

1 ZMP LQR Riccati Equation

Using $z(t)$ as the 2D position of the ZMP, we formulate:

$$\begin{aligned} & \underset{u(t)}{\text{minimize}} \quad \int_0^{t_f} k z(t) \dot{z}_d(t) k_2^2 + k u(t) k_R^2 dt; \\ & \text{subject to} \quad R = R^T > 0; \\ & \quad z_d(t) = z_d(t_f); \quad \forall t \in [0, t_f] \\ & \quad \dot{x}(t) = Ax(t) + Bu(t); \quad z(t) = Cx(t) + Du(t) \\ & \quad A = \begin{bmatrix} 0_{2 \times 2} & I_{2 \times 2} \\ 0_{2 \times 2} & 0_{2 \times 2} \end{bmatrix}; \quad B = \begin{bmatrix} 0_{2 \times 2} \\ I_{2 \times 2} \end{bmatrix} \\ & \quad C = \begin{bmatrix} I_{2 \times 2} & 0_{2 \times 2} \end{bmatrix}; \quad D = \frac{h}{g} I_{2 \times 2} \end{aligned}$$

This can be rewritten as a cost on state, *in coordinates relative to the final conditions*, $x = x - z_d^T(t_f) \begin{bmatrix} 0 & 0 \end{bmatrix}^T$, $z_d(t) = z_d(t) - z_d(t_f)$:

$$\begin{aligned} & \underset{u(t)}{\text{minimize}} \quad \int_0^{t_f} x^T Q_1 x + x^T q_2(t) + q_3(t) + u^T R_1 u + u^T r_2(t) + 2x^T N u \\ & \text{subject to} \quad Q_1 = \text{diag}(1 \quad 1 \quad 0 \quad 0); \quad q_2(t) = \begin{bmatrix} 2z_d(t) \\ 0_{2 \times 1} \end{bmatrix}; \quad q_3(t) = k z_d(t) k_2^2 \\ & \quad R_1 = R + \frac{h}{g} I_{2 \times 2}; \quad r_2(t) = 2z_d(t) \frac{h}{g}; \quad N = \frac{h}{g} \begin{bmatrix} I_{2 \times 2} \\ 0_{2 \times 2} \end{bmatrix} \\ & \quad \dot{x}(t) = Ax(t) + Bu(t) \\ & \quad A = \begin{bmatrix} 0_{2 \times 2} & I_{2 \times 2} \\ 0_{2 \times 2} & 0_{2 \times 2} \end{bmatrix}; \quad B = \begin{bmatrix} 0_{2 \times 2} \\ I_{2 \times 2} \end{bmatrix} \end{aligned}$$

Note that this implies that $x(t_f) = 0$ in order for the cost to be finite.

The resulting cost-to-go is given by

$$J = x^T S_1(t) x + x^T s_2(t) + s_3(t);$$

with the corresponding Riccati differential equation given by

$$\begin{aligned} \dot{S}_1 &= -Q_1 - (N + S_1 B) R^{-1} (B^T S_1 + N^T) + S_1 A + A^T S_1 \\ \dot{s}_2 &= q_2(t) - 2(N + S_1 B) R^{-1} r_s(t) + A^T s_2; \quad r_s(t) = \frac{1}{2}(r_2(t) + B^T s_2(t)) \\ \dot{s}_3 &= q_3(t) - r_s(t)^T R^{-1} r_s(t) \end{aligned}$$

Note that S_1 has no time-dependent terms, and therefore $S_1(t)$ is a constant, given by the steady-state solution of the algebraic Riccati equation (e.g. from time-invariant LQR). Similarly, the feedback controller is given by

$$u(t) = K_1(t)x + k_2(t);$$

and again the feedback $K_1(t)$ is a constant (derived from the infinite horizon LQR with Q , R , and N set as above).

1.1 Solving for $s_2(t)$

Given this, the affine terms in the Riccati differential equation are given by the linear differential equations:

$$\dot{s}_2(t) = A_2 s_2(t) + B_2 z_d(t); \quad s_2(t_f) = 0$$

with

$$A_2 = (N + S_1 B) R^{-1} B^T - A^T; \quad B_2 = \frac{2I_2}{0_2} + 2\frac{h}{g}(N + S_1 B) R^{-1}$$

Assuming $z_d(t)$ is described by a *continuous* piecewise polynomial of degree k with $n + 1$ breaks at t_j (with $t_0 = 0$ and $t_n = t_f$):

$$z_d(t) = \sum_{i=0}^k c_{j,i}(t - t_j)^i; \quad \text{for } j = 0; \dots; n-1; \text{ and } t \in [t_j; t_{j+1});$$

this system has a closed-form solution given by:

$$s_2(t) = e^{A_2(t - t_j)} \gamma_j + \sum_{i=0}^k \gamma_{j,i}(t - t_j)^i; \quad t \in [t_j; t_{j+1});$$

with γ_j and $\gamma_{j,i}$ vector parameters to be solved for. Taking

$$\begin{aligned} s_2(t) &= A_2 e^{A_2(t - t_j)} \gamma_j + \sum_{i=0}^k A_2 \gamma_{j,i}(t - t_j)^i + \sum_{i=0}^k B_2 c_{j,i}(t - t_j)^i \\ &= A_2 e^{A_2(t - t_j)} \gamma_j + \sum_{i=1}^k i \gamma_{j,i}(t - t_j)^{i-1} \end{aligned}$$

forces that

$$\begin{aligned} A_2 \gamma_{j,i} + B_2 c_{j,i} &= (i+1) \gamma_{j,i+1}; \quad \text{for } i = 0; \dots; k-1 \\ A_2 \gamma_{j,k} + B_2 c_{j,k} &= 0; \end{aligned}$$

Note: need to prove that A_2 is full rank (it appears to be in practice). Solve backwards ($i = k; k-1; \dots; 0$) for $\gamma_{j,i}$. Finally, the continuity and the terminal boundary condition $s(t_f) = 0$ gives

$$e^{A_2(t_{j+1} - t_j)} \gamma_j + \sum_{i=0}^k \gamma_{j,i}(t_{j+1} - t_j)^{i+1} = s(t_{j+1});$$

1.2 Reading out $k_2(t)$

The remaining term for the controller is a simple read-out given the solution to $s_2(t)$:

$$k_2(t) = \frac{h}{g} R^{-1} z_d(t) - \frac{1}{2} R^{-1} B^T s_2(t)$$

which can be written as

$$k_2(t) = L e^{A_2(t-t_j)} \underset{j;R}{j;R} + \sum_{i=0}^{\infty} \underset{j;i}{j;i}(t-t_j)^i$$

with

$$L = -\frac{1}{2} R^{-1} B^T$$

$$\underset{j;R}{j;R} = \underset{j}{j}$$

$$\underset{j;i}{j;i} = \frac{h}{g} R^{-1} \underset{j;i}{c_{j;i}} - \frac{1}{2} R^{-1} B^T \underset{j;i}{j;i}$$

1.3 Solving for $x_{com}(t)$

The resulting system is

$$\dot{x} = Ax + B(K_1 x + k_2(t)) = (A + BK_1)x + Bk_2(t);$$

where $x = [x_{com}; y_{com}; \underset{x}{x}_{com}; \underset{y}{y}_{com}]^T$: Since the solution $k_2(t)$ is the result of another linear system (cascaded in front of this one), it is easiest for me to solve

jointly, using $y = \frac{x}{s_2}$:

$$\dot{y} = A_y y + B_y z_d$$

$$A_y = \begin{bmatrix} A + BK_1 & -\frac{1}{2} B R^{-1} B^T \\ 0 & A_2 \end{bmatrix}; \quad B_y = \begin{bmatrix} \frac{h}{g} B R^{-1} \\ B_2 \end{bmatrix}$$

$$y(t) = e^{A_y(t-t_j)} \underset{j}{a_j} + \sum_{i=0}^{\infty} \underset{j;i}{b_{j;i}}(t-t_j)^i$$

i can solve for b using the same technique as above (and re-using the sol),

