

元月 6 月 2024.

Differentiated MHT using PVP developed for UC problem.

$$\min_{X, W} \left[ \frac{1}{2} \text{Tr} \|X_0 - \hat{X}_0\|_P^2 + \text{Tr} \sum_{k=0}^{N-1} \left( \frac{1}{2} \begin{bmatrix} X_k^T \\ W_k^T \end{bmatrix} \begin{bmatrix} L_k^{xx} & L_k^{xw} \\ L_k^{wx} & L_k^{ww} \end{bmatrix} \begin{bmatrix} X_k \\ W_k \end{bmatrix} + \frac{1}{2} \text{Tr} X_k^T L_N^{xx} X_k + \begin{bmatrix} L_N^{x0} \\ L_N^{w0} \end{bmatrix}^T \begin{bmatrix} X_k \\ W_k \end{bmatrix} \right) + \text{Tr} \left( \frac{1}{2} X_N^T L_N^{xx} X_N + \begin{bmatrix} L_N^{x0} \\ L_N^{w0} \end{bmatrix}^T \begin{bmatrix} X_N \\ W_N \end{bmatrix} \right) \right]$$

$$\text{s.t. } X_{k+1} = F_k X_k + G_k W_k, \quad k=0, \dots, N-1.$$

$$L = J + \sum_{k=0}^{N-1} \Lambda_{k+1}^T (F_k X_k + G_k W_k - X_{k+1})$$

Compared to the initial state in control, this arrival cost can be viewed as a soft penalty on the initial state.

$$\nabla_{X_0} L = P X_0 - P \hat{X}_0 + \bar{L}_0^{xx} X_0 + \bar{L}_0^{xw} W_0 + \bar{L}_0^{w0} + F_0^T \Lambda_1 = 0. \text{ boundary } 0.$$

$$\nabla_{X_k} L = \bar{L}_k^{xx} X_k + \bar{L}_k^{xw} W_k + \bar{L}_k^{w0} + F_k^T \Lambda_{k+1} - \Lambda_k = 0, \quad k=1, \dots, N-1.$$

$$\nabla_{W_k} L = \bar{L}_k^{wx} X_k + \bar{L}_k^{ww} W_k + \bar{L}_k^{w0} + G_k^T \Lambda_{k+1} = 0, \quad k=0, \dots, N-1.$$

$$\nabla_{\Lambda_k} L = F_k X_k + G_k W_k - X_{k+1} = 0.$$

$$\nabla_{X_N} L = \bar{L}_N^{xx} X_N + \bar{L}_N^{x0} - \Lambda_N = 0. \text{ boundary } N$$

$$\Lambda_N = \underbrace{\bar{L}_N^{xx} X_N}_{P_N} + \underbrace{\bar{L}_N^{x0}}_{O_N}, \quad \Lambda_k = P_k X_k + O_k, \quad k=1, \dots, N-1$$

$$W_k = -(\bar{L}_k)^{-1} (\bar{L}_k^{wx} X_k + G_k^T \Lambda_{k+1} + \bar{L}_k^{w0})$$

$$\begin{aligned} \Lambda_k &= \bar{L}_k^{xx} X_k - \bar{L}_k^{xw} (\bar{L}_k)^{-1} (\bar{L}_k^{wx} X_k + G_k^T \Lambda_{k+1} + \bar{L}_k^{w0}) + \bar{L}_k^{w0} + F_k^T \Lambda_{k+1} \\ &= \underbrace{(\bar{L}_k^{xx} - \bar{L}_k^{xw} (\bar{L}_k)^{-1} \bar{L}_k^{wx})}_{S_k} X_k + \underbrace{(F_k^T - \bar{L}_k^{xw} (\bar{L}_k)^{-1} G_k^T)}_{\bar{F}_k^T} \Lambda_{k+1} + \underbrace{(\bar{L}_k^{w0} - \bar{L}_k^{xw} (\bar{L}_k)^{-1} \bar{L}_k^{w0})}_{T_k} \end{aligned}$$

$$\Lambda_k = S_k X_k + \bar{F}_k^T \Lambda_{k+1} + T_k$$

$$\begin{aligned} X_{k+1} &= F_k X_k - G_k (\bar{L}_k)^{-1} \bar{L}_k^{wx} X_k - \underbrace{G_k (\bar{L}_k)^{-1} G_k^T}_{M_k} \Lambda_{k+1} - \underbrace{G_k (\bar{L}_k)^{-1} \bar{L}_k^{w0}}_{N_k} \\ &= \bar{F}_k X_k - M_k \Lambda_{k+1} - N_k \end{aligned}$$

$$X_{k+1} = \bar{F}_k X_k - M_k P_{k+1} X_{k+1} - M_k O_{k+1} - N_k$$

$$(I + M_k P_{k+1}) X_{k+1} = \bar{F}_k X_k - M_k O_{k+1} - N_k \quad X_{k+1} = (I + M_k P_{k+1})^{-1} (\bar{F}_k X_k - M_k O_{k+1} - N_k) \quad k=0, \dots, N-1.$$

$$W_k = -(\bar{L}_k)^{-1} (\bar{L}_k^{wx} X_k + G_k^T P_{k+1} X_{k+1} + G_k^T O_{k+1} + \bar{L}_k^{w0})$$

$$= -(\bar{L}_k)^{-1} (\bar{L}_k^{wx} X_k + G_k^T P_{k+1} (I + M_k P_{k+1})^{-1} (\bar{F}_k X_k - M_k O_{k+1} - N_k) + G_k^T O_{k+1} + \bar{L}_k^{w0})$$

$$\Lambda_k = S_k X_k + \bar{F}_k^T P_{k+1} X_{k+1} + \bar{F}_k^T O_{k+1} + T_k$$

Bingheng Wang

$$= S_k X_k + \bar{F}_k^T P_{k+1} (I + M_k P_{k+1})^{-1} \bar{F}_k X_k - \bar{F}_k^T P_{k+1} (M_k O_{k+1} + N_k) + \bar{F}_k^T O_{k+1} + T_k$$

$$= \underbrace{\left[ S_k + \bar{F}_k^T P_{k+1} (I + M_k P_{k+1})^{-1} \bar{F}_k \right]}_{P_k} X_k + \underbrace{\left[ T_k + \bar{F}_k^T O_{k+1} - \bar{F}_k^T P_{k+1} (M_k O_{k+1} + N_k) \right]}_{O_k}$$

$$\Lambda_1 = P_1 X_1 + O_1$$

Bingheng Wang



$$\begin{aligned} W_0 &= -(\bar{L}_0)^{-1} (\bar{L}_0^{\text{xx}} X_0 + G_0^T P_1 (I + M_0 P_1)^{-1} (\bar{F}_0 X_0 - M_0 D_1 - N_0) + G_0^T D_1 + \bar{L}_0^{\text{w0}}) \\ &= -(\bar{L}_0)^{-1} \left[ (\bar{L}_0^{\text{xx}} + G_0^T P_1 (I + M_0 P_1)^{-1} \bar{F}_0) X_0 - G_0^T P_1 (I + M_0 P_1)^{-1} (M_0 D_1 + N_0) + G_0^T D_1 + \bar{L}_0^{\text{w0}} \right] \\ &= -(\bar{L}_0)^{-1} (\bar{L}_0^{\text{xx}} + G_0^T P_1 (I + M_0 P_1)^{-1} \bar{F}_0) X_0 + (\bar{L}_0)^{-1} G_0^T P_1 (I + M_0 P_1)^{-1} (M_0 D_1 + N_0) - (\bar{L}_0)^{-1} (G_0^T D_1 + \bar{L}_0^{\text{w0}}) \end{aligned}$$

$$\Lambda_1 = P_1 X_1 + D_1$$

$X_0$  is unknown and needs to be solved from  $\frac{\partial L}{\partial X_0}$

( $X_0 = 0$  holds for optimal control)

$$= P_1 X_1 = (I + M_0 P_1)^{-1} (\bar{F}_0 X_0 - M_0 D_1 - N_0)$$

$$\Lambda_1 = P_1 (I + M_0 P_1)^{-1} \bar{F}_0 X_0 - P_1 (I + M_0 P_1)^{-1} (M_0 D_1 + N_0) + D_1$$

~~Since  $X_0 = X(0) \Rightarrow \frac{\partial X_0}{\partial \theta} = 0$~~   
because  $X_0 = X(0) \Rightarrow \frac{\partial X_0}{\partial \theta} = 0$   
constant

$$\bar{X}_0 = \frac{\partial \bar{X}_0}{\partial \theta}$$

Below is the solution of  $X_0$ .

$$\nabla_{X_0} L = P X_0 - P \hat{X}_0 + \bar{L}_0^{\text{xx}} X_0 + \bar{L}_0^{\text{x0}} W_0 + \bar{L}_0^{\text{w0}} + \bar{F}_0^T \Lambda_1 = 0.$$

$$= \underbrace{(P + \bar{L}_0^{\text{xx}})}_{\bar{L}_0^{\text{xx}}} X_0 - P \hat{X}_0 + \bar{L}_0^{\text{x0}} W_0 + \bar{L}_0^{\text{w0}} + \bar{F}_0^T \Lambda_1 = 0. \quad \begin{array}{l} \rightarrow \text{a function of } X_0. \\ W_0 \text{ is a function of } X_0. \\ X_1 \text{ is a function of } X_0. \end{array}$$

$$\begin{aligned} &= \bar{L}_0^{\text{xx}} X_0 - P \hat{X}_0 - \bar{L}_0^{\text{x0}} (\bar{L}_0)^{-1} (\bar{L}_0^{\text{xx}} X_0 + G_0^T P_1 (I + M_0 P_1)^{-1} \bar{F}_0 X_0 - G_0^T P_1 (I + M_0 P_1)^{-1} (M_0 D_1 + N_0) - (\bar{L}_0)^{-1} (G_0^T D_1 + \bar{L}_0^{\text{w0}})) \\ &\quad + \bar{L}_0^{\text{w0}} + \bar{F}_0^T P_1 (I + M_0 P_1)^{-1} \bar{F}_0 X_0 - \bar{F}_0^T P_1 (I + M_0 P_1)^{-1} (M_0 D_1 + N_0) + \bar{F}_0^T D_1 = 0 \end{aligned}$$

We can solve  $X_0$  from the above equation.

$$\bar{L}_0^{\text{xx}} X_0 - \bar{L}_0^{\text{x0}} (\bar{L}_0)^{-1} \bar{L}_0^{\text{xx}} X_0 - \bar{L}_0^{\text{x0}} (\bar{L}_0)^{-1} G_0^T P_1 (I + M_0 P_1)^{-1} \bar{F}_0 X_0 + \bar{F}_0^T P_1 (I + M_0 P_1)^{-1} \bar{F}_0 X_0 = \dots$$

$$\underbrace{(\bar{L}_0^{\text{xx}} - \bar{L}_0^{\text{x0}} (\bar{L}_0)^{-1} \bar{L}_0^{\text{xx}})}_{\bar{S}_0} X_0 + \underbrace{(\bar{F}_0^T - \bar{L}_0^{\text{x0}} (\bar{L}_0)^{-1} G_0^T P_1 (I + M_0 P_1)^{-1} \bar{F}_0)}_{\bar{F}_0} X_0 = \dots$$

$$\bar{F}_0 = \bar{F}_0 - \bar{L}_0^{\text{x0}} (\bar{L}_0)^{-1} G_0^T P_1 (I + M_0 P_1)^{-1} \bar{F}_0$$

$$\bar{S}_0 X_0 + \bar{F}_0^T P_1 (I + M_0 P_1)^{-1} \bar{F}_0 X_0 = \dots$$

$$[\bar{S}_0 + \bar{F}_0^T P_1 (I + M_0 P_1)^{-1} \bar{F}_0] X_0 = \dots$$

Summary: We can solve the auxiliary linear MHE using PDP-based method but with some modifications, e.g., solving for  $X_0$ !

Such a PDP-based method, however, is inconsistent with the continuous-time linear MHE where  $\lambda_T = 0$ .

Follow the PDP method

state:  $\Lambda_k \quad k=1, \dots, N \quad \Lambda_{k+1}^T = (F_k X_k + G_k u_k - X_{k+1})$

with  $\Lambda_N = \bar{L}_N^{\text{xx}} X_N + \bar{L}_N^{\text{x0}}$

$P_N$  from the terminal cost.

We need to solve for  $X_0$ ,

so direct application of PDP is impossible.

inconsistent with the continuous-time MHE

if  $N=0$ , then only  $\Lambda_1$  remains, in this case,

how can we determine its value?

There are two unknown variables, but only one equation!

$$\nabla_{X_0} L = P X_0 - P \hat{X}_0 + \bar{L}_0^{\text{xx}} X_0 + \bar{L}_0^{\text{x0}} + \bar{F}_0^T \Lambda_1 = 0.$$

Use Kalman-filter

co-state:  $\Lambda_k \quad k=0, \dots, N-1 \quad \Lambda_{k+1}^T = (X_{k+1} - F_k X_k - G_k u_k)$

with  $\Lambda_N = 0$

$P_0$  is determined by the weight matrix in the control cost

(We also need to solve for  $X_0$  using a Kalman-filter and  $\Lambda_0$ )

consistent with the continuous-time MHE, as both of terminal co-states are zero!

Explore the solution structure.

When  $N=0$ , there is not such an issue, as  $\Lambda_0 = 0$  by definition. So  $X_0$  can be easily obtained from its KKT condition:

$$\nabla_{X_0} L = P X_0 - P \hat{X}_0 + \bar{L}_0^{\text{xx}} X_0 + \bar{L}_0^{\text{x0}} = 0.$$