Cheetah MPC

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1 Conventions

Notation	
\overline{i}	ith leg
$\mathcal B$	left subscript for body coordinate system
$[\mathbf{x}]_{\times} \in \mathbb{R}^{3 \times 3}$	skew-symmetric matrix, $[\mathbf{x}] \times \mathbf{y} = \mathbf{x} \times \mathbf{y}$
$\mathbf{\Theta} = [\phi, \theta, \psi]^T$	orientation as ZYX Euler angles
$\mathbf{J}_i \in \mathbb{R}^{3 imes 3}$	foot Jacobain
$oldsymbol{\Lambda}_i \in \mathbb{R}^{3 imes 3}$	operational space inertia matrix
$\mathbf{K}_p, \mathbf{K}_d \in \mathbb{R}^{3 imes 3}$	diagonal positive definite pd gain matrices

2 State Space Model

$$\frac{\mathrm{d}}{\mathrm{d}t}\hat{\mathbf{\Theta}} = R_z(\psi)\hat{\mathbf{\Theta}} \tag{1}$$

$$\frac{\mathrm{d}}{\mathrm{d}t}\hat{\mathbf{p}} = \hat{\dot{\mathbf{p}}} \tag{2}$$

$$\frac{\mathrm{d}}{\mathrm{d}t}\hat{\omega} = \hat{\mathbf{I}}^{-1} \sum_{i=1}^{n} \mathbf{r}_{i} \times \mathbf{f}_{i} = \hat{\mathbf{I}}^{-1} \left(\left[\mathbf{r}_{1} \right]_{\times} \mathbf{f}_{1} + \dots + \left[\mathbf{r}_{n} \right]_{\times} \mathbf{f}_{n} \right)$$
(3)

$$\frac{\mathrm{d}}{\mathrm{d}t}\hat{\mathbf{p}} = \frac{\sum_{i=1}^{n} \mathbf{f}_{i}}{m} - \mathbf{g} = \frac{\mathbf{f}_{1} + \dots + \mathbf{f}_{n}}{m} - \mathbf{g}$$
(4)

$$\frac{\mathrm{d}}{\mathrm{d}t} \begin{bmatrix} \hat{\mathbf{\Theta}} \\ \hat{\mathbf{p}} \\ \hat{\omega} \\ \hat{\mathbf{p}} \\ \mathbf{g} \end{bmatrix} = \begin{bmatrix} \mathbf{0}_{3} & \mathbf{0}_{3} & \mathbf{R}_{z}(\psi) & \mathbf{0}_{3} & \mathbf{0}_{3\times1} \\ \mathbf{0}_{3} & \mathbf{0}_{3} & \mathbf{0}_{3} & \mathbf{I}_{3} & \mathbf{0}_{3\times1} \\ \mathbf{0}_{3} & \mathbf{0}_{3} & \mathbf{0}_{3} & \mathbf{0}_{3} & \mathbf{0}_{3\times1} \\ \mathbf{0}_{3} & \mathbf{0}_{3} & \mathbf{0}_{3} & \mathbf{0}_{3} & [0 \ 0 \ 1]^{T} \end{bmatrix} \begin{bmatrix} \hat{\mathbf{\Theta}} \\ \hat{\mathbf{p}} \\ \hat{\omega} \\ \hat{\mathbf{p}} \\ \mathbf{g} \end{bmatrix}$$

$$\begin{bmatrix} \mathbf{0}_{3} & \cdots & \mathbf{0}_{3} \\ \mathbf{0}_{2} & \cdots & \mathbf{0}_{2} \end{bmatrix} \begin{bmatrix} \mathbf{f}_{1} \end{bmatrix}$$

$$+ \left[egin{array}{cccc} \mathbf{0}_3 & \cdots & \mathbf{0}_3 \ \mathbf{0}_3 & \cdots & \mathbf{0}_3 \ \hat{\mathbf{I}}^{-1} \left[\mathbf{r}_1
ight]_{ imes} & \cdots & \hat{\mathbf{I}}^{-1} \left[\mathbf{r}_4
ight]_{ imes} \ \mathbf{I}_3/m & \cdots & \mathbf{I}_3/m \ \mathbf{0}_{1 imes 3} & \cdots & \mathbf{0}_{1 imes 3} \end{array}
ight] \left[egin{array}{c} \mathbf{f}_1 \ dots \ \mathbf{f}_4 \end{array}
ight]$$

3 ZOOH Discretization

Given a linear system $\dot{x} = Ax + Bu$ and its solution

$$\mathbf{x}(t) = e^{\mathbf{A}(t-t_0)}\mathbf{x}(t_0) + \int_{t_0}^{t} e^{\mathbf{A}(t-\tau)}\mathbf{B}\mathbf{u}(\tau)d\tau$$

Let $t_0 = t_k, t = t_{k+1}$, and $t_{k+1} - t_k = T$. Therefore,

$$\begin{aligned} \mathbf{x}_{k+1} &= e^{\mathbf{A}T} \mathbf{x}_k + \int_{t_k}^{t_{k+1}} e^{\mathbf{A}(t_{k+1} - \tau)} \mathbf{B} \mathbf{u}(\tau) d\tau \\ &= e^{\mathbf{A}T} \mathbf{x}_k + \int_{t_k}^{t_{k+1}} e^{\mathbf{A}(t_{k+1} - \tau)} \mathbf{B} d\tau \mathbf{u}_k \end{aligned}$$

Let $\tau' = t_{k+1} - \tau$, we have

$$\int_{t_{k}}^{t_{k+1}} e^{\mathbf{A}(t_{k+1}-\tau)} \mathbf{B} d\tau = \int_{T}^{0} e^{\mathbf{A}\tau'} \left(-d\tau'\right) \mathbf{B} = \int_{0}^{T} e^{\mathbf{A}\tau} d\tau \mathbf{B}$$

Thus, $\mathbf{x}_{k+1} = \mathbf{A}_d \mathbf{x}_k + \mathbf{B}_d \mathbf{u}_k$, where $\mathbf{A}_d = e^{\mathbf{A}T}$ and $\mathbf{B}_d = \int_0^T e^{\mathbf{A}\tau} d\tau \mathbf{B}$

4 QP Formulation

$$egin{aligned} oldsymbol{x_k} &= \mathbf{A}^k oldsymbol{x}_0 + \sum_{i=0}^{k-1} \mathbf{A}^{k-1-i} \mathbf{B} \mathbf{u}_i \ egin{aligned} egin{aligned} oldsymbol{x}_1 \ oldsymbol{x}_2 \ dots \ oldsymbol{x}_k \end{aligned} = egin{bmatrix} \mathbf{A}^1 \ \mathbf{A}^2 \ dots \ oldsymbol{x}_0 + egin{bmatrix} \mathbf{0} & & & \cdots & \mathbf{0} \ \mathbf{B} & \mathbf{0} & & \cdots & \mathbf{0} \ \mathbf{A} \mathbf{B} & \mathbf{B} & \mathbf{0} & \cdots & \mathbf{0} \ oldsymbol{x}_0 & & \ddots & dots \ oldsymbol{x}_0 + egin{bmatrix} \mathbf{u}_0 \ \mathbf{u}_1 \ dots \ oldsymbol{x}_{k-1} \ \mathbf{B} & \mathbf{A}^{k-1} \mathbf{B} & \mathbf{0} & \cdots & \mathbf{B} \end{aligned} \end{bmatrix} egin{bmatrix} \mathbf{u}_0 \ \mathbf{u}_1 \ dots \ \mathbf{u}_{k-1} \end{bmatrix}$$

5 Ground Normal Estimation

To enable the ability to traverse stairs and sloped terrain without vision, we use measurements of each footstep location

$$\boldsymbol{p}_i = (p_i^x, p_i^y, p_i^z)$$

to approximate the local slope of the walking surface.

The walking surface is modeled as a plane:

$$z(x,y) = a_0 + a_1 x + a_2 y$$

Coefficients $\mathbf{a} = (a_0, a_1, a_2)^T$ are obtained through the solution of the least squares problem

$$oldsymbol{a} = \left(oldsymbol{W}^Toldsymbol{W}^Toldsymbol{p}^z\ oldsymbol{W} = \left[egin{array}{ccc} oldsymbol{1} & oldsymbol{p}^x & oldsymbol{p}^y \end{array}
ight]_{4 imes 3}$$

where $\mathbf{p}^x = (p_1^x, p_2^x, p_3^x, p_4^x)$