

Receding Horizon Control for MJS

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Given the control system

$$x_k = A_{k-1}x_{k-1} + B_{k-1}u_{k-1}$$

where u is dependent on x , we can recursively calculate the highest order term in terms of all the lower ones: $\forall k \in \mathbb{N}$

$$\begin{aligned}x_1 &= A_0x_0 + B_0u_0 \\x_2 &= A_1(A_0x_0 + B_0u_0) + B_1u_1 \\&= A_1A_0x_0 + A_1B_0u_0 + B_1u_1 \\x_3 &= A_2x_2 + B_2u_2 \\&= A_2(A_1A_0x_0 + A_1B_0u_0 + B_1u_1) + B_2u_2 \\&= A_2A_1A_0x_0 + A_2A_1B_0u_0 + A_2B_1u_1 + B_2u_2 \\&\vdots \\x_k &= (A_{k-1} \dots A_1A_0)x_0 + (A_{k-2} \dots A_2A_1)B_0u_0 + (A_{k-3} \dots A_3A_2)B_1u_1 + \dots \\&\quad + (A_{k-2}A_{k-1})B_{k-3}u_{k-3} + A_{k-1}B_{k-2}u_{k-2} + B_{k-1}u_{k-1}\end{aligned}$$

$$x_{k+1} = \left[\prod_{n=k}^0 A_n \right] x_0 + \sum_{j=1}^k \left[\prod_{n=k}^j A_n \right] B_{j-1}u_{j-1} + B_ku_k$$

So consider the following cost function:

$$\begin{aligned}J(x_0) &= \sum_{k=1}^L x_k^T R_k x_k + \sum_{k=0}^{L-1} u_k^T Q_k u_k \\J(x_0) &= \sum_{k=1}^L x_k^T R_k x_k + u_{k-1}^T Q_{k-1} u_{k-1}\end{aligned}$$

Notice that there is no well defined R_0 or Q_L , and x_0 and u_L are not taken into account.

To minimize the cost function with respect to the control vector, u , we take it's derivative with respect to the highest order term, since it has no other variables that depend on it. So then, because

$$\frac{\partial}{\partial u_{k-1}} [x_k] = \frac{\partial}{\partial u_{k-1}} [A_{k-1}x_{k-1} + B_{k-1}u_{k-1}] = B_{k-1}$$

the chain rule gives

$$\frac{\partial}{\partial u_{L-1}} [J(x_0)] = 2 (x_L^T R_L B_{L-1} + u_{L-1}^T Q_{L-1})$$

Set this equal to zero to find critical points.

$$2 (x_L^T R_L B_{L-1} + u_{L-1}^T Q_{L-1}) = 0$$

$$x_L^T R_L B_{L-1} + u_{L-1}^T Q_{L-1} = 0$$

$$u_{L-1}^T Q_{L-1} = -x_L^T R_L B_{L-1}$$

$$u_{L-1}^T = -x_L^T R_L B_{L-1} Q_{L-1}^{-1}$$

$$u_{L-1} = -Q_{L-1}^{-1} B_{L-1}^T R_L x_L$$

$$u_{L-1} = -Q_{L-1}^{-1} B_{L-1}^T R_L (A_{L-1}x_{L-1} + B_{L-1}u_{L-1})$$

$$Q_{L-1}u_{L-1} = -B_{L-1}^T R_L A_{L-1}x_{L-1} - B_{L-1}^T R_L B_{L-1}u_{L-1}$$

$$B_{L-1}^T R_L B_{L-1}u_{L-1} + Q_{L-1}u_{L-1} = -B_{L-1}^T R_L A_{L-1}x_{L-1}$$

$$(B_{L-1}^T R_L B_{L-1} + Q_{L-1})u_{L-1} = -B_{L-1}^T R_L A_{L-1}x_{L-1}$$

So assuming that $(B_{L-1}^T R_L B_{L-1} + Q_{L-1})$ is invertible for all $L \in \mathbb{N}$, then

$$u_{L-1} = -(B_{L-1}^T R_L B_{L-1} + Q_{L-1})^{-1} B_{L-1}^T R_L A_{L-1}x_{L-1}$$

Now equivalently, for all $k \in \mathbb{Z}^+$, we have assumed that $(B_k^T R_{k+1} B_k + Q_k)$ is invertible. So call

$$G_k = -(B_k^T R_{k+1} B_k + Q_k)^{-1} B_k^T R_{k+1} A_k$$

so that

$$u_{L-1} = G_{L-1}x_{L-1}$$

Starting with $L = 1$ gives

$$u_0 = G_0x_0$$

and finding the next predicted position:

$$x_1 = A_0x_0 + B_0u_0$$

$$x_1 = A_0x_0 + B_0G_0x_0$$

$$x_1 = (A_0 + B_0G_0)x_0$$

Increase the prediction distance to $L = 2$ and reuse the prediction for x_1 to solve for the next control:

$$\begin{aligned} u_1 &= G_1 x_1 \\ u_1 &= G_1 (A_0 + B_0 G_0) x_0 \end{aligned}$$

Predict forward one more time to $L = 3$ and the pattern will emerge:

$$\begin{aligned} x_2 &= A_1 x_1 + B_1 u_1 \\ x_2 &= A_1 (A_0 + B_0 G_0) x_0 + B_1 G_1 (A_0 + B_0 G_0) x_0 \\ x_2 &= (A_1 + B_1 G_1) (A_0 + B_0 G_0) x_0 \end{aligned}$$

solving for u_2 just adds a G_2 term on the front.

$$\begin{aligned} u_2 &= G_2 x_2 \\ u_2 &= G_2 (A_1 + B_1 G_1) (A_0 + B_0 G_0) x_0 \end{aligned}$$

Theorem 0.1. For any $k \in \mathbb{N}$, $k \leq L$, if the predictions for x_1, \dots, x_{k-1} are used recursively, and the control vector u_k is found by minimizing the cost function $J(x_0)$ at each step, then

$$x_k = \left[\prod_{n=k-1}^0 A_n + B_n G_n \right] x_0$$

Proof. Notice that for $k \in \mathbb{N}$, if $k = 1$, then as noted above:

$$x_1 = (A_0 + B_0 G_0) x_0$$

Assume that for $k \geq 1$,

$$x_k = \left[\prod_{n=k-1}^0 A_n + B_n G_n \right] x_0$$

Then by the derived control formula above: $\forall k \in \mathbb{N}$

$$u_{k-1} = G_{k-1} x_{k-1}$$

and using the original system equation, it follows that,

$$\begin{aligned} x_{k+1} &= A_k x_k + B_k u_k \\ x_{k+1} &= A_k x_k + B_k G_k x_k \\ x_{k+1} &= (A_k + B_k G_k) x_k \\ x_{k+1} &= (A_k + B_k G_k) \left[\prod_{n=k-1}^0 A_n + B_n G_n \right] x_0 \\ x_{k+1} &= \left[\prod_{n=k}^0 A_n + B_n G_n \right] x_0 \end{aligned}$$

So by induction, the formula must be valid for all $k \in \mathbb{N}$

□

Corollary 0.1. For any $k, L \in \mathbb{N}$, $L > 2$, and $k < L$; if the predictions for x_1, \dots, x_{k-1} are used recursively, and the control vector u_k is found by minimizing the cost function $J(x_0)$ at each step, then

$$u_k = G_k \left[\prod_{n=k-1}^0 A_n + B_n G_n \right] x_0$$

Proof. This follows directly from the theorem, and the derived control formula above (after shifting up one index): $\forall k \in \mathbb{N}$

$$u_k = G_k x_k$$

$$u_k = G_k \left[\prod_{n=k-1}^0 A_n + B_n G_n \right] x_0$$

□