

# PID control. Practical issues

- Smith Predictor (NOT PID...)
- PID Controller forms
- Ziegler-Nichols tuning
- Windup
- Digital implementation

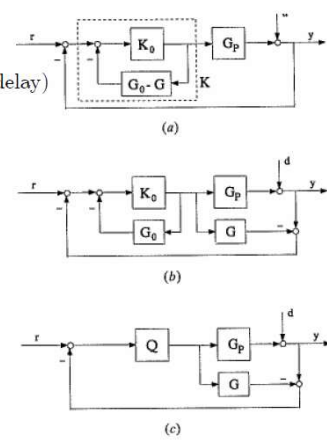
## Smith Predictor

Actual plant:  $G_p$   
 Model:  $G$   
 Delay-free model:  $G_0$   
 "Smith predictor":  $G_0 - G$  (predicts  $y$  when  $G$  has delay)  
 Conventional feedback controller:  $K$   
 $K_0$ : designed for plant without delay

Example  
 $G = k \frac{e^{-\theta s}}{\tau s + 1}$   
 Delay-free model:  $G_0 = k \frac{1}{\tau s + 1}$   
 $G_0 - G = \frac{k}{\tau s + 1} (1 - e^{-\theta s})$   
 Then  

$$K = \frac{K_0}{1 + K_0 \frac{k}{\tau s + 1} (1 - e^{-\theta s})}$$
 which with  $K_0 = \frac{1}{k} \frac{\tau s + 1}{\tau_c s}$   
 (SIMC-PI for delay free  $G_0$ )  
 gives "Smith predictor controller"  

$$K = \frac{\tau s + 1}{\tau_c s + 1 - e^{-\theta s}}$$
 (see also SIMC derivation)



a) Smith predictor control structure; (b) rearranged Smith predictor; (c) IMC structure.

SP looks good in theory. BUT: We have found that well-tuned PID (with  $\tau_D = \theta/3$ ) is more robust and almost always better than Smith predictor controller  $\Rightarrow$  FORGET SP!

## PID controller

$$\text{"Ideal" form: } u(t) = u_0 + \underbrace{K_c \left[ e(t) + \frac{1}{\tau_I} \int_0^t e(t) dt + \tau_D \frac{de(t)}{dt} \right]}_{\Delta u}$$

- $e(t) = y_s - y_m(t)$
- P-part: MV ( $\Delta u$ ) proportional to error
  - This is main part of the controller!
  - Problem. Gives steady-state offset =  $100\% / (1 + K_c k)$
- I-part: To avoid offset, add contribution proportional to integrated error.
  - Integral keeps changing as long as  $e \neq 0$
  - -> Will eventually make  $e=0$  (no steady-state offset!)
- Possible D-part: Add contribution proportional to change in (derivative of) error
  - Can improve control for high-order (S-shaped) response, but sensitive to measurement noise

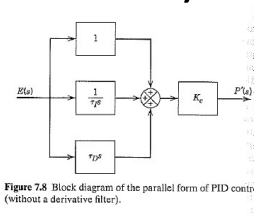
## "Ideal" PID (parallel form)

The *parallel form* of the PID control algorithm (without a derivative filter) is given by

$$p(t) = \bar{p} + K_c \left[ e(t) + \frac{1}{\tau_I} \int_0^t e(t^*) dt^* + \tau_D \frac{de(t)}{dt} \right] \quad (7-13)$$

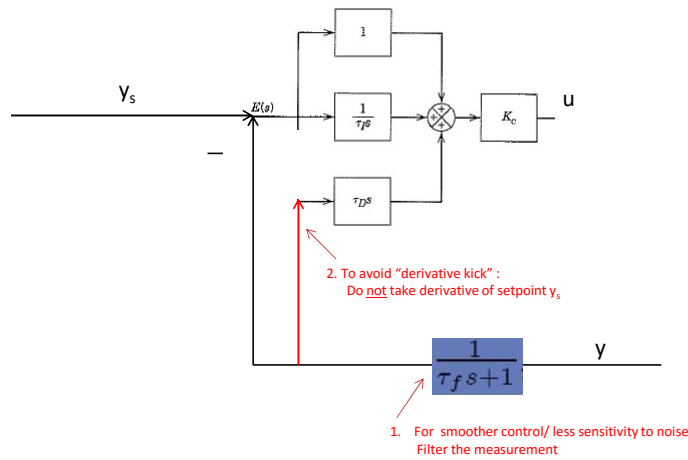
The corresponding transfer function is

$$\frac{P'(s)}{E(s)} = K_c \left[ 1 + \frac{1}{\tau_I s} + \tau_D s \right] \quad (7-14)$$



1. To get realizable controller and less sensitivity to measurement noise: Add filter on the measurement:  $f(s) = \frac{1}{\tau_f s + 1}$ , where  $\tau_f = \alpha \tau_d$ . Typical:  $\alpha = 0.1$
2. To avoid "derivative kick" do not take derivative of setpoint  
 illustrate the elimination of derivative kick, consider the parallel form of PID control in Eq. 7-13. Replacing  $de/dt$  by  $-dy_m/dt$  gives
 
$$p(t) = \bar{p} + K_c \left[ e(t) + \frac{1}{\tau_I} \int_0^t e(t^*) dt^* - \tau_D \frac{dy_m(t)}{dt} \right] \quad (7-17)$$
3. To avoid integral "windup" when input saturates (at max or min), use "anti" windup.  
 Simplest: Stop integration while input saturates

## Block diagram of practical "ideal" PID

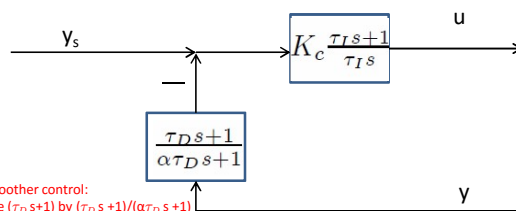


## Series (cascade) PID

Commercial versions of the series-form controller have a derivative filter that is applied to either the derivative term, as in Eq. 7-12, or to the PD term, as in Eq. 7-15:

$$\frac{P'(s)}{E(s)} = K_c \left( \frac{\tau_I s + 1}{\tau_I s} \right) \left( \frac{\tau_D s + 1}{\alpha \tau_D s + 1} \right) \quad (7-15)$$

Typical:  $\alpha=0.1$



1. For smoother control:  
Replace  $(\tau_D s + 1)$  by  $(\tau_D s + 1)/(\alpha \tau_D s + 1)$
2. To avoid "derivative kick"  
do not take derivative of setpoint

## Series to ideal form

Series (cascade) PID:

$$c(s) = K_c \frac{(\tau_I s + 1)(\tau_D s + 1)}{\tau_I s} = \frac{K_c}{\tau_I s} (\tau_I \tau_D s^2 + (\tau_I + \tau_D)s + 1)$$

The settings given in this paper ( $K_c, \tau_I, \tau_D$ ) are for the series (cascade, “interacting”) form PID controller in (1). To derive the corresponding settings for the ideal (parallel, “non-interacting”) form PID controller

$$\text{Ideal PID : } c'(s) = K_c' \left( 1 + \frac{1}{\tau_I' s} + \tau_D' s \right) = \frac{K_c'}{\tau_I' s} (\tau_I' \tau_D' s^2 + \tau_I' s + 1) \quad (35)$$

we use the following translation formulas

$$K_c' = K_c \left( 1 + \frac{\tau_D}{\tau_I} \right); \quad \tau_I' = \tau_I \left( 1 + \frac{\tau_D}{\tau_I} \right); \quad \tau_D' = \frac{\tau_D}{1 + \frac{\tau_D}{\tau_I}} \quad (36)$$

Derivation: See exercise

Note: The reverse transformation (from ideal to series) is not always possible because the ideal controller may have complex zeros.

**Table 7.1** Common PID Controllers

Controller Type	Other Names Used	Controller Equation	Transfer Function
Parallel	Ideal, additive, ISA form	$p(t) = \bar{p} + K_c \left( e(t) + \frac{1}{\tau_I} \int_0^t e(t^*) dt^* + \tau_D \frac{de(t)}{dt} \right)$	$\frac{P'(s)}{E(s)} = K_c \left( 1 + \frac{1}{\tau_I s} + \tau_D s \right)$
Parallel with derivative filter	Ideal, realizable, ISA standard	See Exercise 7.10(a)	$\frac{P'(s)}{E(s)} = K_c \left( 1 + \frac{1}{\tau_I s} + \frac{\tau_D s}{\alpha \tau_D s + 1} \right)$
Series	Multiplicative, interacting	See Exercise 7.11	$\frac{P'(s)}{E(s)} = K_c \left( \frac{\tau_I s + 1}{\tau_I s} \right) (\tau_D s + 1)$
Series with derivative filter	Physically realizable	See Exercise 7.10(b)	$\frac{P'(s)}{E(s)} = K_c \left( \frac{\tau_I s + 1}{\tau_I s} \right) \left( \frac{\tau_D s + 1}{\alpha \tau_D s + 1} \right)$
Expanded	Noninteracting	$p(t) = \bar{p} + K_c e(t) + K_I \int_0^t e(t^*) dt^* + K_D \frac{de(t)}{dt}$	$\frac{P'(s)}{E(s)} = K_c + \frac{K_I}{s} + K_D s$
Parallel, with proportional and derivative weighting	Ideal $\beta, \gamma$ controller	$p(t) = \bar{p} + K_c \left( e_P(t) + \frac{1}{\tau_I} \int_0^t e(t^*) dt^* + \tau_D \frac{de_D(t)}{dt} \right)$ where $e_P(t) = \beta y_{sp}(t) - y_m(t)$ $e(t) = y_{sp}(t) - y_m(t)$ $e_D(t) = \gamma y_{sp}(t) - y_m(t)$	$P'(s) = K_c \left( E_P(s) + \frac{1}{\tau_I s} E(s) + \tau_D s E_D(s) \right)$ where $E_P(s) = \beta Y_{sp}(s) - Y_m(s)$ $E(s) = Y_{sp}(s) - Y_m(s)$ $E_D(s) = \gamma Y_{sp}(s) - Y_m(s)$

+ many more (see manual for your control system...)

## Optimal PID settings

- Can find optimal settings using optimization
- SIMC-rules are close to IAE-optimal for combined setpoints and disturbances\*

### 11.3.2 Tuning Relations Based on Integral Error Criteria

Controller tuning relations have been developed that optimize the closed-loop response for a simple process model and a specified disturbance or set-point change. The optimum settings minimize an *integral error criterion*. Three popular integral error criteria are

1. *Integral of the absolute value of the error (IAE)*

$$\text{IAE} = \int_0^{\infty} |e(t)| dt \quad (11-35)$$

where the error signal  $e(t)$  is the difference between the set point and the measurement.

2. *Integral of the squared error (ISE)*

$$\text{ISE} = \int_0^{\infty} e^2(t) dt \quad (11-36)$$

3. *Integral of the time-weighted absolute error (ITAE)*

$$\text{ITAE} = \int_0^{\infty} t|e(t)| dt \quad (11-37)$$

\*Chriss Grimholt and Sigurd Skogestad, "Optimal PI-Control and Verification of the SIMC Tuning Rule", Proceedings IFAC conference on Advances in PID control (PID'12), Brescia, Italy, 28-30 March 2012.

## Methods for online tuning of PID controllers

- I. Trial and error
- II. Ziegler Nichols
  - Oscillating P-control
  - Relay method to get oscillations
- III. Closed-loop response with P-control
  - Shams method

On-line tuning: Avoids an open-loop experiment, like a step input change.

Advantage on-line: Process is always "under control"

In practice: Both "open-loop" and "closed-loop" (online) methods are used

Tuning of your PID controller

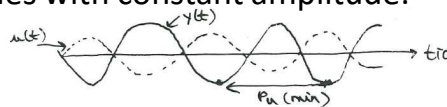
## I. "Trial & error" approach (online)

- P-part: Increase controller gain ( $K_c$ ) until the process starts oscillating or the input saturates
- Decrease the gain ( $\sim$  factor 2)
- I-part: Reduce the integral time ( $\tau_I$ ) until the process starts oscillating
- Increase a bit ( $\sim$  factor 2)
- Possible D-part: Increase  $\tau_D$  and see if there is any improvement

Very common approach,  
BUT: Time consuming and does not give good tunings: NOT recommended

## II. Ziegler-Nichols closed-loop method (1942)

- P-control only: Increase controller gain ( $K_c$ ) until the process cycles with constant amplitude:



- Write down the corresponding "ultimate" period ( $P_u$ ) and controller gain ( $K_u$ ).
- Based on this "process information" obtain PID settings:

**Table 11.4** Controller Settings based on the Continuous Cycling Method

Ziegler-Nichols	$K_c$	$\tau_I$	$\tau_D$
P	$0.5K_{cu}$	—	—
PI	$0.45K_{cu}$	$P_u/1.2$	—
PID (ideal)	$0.6K_{cu}$	$P_u/2$	$P_u/8$
Tyres-Luyben <sup>†</sup>	$K_c$	$\tau_I$	$\tau_D$
PI	$0.31K_{cu}$	$2.2P_u$	—
PID	$0.45K_{cu}$	$2.2P_u$	$P_u/6.3$

← ZN is often a bit aggressive.

← TL-modification is smoother (smaller  $K_c$  and larger  $\tau_I$ ).

<sup>†</sup> Luyben and Luyben (1997).

**TIGHT CONTROL**

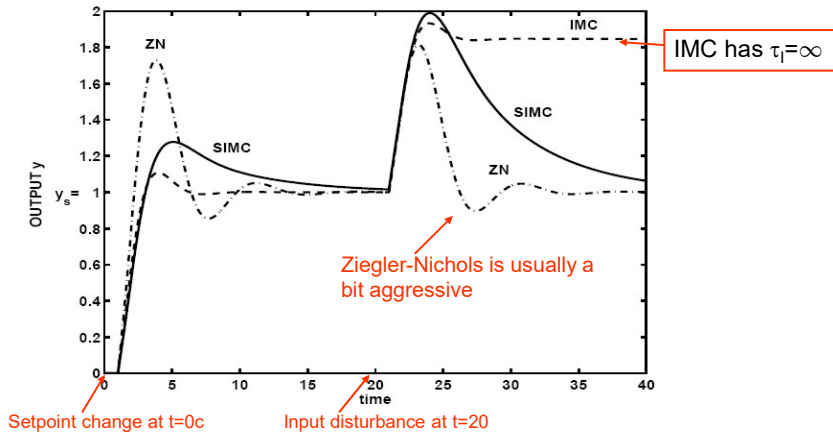
Example. Integrating process with delay=1.  $G(s) = e^{-s}/s$ .

Model:  $k'=1, \theta=1, \tau_1=\infty$

SIMC-tunings with  $\tau_c$  with  $\theta=1$ :

$$K_c = \frac{1}{k'} \frac{1}{\tau_c + \theta} = 1 \cdot \frac{1}{1+1} = 0.5$$

$$\tau_I = \min(\tau_1, 4(\tau_c + \theta)) = \min(\infty, 8) = 8$$



**TIGHT CONTROL**

EXAMPLE: Process from Astrom et al. (1998)

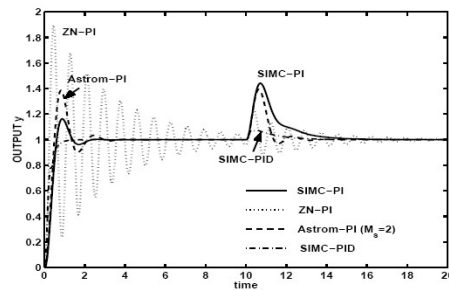


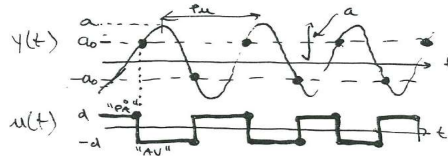
Figure 3: Load disturbance of magnitude 2 occurs at  $t = 10$ .

$$g_0(s) = \frac{1}{(s + 1)(0.2s + 1)(0.04s + 1)(0.008s + 1)}$$

1. Approximate as first-order model with  $k=1, \tau_1 = 1+0.1=1.1, \theta=0.1+0.04+0.008 = 0.148$   
Get SIMC PI-tunings ( $\tau_c=0$ ):  $K_c = 1 \cdot 1.1/(2 \cdot 0.148) = 3.71, \tau_I = \min(1.1, 8 \cdot 0.148) = 1.1$
2. Approximate as second-order model with  $k=1, \tau_1 = 1, \tau_2=0.2+0.02=0.22, \theta=0.02+0.008 = 0.028$   
Get SIMC PID-tunings ( $\tau_c=0$ ):  $K_c = 1 \cdot 1/(2 \cdot 0.028) = 17.9, \tau_I = \min(1.8, 8 \cdot 0.028) = 0.224, \tau_D = 0.22$

## Åstrøm relay method (1984): Alternative approach to obtain cycling (and $K_u + P_u$ )

- Avoids operating at limit to instability
- Use ON/OFF controller (=relay) where input  $u(t)$  varies  $\pm d$  (around nominal)
- Switch when output  $y(t)$  reaches  $\pm a_0$  (deadband) (around setpoint)
- Example: Thermostat in your home



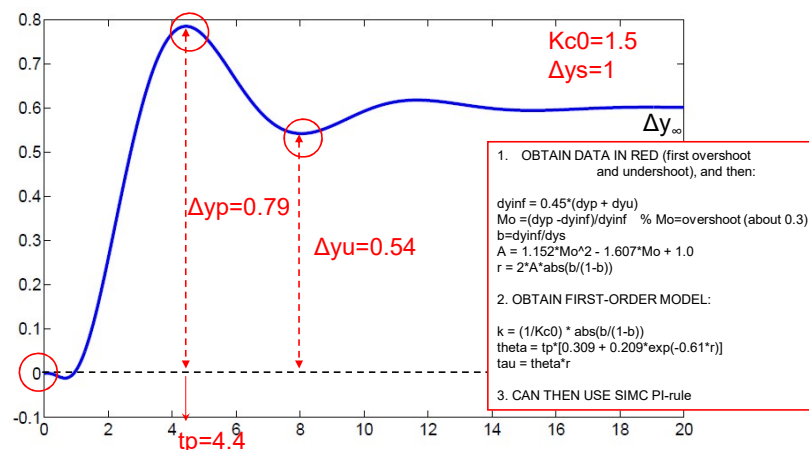
- From this obtain  $P_u$  and

$$K_u = \frac{4}{1/4} \frac{d}{a}$$

← d: amplitude u(t) (set by user)  
← a: amplitude y(t) (from experiment)

Alternative to Ziegler-Nichols closed-loop experiment that avoids cycling.

### III. Shams' method: Closed-loop setpoint response with P-controller with about 20-40% overshoot



Example 2: Get  $k=0.99$ ,  $\theta=1.68$ ,  $\tau=3.03$

Ref: Shamsuzzoha and Skogestad (JPC, 2010)  
 + modification by C. Grimholt (Project, NTNU, 2010; see also PID-book 2012)



*Example E2 (Further continued)* We want to derive PI- and PID-settings for the process

$$g_0(s) = \frac{(-0.3s + 1)(0.08s + 1)}{(2s + 1)(1s + 1)(0.4s + 1)(0.2s + 1)(0.05s + 1)^3}$$

using the SIMC tuning rules with the “default” recommendation  $\tau_c = \theta$ . From the closed-loop setpoint response, we obtained in a previous example a first-order model with parameters  $k = 0.994, \theta = 1.67, \tau_1 = 3.00$  (5.10). The resulting SIMC PI-settings with  $\tau_c = \theta = 1.67$  are

$$\text{PI}_{\text{cl}} : K_c = 0.904, \quad \tau_I = 3.$$

From the full-order model  $g_0(s)$  and the half rule, we obtained in a previous example a first-order model with parameters  $k = 1, \theta = 1.47, \tau_1 = 2.5$ . The resulting SIMC PI-settings with  $\tau_c = \theta = 1.47$  are

$$\text{PI}_{\text{half-rule}} : K_c = 0.850, \quad \tau_I = 2.5.$$

From the full-order model  $g_0(s)$  and the half rule, we obtained a second-order model with parameters  $k = 1, \theta = 0.77, \tau_1 = 2, \tau_2 = 1.2$ . The resulting SIMC PID-settings with  $\tau_c = \theta = 0.77$  are

$$\text{Series PID} : K_c = 1.299, \quad \tau_I = 2, \quad \tau_D = 1.2.$$

The corresponding settings with the more common ideal (parallel form) PID controller are obtained by computing  $f = 1 + \tau_D/\tau_I = 1.60$ , and we have

$$\text{Ideal PID} : K_c' = K_c f = 1.69, \quad \tau_I' = \tau_I f = 3.2, \quad \tau_D' = \tau_D / f = 0.75. \quad (5.30)$$

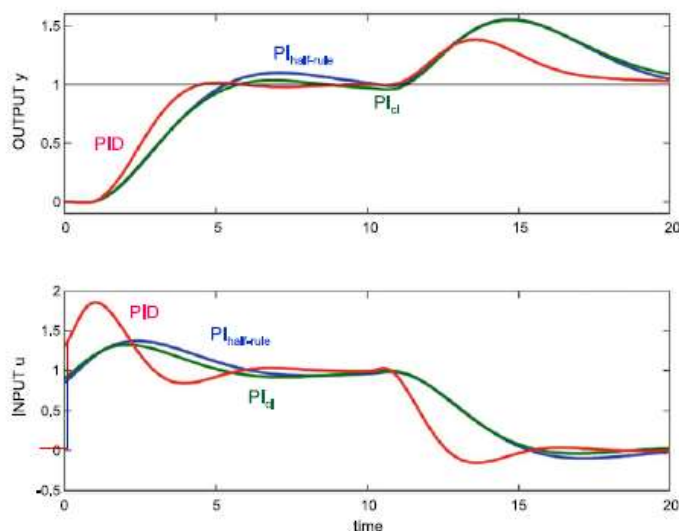
April 4-8, 2004

KFUPM-DISTANCE CONTROL COURSE

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S. Skogestad and C. Grimholt



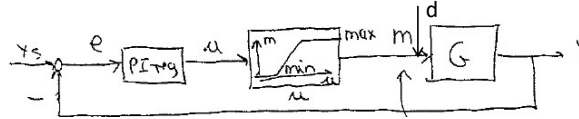
**Fig. 5.6** Closed-loop responses for process E2 using SIMC PI- and PID-tunings with  $\tau_c = \theta$ . Setpoint change at  $t = 0$  and input (load) disturbance at  $t = 10$ . For the PID controller, D-action is only on the feedback signal, i.e., not on the setpoint  $y_s$

Apr

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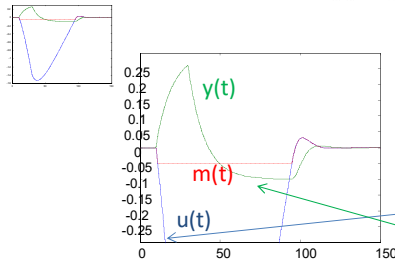
## Integral windup

- Problem: Integrator “winds up”  $u(t)$  when actual input has saturated



Actual input is  $m$ .  
 $m = u$  if no saturation

$$u(t) = u_0 + K_c e(t) + \underbrace{\frac{K_c}{\tau_I} \int_0^t e(t) dt}_{\text{Keeps changing when } e(t) \neq 0}$$

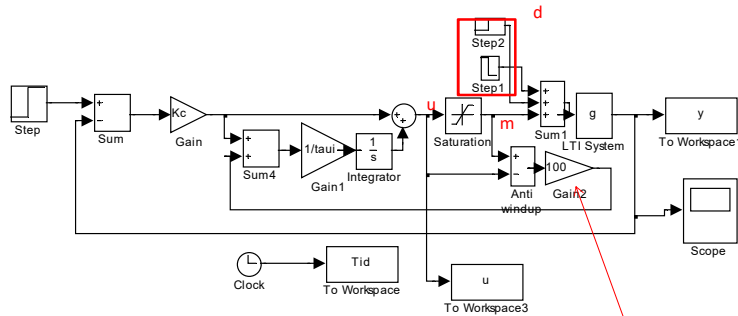


- $t=10$ : Disturbance  $d$  starts
- $t \approx 12$ : Reach saturation in  $m$  (actual input)
- oops! Integration makes  $u$  (desired input) “wind up” ...
- ...so when disturbance ends (at  $t=30$ )  
 $y(t)$  has to overshoot on the negative side to bring  $u(t)$  back.

## Anti-windup

- Approaches to avoid windup
  - Stop integration (e.g. set  $\tau_I = 9999$ ) when saturation in input occurs (*requires logic*)
  - Force integrator to follow true input using high-gain feedback correction (*see Example*)
  - Use discrete controller in *velocity form*

# Example anti-windup

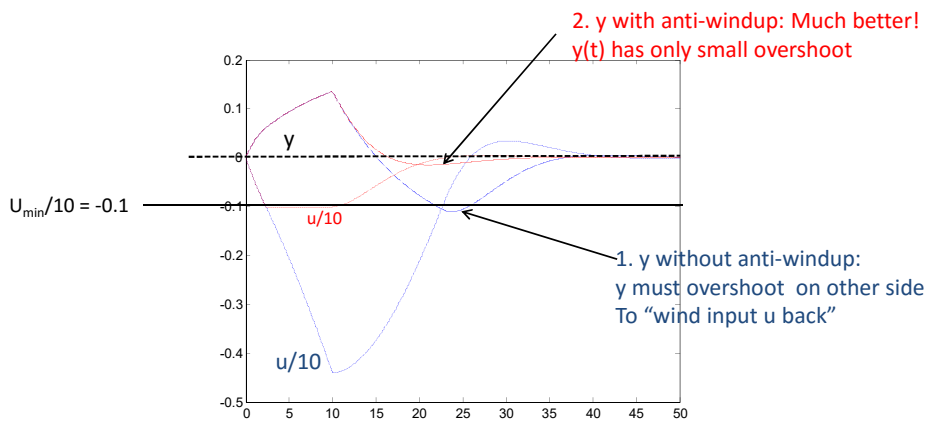


$g = 0.2/(10*s+1)$   
 $\tau_{auc}=1: K_c=12.5, \tau_{ai}=4$

Input: max=1, min=-1  
 Disturbance: Pulse from 0 to 2 and back to 0 at t=10

Approach 2:  
 High-gain feedback correction  
 which makes u follow m.  
 Gain should be 1 or larger.  
 (Without anti-windup:  
 Set this gain to zero)

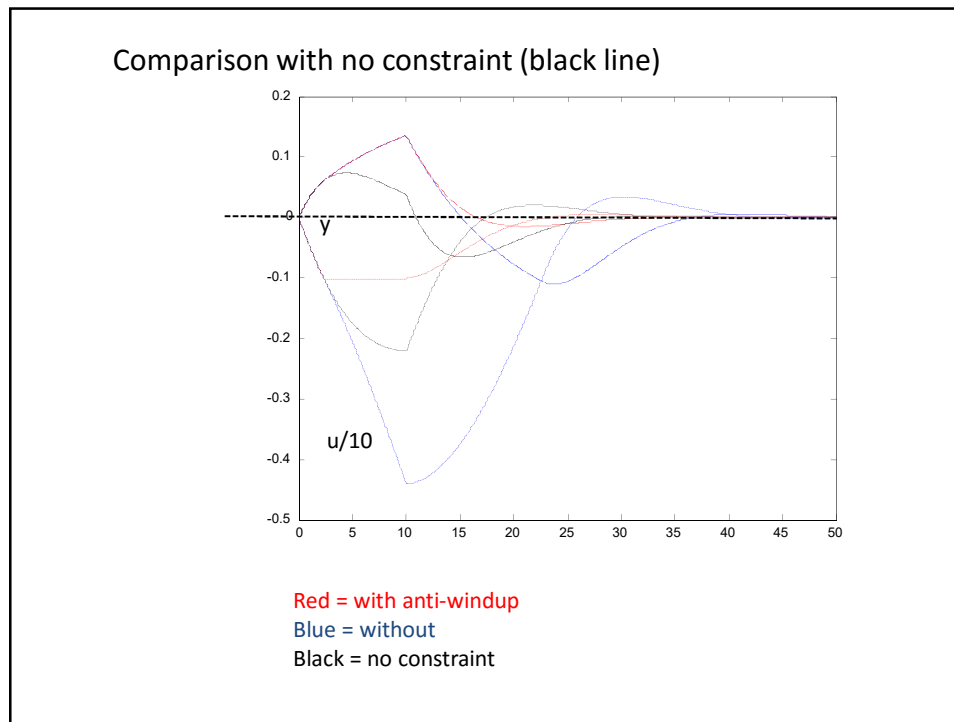
File: tunePIDantiwindup.mdl



t=0: Disturbance starts  
 t=10: Disturbance ends

1. Blue = without anti-windup  
 2. Red = with anti-windup

y = output process  
 u = output controller (the real process input cannot go below  $u/10=-0.1$ )  
 m = sat(u) = input process

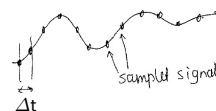


## Bumpless transfer

- We want a “soft” transition when the controller is switched between “manual” (1) and “auto” (2)
  - or back from auto to manual
  - or when controller is retuned
- Simple solution: reset  $u_0$  as you switch, so that  $u_1(t) = u_2(t)$ .

$$u(t) = u_0 + \underbrace{K_c \left[ e(t) + \frac{1}{\tau_I} \int_0^t e(t) dt + \tau_D \frac{de(t)}{dt} \right]}_{\Delta u}$$

# Effect of sampling



- All real controllers are digital, based on sampling
- $\Delta t$  = sampling time (typical 1 sec. in process control, but could be MUCH faster)
- Max sampling time (Shannon):  $\Delta t < \tau_c/2$ , but preferably much smaller ( $\tau_c$  = closed-loop response time)
- With continuous methods: Approximate sampling time as effective delay  $\theta \approx \Delta t/2$
- Strange things can happen if  $\Delta t$  is too large:

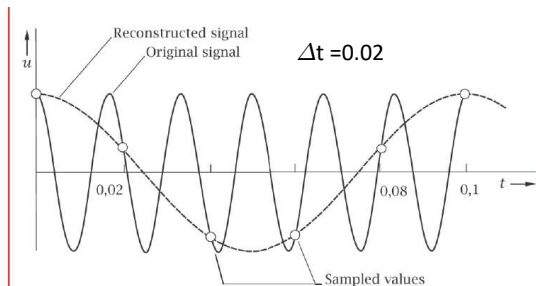


Figure 6-8: Falsification due to undersampling

Figure 6-8 illustrates in a particularly drastic manner the consequences of a violation of the Shannon theorem. A sinusoidal original signal with a frequency of 60 Hz is sampled with a frequency of 50 Hz, although the sampling frequency should be higher than 120 Hz. The consequence

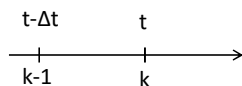
"How to program a PID controller in 5 minutes." (In addition you should filter the measurement)

## Discrete (digital) implementation (practical in computer) of PID controller

Continuous (not possible in computer):

$$u(t) = u_0 + \underbrace{\frac{K_c}{\tau_I} \int_0^t e(t) dt + K_c e(t) + K_c \tau_D \frac{de(t)}{dt}}_{\bar{u}(t)}$$

$e(t) = y_s - y$  but use  $-dy/dt$  to avoid differentiation of setpoint  
 $y$  = filtered measurement



Digital PID implementation:

$$\bar{u}_k = \bar{u}(t) \approx \bar{u}_{k-1} + \frac{K_c}{\tau_I} e_k \Delta t \text{ (Backward Euler)}$$

$$u_k = \bar{u}_k + K_c e_k - K_c \tau_D \frac{y_k - y_{k-1}}{\Delta t}$$

This is Sigurd's recommendation = "Alt. 3" (see next page)

To avoid windup and to get bumpless transfer:  
Adjust bias  $\bar{u}_k$  so that  $u_k$  = actual input  
(Comment: Can "clamp"  $\bar{u}_k$  as we enter saturation (= stop integration), but may then take a little longer to get out of saturation, because  $e$  must return to value it has as we entered saturation.)

### Comparison with book: Digital implementation of PID controllers

**Alt. 1 (position form)**

$$p(t) = \bar{p} + K_c \left[ e(t) + \frac{1}{\tau_I} \int_0^t e(t^*) dt + \tau_D \frac{de(t)}{dt} \right] \quad (7-13)$$

Note:  $p$  = output from controller

Finite difference approximation:

$$\int_0^t e(t^*) dt \approx \sum_{j=1}^k e_j \Delta t \quad (7-24)$$

$$\frac{de}{dt} \approx \frac{e_k - e_{k-1}}{\Delta t} \quad (7-25)$$

where

$\Delta t$  = the sampling period (the time between successive measurements of the controlled variable)  
 $e_k$  = error at the  $k$ th sampling instant for  $k = 1, 2, \dots$

Substituting Eqs. 7-24 and 7-25 into (7-13) gives the position form,

$$p_k = \bar{p} + K_c \left[ e_k + \frac{\Delta t}{\tau_I} \sum_{j=1}^k e_j + \frac{\tau_D}{\Delta t} (e_k - e_{k-1}) \right] \quad \text{Alt. 1}$$

$e_k$  = present sampled value =  $e(t)$   
 $e_{k-1}$  = previous sample =  $e(t-\Delta t)$   
 $e_{k-2}$  =  $e(t-2\Delta t)$

**Alt. 3 (Sigurd's with bias as extra state, better than Alt. 1 and Alt. 2)**

Alternative to Alt. 1: Store integral action in the bias  $\bar{p}_k$

$$p_k = \bar{p}_k + K_c \left[ e_k + \frac{\tau_D}{\Delta t} (e_k - e_{k-1}) \right]$$

where we update ("reset") the bias:  $\bar{p}_k = \bar{p}_{k-1} + K_c \frac{\Delta t}{\tau_I} e_k$

To avoid windup and to get bumpless transfer:

Adjust bias  $\bar{p}_k$  so that  $p_k$  = actual input

**Alt. 2 (velocity form)**

In the velocity form, the change in controller output is calculated. The velocity form can be derived by writing Eq. 7-26 for the  $(k-1)$  sampling instant:

$$p_{k-1} = \bar{p} + K_c \left[ e_{k-1} + \frac{\Delta t}{\tau_I} \sum_{j=1}^{k-1} e_j + \frac{\tau_D}{\Delta t} (e_{k-1} - e_{k-2}) \right] \quad (7-27)$$

Note that the summation still begins at  $j = 1$ , because it is assumed that the process is at the desired steady state for  $j \approx 0$ , and thus  $e_j = 0$  for  $j \approx 0$ . Subtracting Eq. 7-27 from (7-26) gives the velocity form of the digital PID algorithm:

$$\Delta p_k = p_k - p_{k-1} = K_c \left[ (e_k - e_{k-1}) + \frac{\Delta t}{\tau_I} e_k + \frac{\tau_D}{\Delta t} (e_k - 2e_{k-1} + e_{k-2}) \right] \quad \text{Velocity form} \quad (7-28)$$

Alt. 2

The velocity form has three advantages over the position form:

1. It inherently contains antireset windup, because the summation of errors is not explicitly calculated.
2. This output is expressed in a form,  $\Delta p_k$ , that can be utilized directly by some final control elements, such as a control valve driven by a pulsed stepping motor.

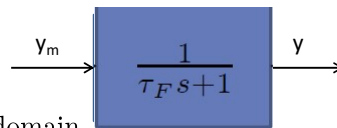
3. For the velocity algorithm, transferring the controller from manual to automatic model does not require any initialization of the output ( $\bar{p}$  in

=Bumpless transfer

A minor disadvantage of the velocity form is that the integral mode *must* be included. When the set point is constant, it cancels out in both the proportional and derivative error terms. Consequently, if the integral mode were omitted, the process response to a disturbance would tend to drift away from the set point.

? This is a major disadvantage of Alt. 2

### Digital implementation of first-order filter of measurement\*



Tuning: Select  $\tau_F < \tau_c/10$

Continuous s-domain

$$y(s) = \frac{1}{\tau_F s + 1} y_m(s) \Rightarrow \tau_F s y(s) + y(s) = y_m(s)$$

Continuous time domain

$$\tau_F \frac{dy(t)}{dt} + y(t) = y_m(t)$$

Discrete (digital) approximations :

$$\frac{\tau_F}{\Delta t} (y_k - y_{k-1}) + y_k = y_{m,k}$$

Rearrange

$$y_k = \alpha y_{m,k} + (1 - \alpha) y_{k-1}$$

where

$$\alpha = \frac{1}{1 + \tau_F / \Delta t}$$

$$\tau_F = 0 \Rightarrow \alpha = 1 \text{ (no filtering)}$$

$$\tau_F = \Delta t \Rightarrow \alpha = 0.5$$

$$\tau_F = 9\Delta t \Rightarrow \alpha = 0.1$$

\*Equivalent to exponential moving average of time series data

## Block diagram symbols

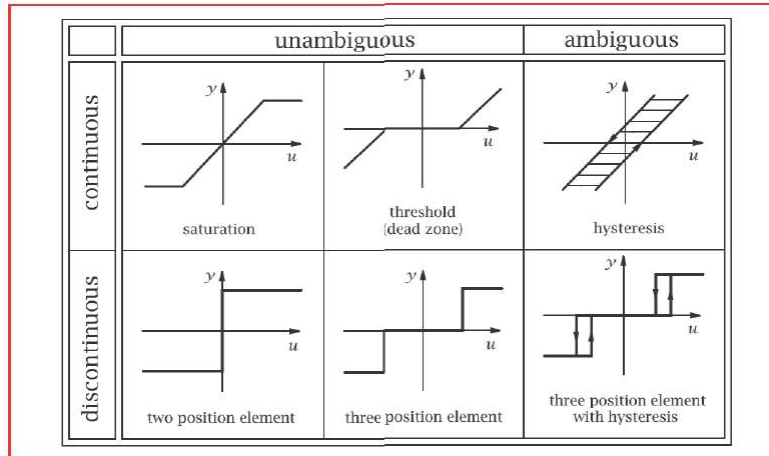


Figure 9-1: Types of characteristic curves