

translational

$${}^{i+1}V_{i+1} = {}^iR \cdot {}^iV_i + {}^iR({}^i\omega_i \times {}^iP_{i+1}) + \dot{\theta} {}^iI \cdot {}^i\hat{z}_{i+1}$$

$${}^{i+1}\omega_{i+1} = {}^iR \cdot {}^i\omega_i$$

rotational

$${}^{i+1}V_{i+1} = {}^iR \cdot {}^iV_i + {}^iR({}^i\omega_i \times {}^iP_{i+1})$$

$${}^{i+1}\omega_{i+1} = {}^iR \cdot {}^i\omega_i + \dot{\theta} {}^iI \cdot {}^i\hat{z}_{i+1}$$

$${}^0J_V(\theta) = \frac{\partial V}{\partial \theta} \quad {}^0J_W(\theta) = \begin{bmatrix} {}^0r_3 & {}^0r_2 & {}^0r_3 \end{bmatrix}$$

$${}^A J(\theta) = \begin{bmatrix} {}^A R & 0 \\ 0 & {}^B R \end{bmatrix} {}^B J(\theta)$$

$$T = J^T F$$

$$\begin{bmatrix} {}^A V_A \\ {}^A W_A \end{bmatrix} = \begin{bmatrix} {}^A R & {}^A P_{BORG} \times {}^A R \\ 0 & {}^B R \end{bmatrix} \begin{bmatrix} {}^B V_B \\ {}^B W_B \end{bmatrix} \quad {}^A V_A = {}^A_B T_V {}^B V_B$$

$$\begin{bmatrix} {}^A F_A \\ {}^A N_A \end{bmatrix} = \begin{bmatrix} {}^A R & 0 \\ {}^A P_{BORG} \times {}^A R & {}^B R \end{bmatrix} \begin{bmatrix} {}^B F_B \\ {}^B N_B \end{bmatrix} \quad {}^A F_A = {}^A_B T_F {}^B F_B$$

$${}^A_B T_V^T = {}^A_B T_F$$

$${}^A I = \begin{bmatrix} \frac{m}{3}(l^2 + h^2) & -\frac{m}{4}wl & -\frac{m}{4}hw \\ -\frac{m}{4}wl & \frac{m}{3}(w^2 + h^2) & -\frac{m}{4}hl \\ -\frac{m}{4}hw & -\frac{m}{4}hl & \frac{m}{3}(l^2 + w^2) \end{bmatrix}$$

$${}^A I = {}^C I + m(P_C^T P_C I_3 - P_C P_C^T)$$

$$F_i = mV_i \quad N_i = {}^C I \omega_i + \omega_i \times {}^C I \omega_i$$

$$T = M(\theta)\ddot{\theta} + V(\theta, \dot{\theta}) + G(\theta)$$

$$k_i = \frac{1}{2}m_i V_{Ci}^T V_{Ci} + \frac{1}{2} {}^i\omega_i^T \alpha_i] {}^i\omega_i$$

$$K(\theta, \dot{\theta}) = \frac{1}{2} \dot{\theta}^T M(\theta) \dot{\theta}$$

$$u_i = -m_i {}^0 g^T {}^0 P_{Ci}$$

$$T = \frac{d}{dt} \frac{dk}{d\theta} - \frac{dk}{d\theta} + \frac{du}{d\theta}$$

$$F = M_x(\theta) \ddot{x} + V_x(\theta, \dot{\theta}) + G_x(\theta)$$

$$M_x(\theta) = J^{-T}(\theta) M(\theta) J^{-1}(\theta)$$

$$V_x(\theta, \dot{\theta}) = J^{-T}(\theta) (V(\theta, \dot{\theta}) - M(\theta) J^{-1}(\theta) j(\theta) \dot{\theta})$$

$$G_x(\theta) = J^{-T}(\theta) G(\theta)$$

$$m\ddot{x} + b\dot{x} + kx = f$$

$$f = \alpha f' + \beta \quad f' = \ddot{x} = -kv \dot{x} - kp x$$

$$kv = 2\sqrt{kp}$$

$$\text{恒等式 } \frac{1}{2} \dot{\theta}^T M(\theta) \dot{\theta} = \dot{\theta}^T V(\theta, \dot{\theta}).$$

$$\ddot{x}_i = f^i = \underbrace{\dot{x}_d + k_v \dot{e} + k_p e}_{\text{servo}} + k_i \int e dt$$

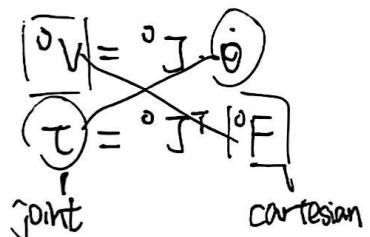
$${}^A_B R = [{}^A \hat{x}_B \quad {}^A \hat{y}_B \quad {}^A \hat{z}_B]$$

piiper's theorem. 六轴IK封闭解，与三轴交于一点

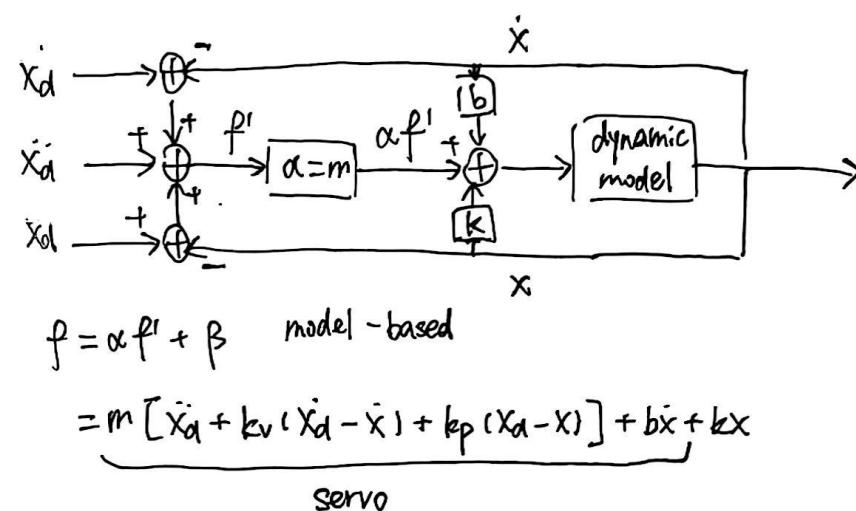
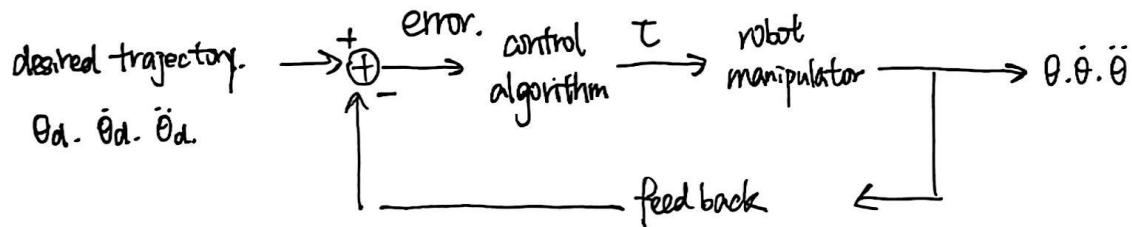
workspace inferior singularities :

充分条件: $|{}^0 J| = 0 / \text{rank}(J) < m$

必要条件: $J\dot{q} = 0 / \text{rank}(J) < n$



feedback control



$$f = \alpha f^i + \beta \quad \text{model-based}$$

$$= m [\dot{x}_d + k_v (\dot{x}_d - \dot{x}) + k_p (x_d - x)] + b\dot{x} + kx$$

$$\text{PID. } \tau' = \ddot{\theta}_d + k_p (\theta_d - \theta) + k_i \int (\theta_d - \theta) dt + k_v (\dot{\theta}_d - \dot{\theta})$$

lyapunov stability theory.

-阶连续导数. $V(x) > 0$ for $x > 0$. $V(0) = 0$.

$$\dot{V}(x) \leq 0. \text{ lyapunov 稳定. } V = \frac{1}{2} \dot{\theta}^T M(\theta) \dot{\theta} + \frac{1}{2} \theta^T K_p \theta.$$

$\dot{V}(x) < 0$. 滚正稳定.

α_{i-1} 从 x_{i-1} 转 z_i .

α_{i-1} 从 x_{i-1} 到 $i-1$ 移到 i

d_i 从 z_i 从 $i-1$ 移到 i

θ_i 从 z_i 转 x .

x_i 从 z_i 指向 z_{i+1} .

$${}^{i-1}_i T = \text{ROT}(x_{i-1}, \alpha_{i-1}) \cdot \text{TRANS}(x_{i-1}^n, \alpha_{i-1}) \cdot$$

$$\text{ROT}(z_i^i, \theta_i) \cdot \text{TRANS}(z_i^i, d_i)$$

动力学精确 \rightarrow 反馈控制

动力学不精确 \rightarrow 自适应控制

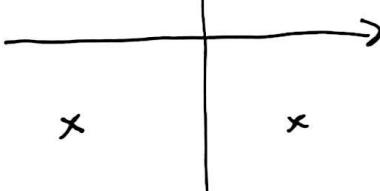
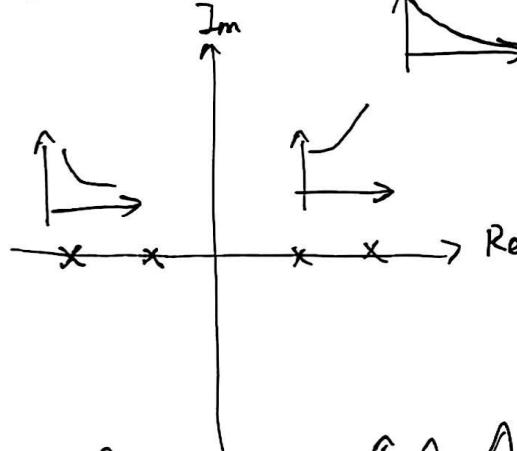
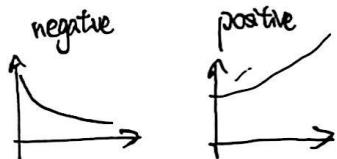
二阶线性.
 $\alpha f^1 + \beta$ 轨迹跟踪
 力矩跟踪
 PID.

$$m\ddot{x} + b\dot{x} + x = 0.$$

$$b^2 - 4mk > 0. \text{ 阻尼.}$$

$$b^2 - 4mk < 0. \text{ 共振.}$$

$$b^2 - 4mk = 0. \text{ 临界阻尼.}$$



控制律分解

$$f = \alpha f^1 + \beta = \text{servo} + \text{model} \quad \text{轨迹跟踪.}$$

$$\text{力矩控制.} \quad \tau = M(\theta) \ddot{\theta} + C(\theta, \dot{\theta}) \dot{\theta} + G(\theta) + k_p(\theta_d - \theta) + k_a(\dot{\theta}_d - \dot{\theta})$$

$$\text{PID 控制.} \quad \tau' = \dot{\theta} + k_p(\theta_d - \theta) + k_i \int (\theta_d - \theta) dt + k_v(\dot{\theta}_d - \dot{\theta}) \quad \alpha = 1, \beta = 0.$$

$$\text{重力补偿 PD 控制.} \quad \tau = k_p e + k_d \dot{e} + G(\theta)$$

adaptive control.

$$M(\theta) \ddot{\theta} + V(\theta, \dot{\theta}) + G(\theta) = \tau = \hat{M}(\theta, \dot{\theta}, \ddot{\theta}) \cdot \theta$$

$$\tau = \hat{M}(\theta) \ddot{\theta} + \hat{C}(\theta, \dot{\theta}) \dot{\theta} + \hat{G}(\theta) + \begin{cases} k_p(\theta_d - \theta) \\ k_d(\dot{\theta}_d - \dot{\theta}) \end{cases} \quad \begin{matrix} \text{model} & \text{servo} \end{matrix}$$

$$\hat{\theta}_p(t+1) = \hat{\theta}_p(t) - \alpha \hat{B}^T(\theta(t), \dot{\theta}(t), \ddot{\theta}(t)) e(t)$$

$$y(\theta, \dot{\theta}, \ddot{\theta}) \hat{\psi} = \hat{M}(\theta) (\ddot{E} + k_v \dot{E} + k_p E) + y(\theta, \dot{\theta}, \ddot{\theta}) \hat{\psi}$$

real model = servo part + estimated model

$$\ddot{E} + k_v \dot{E} + k_p E = \hat{M}^{-1}(\theta) y(\theta, \dot{\theta}, \ddot{\theta}) \hat{\psi}$$

$$\dot{\tilde{E}} = AE + B\hat{M}^{-1}(\theta) y(\theta, \dot{\theta}, \ddot{\theta}) \hat{\psi} \quad \tilde{E} = \begin{bmatrix} E \\ \dot{E} \end{bmatrix} \quad A = \begin{bmatrix} 0 & I \\ -k_p & -k_v \end{bmatrix} \quad B = \begin{bmatrix} 0 \\ I \end{bmatrix}$$

$$\text{select Lyapunov function.} \quad V = \tilde{E}^T P \tilde{E} + \hat{\psi}^T \hat{\psi}$$

$$\tau = M(\theta) \ddot{\theta} + V(\theta, \dot{\theta}) + G(\theta).$$

$$\ddot{\theta} = M^{-1}(\theta) [\tau - V(\theta, \dot{\theta}) - G(\theta)] \quad \dot{\theta}(t + \Delta t) = \dot{\theta}(t) + \ddot{\theta}(t) \Delta t$$

$$\theta(t + \Delta t) = \theta(t) + \dot{\theta}(t) \Delta t + \frac{1}{2} \ddot{\theta}(t) \Delta t^2$$