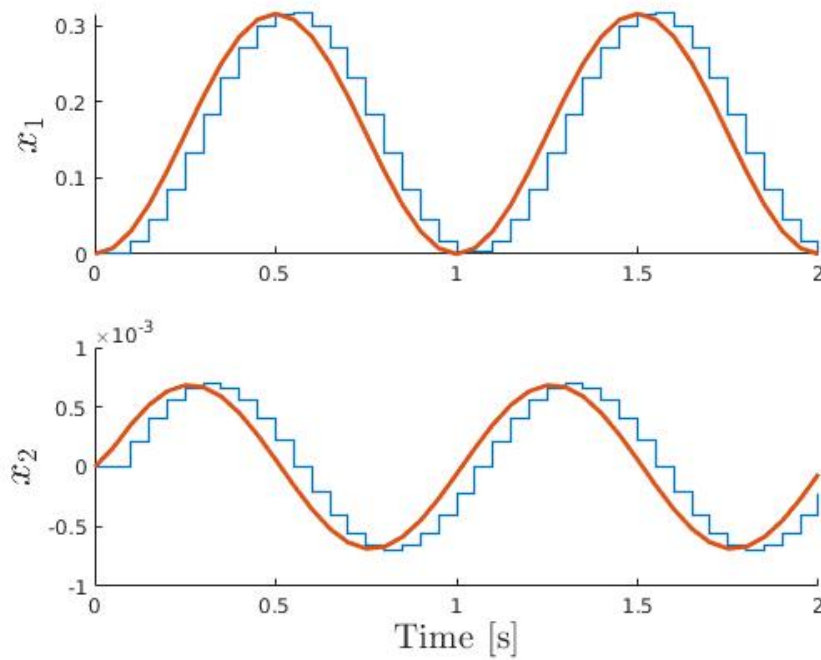


Homework 2 - Model Predictive Control

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1 Mini-Project - Elevator

In order to verify that the discrete states have the same physical meaning as their continuous counterparts, a simulation was done in order to compare both states after discretization for a sinusoidal control input.



The code to verify the discretization can be found here: <https://github.com/EriveltonGualter/ESC-794-Model-Predictive-Control/tree/master/HW2>.

1.1 Tuning Control System

In order to help the task to control the system, it was developed a interface, which is despited in the following figure, to accelerate and simplify the process to tune the parameters.

GUI_HW2

Select Code: Question 2 - Elevator Control

Question 1 Parameters

lamba: 2

Simulation Horizon: 8

Prediction Horizon: 5

Control Horizon: 4

Abs. Value of Control Input: 0.5

☐ Hold On ?

RUN

STATUS

Elevator Parameters

Jt: 20

a: 12

Rm: 0.1

R: 1

mc: 1000

mw: 750

g: 9.81

Max Volt.: 40

Max Acc.: 0.2

Max Vel.: 3

LambdA: 0.1

LambdU: 0.5

dUMAX: 1000

Floor: 30

ny: 10

nu: 9

N: 1250

Figure 1: Interface for Homework 2

After some tests varying the parameters of the MPC controller (λ and λ_u), it was clear that for the current specifications of the problem we ca not achieve it. In order to reach all the requirements, it is necessary to alter some of the constraints for example.

Next figure contains the results of the controller.

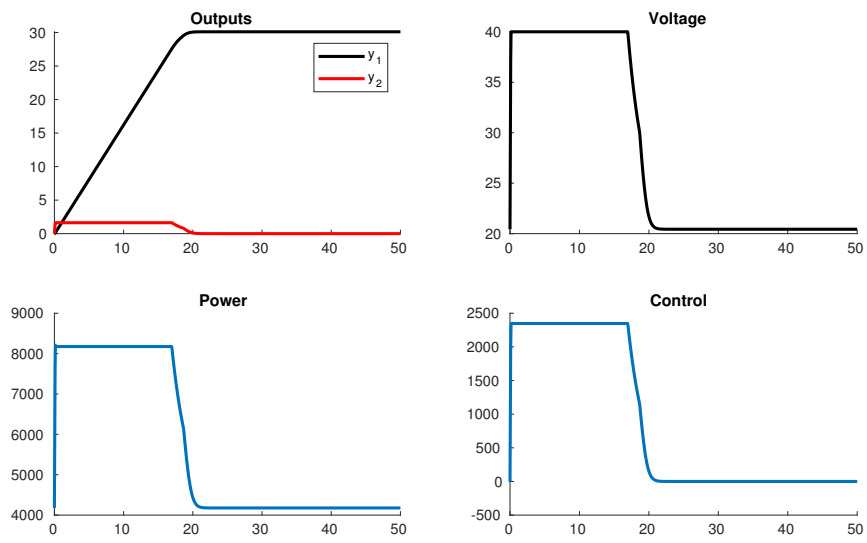


Figure 2: Results

References

1. <https://www.thyssenkruppelevator.com/Tools/energy-calculator>

Homework 2 - Question 1

1. For $x(k+1) = A_d x(k) + B_d u(k) + \delta$

$$y(k) = C x(k) + D u(k) \quad \text{where}$$

δ : constant disturbance . Obtain the optimisation cost function.

State Prediction: " $x_1 = x(k+1)$ and $x_0 = x(k)$ "

For $x_1 = A_d x_0 + B_d u_0 + \delta$, So

$$x_1 = A_d x_0 + B_d u_0$$

$$x_2 = A_d (A_d x_0 + B_d u_0) + B_d u_1 = A_d^2 x_0 + A_d B_d u_0 + B_d u_1$$

$$x_3 = A_d x_2 + B_d u_2 = A_d^3 x_0 + A_d^2 B_d u_0 + A_d B_d u_1 + B_d u_2$$

:

$$x_{ny} = A_d^{ny} x_0 + A_d^{ny-1} B_d u_0 + \dots + B_d u_{ny-1}$$

therefore:

$$\hat{x} = \begin{bmatrix} A_d \\ \vdots \\ A_d^{ny} \end{bmatrix} x_0 + \begin{bmatrix} B_d & \dots & 0 \\ \vdots & \ddots & \vdots \\ A_d^{ny-1} B_d & B_d & B_d \end{bmatrix} \hat{u}$$

$\begin{matrix} P_{xx} & H_x \\ nny \times 1 & nny \times nny \end{matrix}$

It results in $\hat{x} = P_{xx} x_0 + X_x \hat{u}$

Output Prediction: $y_i = C x_i + D u_i + \delta$

$$y_1 = C (A_d x_0 + B_d u_0) + D u_1 + \delta = C A_d x_0 + C B_d u_0 + D u_1 + \delta$$

$$y_2 = C A_d^2 x_0 + C A_d B_d u_0 + C B_d u_1 + D u_2 + \delta$$

$$y_{ny} = C A_d^{ny} x_0 + C A_d^{ny-1} B_d u_0 + \dots + C B_d u_{ny-1} + D u_{ny} + \delta$$

therefore $\hat{y} = P x_0 + H \hat{u} + \Delta$

Error Predictions : $\hat{e} = \hat{r} - \hat{y} = \hat{r} - (P\hat{x}_0 + H\hat{u} + \Delta)$

Object Function: $J = \sum_{i=1}^{N-1} e^T e + \lambda \hat{u}^T \hat{u} = \hat{e}^T \hat{e} + \lambda \hat{u}^T \hat{u}$

$$\begin{aligned} &= (\hat{r} - P\hat{x}_0 - H\hat{u} - \Delta)^T (\hat{r} - P\hat{x}_0 - H\hat{u} - \Delta) + \lambda \hat{u}^T \hat{u} \\ &= (\hat{r}^T - \hat{x}_0^T P^T - \hat{u}^T H^T - \Delta^T) (\hat{r} - P\hat{x}_0 - H\hat{u} - \Delta) + \lambda \hat{u}^T \hat{u} \\ &= \hat{r}^T \hat{r} - \hat{r}^T P\hat{x}_0 - \hat{r}^T H\hat{u} - \hat{r}^T \Delta - \hat{x}_0^T P^T \hat{r} + \hat{x}_0^T P^T P\hat{x}_0 + \\ &\quad + \hat{x}_0^T P^T H\hat{u} + \hat{x}_0^T P^T \Delta - \hat{u}^T H^T \hat{r} + \hat{u}^T H^T P\hat{x}_0 + \hat{u}^T H^T H\hat{u} + \\ &\quad \hat{u}^T H^T \Delta + \Delta^T \hat{r} + \Delta^T P\hat{x}_0 + \Delta^T H\hat{u} + \Delta^T \Delta + \lambda \hat{u}^T \hat{u} \end{aligned}$$

Note: $-\hat{r}^T H\hat{u} - \hat{u}^T H^T \hat{r} = -2\hat{r}^T H\hat{u}$
 $\hat{x}_0^T P^T H\hat{u} + \hat{u}^T H^T P\hat{x}_0 = 2\hat{x}_0^T P^T H\hat{u}$
 $\hat{u}^T H^T \Delta + \Delta^T H\hat{u} = 2\Delta^T H\hat{u}$

Therefore we can write as:

$$J = J_0 + \hat{u}^T (H^T H + \lambda I) \hat{u} + 2(\hat{x}_0^T P^T - \hat{r}^T + \Delta^T) H \hat{u}$$

$$\nabla J = \frac{\partial J}{\partial \hat{u}} = 2(H^T H + \lambda I) \hat{u} + 2(\hat{x}_0^T P^T - \hat{r}^T + \Delta^T) H = 0$$

Question 2

For running cost: $l = c^2 + \lambda P + \lambda u \Delta u^2$

Power: $P = VI = V \left(\frac{V - a \dot{y}/R}{R_m} \right)$ and $V = \frac{R_m}{aR} (u + R^2 m_{cw} g)$

First: $P = \frac{V^2 - V a \dot{y}/R}{R_m}$ (I)

For $V^2 \rightarrow V^2 = \frac{R_m^2}{(aR)^2} (u^2 + 2uR^2 m_{cw} g + R^4 m_{cw}^2 g^2)$ (II)

Replacing II and III to I:

$$P = \frac{R_m}{(aR)^2} (u^2 + 2uR^2 m_{cw} g + R^4 m_{cw}^2 g^2) - \frac{\dot{y}}{R^2} (u + R^2 m_{cw} g)$$

$$= \frac{R_m u^2}{a^2 R^2} + \frac{2 R_m u m_{cw} g}{a^2} + \frac{R^2 m_{cw}^2 g^2}{a^2} - \frac{\dot{y} u}{R^2} - \dot{y} m_{cw} g \quad (IV)$$

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Error: $\hat{e} = \hat{y}_f - \hat{y}_{pos} = \hat{y}_f - P_{pos} x_a - H_{pos} \Delta \hat{u} \quad (V)$
 where $x_a = \begin{bmatrix} \pi \\ u \end{bmatrix}$

Therefore replacing (IV) and (V) to the running cost

$$J = \underbrace{(\hat{y}_f - P_{pos} x_a - H_{pos} \Delta \hat{u})^T (\hat{y}_f - P_{pos} x_a - H_{pos} \Delta \hat{u})}_{\text{Term I}} + \dots$$

$$+ \lambda \underbrace{\left(\frac{R_m u^2}{a^2 R^2} + \frac{2 R_m u m_{cw} g}{a^2} + \frac{R^2 m_{cw}^2 g^2}{a^2} - \frac{\dot{y} u}{R^2} - \dot{y} m_{cw} g \right)}_{\text{Term II}} + \lambda u \Delta u^2$$

First, solving Term I:

$$= (\hat{y}_f^T - x_a^T P_{pos}^T - \Delta \hat{u}^T H_{pos}^T) (\hat{y}_f - P_{pos} x_a - H_{pos} \Delta \hat{u})$$

$$= \hat{y}_f^T \hat{y}_f - \hat{y}_f^T P_{pos} x_a - \hat{y}_f^T H_{pos} \Delta \hat{u} - x_a^T P_{pos}^T \hat{y}_f + x_a^T P_{pos}^T P_{pos} x_a$$

$$+ x_a^T P_{pos}^T H_{pos} \Delta \hat{u} - \Delta \hat{u}^T H_{pos}^T \hat{y}_f + \Delta \hat{u}^T H_{pos}^T P_{pos} x_a + \Delta \hat{u}^T H_{pos}^T H_{pos} \Delta \hat{u}$$

Note: $-\hat{y}_f^T H_{pos} \Delta \hat{u} - \Delta \hat{u}^T H_{pos}^T \hat{y}_f = -2 \hat{y}_f^T H_{pos} \Delta \hat{u}$

$x_a^T P_{pos}^T H_{pos} \Delta \hat{u} + \Delta \hat{u}^T H_{pos}^T P_{pos} x_a = 2 x_a^T P_{pos}^T H_{pos} \Delta \hat{u}$

Therefore, Term I can be simplified as:

$$\text{Term I} = \underbrace{(\text{Term I})}_{\text{constants}} + 2 \left(\mathbf{x}_a^T \mathbf{P}_{\text{pos}}^T - \hat{\mathbf{y}}_f^T \right) \mathbf{H}_{\text{pos}} \Delta \hat{\mathbf{u}} + \Delta \hat{\mathbf{u}}^T \mathbf{H}_{\text{pos}}^T \mathbf{H}_{\text{pos}} \Delta \hat{\mathbf{u}}$$

For "Term II", note we have the differentiation of $\mathbf{y} : \dot{\mathbf{y}}$, which correspond to the velocity state. So this can be written as:

$$\hat{\mathbf{y}}_{\text{vel}}^T = \mathbf{x}_a^T \mathbf{P}_{\text{vel}}^T + \Delta \hat{\mathbf{u}}^T \mathbf{H}_{\text{vel}}^T$$

or

$$\hat{\mathbf{y}}_{\text{vel}} = \mathbf{P}_{\text{vel}} \mathbf{x}_a + \mathbf{H}_{\text{vel}} \Delta \hat{\mathbf{u}}$$

Additionally, we can write $\hat{\mathbf{u}}$ as

$$\hat{\mathbf{u}} = \mathbf{P}_u \mathbf{x}_a + \mathbf{H}_u \Delta \hat{\mathbf{u}}$$

Before simplify the "term II" let's solve some of the subterms:

$$\begin{aligned} \hat{\mathbf{u}}^T \mathbf{u} &= (\mathbf{x}_a^T \mathbf{P}_u^T + \Delta \hat{\mathbf{u}}^T \mathbf{H}_u^T) (\mathbf{P}_u \mathbf{x}_a + \mathbf{H}_u \Delta \hat{\mathbf{u}}) \\ &= \mathbf{x}_a^T \mathbf{P}_u^T \mathbf{P}_u \mathbf{x}_a + \mathbf{x}_a^T \mathbf{P}_u^T \mathbf{H}_u \Delta \hat{\mathbf{u}} + \Delta \hat{\mathbf{u}}^T \mathbf{H}_u^T \mathbf{P}_u \mathbf{x}_a + \Delta \hat{\mathbf{u}}^T \mathbf{H}_u^T \mathbf{H}_u \Delta \hat{\mathbf{u}} \\ &= \underbrace{\mathbf{x}_a^T \mathbf{P}_u^T \mathbf{P}_u \mathbf{x}_a}_{\text{constants}} + 2 \mathbf{x}_a^T \mathbf{P}_u^T \mathbf{H}_u \Delta \hat{\mathbf{u}} + \Delta \hat{\mathbf{u}}^T \mathbf{H}_u^T \mathbf{H}_u \Delta \hat{\mathbf{u}} \end{aligned}$$

"Note we have a constant term here and others terms are in function of $\Delta \mathbf{u}$. Since we don't need the constants values for J, for next derivation it will be not performed"

$$\text{For } \frac{2 R_m m_{\text{ewg}} g}{a^2} \hat{\mathbf{u}} = \frac{2 R_m m_{\text{ewg}} g}{a^2} \left(\mathbf{P}_u \mathbf{x}_a + \mathbf{H}_u \Delta \hat{\mathbf{u}} \right)$$

constant

$$\text{For } \frac{\hat{\mathbf{y}}_{\text{vel}}^T \hat{\mathbf{u}}}{R^2} = \frac{1}{R^2} (\mathbf{x}_a^T \mathbf{P}_{\text{vel}}^T + \Delta \hat{\mathbf{u}}^T \mathbf{H}_{\text{vel}}^T) (\mathbf{P}_u \mathbf{x}_a + \mathbf{H}_u \Delta \hat{\mathbf{u}})$$

$$= \frac{1}{R^2} \left(\overset{\text{constant}}{\cancel{x_a^T P_{vel}^T P_u x_a}} + \cancel{x_a^T P_{vel}^T H_u \Delta \hat{u}} + \Delta \hat{u}^T H_{vel}^T P_u x_a + \Delta \hat{u}^T H_{vel}^T H_u \Delta \hat{u} \right)$$

$$\text{For } m_{cw} \hat{y}_{vel} = m_{cw} g \left(\overset{\text{constant}}{\cancel{P_{vel} x_a}} + H_{vel} \Delta \hat{u} \right)$$

Therefore, the second term can be written as:

$$\begin{aligned} \text{Term II} = & \left(\text{Term I} \right) + \lambda \left[\underset{\text{constants}}{\frac{R_m}{a^2 R^2}} \left(2 x_a^T P_u^T H_u \Delta \hat{u} + \Delta \hat{u}^T H_u^T H_u \Delta \hat{u} \right) + \dots \right. \\ & + \frac{2 R_m m_{cw} g H_u \Delta \hat{u}}{a^2} - \frac{x_a^T P_{vel}^T H_u \Delta \hat{u}}{R^2} - \frac{x_a^T P_u^T H_{vel} \Delta \hat{u}}{R^2} - \frac{\Delta \hat{u}^T H_{vel}^T H_u \Delta \hat{u}}{R^2} + \dots \\ & \left. - m_{cw} g H_{vel} \Delta \hat{u} \right] \end{aligned}$$

Therefore, cost function can be written as:

$$J = J_0 + \text{Term I} + \text{Term II} + \lambda_u \Delta u^T \Delta u$$

$$\begin{aligned} J = & J_0 + \Delta \hat{u}^T \left(H_{pos}^T H_{pos} + \frac{\lambda R_m}{a^2 R^2} H_u^T H_u - \frac{H_{vel}^T H_u}{R^2} + \lambda_u I \right) \Delta \hat{u} + \dots \\ & + \left[2 \left(x_a^T P_{pos}^T - \hat{y}_t^T \right) H_{pos} + \frac{2 \lambda R_m x_a^T P_u^T H_u \Delta \hat{u}}{a^2 R^2} + \frac{2 R_m m_{cw} g H_u \Delta \hat{u}}{a^2} \right. \\ & \left. - \frac{\lambda x_a^T P_{vel}^T H_u}{R^2} - \frac{\lambda x_a^T P_u^T H_{vel}}{R^2} - \lambda m_{cw} g H_{vel} \right] \Delta \hat{u} \end{aligned}$$