

The True Role of Accelerometer Feedback in Quadrotor Control

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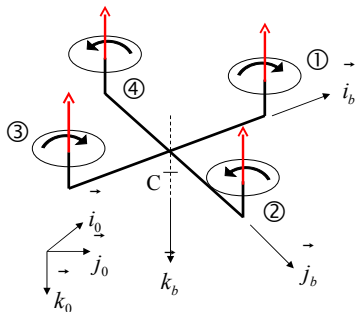
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Two puzzling questions

Quadrotor:

- 2 CW, 2 CCW props
- $(\vec{i}_b, \vec{j}_b, \vec{k}_b)$ body axes



Newton's law for usual model

$$\dot{\vec{V}}_C = \vec{g} + \frac{T}{m} \vec{k}_b$$

Triaxial accelero located at C

$$\vec{a} = \dot{\vec{V}}_C - \vec{g} \quad (\text{in body axes})$$

This implies $\vec{a} = \frac{T}{m} \vec{k}_b$, hence

$$(a_x, a_y) = (\vec{a} \cdot \vec{i}_b, \vec{a} \cdot \vec{j}_b) = (0, 0)$$

What is accelero feedback good for?

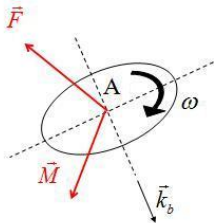
Usual approximation: $\vec{a} \approx -\vec{g}$, hence

$$(a_x, a_y) \approx (g \sin \theta, -g \sin \phi \cos \theta)$$

Meaning of the approximation?

Single propeller

- angular velocity $\varepsilon\omega$ around axis \vec{k}_b
 $\omega \geq 0, \varepsilon = \pm 1$ (CCW/CW)
- \vec{V}_A linear velocity of prop center A
- $\vec{\Omega}$ angular velocity of “rotor plane” $\perp \vec{k}_b$



Aerodynamic efforts “near” hovering

$$\vec{F} = -a\omega^2\vec{k}_b - \omega(\lambda_1\vec{V}_A^\perp - \lambda_2\vec{\Omega} \times \vec{k}_b) + \varepsilon\omega(\lambda_3\vec{V}_A \times \vec{k}_b - \lambda_4\vec{\Omega}^\perp)$$

$$\vec{M} = -b\varepsilon\omega^2\vec{k}_b - \omega(\mu_1\vec{V}_A^\perp + \mu_2\vec{\Omega} \times \vec{k}_b) - \varepsilon\omega(\mu_3\vec{V}_A \times \vec{k}_b + \mu_4\vec{\Omega}^\perp)$$

$a, b, \lambda_1, \lambda_2, \lambda_3, \lambda_4, \mu_1, \mu_2, \mu_3, \mu_4$ positive constants

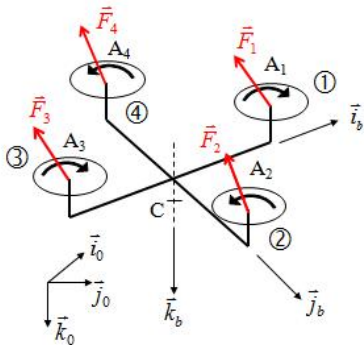
projection on rotor plane $\vec{U}^\perp := \vec{k}_b \times (\vec{U} \times \vec{k}_b) = \vec{U} - (\vec{U} \cdot \vec{k}_b)\vec{k}_b$

Classical blade element theory, with two extra simplifications:

- higher-order linear and angular velocity terms neglected
- linear and angular accelerations neglected

Complete quadrotor:

- quadrotor \mathcal{B} = frame \mathcal{B}_0
+ 4 × (prop + motor) \mathcal{B}_i
- 2 CW, 2 CCW props
- $(\vec{i}_b, \vec{j}_b, \vec{k}_b)$ body axes



Motion equations (dimension 13)

$$\dot{\vec{V}}_C = \vec{g} + \frac{1}{m} \sum_{i=1}^4 \vec{F}_i$$

$$\dot{\vec{\sigma}}_C^{\mathcal{B}} = \sum_{i=1}^4 \vec{CA}_i \times \vec{F}_i + \vec{M}_i$$

$$\dot{\vec{\sigma}}_{A_i}^{\mathcal{B}_i} \cdot \vec{k}_b = \vec{M}_i \cdot \vec{k}_b + \varepsilon_i \Gamma_i, \quad i = 1, 2, 3, 4$$

$$\dot{R}_{\phi, \theta, \psi} = \vec{\Omega} \times R_{\phi, \theta, \psi}$$

Inertial sensors (in body axes)

$$\text{accelero: } \vec{a} = \dot{\vec{V}}_C - \vec{g} = \frac{1}{m} \sum_{i=1}^4 \vec{F}_i$$

$$\text{gyro: } \vec{\Omega}$$

with some simplifications (near hovering and mostly rigid prop)

$$\dot{\vec{V}}_C \approx \vec{g} - \frac{a}{m} \left(\sum_{i=1}^4 \omega_i^2 \right) \vec{k}_b - \frac{\lambda_1}{m} \left(\sum_{i=1}^4 \omega_i \right) \vec{V}_C^\perp$$

$$\dot{\vec{\sigma}}_C^B \approx a(\omega_4^2 - \omega_2^2) \vec{i}_b + a(\omega_1^2 - \omega_3^2) \vec{j}_b - b \left(\sum_{i=1}^4 \varepsilon_i \omega_i^2 \right) \vec{k}_b$$

$$- \left(\sum_{i=1}^4 \omega_i \right) (\lambda_1 l^2 r \vec{k}_b + \mu'_1 \vec{V}_C \times \vec{k}_b + \mu'_2 \vec{\Omega}^\perp)$$

$$\dot{\vec{\sigma}}_{A_i}^{B_i} \cdot \vec{k}_b \approx \varepsilon_i \Gamma_i - \left(\sum_{i=1}^4 \omega_i \right) \lambda_1 l^2 r, \quad i = 1, 2, 3, 4$$

$$\dot{R}_{\phi, \theta, \psi} = \vec{\Omega} \times R_{\phi, \theta, \psi}$$

$$\vec{a} \approx -\frac{a}{m} \left(\sum_{i=1}^4 \omega_i^2 \right) \vec{k}_b - \frac{\lambda_1}{m} \left(\sum_{i=1}^4 \omega_i \right) \vec{V}_C^\perp$$

Linearized model splits into four independent subsystems:

- longitudinal (input $\Gamma_q := \frac{\Gamma_1 - \Gamma_3}{J_r}$, states $u, \theta, q, \omega_q := \omega_1 - \omega_3$)

$$\dot{u} = -g\theta - f_1 u \quad (\text{longitudinal velocity})$$

$$\dot{\theta} = q \quad (\text{pitch angle})$$

$$\dot{q} = f_4 \omega_q + f_2 u - f_3 q \quad (\text{pitch rate})$$

$$\dot{\omega}_q = \Gamma_q - f_5 \omega_q \quad (\text{prop difference})$$

Measurements $a_x = -f_1 u$ and $g_y = q$

- lateral (input $\Gamma_4 - \Gamma_2$, states $v, \phi, p, \omega_4 - \omega_2$)
- vertical (input $\sum_{i=1}^4 \Gamma_i$, states $w, \sum_{i=1}^4 \omega_i$)
- heading (input $\sum_{i=1}^4 \varepsilon_i \Gamma_i$, states $\psi, r, \sum_{i=1}^4 \varepsilon_i \omega_i$)

$$(f_1, f_2, f_3, f_4, f_5) := \left(\frac{4\lambda_1 \bar{w}}{m}, \frac{4\mu'_1 \bar{w}}{l}, \frac{4\mu''_2 \bar{w}}{l}, \frac{2a l \bar{w}}{l}, \frac{2b \bar{w}}{J_r} \right)$$

Experimental setup: home-built “Quadricopter” + radio data link

- “true” Earth velocity V_x, V_y, V_z and orientation ϕ_m, θ_m, ψ_m given by MIDG2 “GPS-aided Inertial Navigation System”
- raw accelero data a_{xm}, a_{ym} also given by MIDG2
- quadrotor flown in back and forth translations for 1 minute
- seek to validate force model only (because of low throughput of radio data link)

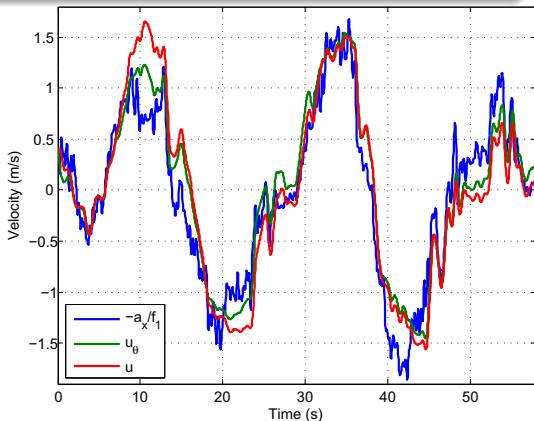
Processing of flight data

- filter all the data with 5th-order Bessel @ 2Hz, which preserves the transfer functions
- compensate data for MIDG2 misalignment ϕ_0, θ_0, ψ_0
- compute “true” body velocities u, v, w
- compute aligned accelero measurements a_x, a_y

Validation of force model

With $(\frac{1}{f_1}, \phi_0, \theta_0, \psi_0) := (4s, 1.2^\circ, -2.4^\circ, 2^\circ)$ good fit between:

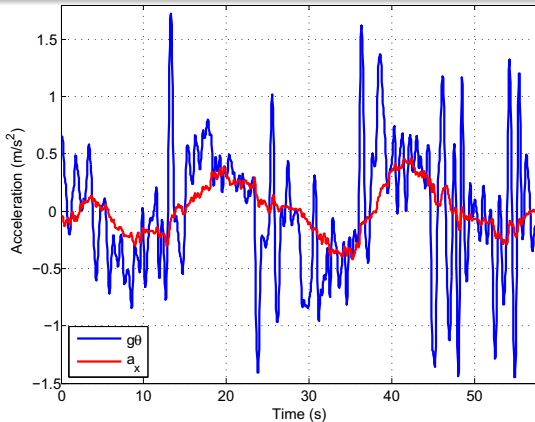
- “true” body velocity u
- “accelerometer-based” velocity $-\frac{a_x}{f_1}$
- velocity u_θ “predicted” by model from “true” pitch angle θ



Invalidation of no-acceleration assumption $\vec{a} \approx -\vec{g}$

$$a_x \approx g \sin \theta \quad ???$$

- more or less ok in average... because quadrotor stabilized by other means
- very wrong transients!



Importance of coefficients wrt “full-state” feedback?

$$\dot{u} = -g\theta - f_1 u$$

$$\dot{\theta} = q$$

$$\dot{q} = f_4 \omega_q + f_2 u - f_3 q$$

$$\dot{\omega}_q = \Gamma_q - f_5 \omega_q$$

$$\Gamma_q = -\frac{k_p}{\varepsilon^2} q - \frac{k_d}{\varepsilon} \dot{q} + \frac{k_p}{\varepsilon^2} q_r \quad (\text{inner loop})$$

$$q_r = k_1 u + k_2 \dot{u} - k_1 u_r \quad (\text{outer loop})$$

After fast inner loop:

$$\dot{u} = -f_1 u - g\theta$$

$$\dot{\theta} = q$$

$$\varepsilon \dot{q} = f_4 \tilde{\omega}_q + \mathcal{O}(\varepsilon)$$

$$\varepsilon \dot{\tilde{\omega}}_q = -k_p q - f_4 k_d \tilde{\omega}_q + k_p q_r + \mathcal{O}(\varepsilon)$$

Singular perturbations ($\tilde{\omega}_q := \varepsilon \omega_q$)

⇒ f_2, f_3, f_5 dominated by feedback

After slow outer loop:

$$\dot{u} = -f_1 u - g\theta$$

$$\dot{\theta} = (k_1 - f_1 k_2) u - g k_2 \theta - k_1 u_r$$

Reasonable settling time

⇒ f_1 dominated by feedback

Conclusion: if u (and q) measured, revisited state model not useful!

Slow control model

$$\dot{u} = -g\theta - f_1 u$$

$$\dot{\theta} \approx q_r \quad (\text{for controller})$$

$$\dot{\theta} = q \quad (\text{for estimator})$$

Outer loop (with “angle estimator”)

$$q_r = k(\theta_r - \hat{\theta})$$

$$\dot{\hat{\theta}} = q + l\left(\frac{a_x}{g} - \hat{\theta}\right)$$

Usual interpretation

$a_x = g\theta \Rightarrow$ closed-loop polynomial $s(s+k)(s+l)$

$$\theta = \frac{k}{s+k} \theta_r$$

$$u = \frac{-gk}{s(s+k)} \theta_r$$

With $k, l > 0$,

$(u, \theta, \hat{\theta}) \rightarrow (\infty, \theta_r, \theta_r)$

Not quite consistent with experience!

Revisited interpretation

$a_x = -f_1 u \Rightarrow$ closed-loop polynomial

$$\Delta = s^3 + (k+l+f_1)s^2 + f_1(k+l)s + f_1kl$$

$$\approx (s+k)(s^2 + f_1s + f_1l) \quad \text{with } l \ll k$$

$$\approx (s+f_1)(s+k)(s+l) \quad \text{with } l \ll f_1$$

$$\theta = \frac{k(s+f_1)(s+l)}{\Delta} \theta_r \approx \frac{k}{s+k} \theta_r$$

$$u = \frac{-gk(s+l)}{\Delta} \theta_r \approx \frac{-gk}{(s+f_1)(s+k)} \theta_r$$

With $k, l > 0$, $(u, \theta, \hat{\theta}) \rightarrow (-\frac{g}{f_1} \theta_r, \theta_r, \theta_r)$

Improve performance with controller based on revisited model

Performance of usual control scheme limited by time constant $\frac{1}{f_1}$
 \Rightarrow better performance with eg “revisited controller-observer”

$$q_r = -k_1 \hat{u} - k_2 \hat{\theta} + \left(k_1 - \frac{f_1 k_2}{g} \right) u_r$$

$$\dot{\hat{u}} = -f_1 \hat{u} - g \hat{\theta} + l_1 (a_x + f_1 \hat{u})$$

$$\dot{\hat{\theta}} = g_y + l_2 (a_x + f_1 \hat{u})$$

