



EL2520

Control Theory and Practice

Lecture 13:

Dealing with Hard Constraints

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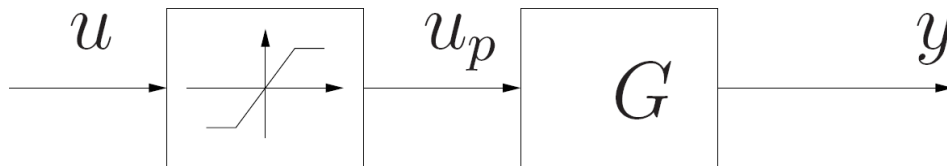
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Input Constraints

Dealing with input constraints:

- Linear control design: punish large control moves, e.g.,
 - LQG: choose large input weight Q_2
 - H_∞ : include e.g, $\|G_{wu}\|_\infty$ in objective function
- But, inputs often have hard constraints

$$u_{min} \leq u_p \leq u_{max}$$



Outline of Lecture

Dealing with hard constraints

- Previous lecture: Constrained Receding Horizon Control / MPC
- This lecture: Anti reset windup
 - State feedback with a nonlinear observer
 - Interpretation and extension to any controller

Model Predictive Control

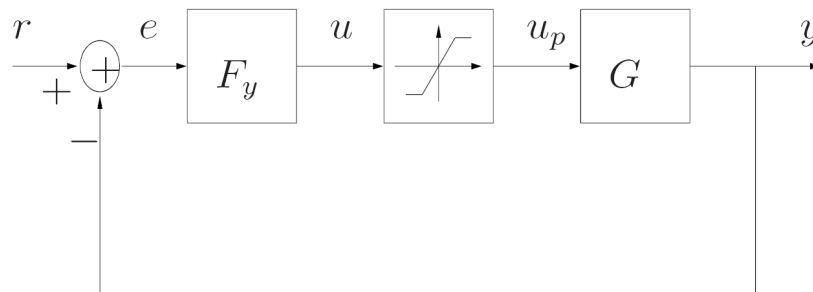
- Finite-horizon discrete time LQR with hard constraints on u and y :

$$\begin{aligned} \text{minimize} \quad & J(U) = \sum_{k=0}^{N-1} (x_k^T Q_1 x_k + u_k^T Q_2 u_k) + x_N^T Q_f x_N \\ \text{subject to} \quad & u_{\min} \leq u_k \leq u_{\max}, \quad k = 0, \dots, N-1 \\ & y_{\min} \leq Cx_k \leq y_{\max}, \quad k = 1, \dots, N \\ & x_{k+1} = Ax_k + Bu_k \end{aligned}$$

- Can be written as a quadratic programming problem in $\{u_0, \dots, u_{N-1}\}$
- Implement only u_0 , let system evolve one sample and redo optimization with new state estimate; results in receding horizon optimization
- Main advantage: constraints can be included in optimization

Anti Reset Windup

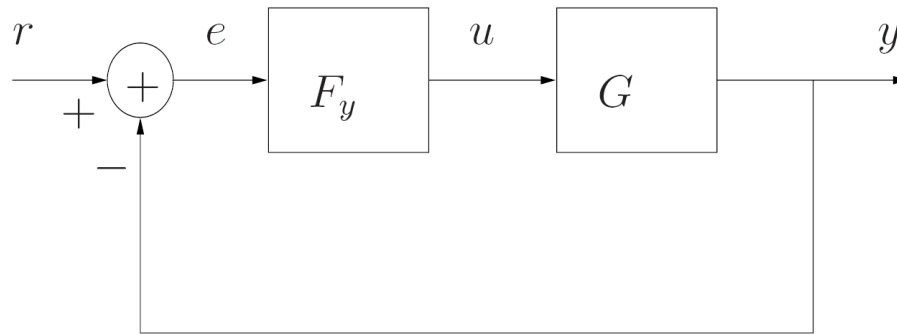
- The problem with saturating input



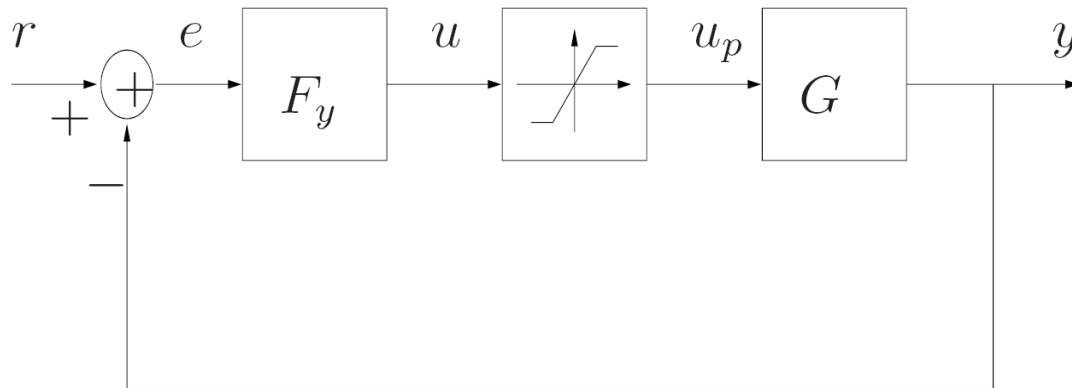
- feedback broken, i.e., system open-loop, when u in saturation
- problem, in particular if F or G unstable
- F usually has integrator (unstable)
- The classical approach to deal with hard constraints on the input is called anti-reset windup

Magnitude limitations on control

Linear model



Actual implementation



Example: DC servo

Servo:

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

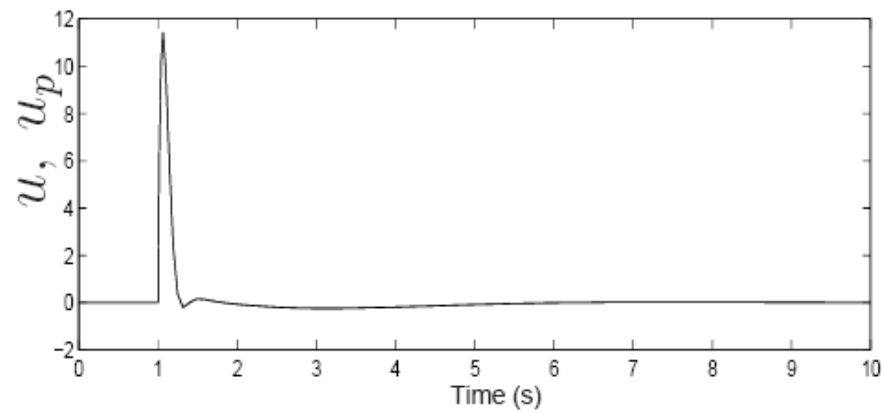
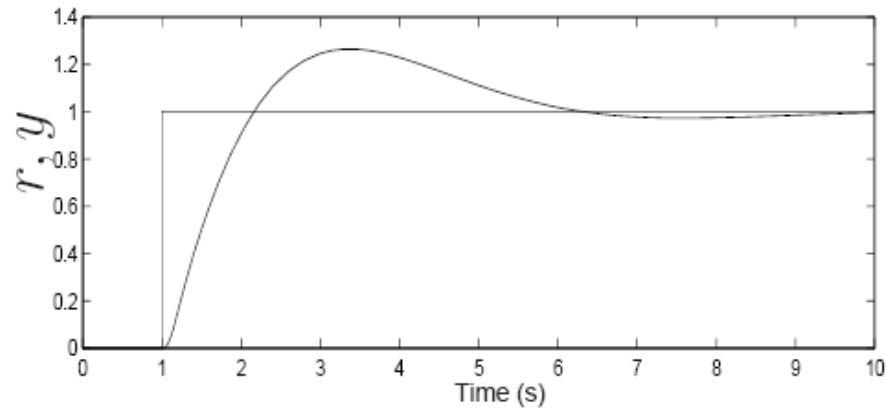
A controller designed using LQG is

$$F_y(s) = \frac{439s^2 + 710.5s + 316.2}{s^3 + 26.47s^2 + 349.8s - 7.13}$$

which has poles in $-13.2444 \pm 13.2255i$, and 0.0204

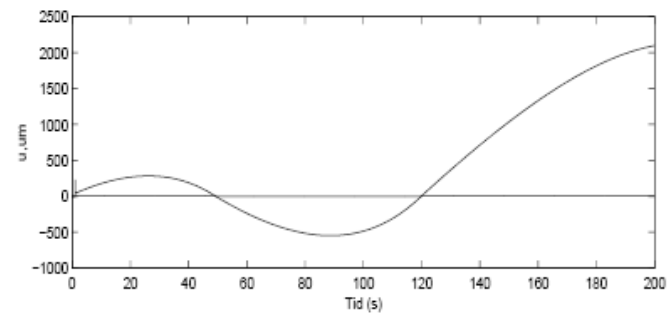
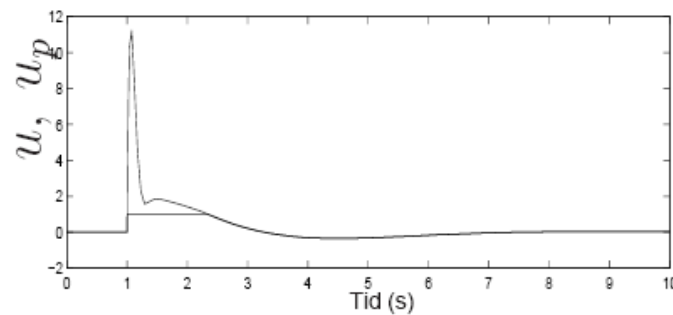
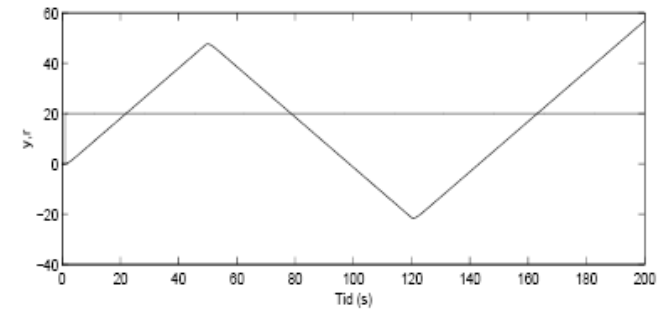
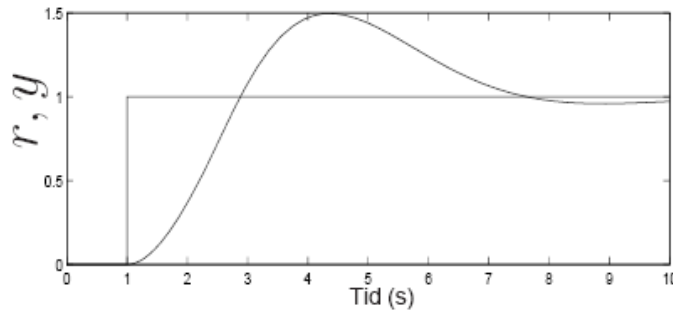
Note: controller is unstable, but closed loop is internally stable!

Step response (no constraints)



Step response with saturated input

$$-1 \leq u_p \leq 1$$



Slower, larger overshoot

unstable

Observer + State Feedback

Many controllers based on feedback from observed states

- Observer:

$$\frac{d}{dt}\hat{x}(t) = A\hat{x}(t) + Bu(t) + K(y(t) - C\hat{x}(t))$$

- Feedback from observed states

$$u = -L\hat{x}$$

- Controller transfer-function

$$U(s) = -F_y(s)Y(s) = -L(sI - A + BL + KC)^{-1}KY(s)$$

A solution: modified observer

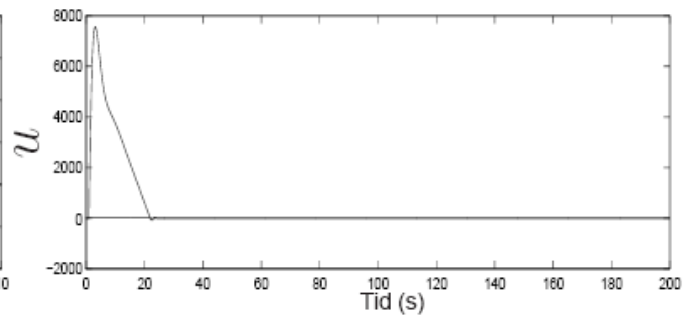
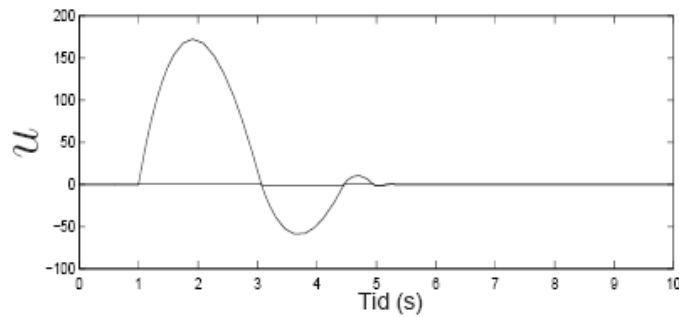
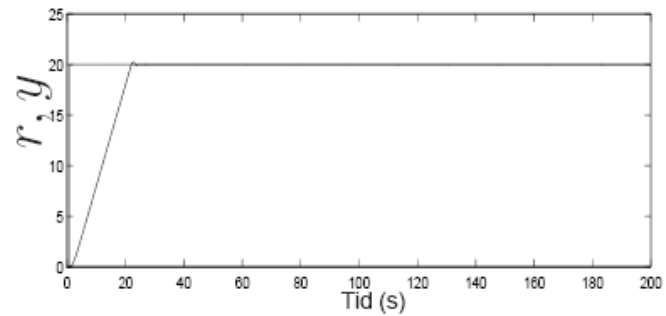
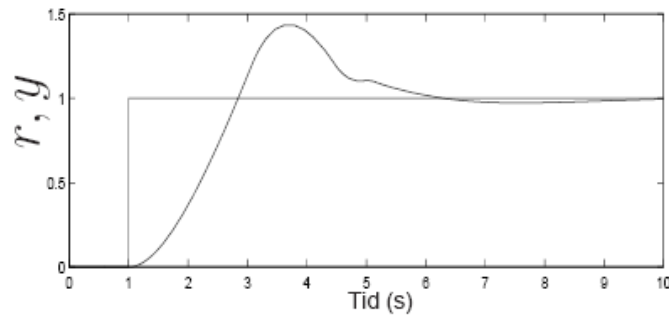
Observer should reflect true dynamics

$$\frac{d}{dt}\hat{x}(t) = A\hat{x}(t) + Bu_p(t) + K(y(t) - C\hat{x}(t))$$

The constrained (actually applied) input is used in observer

- a nonlinear observer!
- based on measuring the actual input or having a model of the constraint

Step responses with modified observer



Analysis: stability also in saturation

$$\begin{aligned}\frac{d}{dt}\hat{x}(t) &= A\hat{x}(t) + Bu_p(t) + K(y(t) - C\hat{x}(t)) = \\ &= (A - KC)\hat{x}(t) + Bu_p(t) + Ky(t)\end{aligned}$$

Controller transfer function

$$U(s) = -L(sI - A + KC)^{-1}KY(s) - L(sI - A + KC)^{-1}BU_p(s)$$

In saturation ($u < u_{\min}$ or $u > u_{\max}$), u_p is constant

Thus, in saturation, the controller dynamics is given by $A-KC$ whose eigenvalues are $-0.5446 \pm 0.7276i$, -1.2106 (i.e. stable)

This modification is known as *anti-reset windup*.

Interpretation: feedback from $u-u_p$

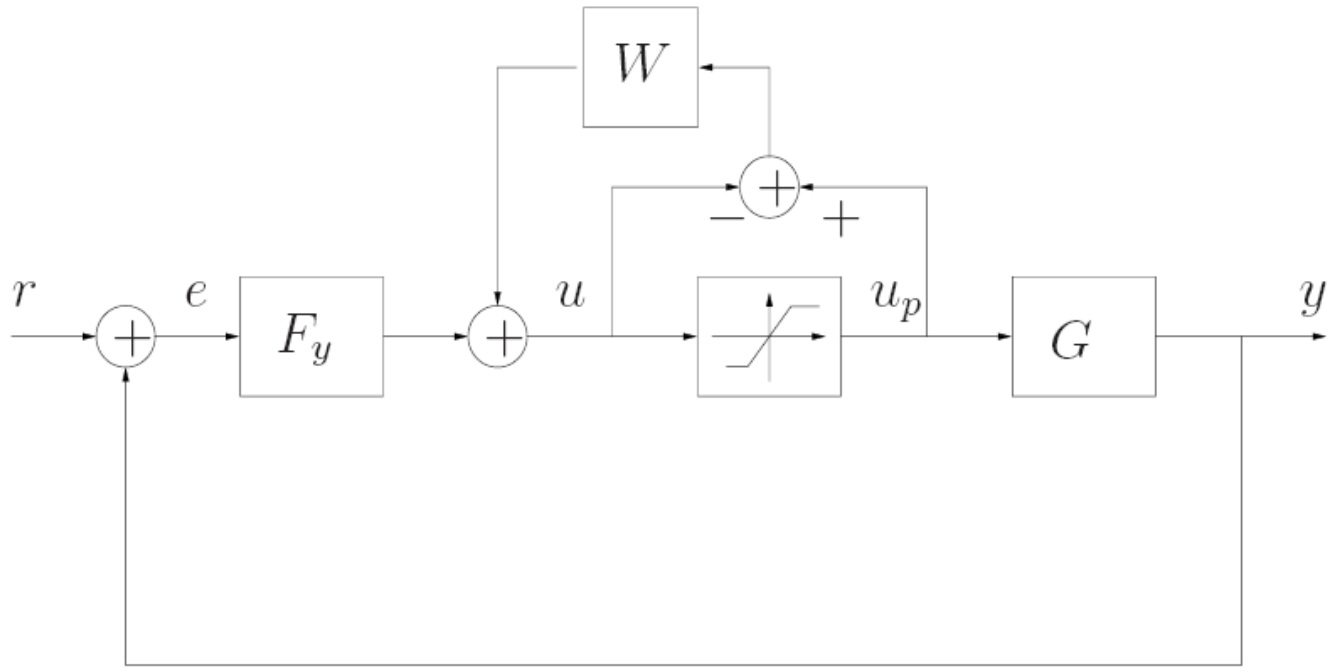
Write controller as

$$\begin{aligned}\frac{d}{dt}\hat{x}(t) &= A\hat{x}(t) + Bu_p(t) + K(y(t) - C\hat{x}(t)) = \\ &= (A - KC)\hat{x}(t) + B(u_p(t) + u(t) - u(t)) + Ky(t) = \\ &= (A - BL - KC)\hat{x}(t) + Ky(t) + B(u_p(t) - u(t))\end{aligned}$$

Taking Laplace transforms

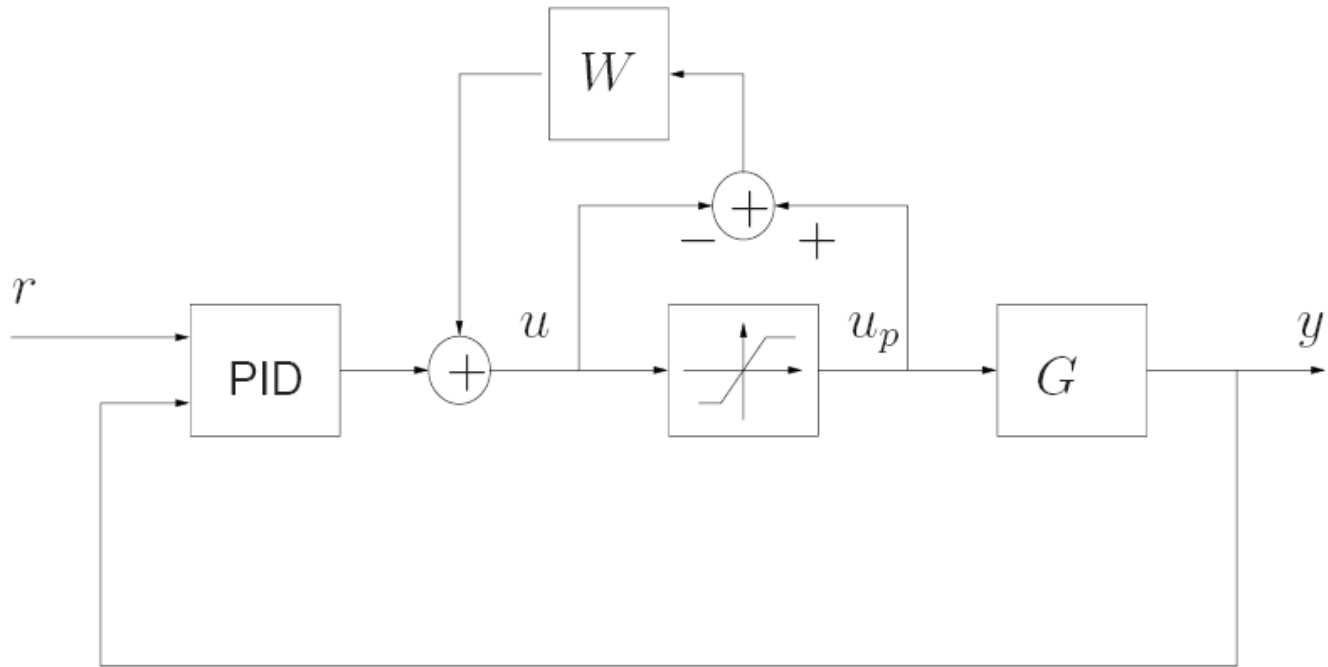
$$\begin{aligned}U(s) &= -L(sI - A + BL + KC)^{-1}KY(s) \\ &\quad - L(sI - A + BL + KC)^{-1}B(U_p(s) - U(s)) = \\ &= -F_y(s)Y(s) + W(s)(U_p(s) - U(s))\end{aligned}$$

In block diagram



- Anti-reset windup is based on tracking input

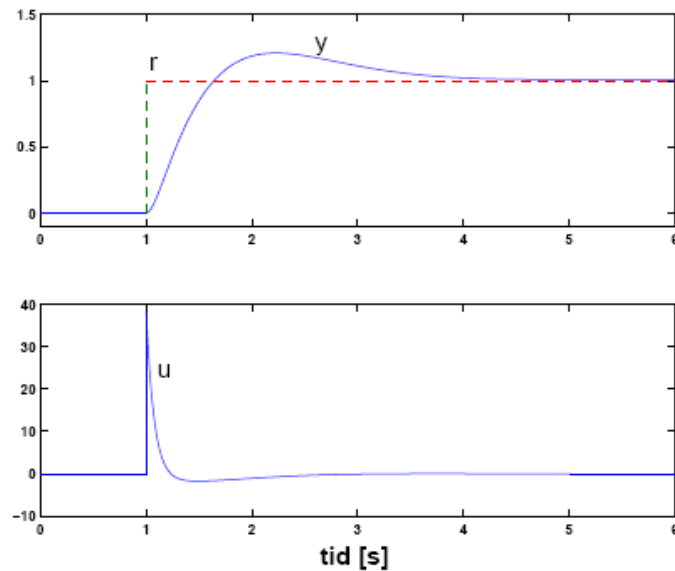
Application to PID controllers



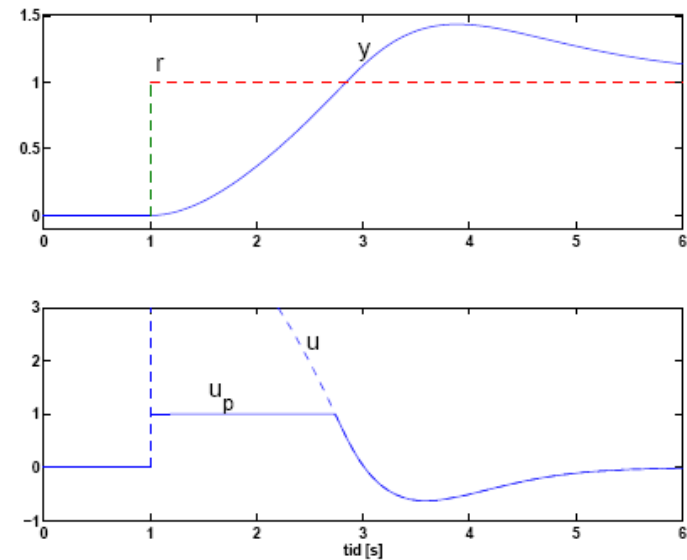
- Common choice: $W(s) = \frac{1}{sT_t}$

DC Servo under PID control

No constraint

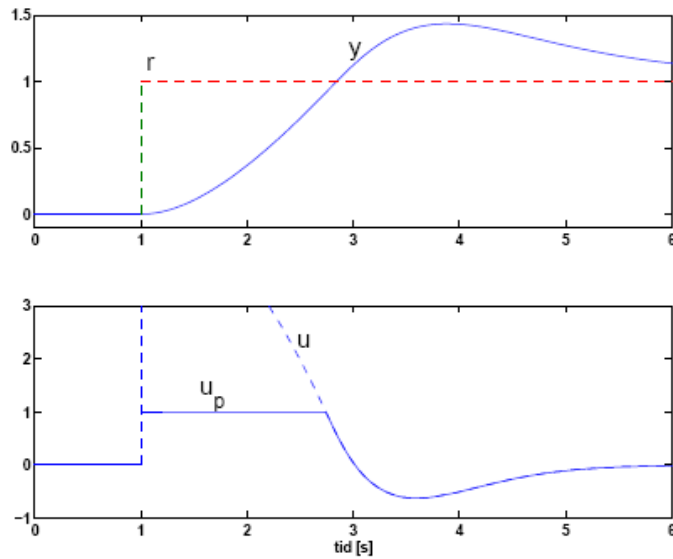


$u_p \in [-1, 1]$, no compensation

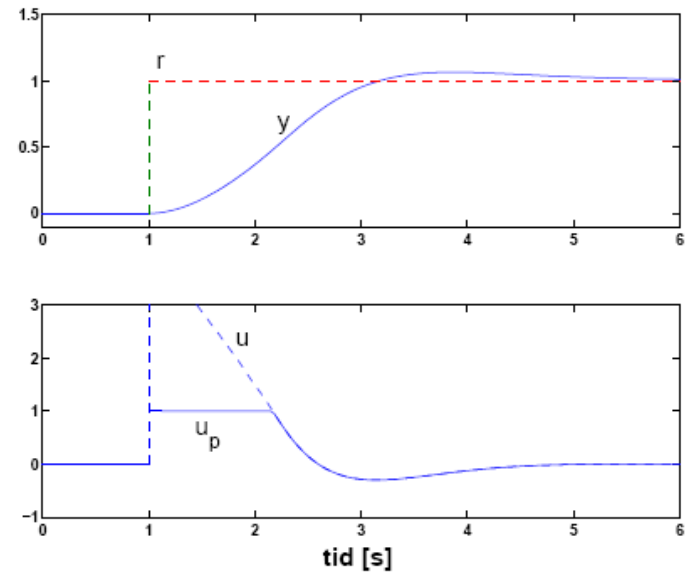


Servo: PID+Anti-reset Windup

$$T_t = 1000$$



$$T_t = T_i = 1.9$$



Summary

- Hard constraints: a nonlinearity essentially always present in real control systems
- Main problem: system drifts off when input in saturation
- Approaches to deal with hard constraints
 - constrained receding horizon LQG control (MPC)
 - anti-reset windup