

translational

$${}^{i+1}V_{i+1} = {}^{i+1}_i R \cdot {}^iV_i + {}^{i+1}_i R ({}^i\omega_i \times {}^iP_{i+1}) + d_{i+1} \cdot \hat{z}_{i+1}$$

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

rotational

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

$${}^{i+1}\omega_{i+1} = {}^{i+1}_i R \cdot {}^i\omega_i + \theta_{i+1} \cdot \hat{z}_{i+1}$$

$${}^0J_v(\theta) = \frac{\partial V}{\partial \theta} \quad {}^0J_\omega(\theta) = \begin{bmatrix} 0 & r_3 & 0 & r_3 & 0 & r_3 \end{bmatrix}$$

$${}^AJ(\theta) = \begin{bmatrix} {}^AR & 0 \\ 0 & {}^BR \end{bmatrix} {}^BJ(\theta)$$

$$\tau = J^T F$$

$$\begin{bmatrix} {}^AV_A \\ {}^A\omega_A \end{bmatrix} = \begin{bmatrix} {}^BR & {}^AP_{BORG} \times {}^BR \\ 0 & {}^BR \end{bmatrix} \begin{bmatrix} {}^BV_B \\ {}^B\omega_B \end{bmatrix} \quad {}^AV_A = {}^A_B T_V {}^BV_B$$

$$\begin{bmatrix} {}^AF_A \\ {}^AN_A \end{bmatrix} = \begin{bmatrix} {}^BR & 0 \\ {}^AP_{BORG} \times {}^BR & {}^BR \end{bmatrix} \begin{bmatrix} {}^BF_B \\ {}^BN_B \end{bmatrix} \quad {}^AF_A = {}^A_B T_f {}^BF_B$$

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

$${}^AI = \begin{bmatrix} \frac{m}{3}(l^2+h^2) & -\frac{m}{4}\omega l & -\frac{m}{4}h\omega \\ -\frac{m}{4}\omega l & \frac{m}{3}(\omega^2+h^2) & -\frac{m}{4}hl \\ -\frac{m}{4}h\omega & -\frac{m}{4}hl & \frac{m}{3}(l^2+\omega^2) \end{bmatrix}$$

$${}^AI = {}^CI + m(P_C^T P_C I_3 - P_C P_C^T)$$

$$F_i = m\dot{V}_i \quad N_i = {}^Ci I \omega_i + \omega_i \times {}^Ci I \omega_i$$

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

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

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

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

$$\tau = \frac{d}{dt} \frac{\partial k}{\partial \dot{\theta}} - \frac{\partial k}{\partial \theta} + \frac{\partial u}{\partial \theta}$$

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

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

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

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

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

$$f = \alpha f' + \beta \quad f' = \ddot{x} = -k_v \dot{x} - k_p x$$

$$k_v = 2\sqrt{k_p}$$

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

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

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

pieper's theorem. 六轴1K封闭解, 三轴交于一点

workspace interior singularities:

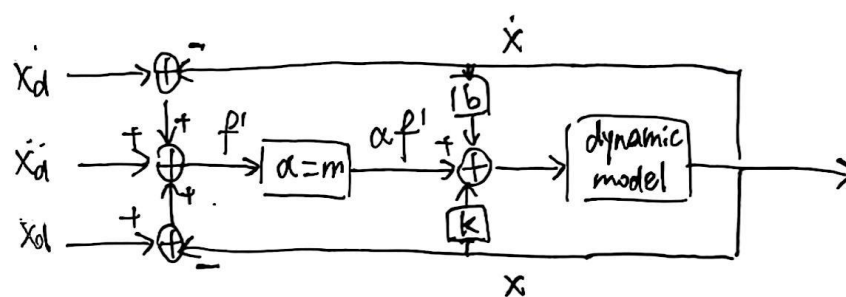
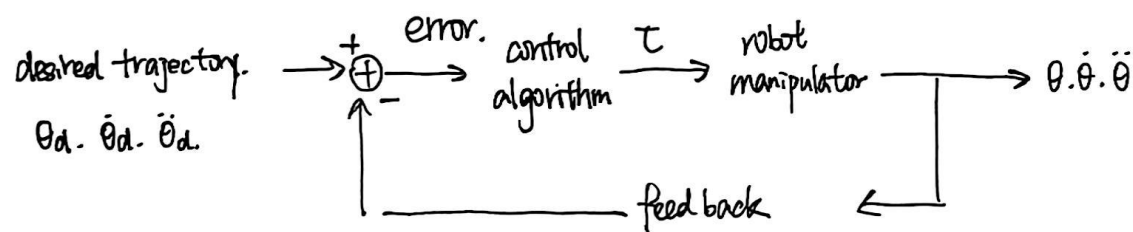
充分条件: $|{}^0J| = 0$ / $\text{rank}(J) < m$

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

$$\begin{aligned} |{}^0V| &= {}^0J \cdot \dot{\theta} \\ \tau &= {}^0J^T F \end{aligned}$$

joint cartesian

feedback control



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

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

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

Lyapunov stability theory.

一阶连续导数. $V(x) > 0$ for $x > 0$. $V(0) = 0$.

$\dot{V}(x) \leq 0$. Lyapunov 稳定.

$\dot{V}(x) < 0$. 渐近稳定.

$$V = \frac{1}{2} \dot{\theta}^T M(\theta) \dot{\theta} + \frac{1}{2} \theta^T k_p \theta$$

α_{i-1} 从 x_{i-1} 转 Z.

a_{i-1} 沿 x_{i-1} 从 $i-1$ 移到 i

d_i 沿 z_i 从 $i-1$ 移到 i

θ_i 从 z_i 转 X.

x_i 轴从 z_i 指向 z_{i+1} .

$${}^{i-1}_i T = \text{ROT}({}^{\hat{x}_{i-1}}_{\hat{x}_{i-1}}, \alpha_{i-1}) \cdot \text{TRANS}({}^{\hat{x}_{i-1}}_{\hat{x}_{i-1}}, a_{i-1}) \cdot$$

$$\text{ROT}(\hat{z}_i, \theta_i) \cdot \text{TRANS}(\hat{z}_i, d_i)$$

动力学精确 \rightarrow 反馈控制

动力学不精确 \rightarrow 自适应控制

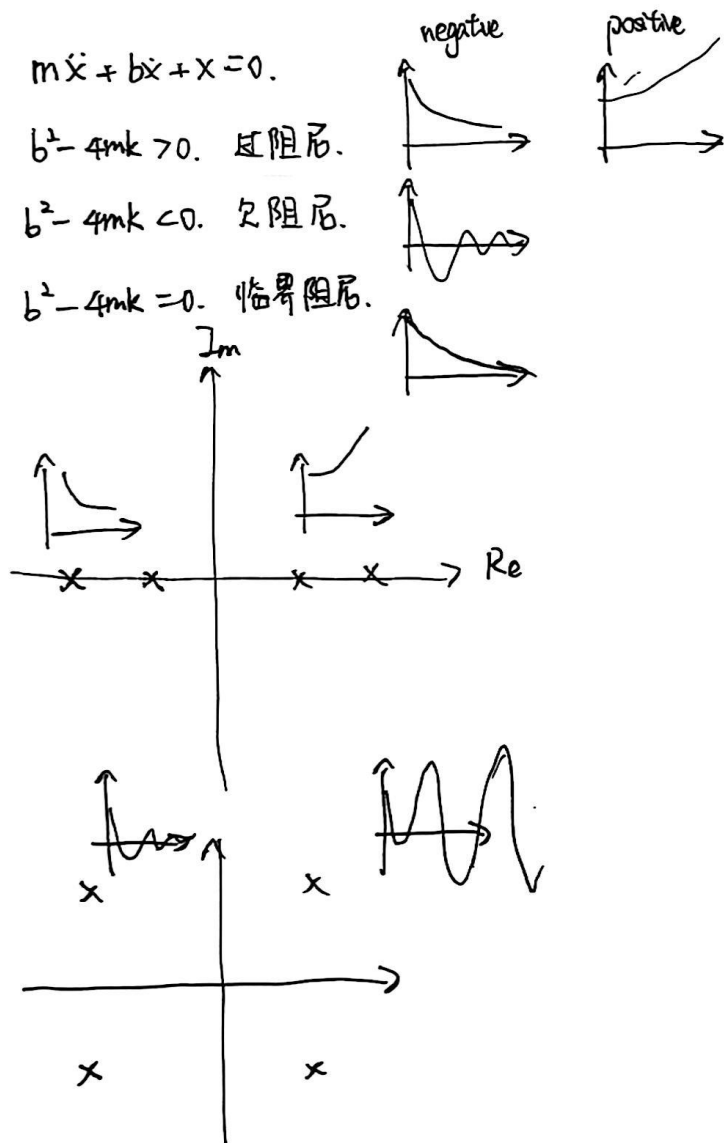
$\left\{ \begin{array}{l} \text{二阶线性} \\ \alpha \ddot{p} + \beta \dot{p} = \text{轨迹跟踪} \\ \text{力矩跟踪} \\ \text{PID} \end{array} \right.$

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

$$b^2 - 4mk > 0. \text{ 过阻尼.}$$

$$b^2 - 4mk < 0. \text{ 欠阻尼.}$$

$$b^2 - 4mk = 0. \text{ 临界阻尼.}$$



控制律分解

$$p = \alpha \ddot{p} + \beta \dot{p} = \text{servo} + \text{model} \quad \text{轨迹跟踪.}$$

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

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

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

adaptive control.

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

$$\tau = \hat{M}(\theta) \ddot{\theta}_d + \hat{C}(\theta, \dot{\theta}) \dot{\theta}_d + \hat{G}(\theta) + k_p(\theta_d - \theta) + k_d(\dot{\theta}_d - \dot{\theta})$$

model servo

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

$$\gamma(\theta, \dot{\theta}, \ddot{\theta}) \tilde{\varphi} = \hat{M}(\theta) (\ddot{E} + k_v \dot{E} + k_p E) + \gamma(\theta, \dot{\theta}, \ddot{\theta}) \tilde{\varphi}$$

real model = servo part + estimated model

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

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

$$\text{select lyapunov function. } V = \tilde{E}^T P \tilde{E} + \tilde{\varphi}^T T^{-1} \tilde{\varphi}$$

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

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

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