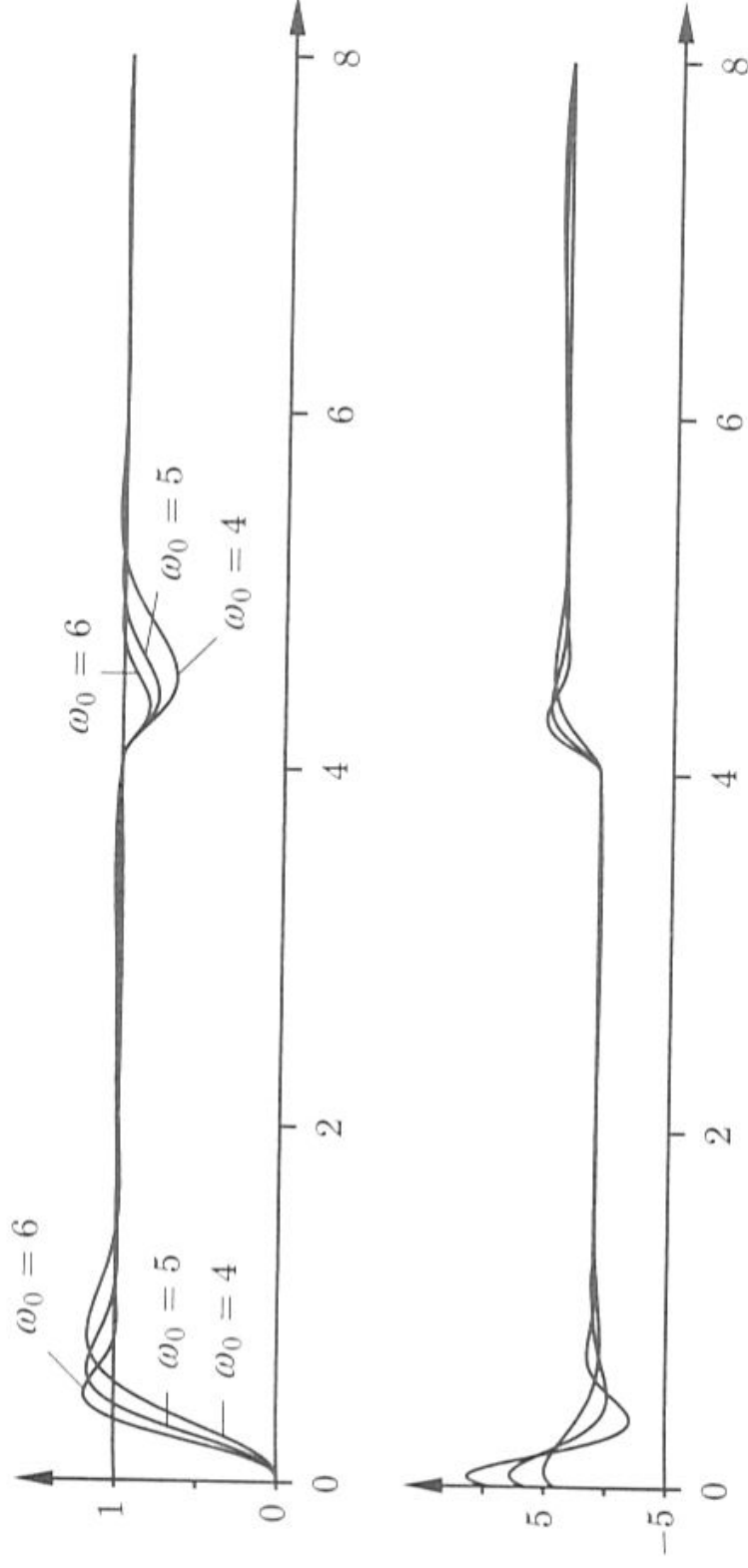


Figure 6.12 Set-point and load disturbance responses of the process with two poles controlled by a PID controller tuned according to Example 6.12. The responses for $\omega_0 = 4, 5,$ and 6 are shown. The upper diagram shows set point $y_{sp} = 1$ and process output y , and the lower diagram shows control signal u .

Optimization in Matlab

- Minimize the cost function by simulating the controlled model
- Use *fminsearch* function to find optimized parameters for the PID controller
- Choose appropriate cost function carefully
- Don't forget local minima

Analytical tuning methods

- λ – Tuning
- processes with long dead time
- desired closed-loop tf:

$$G_0 = \frac{e^{-sL}}{1 + s\lambda T}$$

- controller transfer function:

$$G_c = \frac{1 + sT}{K_p (1 + s\lambda T - e^{-sL})}$$

- $\lambda < 1$ faster response (smaller time constant)

Optimization in Matlab

1. Guess initial values for K_p^0, K_i^0, K_d^0 , Set $n = 0$
2. Solve $\mathbf{y}(t)$ using simulation (SIMULINK) and also the value of cost function $J(K_p^n, K_i^n, K_d^n)$
3. Use optimization algorithm (FMINSEARCH or FMINUNC) to update the control parameters:

$$K_p^{n+1} = K_p^n + \text{correction} \text{ (depends on optimization algorithm)}$$

$$K_i^{n+1} = K_i^n + \text{correction} \text{ (depends on optimization algorithm)}$$

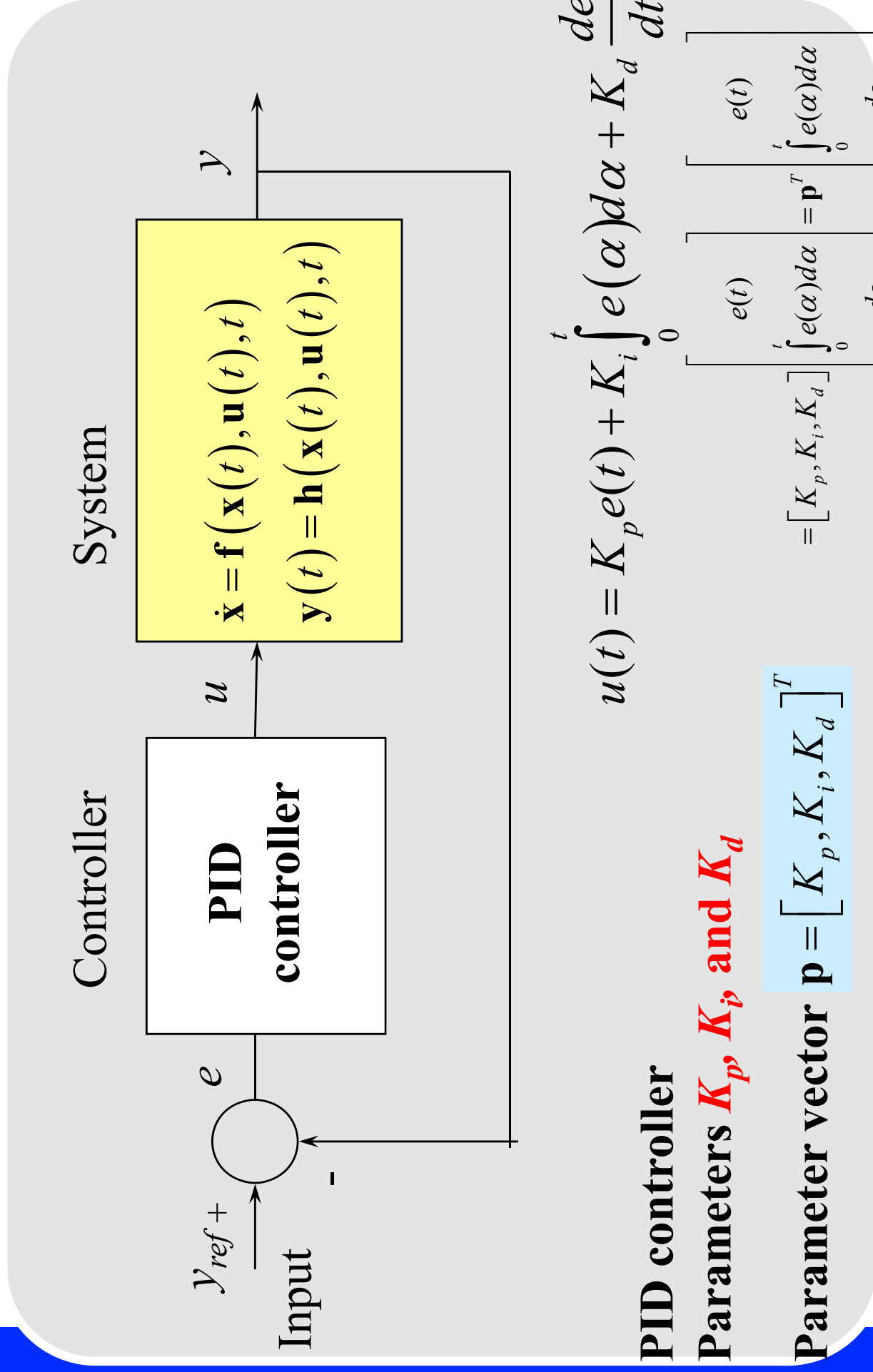
$$K_d^{n+1} = K_d^n + \text{correction} \text{ (depends on optimization algorithm)}$$

4. Check if minimum is reached

$$|J^{n+1} - J^n| < \text{tolerance}$$

If not, set $n = n+1$ and go back to 2.

Optimization in Matlab



Cost criteria

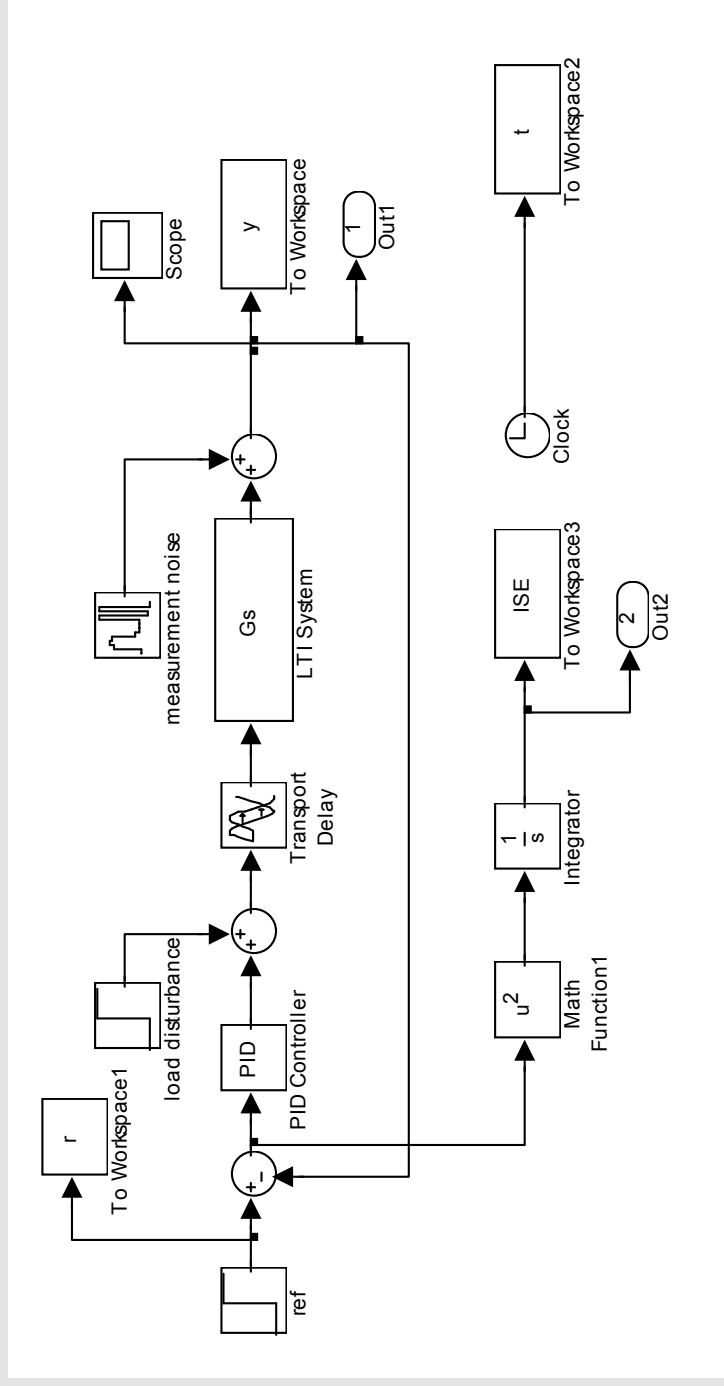
$$ITAE = \int_0^{\infty} t |e(t)| dt$$

$$ITE = \int_0^{\infty} te(t) dt$$

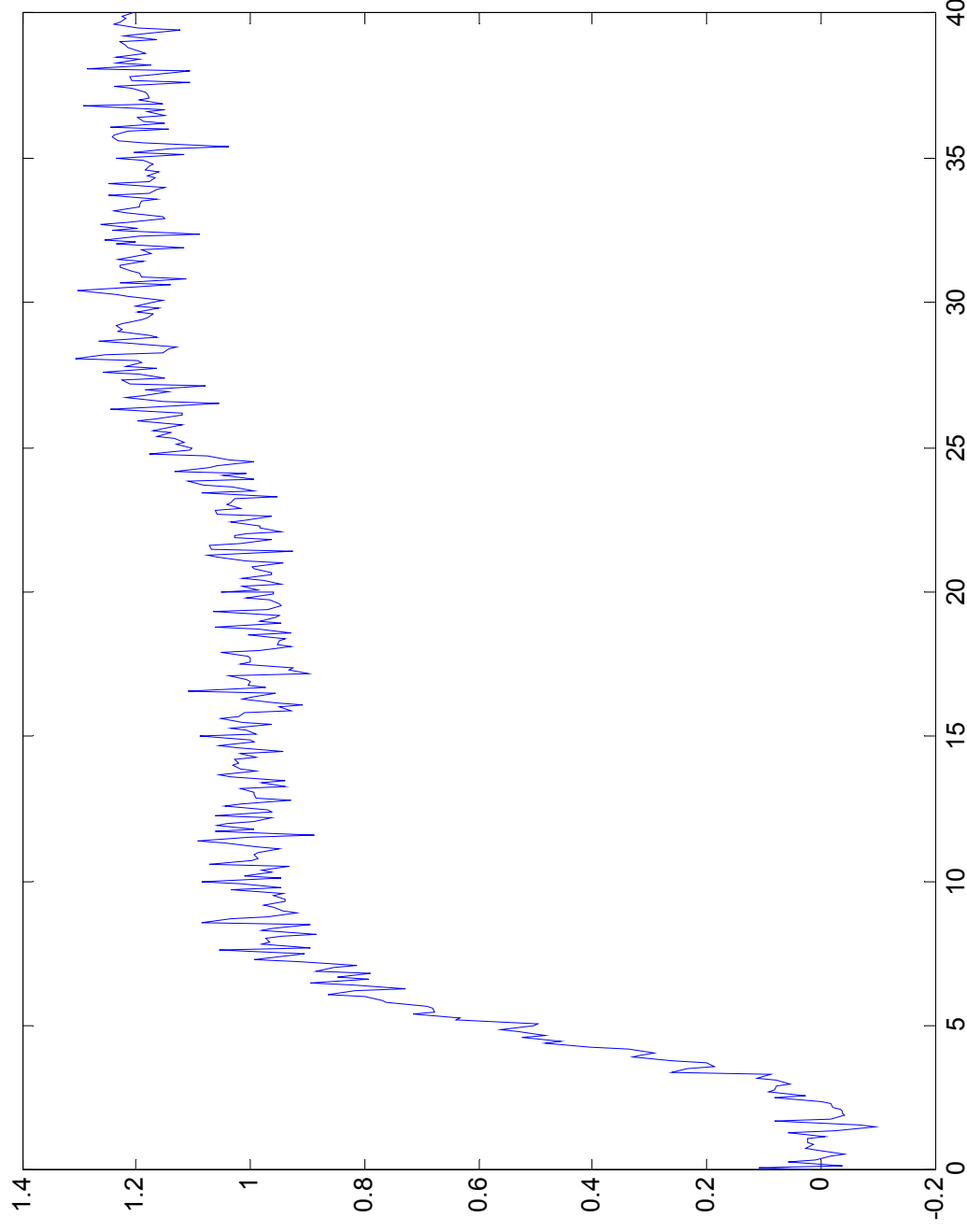
$$ITSE = \int_0^{\infty} te(t)^2 dt$$

$$ISTE = \int_0^{\infty} t^2 e(t)^2 dt$$

Example model



Open loop response

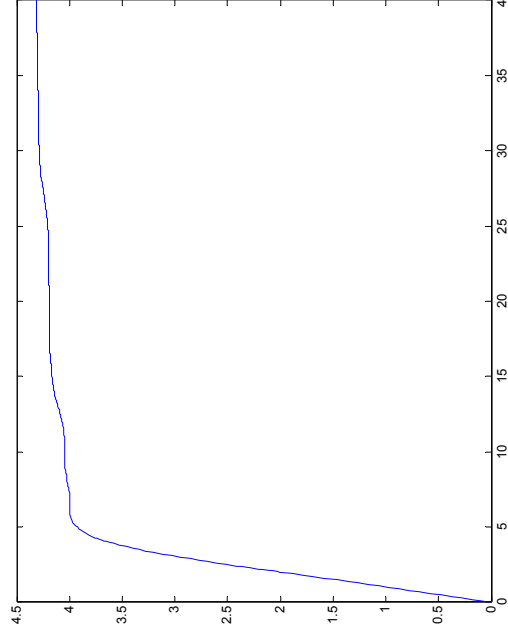
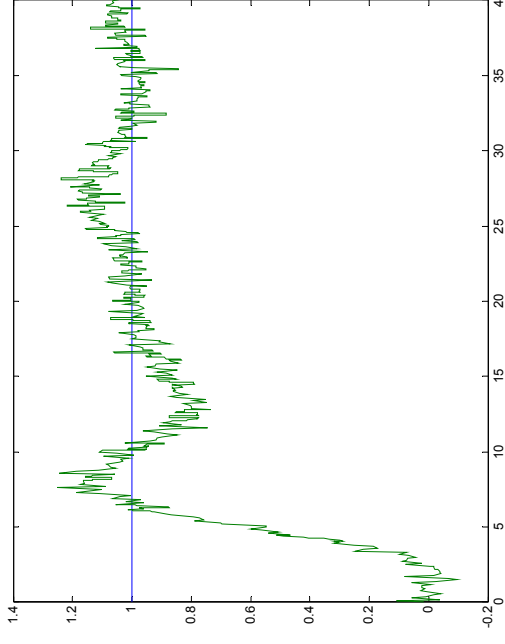


Cost function in Matlab

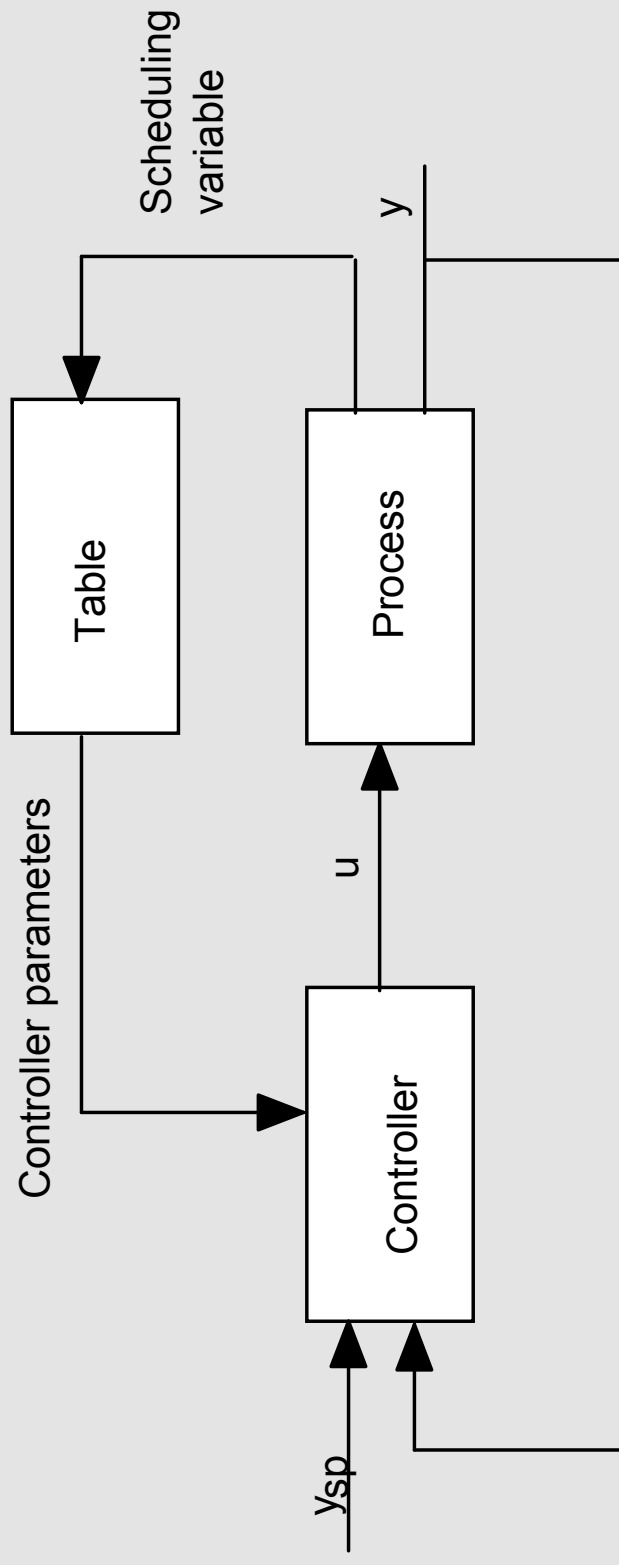
- %evaluate cost function from the simulink model
- function J = cost(inputs)
- %parameters
- Tsim=40; %simulation time
- model= 'pidmodel'; %simulink model
- %assign new parameters to workspace for simulation
- assignin('base','Kp',inputs(1));
- assignin('base','Ti',inputs(2));
- %simulate
- [t,x,y]=sim(model,Tsim);
- %y(2) corresponds with cost
- J=max(y(:,2));

IN MATLAB:

- fminsearch('cost',[1 10])



Gain scheduling



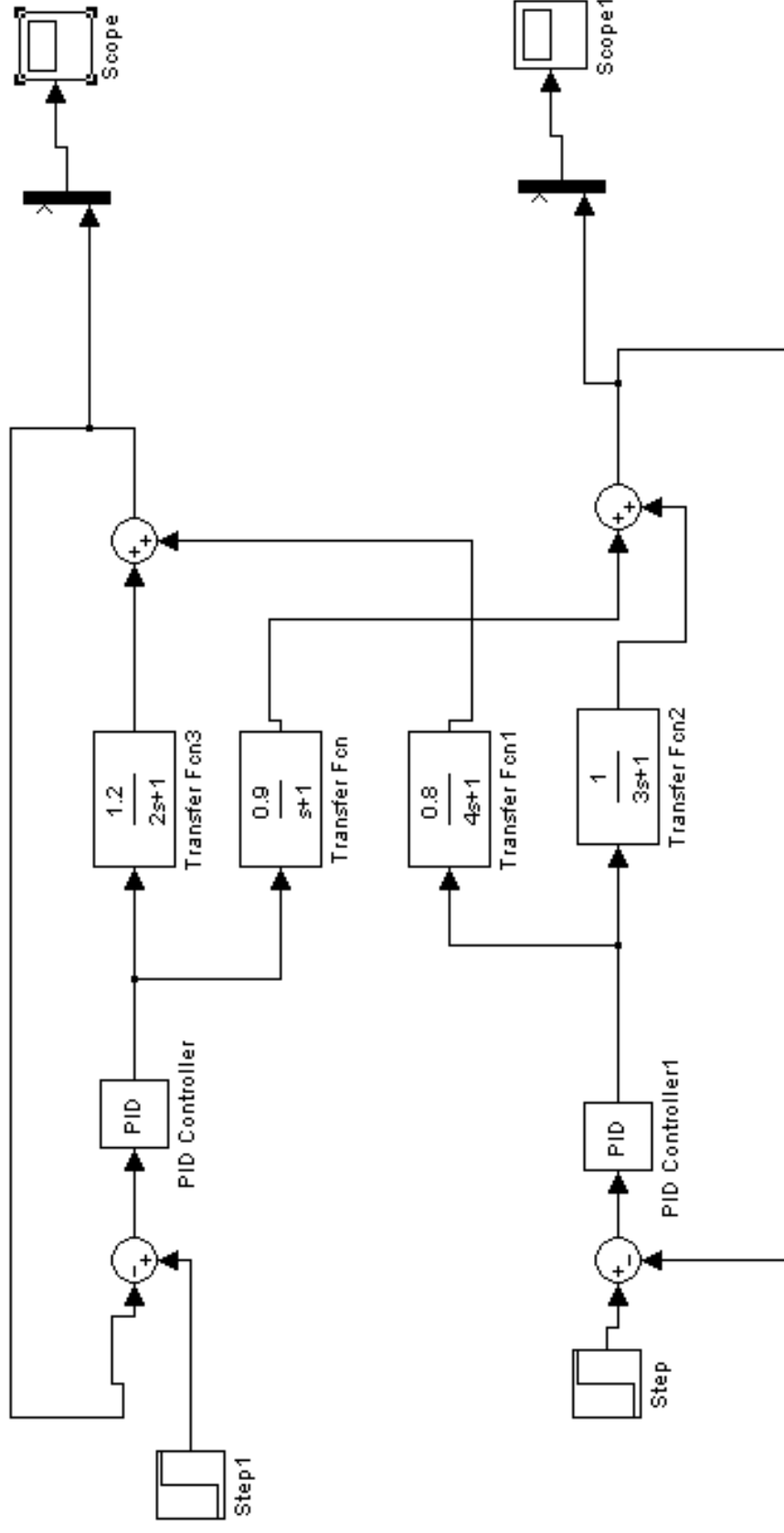
Rule-Based Methods

	Speed	Stability
K increases	increases	reduces
T_i increases	reduces	increases
T_d increases	increases	increases

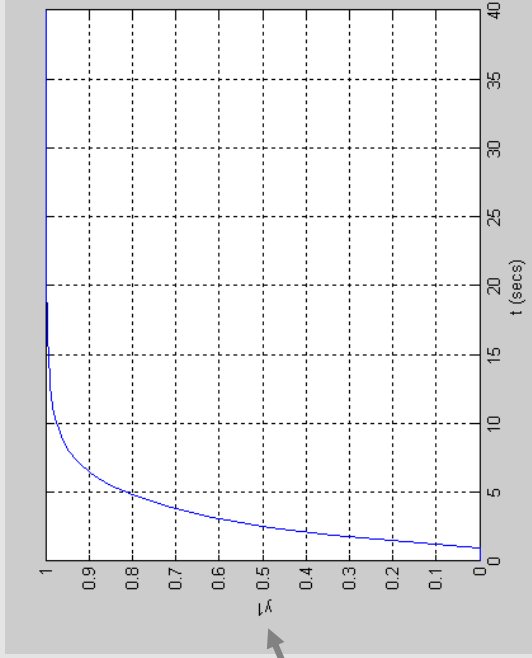
MIMO PID control- Decentralized

- Multiple Input Multiple Output (MIMO)

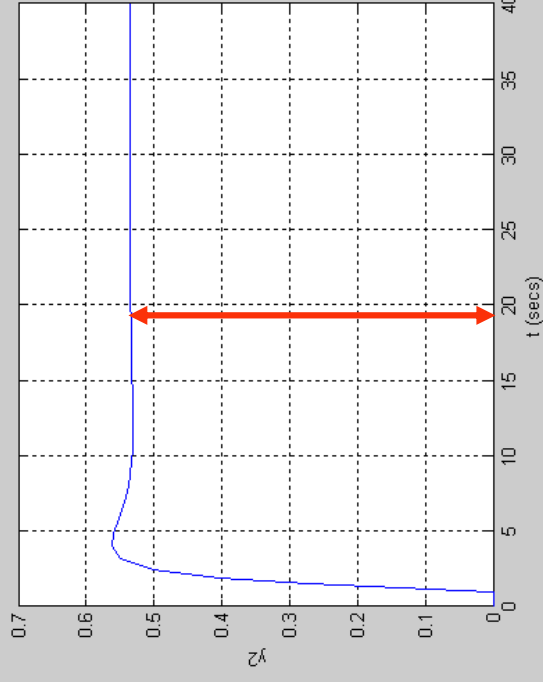
MIMO PID control- Decentralized



MIMO PID control- Decentralized

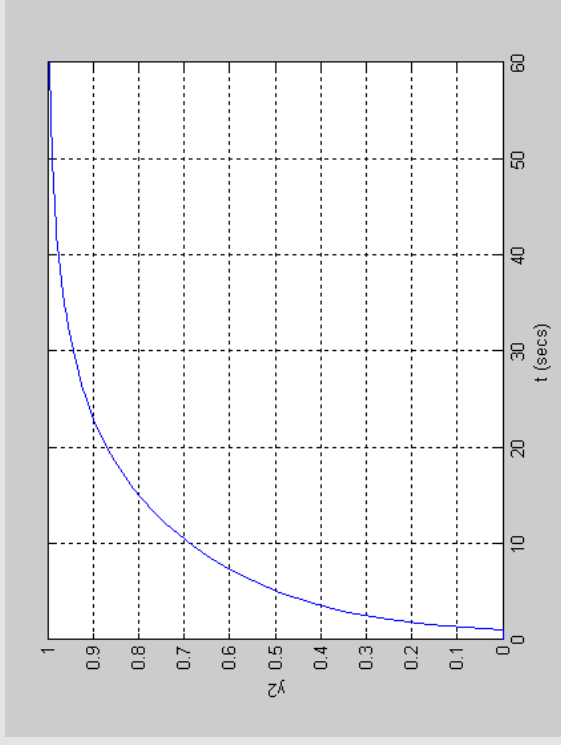


Lieslehto
 $K=0.83$, $T_i=2$, $I=0.5$
Unit step in ref1



Amount of
 interaction

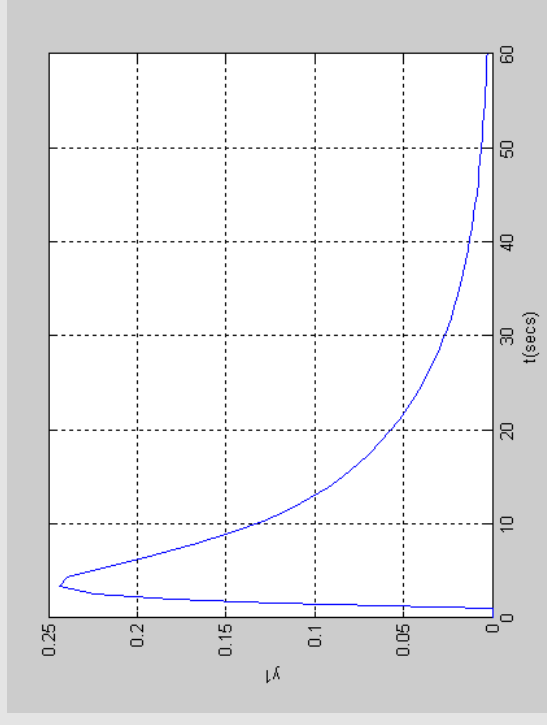
MIMO PID control- Decentralized



Lieslehto

$K=1.5$, $T_i=2$, $I=0.5$

Unit step in ref2



**Amount of
interaction**

MIMO PID control - Centralized

State-space equation

$$\dot{\mathbf{x}} = \mathbf{A}\mathbf{x} + \mathbf{B}\mathbf{u}$$

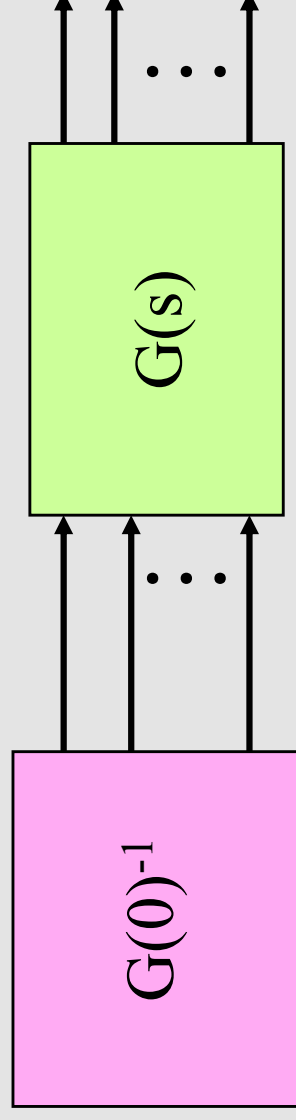
$$\mathbf{y} = \mathbf{C}\mathbf{x}$$

Transfer function

$$\mathbf{G}(s) = \mathbf{C}(s\mathbf{I} - \mathbf{A})^{-1}\mathbf{B}$$

MIMO PID control - Centralized

Decoupling at steady state



MIMO PID control – Integral gain

$$G(0) = \begin{bmatrix} 1.2 & 0.9 \\ 0.8 & 1 \end{bmatrix}$$

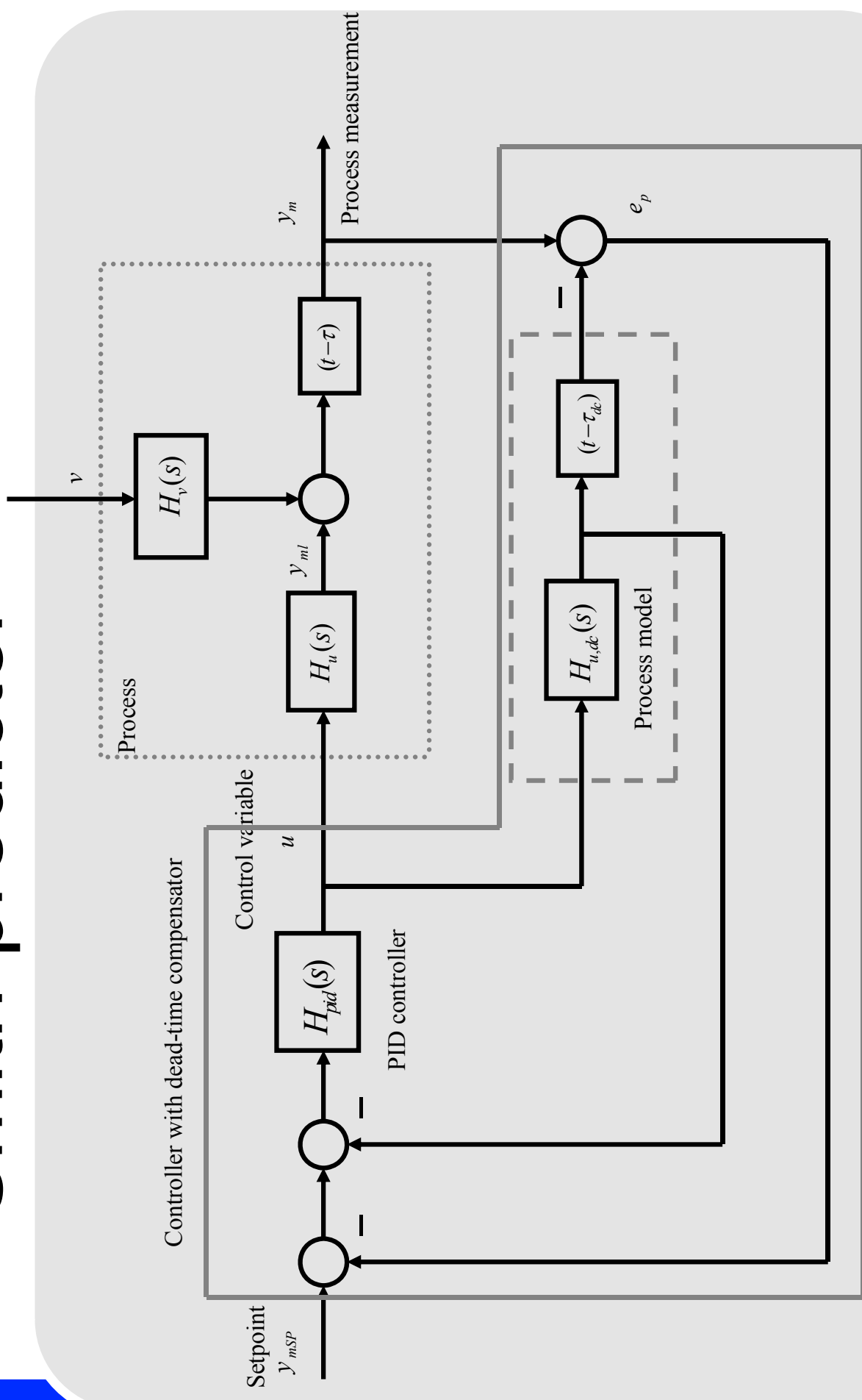
$$K_i = \varepsilon (G(0))^{-1} = \varepsilon \begin{bmatrix} 1.2 & 0.9 \\ 0.8 & 1 \end{bmatrix}^{-1} = \varepsilon \begin{bmatrix} 2.1 & -1.9 \\ -1.7 & 2.5 \end{bmatrix}$$

MIMO PID control – Proportional gain

$$\lim_{s \rightarrow 0} sG(s) = \lim_{s \rightarrow 0} s \begin{bmatrix} \frac{1.2}{2s+1} & \frac{0.9}{s+1} \\ \frac{0.8}{4s+1} & \frac{1}{3s+1} \end{bmatrix} = \begin{bmatrix} 0.6 & 0.9 \\ 0.2 & 0.3 \end{bmatrix}$$

$$K_p = \delta \left(\lim_{s \rightarrow 0} sG(s) \right)^{-1} = \delta \begin{bmatrix} 0.6 & 0.9 \\ 0.2 & 0.33 \end{bmatrix}^{-1} = \delta \begin{bmatrix} 18.3 & -50 \\ -11 & 33 \end{bmatrix}$$

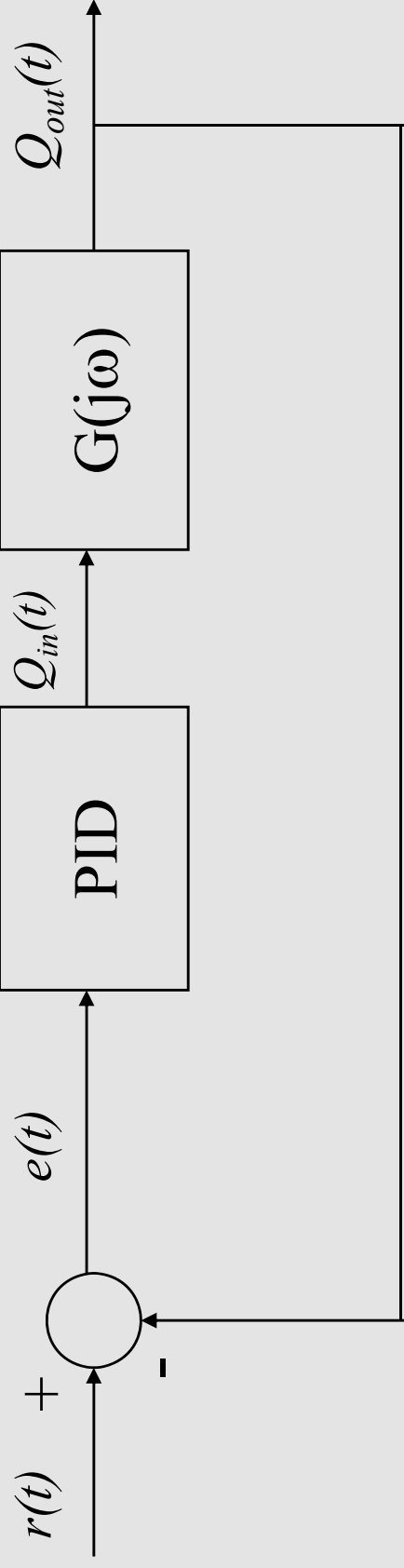
Smith-predictor



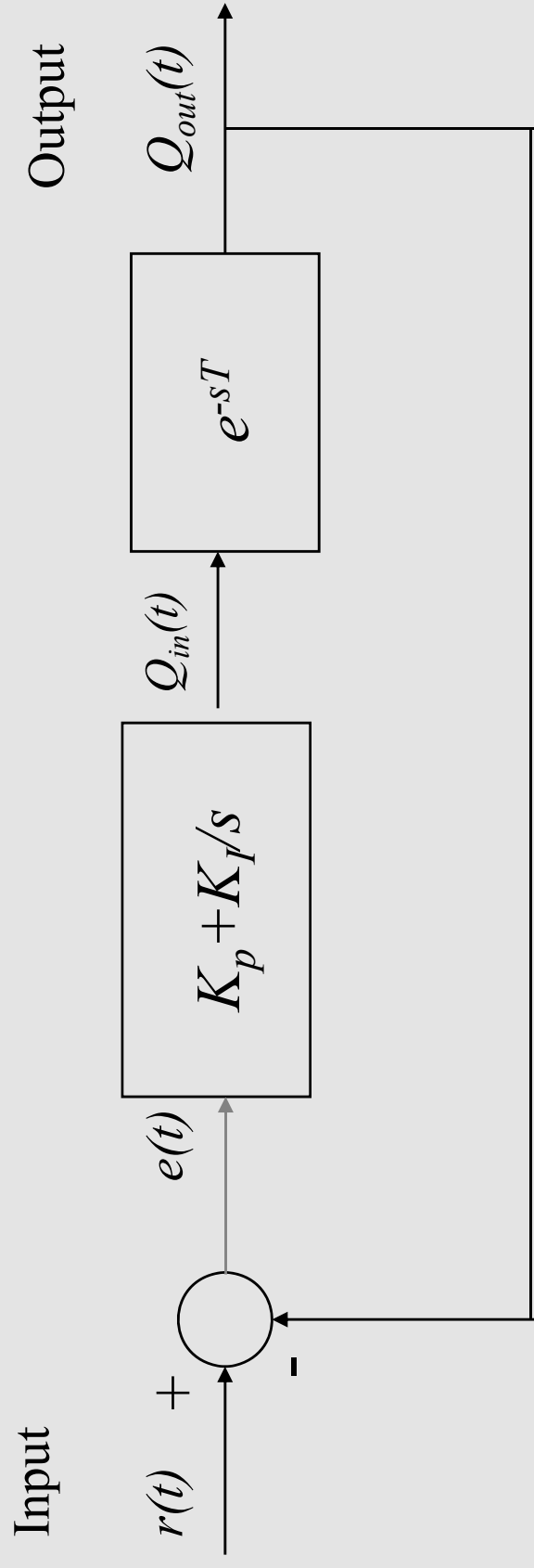
References

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Tuning (3rd ed) (2005)

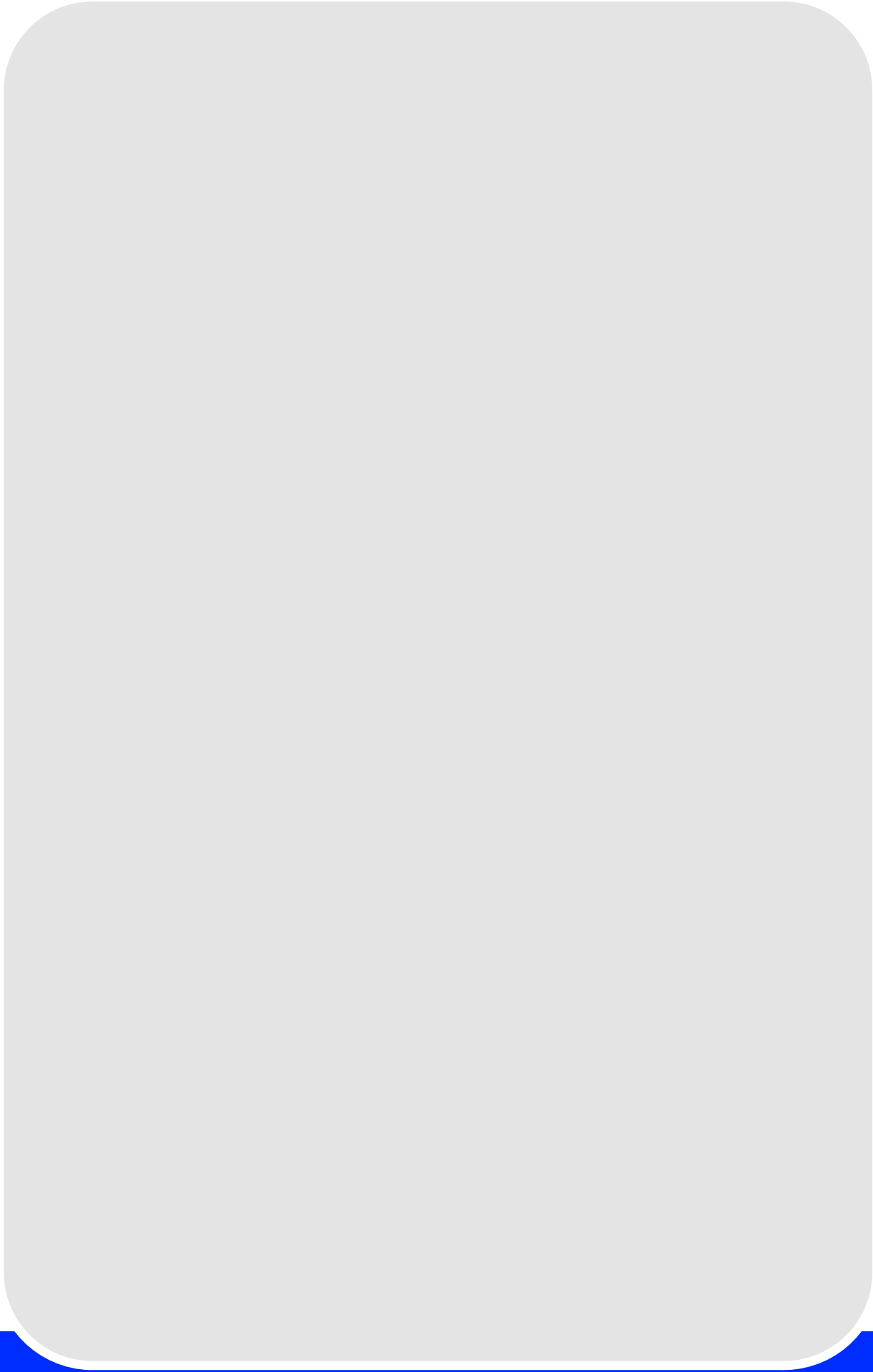
Linear system

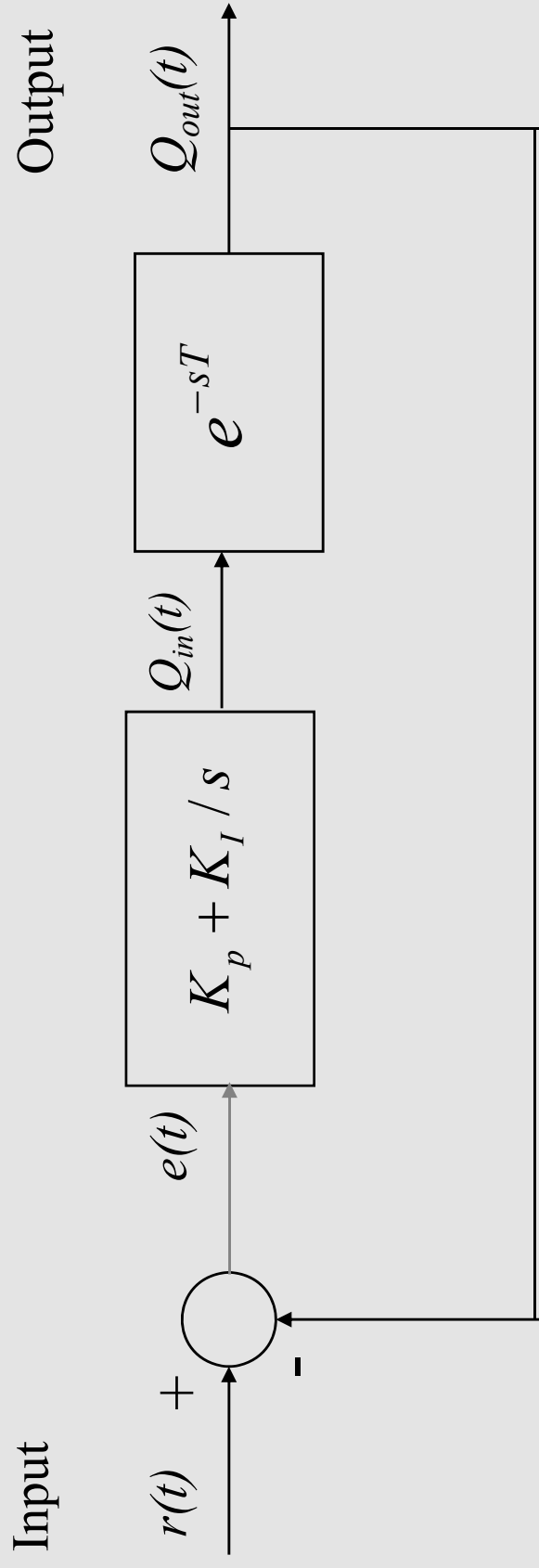


$$K_p + K_I / s$$



28.1.2009





Gain scheduling

28.1.2009

