

Control concepts Dynamic Systems

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Goals

As a student you should be able to:

- Explain the basic concepts of control
- Understand different control structures and how they can be used to solve a control problem
- Understand practical limitations and know how to account for them

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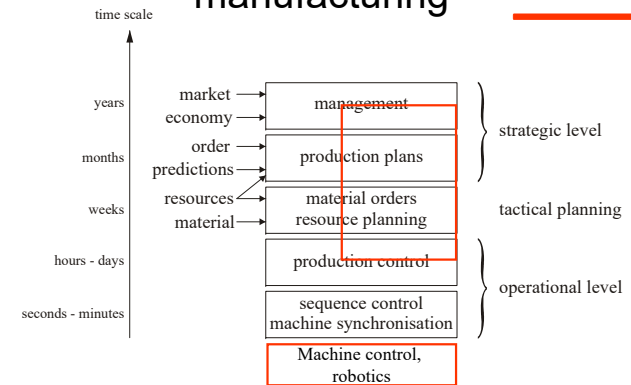
Control of continuous systems

Examples

- In process industry
 - Reactions, pressure, flow rates
- In vehicle control
 - Aircrafts, space, missiles, cars
- In manufacturing
 - Robotics, machine control

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Continuous in discrete manufacturing



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Components of control

- Process / machine incl. controlled variable/state
- Sensor
- **Controller**
- Actuator
- Communication
- Data handling (SCADA) and storage (data bases)

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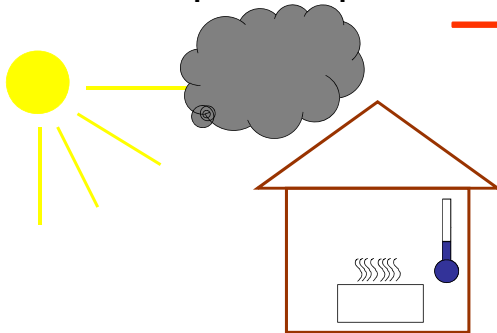
Important controller concepts

- Open loop
- Closed loop
- Feed-forward
- Feedback
- Stationary error
- Stability
- Time delays

Fundamental concepts of control that a control or automation engineer must know!!!

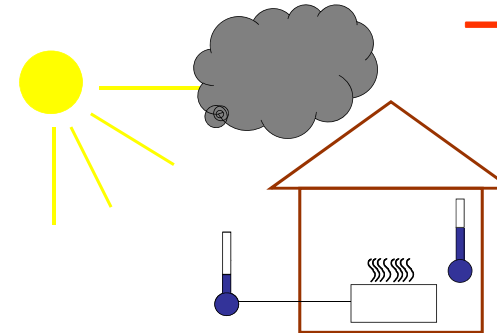
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Open loop



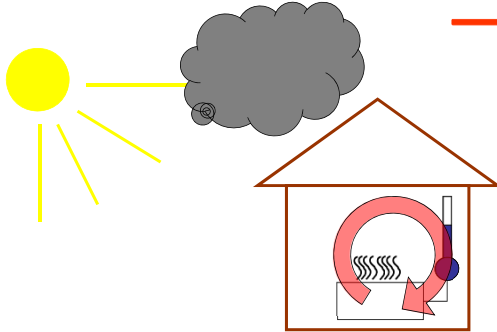
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Feed-forward



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Closed loop - feedback



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Content - some control ideas

- Simple controllers – PID
- Cascade control
- State feedback
- Feed forward
- More advanced control

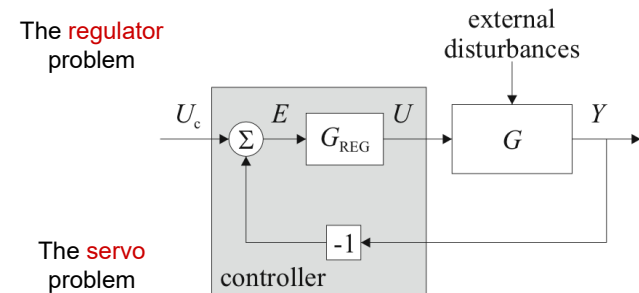
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Feedback

- Concept of feedback used everywhere
 - Society
 - Economy – economic data is fed to ECB, which controls the interest rate in order to keep the inflation at a certain level
 - Nature
 - Human body – the nerve cells senses the temperature and the brain controls the muscles to restrict the skin capillaries
- Feedback – corrective action

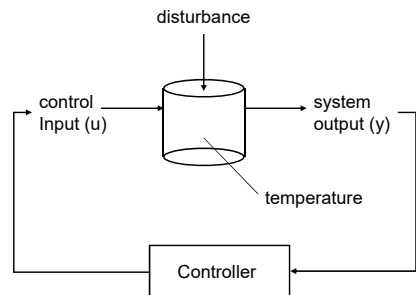
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Feedback controller



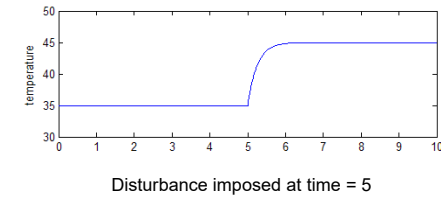
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Test model – disturbance rejection



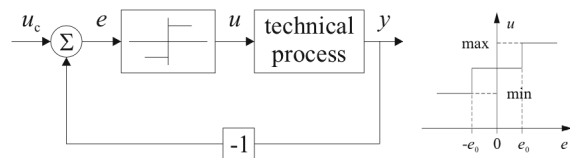
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Without control



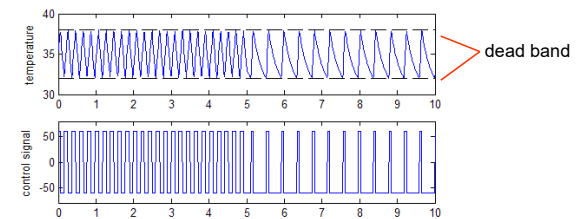
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On-off controller



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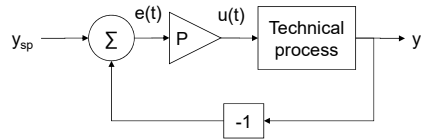
On-off control



Used much in practice
e.g. level control
Very hard on equipment!

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Proportional (P) control

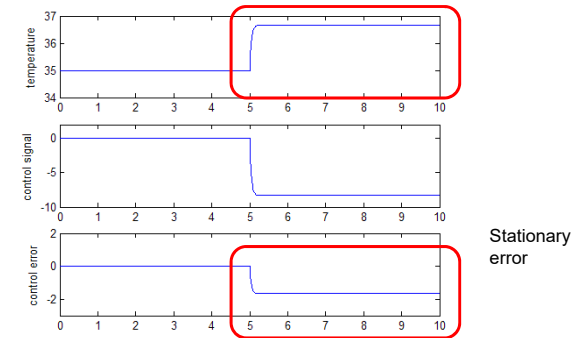


$$u(t) = P \cdot e(t)$$

$$e(t) = y_{sp} - y$$

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P-control



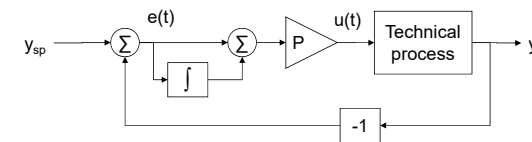
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Stationary error

- When $e \Rightarrow 0$, $u \Rightarrow 0$
- One option: increase gain (P)
 - May lead to instability
- Better option: introduce integral action

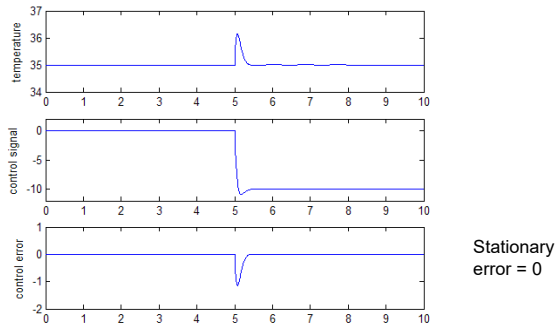
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Integral action



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PI-control



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The PID controller

$$u = u_0 + K_P \left(e + \frac{1}{T_i} \int e dt + T_D \frac{de}{dt} \right)$$

Control signal

Off-set

Proportional gain

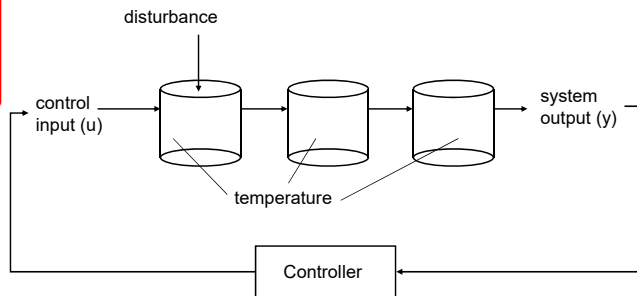
Error $y_{sp} - y$

Integral time

Derivative time

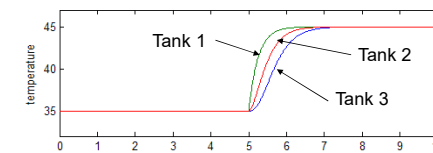
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Test model – 3 tanks in series



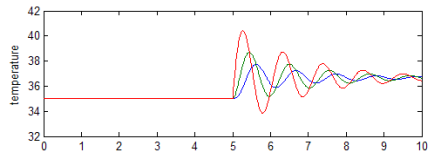
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Tanks in series – no control

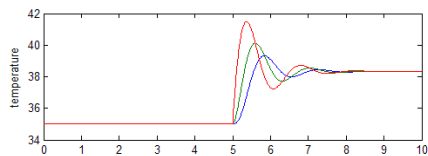


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P-control



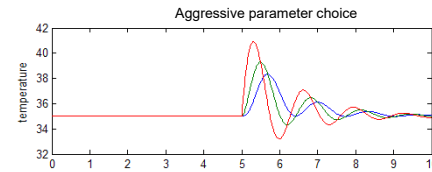
Same gain
as in 1-tank



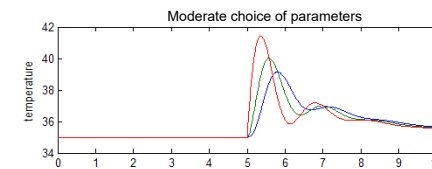
Smaller gain:
Larger stationary
error

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PI-control



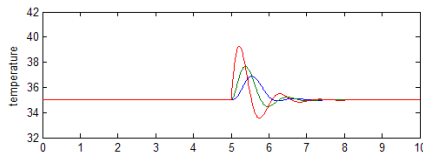
No stationary
error but
oscillating



Less oscillating
but slow

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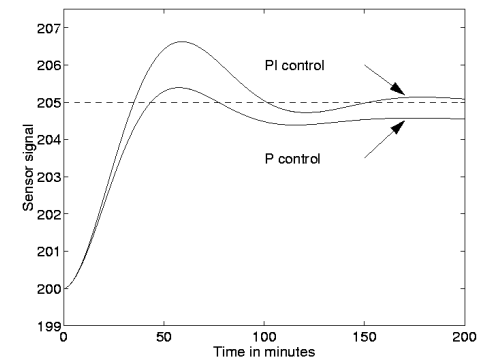
PID-control



No (or marginal)
oscillations
but slow

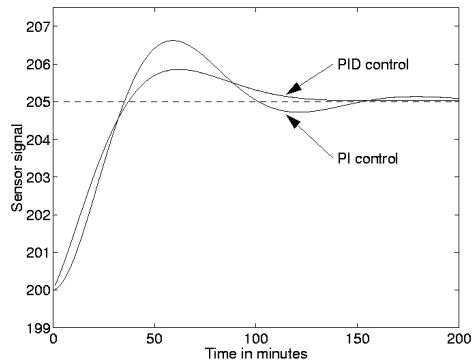
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P vs PI control – step change



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PI vs PID control – step change



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Ziegler Nichols tuning

1. Turn the controller to P-only mode, i.e. turn both the Integral and Derivative modes off.
2. Note if control gain is positive or negative.
3. Apply the input.
4. Change K in step increments and wait for a steady state.
5. Increase until sustained periodic oscillation and note the critical K_u
6. Note the period of the oscillation, P_u

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Ziegler Nichols tuning

	K	T_i	T_d
P	$K_u/2$	-	-
PI	$K_u/2.2$	$P_u/1.2$	-
PID	$K_u/1.7$	$P_u/2$	$P_u/8$

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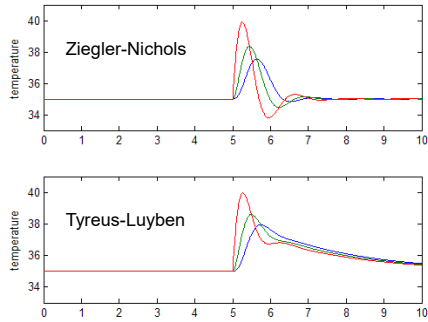
Tyreus-Luyben Tuning

- Same procedure to find K_u and P_u
- Less aggressive – less oscillations

	K	T_i	T_d
P	$K_u/2$	-	-
PI	$K_u/3.2$	$2.2 \cdot P_u$	-
PID	$K_u/2.2$	$2.2 \cdot P_u$	$P_u/6.3$

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Tuning



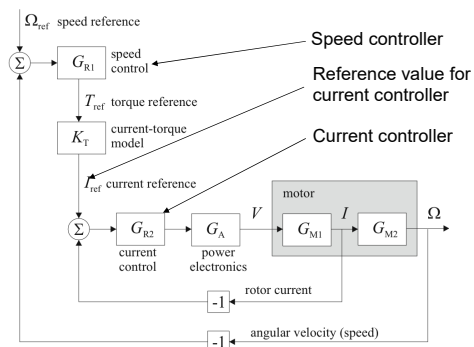
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Improving the performance

- Cascade control
- State feedback
- Feed-forward

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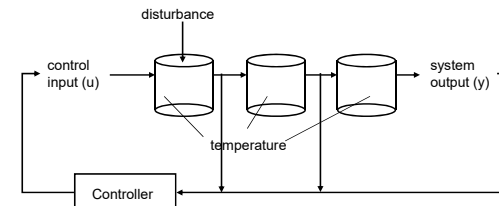
Cascade control



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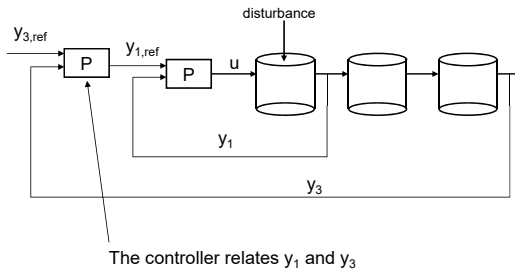
Cascade control

- Assume that we can measure in the test process



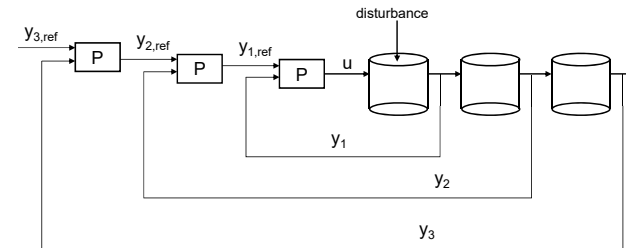
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Cascade control



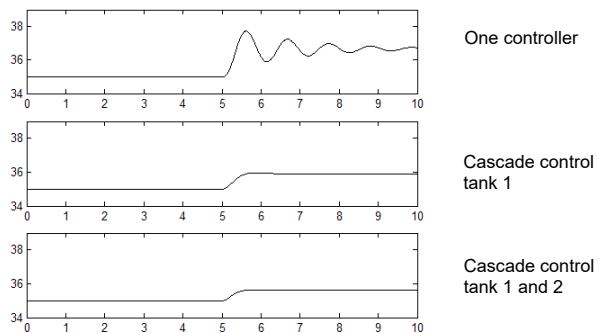
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Cascade control



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Cascade control



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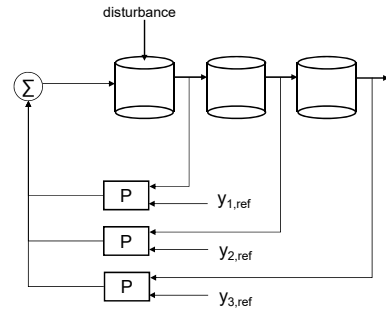
Cascade control

- The stationary error is easily removed by integral action on the outer loop
- Only one integral action
 - If the time constants of the loops are close they can start working against each other
- More than one integral action possible if the time constants are different
- In practice: the inner loops are typically sampled faster

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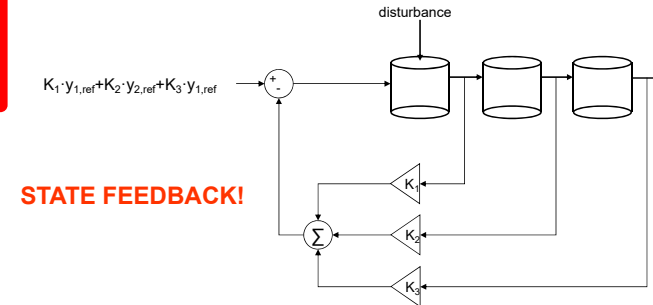
Cascade control alternative representation

Note: the parameters of the controllers are different from the other representation.



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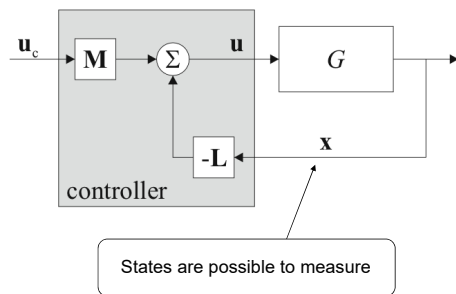
Cascade control alternative representation



STATE FEEDBACK!

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State feedback



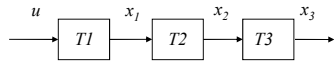
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State feedback (or cascade control)

- Feed back states to controller
- Pole placement used for controller design
 - Hand calculations for small (2nd order) systems
 - Matlab for larger systems
- If all states measurable
 - Control can be made fast (if no limits on the control action)
- Price: many sensors

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Concentration in 3 tanks



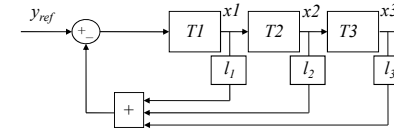
$\dot{x} = Ax + Bu$
 $y = Cx$

State space description of the system

$$A = \begin{pmatrix} -\frac{q}{V_1} & 0 & 0 \\ \frac{q}{V_2} & -\frac{q}{V_2} & 0 \\ 0 & \frac{q}{V_3} & -\frac{q}{V_3} \end{pmatrix} \quad B = \begin{pmatrix} \frac{q}{V_1} \\ 0 \\ 0 \end{pmatrix}$$

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State feedback 3 tanks in series



$$\dot{x} = (A - BL)x + By_{ref}$$

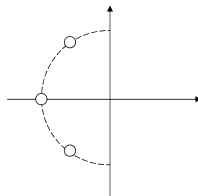
$$A - BL = \begin{pmatrix} -\frac{q}{V_1} - l_1 & \frac{q}{V_1} & -l_2 \frac{q}{V_1} \\ \frac{q}{V_2} & -\frac{q}{V_2} & 0 \\ 0 & \frac{q}{V_3} & -\frac{q}{V_3} \end{pmatrix}$$

The eigenvalues of $A - BL$ will determine the system

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Example: control design

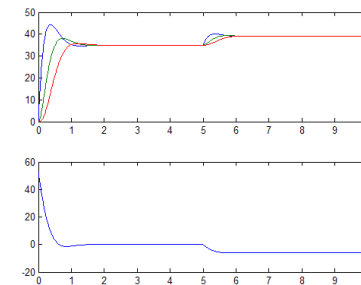
- Set L
 - Difficult to say anything of the system characteristics
- Pole placement
 - According to Butterworth – poles are placed on a circle
- Three cases
 - $L = [0.5 \ 0.5 \ 0.5] \Rightarrow \lambda = [-6.9 \ -3.5 + 3.2i \ -3.5 - 3.2i]$
 - $\lambda = [-1 \ -0.5 + 0.86i \ -0.5 - 0.86i] \cdot 6 \Rightarrow L = [0 \ 1.47 \ 0.86]$
 - $\lambda = [-1 \ -0.5 + 0.86i \ -0.5 - 0.86i] \cdot 12 \Rightarrow L = [3.0 \ 8.9 \ 13.8]$



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Example:

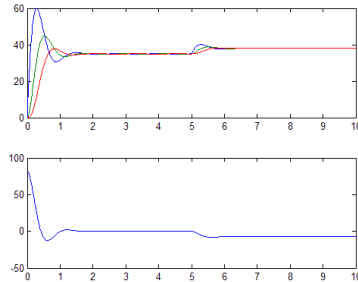
$$L = [0.5 \ 0.5 \ 0.5] \Rightarrow \lambda = [-6.9 \ -3.5 + 3.2i \ -3.5 - 3.2i]$$



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Example:

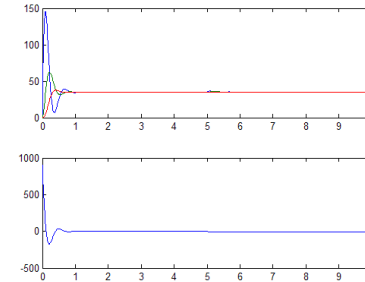
$$\lambda = [-1 -0.5+0.86i -0.5-0.86i] \cdot 6 \Rightarrow L=[0 \ 1.47 \ 0.86]$$



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Example:

$$\lambda = [-1 -0.5+0.86i -0.5-0.86i] \cdot 12 \Rightarrow L=[3.0 \ 8.9 \ 13.8]$$



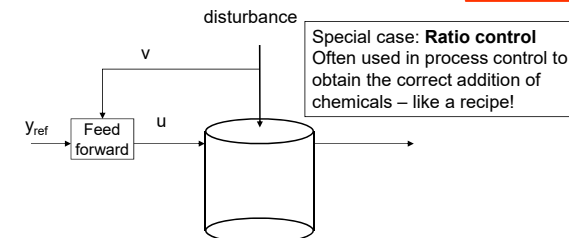
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Feed-forward

- **Measure** the reference value or the disturbance before it hits the plant
- **Compensate** for the disturbance before it has affected the plant
- **The price:** must supply a model of the influence of the disturbance

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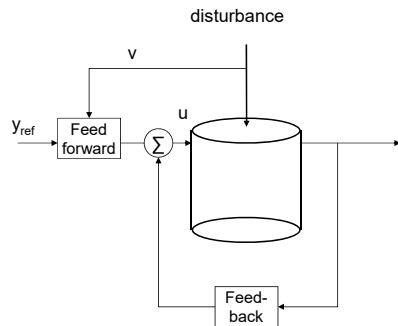
Feed forward



- No feedback – no guarantee that y_{ref} is achieved
- Need a perfect model between v and y
- Often sufficient for control

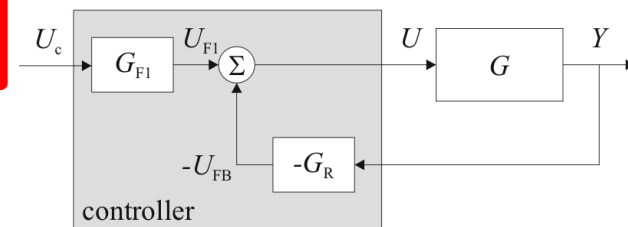
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Feed forward + feedback



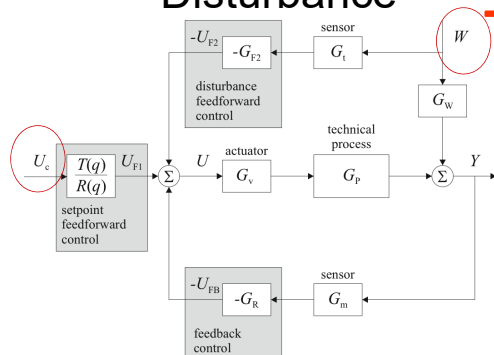
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Feedforward from reference + feedback



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FF from both Ref and Disturbance



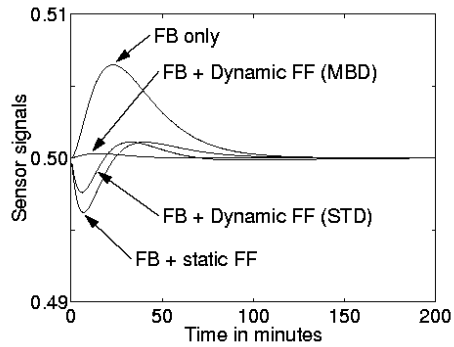
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Feedforward design

- **Measure** dynamics of manipulated variable and disturbance
- Different sophistication level on model from disturbance to process (static, dynamic)
- **Check** realisability
 - Manipulated variable dynamics faster than disturbance dynamics
- **Implement** full or partial controller

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Feedforward performance



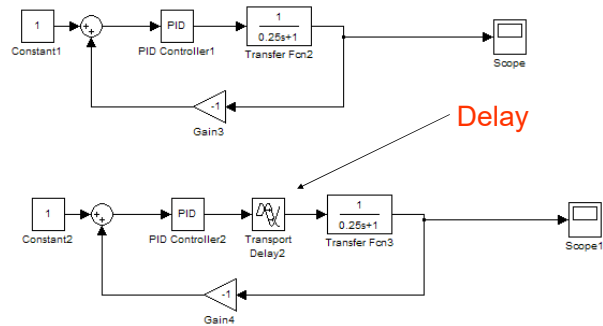
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Time delays

- Time delays in the process
 - Delay in the physical process – transport, slow dynamics
 - Delay in the measurement of the variable
 - Delay in the control action
- May introduce oscillations
- Makes the control slow
 - Control parameters must be moderate

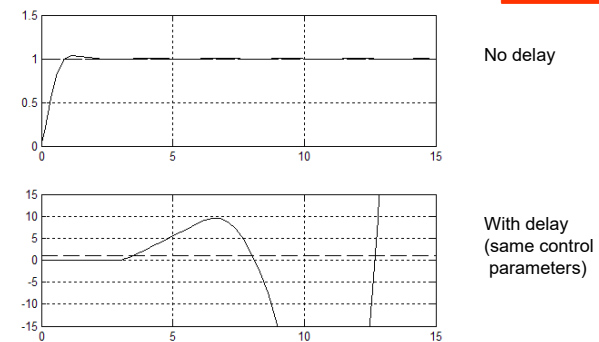
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Time delays



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Time delays



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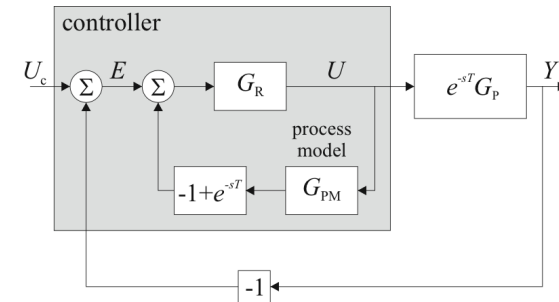
Time delays

Solutions:

- Use moderate control parameters
 - Slow controller
- Use controller with delay compensation
 - Smith predictor
 - If model perfect – results the same as the response without delay but time lagged!

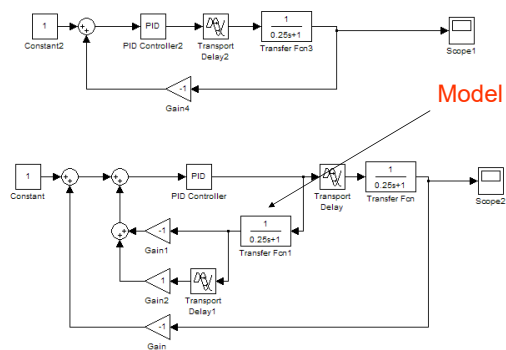
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Smith predictor



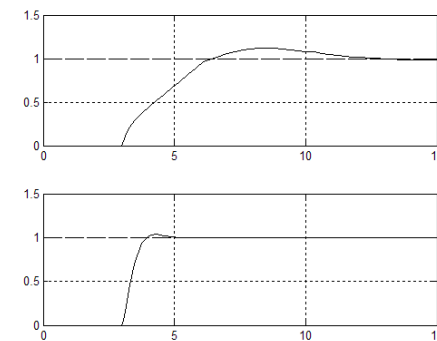
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Smith predictor



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Smith predictor



With delay
(moderate control
parameters)

With delay
(Smith predictor)

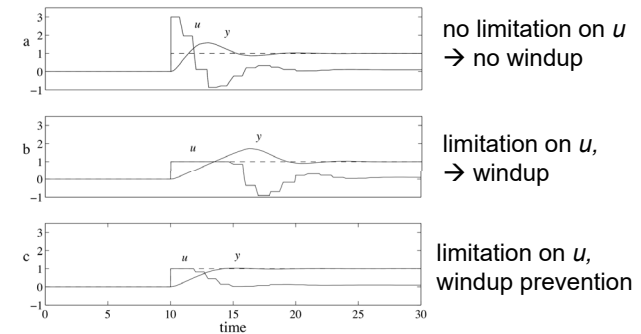
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Some practical limitations

- Control actions
 - Amplitude
 - Direction
- Measurement
 - States – are they possible to measure?
 - Disturbances – are they possible to measure?
 - Noise
- Time varying process
 - Process behaviour not constant

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The windup problem



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Changing the D part

$$u = u_0 + K_P \left(e + \frac{1}{T_i} \int e dt - T_D \frac{dy}{dt} \right)$$

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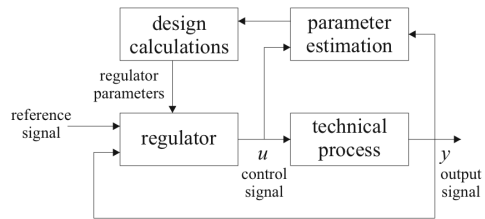
Noise filtering of the D part

$$-T_D \frac{dy}{dt} \rightarrow -T_D s Y(s)$$

$$\rightarrow -T_D \frac{s}{1 + T_f s} Y(s)$$

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Adaptive controller



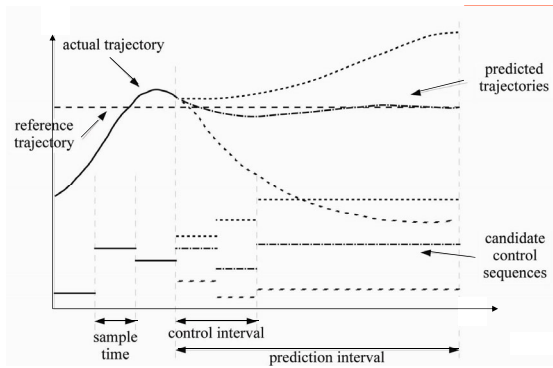
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Predictive control

- General principle of operation:
 - use **past control actions** (and predicted disturbances) with the model to **predict future** measured variables
 - **compare** to the goals and constraints
 - **determine** appropriate control actions to take

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Predictive control



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