

Introduction to Model Predictive Control

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Control theory

History

PID control

Model predictive control

Optimization background

classical control
(transfer functions)

PID control

cascade control

modern control
(state space models)
linear × nonlinear

Robust control

Optimal control

Predictive control

Intelligent control

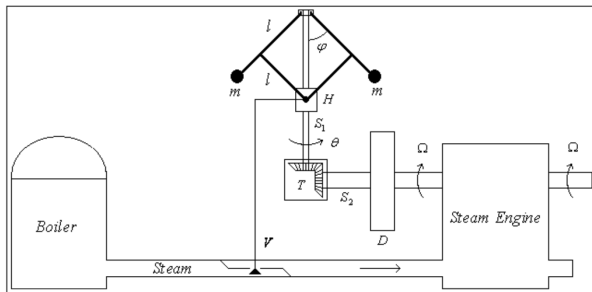
\hat{A}





Centrifugal Governor:

- ▶ Watt & Boulton in 1788
- ▶ First popular feedback controller
- ▶ **Speed controller**
- ▶ Negative feedback
- ▶ output too high -> decrease input



Formalization of the idea of control error suppression:

$$u_t = K_P e_t + K_I \int e_t dt + K_D \frac{de_t}{dt},$$

control gains, K_P , K_I , K_D .

P: Proportional:

- ▶ large - instability,
- ▶ small - rejection of small errors,

I: Integral:

- ▶ removes steady state error,
- ▶ causes overshoot,

D: differential

- ▶ improves settling time,
- ▶ rarely used (20%) - instability

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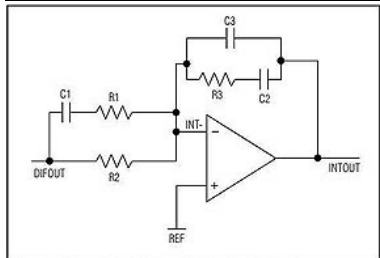
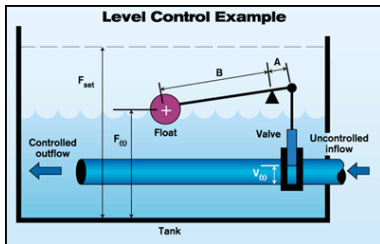
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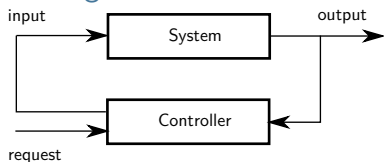
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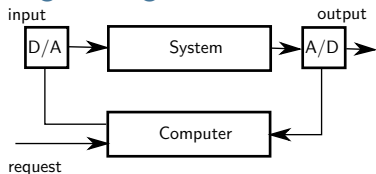


Proportional Integral Derivative Controller

Analog: direct connection

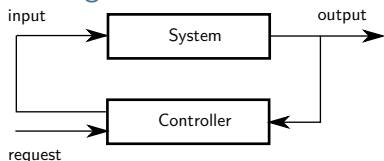


Digital: digital comm.

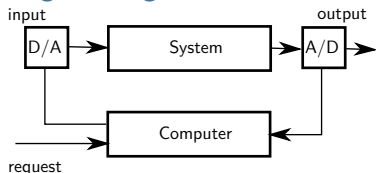


- ▶ far more flexible
- ▶ can we take advantage of it?

Analog: direct connection



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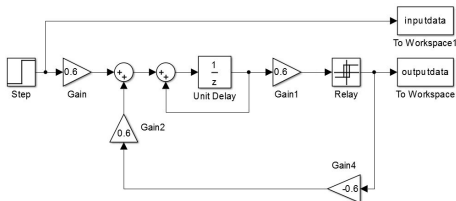
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Design change:

- ▶ Laplace transform \rightarrow Z transform
- ▶ Discrete time controllers:

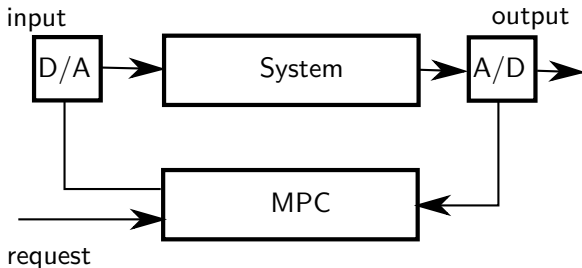
$$u_t = K_p e_t + K_I \sum_t e_t + K_D (e_t - e_{t-1}),$$

Block diagrams:



Classical analog intuition and thinking.

Block diagram:



- ▶ no z transform,
- ▶ no links,
- ▶ no intuition?
 - ▶ Different language (optimal control theory)
 - ▶ Unique solution

- ▶ cost function
- ▶ minimum, argument minima
- ▶ constraint
- ▶ finite \times continuous control set

Keep current i_t of series RL circuit close to the requested value i_t^* .
Closeness is measured by square distance

$$(i_t - i_t^*)^2$$

We can not change the measured value of the current, but the future. The model of RL circuit is:

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$$L \frac{di}{dt} + Ri = u \qquad i_{t+1} = \left(1 - \frac{\Delta t R}{L}\right) i_t + \frac{\Delta t}{L} u_t$$

simplified to

$$i_{t+1} = ai_t + u_t,$$

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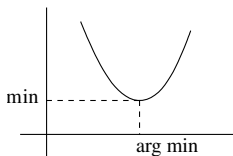
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Using the finite control set approach:

$$u_t^* = \arg \min_{u_t} ((i_{t+1} - i^*)^2),$$

$$s.t. : u_t \in \{-1, 0, 1\}, i_{t+1} = ai_t + u_t,$$

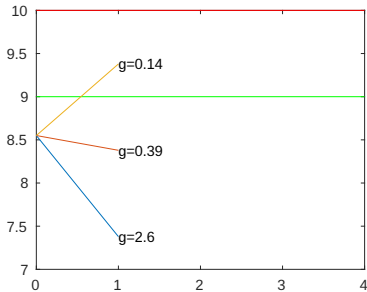


$$i_t = 8.55$$

$$i_{t+1} = 0.98i_t + u_t,$$

$$u_t = \arg \min_{\underbrace{g}_{g}} ((i_{t+1} - i^*)^2)$$

$$u_t \in \{-1, 0, 1\}$$



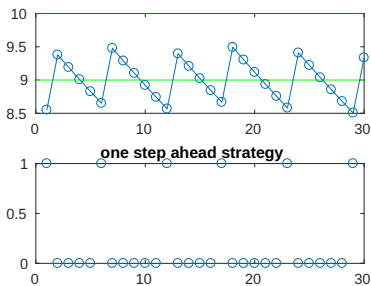
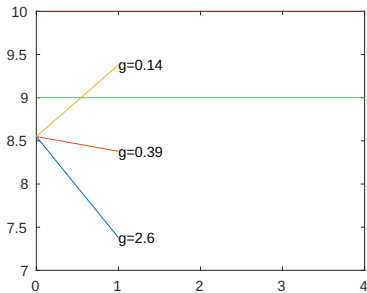
u_0	i_1	g_1
1	9.3	0.14
0	8.3	0.39
-1	7.3	2.6

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$$u_t \in \{-1, 0, 1\}$$



```
clear all
i(1) = 8.55; T = 30; u=-1:1;

for t=1:T
    % model predictive controller
    for j = 1:3
        ip = 0.98*i(t) + u(j);
        g(j) = (ip-9)^2;
    end
    [mi, imi]=min(g);
    U(t) = u(imj);

    % simulation of the real plant
    i(t+1) = 0.98*i(t) + u(imj);
end
```

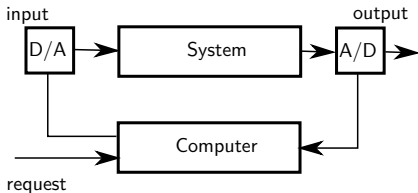
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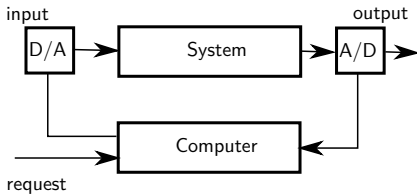
- ▶ Can we save some operations?

Zero-order hold



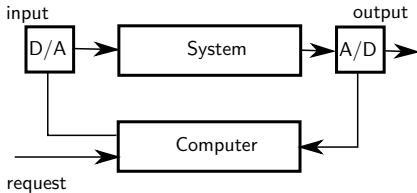
1. investigate spectrum of the generated current

Zero-order hold



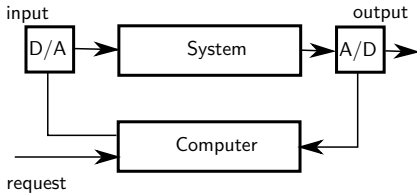
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2. try to track a sine wave

Zero-order hold



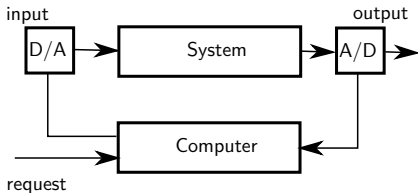
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3. Introduce current limit, $i_t \leq 9.2$

Zero-order hold



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4. change of switching frequency (penalization)

Zero-order hold

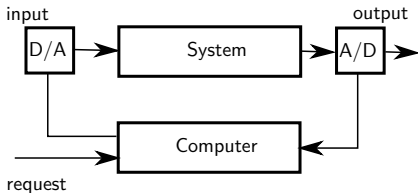


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$$u_t^* = \arg \min_{u_t} ((i_{t+1} - i^*)^2 + \alpha |u_t - u_{t-1}|),$$

Any control modification is achieved by change of model or cost.

Zero-order hold



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Any control modification is achieved by change of model or cost.

► can the cost be $(\omega - \omega^*)^2$?

Practicalities: time delay & delay compensation

- ▶ We can not compute the response arbitrarily fast
- ▶ When the controller computes action, the real system is still running

- ▶ We can not compute the response arbitrarily fast
- ▶ When the controller computes action, the real system is still running
- ▶ We need to assume that we compute future control action u_{t+1} that will have effect on the next current i_{t+2} .
 - ▶ The state i_{t+1} is computed using the model

$$i_{t+1} = ai_t + u_t,$$

where u_t is known from previous step.

- ▶ the optimization is performed for

$$\begin{aligned} u_{t+1}^* &= \arg \min_{u_{t+1}} ((i_{t+2} - i^*)^2), \\ \text{s.t. : } u_t &\in \{-1, 0, 1\}, i_{t+2} = ai_{t+1} + u_{t+1} \end{aligned}$$

Exercise: RL control with delay compensation

```
clear all
i(1) = 8.55; T = 30; u=-1:1;

for t=2:T
    % model predictive controller
    for j = 1:3
        ip = 0.98*i(t-1) + u(j);
        g(j) = (ip-9)^2;
    end
    [mi, imi]=min(g);
    U(t-1) = u(imj);

    % simulation of the real plant
    i(t) = 0.98*i(t-1) + u(imj);
end
```

- ▶ Implement with 1-step time delay compensation

- ▶ Why one-step-ahead may not be enough?

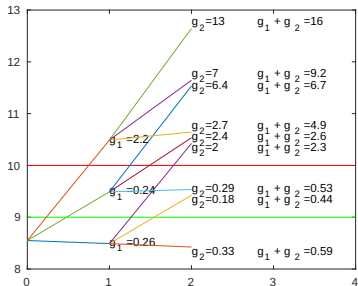
- ▶ Why one-step-ahead may not be enough?
 - ▶ delay
 - ▶ high order dynamics,
- ▶ Correct solution for additive cost via dynamic programming
- ▶ Computational burden: grows exponentially
- ▶ Approximate MPC
 - ▶ preselection strategies
 - ▶ sphere decoding

$$i_t = 8.55$$

$$i_{t+1} = 1.11i_t + u_t,$$

$$u_t, u_{t+1} = \arg \min((i_{t+1} - i^*)^2 + (i_{t+2} - i^*)^2)$$

$$u_t \in \{-1, 0, 1\}, u_{t+1} \in \{-1, 0, 1\}$$

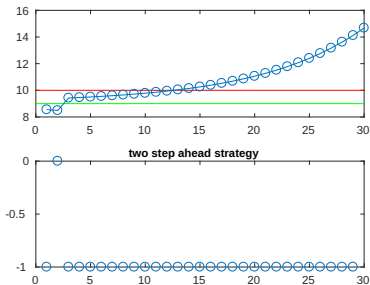
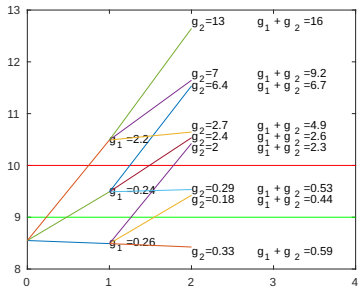


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Find a control action u^* from the set of admissible values \mathcal{U} minimizing cost

$$\begin{aligned} u_{t:t+h}^* &= \arg \min g(y^*, x_t) \\ \text{s.t. } x_{t+1} &= f(x_t, u_t), \\ u_\tau &\in \mathcal{U}, x_\tau \in \mathcal{X}, \forall \tau = t, \dots, t+h \end{aligned}$$

where:

- ▶ \mathcal{U} can be a finite set of states (FCS) or infinite (continuous) set,
- ▶ \mathcal{X} is a set of admissible states
- ▶ h is the prediction horizon.

Model in flux-linked d-q frame ($\Psi_{rq} = 0$)

$$\begin{aligned}\frac{di_{sd}}{dt} &= -\alpha i_{sd} + \omega_s \Psi i_{sq} + \beta \Psi_{rd} + \delta u_{sd}, \\ \frac{di_{sq}}{dt} &= -\omega_s \Psi i_{sd} - \mu i_{sd} - \eta \omega_s \Psi \Psi_{rd} + \delta u_{sq}, \\ \frac{d\Psi_{rd}}{dt} &= R_r \frac{L_m}{L_r} i_{sd} - \frac{R_r}{L_r} \Psi_{rd},\end{aligned}$$

with parameters $\lambda = L_{s\sigma} + L_{r\sigma} \frac{L_m}{L_r}$, $\alpha = \frac{R_s + R_r (L_m^2 / L_r^2)}{\lambda}$, $\beta = \frac{R_r (L_m / L_r^2)}{\lambda}$, $\delta = \frac{1}{\lambda}$, $\mu = \frac{R_s}{\lambda}$, $\eta = \frac{L_m / L_r}{\lambda}$, $L_r = L_m + L_{r\sigma}$ are transformations of physical parameters of the drive (stator resistance R_s , stator leakage inductance $L_{s\sigma}$, magnetizing inductance L_m , rotor inductance L_r , rotor leakage inductance $L_{r\sigma}$, and rotor resistance R_r), and $\omega_s \Psi$ is angular speed of the rotor magnetic flux vector Ψ_r with respect to stator coordinates.

Position of the rotor flux ϑ_r is computed using

$$\vartheta_r = p_p \vartheta_m + \int_t \frac{L_m}{T_r} \frac{i_{sq}}{\Psi_{rd}} dt, \quad (1)$$

where rotor time constant $T_r = L_r/R_r$, p_p is the number of pole pairs, and ϑ_m is the mechanical position of the rotor.

Cost function

$$g_{is}(x_{\text{pred}}, S_i) = |i_{sd}^* - \hat{i}_{sd,t+2}(S_i)|^2 + |i_{sq}^* - \hat{i}_{sq,t+2}(S_i)|^2, \quad (2)$$

where $S_i \in \mathcal{S} = \{++, ++-, +-+, -+++, +--, -+-, --+, ---\}$,
is the switching combination.

- ▶ define input voltage vectors $u_{sd}(S_i) =$
- ▶ predict currents i_{t+1}
- ▶ update state (flux magnitude and angle)