

Robust Adaptive Control of Multi-Rotor UAVs

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Unmanned Aerial Vehicles – UAVs

Typical **UAV tasks** include:

- Surveillance;
- Crop monitoring;
- Search and rescue;

Mission scenarios usually

- Occur in **open space**;
- Require **calm weather** conditions;
- Assume **known vehicle and payload properties**

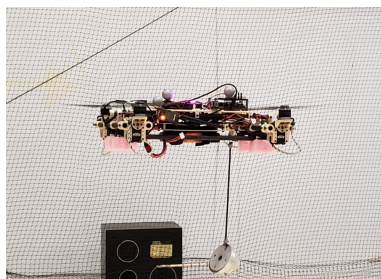


These are **passive missions** in the environment

UAVs of the Future

Intelligent UAVs of the future will **actively interact with the environment** performing tasks such as

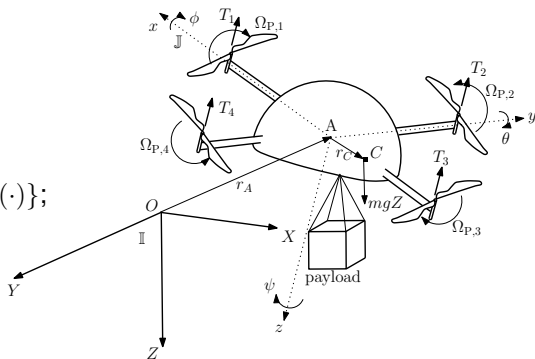
- Maneuvering **unknown objects**;
- Interacting with walls or hard surfaces;
 - **Aerodynamic wall effects**;
- **Pulling or pushing** items such as cables



Novel theoretical frameworks needed to **design** suitable **autopilots**

Modeling Assumptions

- Center of mass is not the reference point $A(\cdot)$;
- Main frame is rigid;
- Inertia matrix is constant
- Inertial reference frame,
 - $\mathbb{I} = \{O; X, Y, Z\}$;
- Body reference frame,
 - $\mathbb{J}(\cdot) = \{A(\cdot); x(\cdot), y(\cdot), z(\cdot)\}$;
- Position of $A(\cdot)$ w.r.t O ,
 - $r_A^{\mathbb{I}} : [t_0, \infty) \rightarrow \mathbb{R}^3$;
- Velocity of $A(\cdot)$ w.r.t \mathbb{I} ,
 - $v_A^{\mathbb{I}} : [t_0, \infty) \rightarrow \mathbb{R}^3$;
- Angular velocity of $\mathbb{J}(\cdot)$ w.r.t \mathbb{I} ,
 - $\omega : [t_0, \infty) \rightarrow \mathbb{R}^3$;



Kinematic Equations

■ Translational Kinematic Equations

$$\dot{r}_A^{\mathbb{I}}(t) = R(\phi(t), \theta(t), \psi(t)) v_A(t), \quad r_A^{\mathbb{I}}(t_0) = r_{A,0}^{\mathbb{I}}, \quad t \geq t_0,$$

$$R(\phi, \theta, \psi) \triangleq \begin{bmatrix} \cos \psi & -\sin \psi & 0 \\ \sin \psi & \cos \psi & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} \cos \theta & 0 & \sin \theta \\ 0 & 1 & 0 \\ -\sin \theta & 0 & \cos \theta \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & \cos \phi & -\sin \phi \\ 0 & \sin \phi & \cos \phi \end{bmatrix},$$

$$(\phi, \theta, \psi) \in [0, 2\pi) \times \left(-\frac{\pi}{2}, \frac{\pi}{2}\right) \times [0, 2\pi)$$

■ Rotational Kinematic Equations

$$\begin{bmatrix} \dot{\phi}(t) \\ \dot{\theta}(t) \\ \dot{\psi}(t) \end{bmatrix} = \Gamma(\phi(t), \theta(t)) \omega(t), \quad \begin{bmatrix} \phi(t_0) \\ \theta(t_0) \\ \psi(t_0) \end{bmatrix} = \begin{bmatrix} \phi_0 \\ \theta_0 \\ \psi_0 \end{bmatrix},$$

$$\Gamma(\phi, \theta) \triangleq \begin{bmatrix} 1 & \sin \phi \tan \theta & \cos \phi \tan \theta \\ 0 & \cos \phi & -\sin \phi \\ 0 & \sin \phi \sec \theta & \cos \phi \sec \theta \end{bmatrix}$$

Dynamic Equations I

■ Translational Dynamic Equations

$$\begin{aligned}
 F_g(\phi(t), \theta(t)) - F_T(t) + F(t) \\
 &= m [\dot{v}_A(t) + \omega^\times(t)v_A(t)] + m [\ddot{r}_C(t) + \dot{\omega}^\times(t)r_C(t) \\
 &\quad + 2\omega^\times(t)\dot{r}_C(t) + \omega^\times(t)\omega^\times(t)r_C(t)], \quad v_A(t_0) = v_{A,0},
 \end{aligned}$$

- Thrust force: $-F_T = [0, 0, u_1]^T$,
- Quadcopter's weight

$$\begin{aligned}
 F_g(\phi, \theta) &= mg[-\sin \theta, \cos \theta \sin \phi, \cos \theta \cos \phi]^T, \\
 (\phi, \theta) &\in [0, 2\pi) \times \left(-\frac{\pi}{2}, \frac{\pi}{2}\right),
 \end{aligned}$$

- Aerodynamic drag: $F : [t_0, \infty) \rightarrow \mathbb{R}^3$

Dynamic Equations II

■ Rotational Dynamic Equations

$$\begin{aligned}
 M_T(t) + M_g(r_C(t), \phi(t), \theta(t)) + M(t) \\
 = mr_C^\times(t) [\dot{v}_A(t) + \omega^\times(t)v_A(t)] + I\dot{\omega}(t) + \omega^\times(t)I\omega(t) \\
 + I_P \sum_{i=1}^4 \begin{bmatrix} 0 \\ 0 \\ \dot{\Omega}_{P,i}(t) \end{bmatrix} + \omega^\times(t)I_P \sum_{i=1}^4 \begin{bmatrix} 0 \\ 0 \\ \Omega_{P,i}(t) \end{bmatrix}, \\
 \omega(t_0) = \omega_0, \quad t \geq t_0
 \end{aligned}$$

- **Moment of the propellers' forces:** $M_T = [u_2, u_3, u_4]^T$,
- **Moment of the weight:** $M_g(r_C, \phi, \theta) \triangleq r_C^\times F_g(\phi, \theta)$,
- **Moment of the aerodynamic drag:** $M : [t_0, \infty) \rightarrow \mathbb{R}^3$

Remark

Translational Dynamic Equations II

$$m\ddot{\mathbf{r}}_A^{\parallel}(t) = R(\phi(t), \theta(t), \psi(t)) \begin{bmatrix} 0 \\ 0 \\ u_1(t) \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ mg \end{bmatrix} + \mathbf{F}^{\parallel}(t) - m\ddot{\mathbf{r}}_C^{\parallel}(t),$$

$$\mathbf{r}_A^{\parallel}(t_0) = \mathbf{r}_{A,0}^{\parallel}, \quad \mathbf{v}_A^{\parallel}(t_0) = \mathbf{v}_{A,0}^{\parallel}, \quad t \geq t_0$$

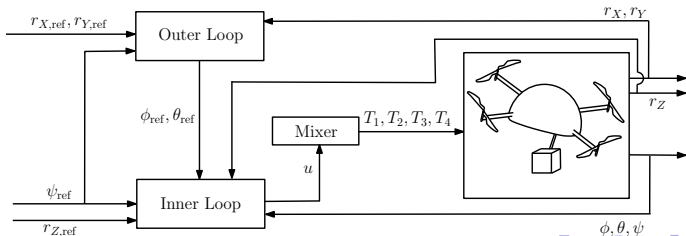
“Mixer”

$$\begin{bmatrix} u_1(t) \\ u_2(t) \\ u_3(t) \\ u_4(t) \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 0 & -l & 0 & l \\ l & 0 & -l & 0 \\ -c_T & c_T & -c_T & c_T \end{bmatrix} \begin{bmatrix} T_1(t) \\ T_2(t) \\ T_3(t) \\ T_4(t) \end{bmatrix},$$

$$T_i(t) = k\Omega_{P,i}^2(t), \quad i = 1, \dots, 4$$

Conventional Autopilot Architecture

- **6 states** $r_A(\cdot), v_A(\cdot), \phi(\cdot), \theta(\cdot), \psi(\cdot)$
- **4 controls** $u_1(\cdot), \dots, u_4(\cdot)$
 - **Quadcopters are underactuated**
- Choose **4 reference values to track**:
 $r_{X,\text{ref}}, r_{Y,\text{ref}}, r_{Z,\text{ref}}, \psi_{\text{ref}} : [t_0, \infty) \rightarrow \mathbb{R}$;
- **Outer loop**: Deduce $\phi_{\text{ref}}, \theta_{\text{ref}} : [t_0, \infty) \rightarrow \mathbb{R}$
- **Inner loop**: Deduce $u_1(\cdot), \dots, u_4(\cdot)$



Outer Loop Design — Example 1

If $\phi(\cdot)$ & $\theta(\cdot)$ are **small** & $r_C(t) \equiv 0$, **then**

$$\begin{bmatrix} \ddot{r}_X(t) \\ \ddot{r}_Y(t) \end{bmatrix} = \frac{u_1(t)}{m} \begin{bmatrix} \sin \psi(t) & \cos \psi(t) \\ -\cos \psi(t) & \sin \psi(t) \end{bmatrix} \begin{bmatrix} \phi(t) \\ \theta(t) \end{bmatrix},$$

PID Controller

$$\begin{bmatrix} \phi_{\text{ref}}(t) \\ \theta_{\text{ref}}(t) \end{bmatrix} = \frac{m}{u_1(t)} \begin{bmatrix} \sin \psi_{\text{ref}}(t) & -\cos \psi_{\text{ref}}(t) \\ \cos \psi_{\text{ref}}(t) & \sin \psi_{\text{ref}}(t) \end{bmatrix} \cdot \left(K_p \begin{bmatrix} r_{X,\text{ref}}(t) - r_X(t) \\ r_{Y,\text{ref}}(t) - r_Y(t) \end{bmatrix} + K_d \begin{bmatrix} \dot{r}_{X,\text{ref}}(t) - \dot{r}_X(t) \\ \dot{r}_{Y,\text{ref}}(t) - \dot{r}_Y(t) \end{bmatrix} + \begin{bmatrix} \ddot{r}_{X,\text{ref}}(t) \\ \ddot{r}_{Y,\text{ref}}(t) \end{bmatrix} \right),$$

Outer Loop Design — Example 2

If $r_C(t) \equiv 0$, then

$$\begin{bmatrix} \ddot{r}_X(t) \\ \ddot{r}_Y(t) \end{bmatrix} = \frac{u_1(t)}{m} \begin{bmatrix} u_X(t) \\ u_Y(t) \end{bmatrix}.$$

- Design $u_X(\cdot)$ & $u_Y(\cdot)$ applying *any* control technique
- Find $\phi_{\text{ref}}(\cdot)$ & $\theta_{\text{ref}}(\cdot)$ s.t.

$$\begin{aligned} \sin \phi_{\text{ref}}(t) &= u_X(t) \sin \psi_{\text{ref}}(t) - u_Y(t) \cos \psi_{\text{ref}}(t), \\ \sin \theta_{\text{ref}}(t) \cos \phi_{\text{ref}}(t) &= u_X(t) \cos \psi_{\text{ref}}(t) + u_Y(t) \sin \psi_{\text{ref}}(t) \end{aligned}$$

Model Reference Adaptive Control (MRAC)

- Consider the **plant** & **reference model**

$$\begin{aligned}\dot{x}(t) &= Ax(t) + B\Lambda [u(t) + \Theta^T \Phi(t, x(t))], & x(t_0) &= x_0, & t &\geq t_0, \\ \dot{x}_{\text{ref}}(t) &= A_{\text{ref}}x_{\text{ref}}(t) + B_{\text{ref}}r(t), & x_{\text{ref}}(t_0) &= x_{\text{ref},0}\end{aligned}$$

$x(t) \in \mathcal{D}$, A , Θ unknown, (A, B) controllable & A_{ref} Hurwitz,

- Assume that $\exists K_x \in \mathbb{R}^{n \times m}$ & $K_r \in \mathbb{R}^{m \times m}$ s.t.

$$A_{\text{ref}} = A + B\Lambda K_x^T, \quad B_{\text{ref}} = B\Lambda K_r^T,$$

- Both K_x & K_r are unknown
- If A & Θ were known, then

$$u_{\text{ideal}} = K_x^T x + K_r^T r - \Theta^T \Phi(t, x)$$

would guarantee asymptotic convergence of $e(t) \triangleq x(t) - x_{\text{ref}}(t)$

Goals

Find $\hat{K}_x(\cdot)$, $\hat{K}_r(\cdot)$ & $\hat{\Theta}(\cdot)$ s.t.

$$u = \hat{K}_x^T(t)x + \hat{K}_r^T(t)r - \hat{\Theta}^T(t)\Phi(t, x)$$

- $e(t) \rightarrow 0$ as $t \rightarrow \infty$
- $e(\cdot)$, $\Delta K_x(\cdot)$, $\Delta K_r(\cdot)$ & $\Delta \Theta(\cdot)$ uniformly bounded

$$\Delta K_x(t) \triangleq \hat{K}_x(t) - K_x,$$

$$\Delta K_r(t) \triangleq \hat{K}_r(t) - K_r,$$

$$\Delta \Theta(t) \triangleq \hat{\Theta}(t) - \Theta,$$

Idea

Control law for $u(\cdot)$ mimics behavior of $u_{\text{ideal}}(\cdot)$

Classical MRAC

- Given $Q = Q^T > 0$, & let $P = P^T > 0$ s.t.

$$0 = A_{\text{ref}}^T P + P A_{\text{ref}} + Q$$

- Let

$$\dot{\hat{K}}_x^T = -\Gamma_x x(t) e^T(t) P B, \quad \hat{K}_x(t_0) = K_{x,0}, \quad t \geq t_0,$$

$$\dot{\hat{K}}_r^T = -\Gamma_r r(t) e^T(t) P B, \quad \hat{K}_r(t_0) = K_{r,0},$$

$$\dot{\hat{\Theta}}^T = \Gamma_{\Theta} \Phi(t, x(t)) e^T(t) P B, \quad \hat{\Theta}(t_0) = \Theta_0$$

- Then, $\lim_{t \rightarrow \infty} e(t) = 0$ uniformly in t_0 & $e(\cdot)$, $\Delta K_x(\cdot)$, $\Delta K_r(\cdot)$ & $\Delta \Theta(\cdot)$ uniformly bounded

e-Modification MRAC

- Consider the plant

$$\begin{aligned}\dot{x}(t) &= Ax(t) + B\Lambda [u(t) + \Theta^T \Phi(t, x(t))] + \xi(t), & x(t_0) &= x_0, \\ \dot{x}_{\text{ref}}(t) &= A_{\text{ref}}x_{\text{ref}}(t) + B_{\text{ref}}r(t), & x_{\text{ref}}(t_0) &= x_{\text{ref},0}\end{aligned}$$

where $\|\xi(t)\| \leq \xi_{\max}$,

- Let

$$\begin{aligned}\dot{\hat{K}}_x^T(t) &= -\Gamma_x x(t) e^T(t) PB - \sigma_x \|e^T(t) PB\| \hat{K}_x^T(t), & \hat{K}_x(t_0) &= K_{x,0}, \\ \dot{\hat{K}}_r^T(t) &= -\Gamma_r r(t) e^T(t) PB - \sigma_r \|e^T(t) PB\| \hat{K}_r^T(t), & \hat{K}_r(t_0) &= K_{r,0}, \\ \dot{\hat{\Theta}}^T(t) &= \Gamma_\Theta \Phi(t, x(t)) e^T(t) PB + \sigma_\Theta \|e^T(t) PB\| \hat{\Theta}^T(t), & \hat{\Theta}(t_0) &= \Theta_0\end{aligned}$$

- Then, $e(\cdot)$ uniformly ultimately bounded & $\Delta K_x(\cdot)$, $\Delta K_r(\cdot)$ & $\Delta \Theta(\cdot)$ uniformly bounded

Inner Loop Design — Example

- If r_C constant & unknown, inner-loop linearized equations:

$$\begin{bmatrix} \dot{z}(t) \\ \dot{\phi}(t) \\ \dot{\theta}(t) \\ \dot{\psi}(t) \\ \dot{v}_z(t) \\ \dot{\omega}(t) \end{bmatrix} = \begin{bmatrix} 0_{4 \times 4} & \mathbf{1}_4 \\ \begin{bmatrix} 0 & -g & 0 & 0 \\ g & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} & 0_{4 \times 4} \end{bmatrix} \begin{bmatrix} z(t) \\ \phi(t) \\ \theta(t) \\ \psi(t) \\ v_z(t) \\ \omega(t) \end{bmatrix} + \begin{bmatrix} 0_{4 \times 4} \\ \mathbf{1}_4 \end{bmatrix} \begin{bmatrix} -m^{-1} & 0 & 0 & 0 \\ 0 & I_x^{-1} & 0 & 0 \\ 0 & 0 & I_y^{-1} & 0 \\ 0 & 0 & 0 & I_z^{-1} \end{bmatrix} \\
 \cdot \left[\mathbf{u}(t) + \begin{bmatrix} 0_{1 \times 3} \\ r_C^\times \end{bmatrix} (F_g(\phi(t), \theta(t)) - m(\dot{v}_A(t) + \omega^\times(t)v_A(t))) \right], \\
 [z(0), \phi(0), \theta(0), \psi(0), v_z(0), \omega(0)]^T = [z_0, \phi_0, \theta_0, \psi_0, v_{z,0}, \omega_0]^T$$

- Same form as MRAC plant

CAD-Based Simulator

Simulation Environment

- Matlab's SimScape environment utilizes **accurate CAD models**;
- EoM not coded: solved from forces applied to CAD model;
- Includes **sensor noise** and **motor models**;
- **Disturbances**: **motor deficiencies** and **simulated gyro failures**;
- Includes **4 control laws**:
 - PID Control
 - Sliding Mode Control
 - **Model Reference Adaptive Control**
 - Adaptive Sliding Mode Control

http://lafflitto.com/Quad_Simulator.html to download simulator

Simulator Layout

Step 1

Initialize Variables

Step 2

Select a trajectory

Spiral Selected

Step 3

Select Control Laws

Adaptive Sliding Mode

Vehicle 1

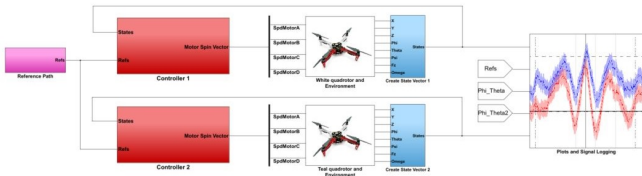
Model Reference Adaptive Control

Vehicle 2

Step 4

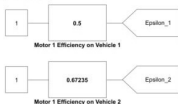
Run

Plots



Motor Deficiencies

Adjust Sliders to set efficiency of Motor 1 on the corresponding vehicles (1 is fully functioning, .1 is 90% efficiency loss). Deficiency occurs at 20 seconds in Simulation time.



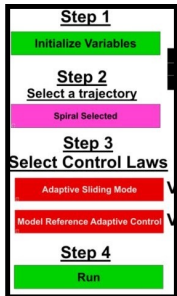
Sensor Failure

Gyro Working

Gyro Working

These two give the option to cause a sort of gyroscope failure. This failure causes the feedback of the phi angle to be 0.

Simulator Layout



Plots

Vehicle 1 Plots



Vehicle 2 Plots



Comparison Plots



3D Plots



Error Plots

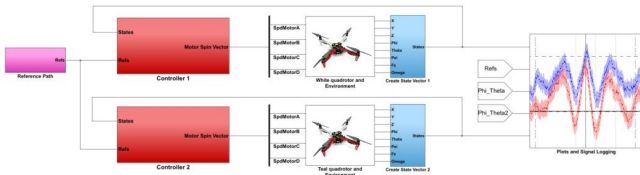


Control Laws Plots



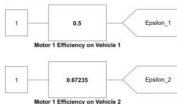
Vehicle 1

Vehicle 2



Motor Deficiencies

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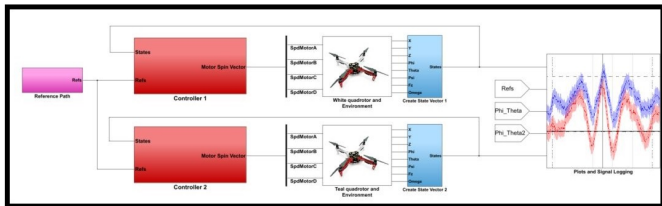
Model Reference Adaptive Control

Vehicle 2

Step 4

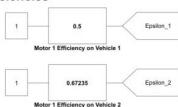
Run

Plots



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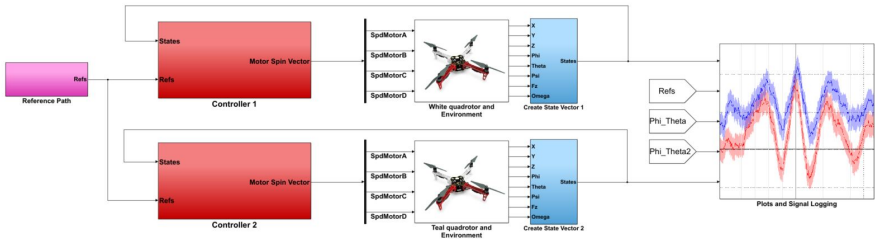
Sensor Failure

Gyro Working

Gyro Working

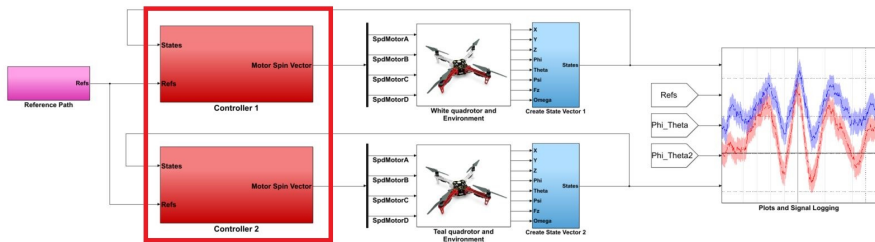
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Simulator Layout II



- Reference trajectory generation;
- Control algorithms;
- Vehicle dynamics;
- Integrate differential equations and add noise

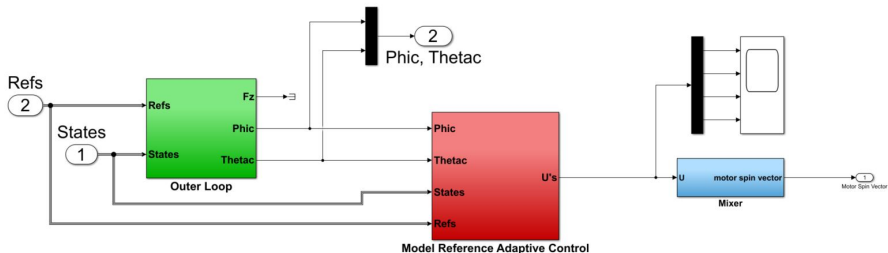
Simulator Layout II



- Reference trajectory generation;
- Control algorithms;
- Vehicle dynamics;
- Integrate differential equations and add noise

Simulator Layout III

Control blocks all have the same layout:



- Outer loop control;
- Inner loop control;
- Mixer

Tips for Tuning MRAC

Reference Model

- Design **frequency response** of **reference model** to **match the real system**;
- For quadcopters, the **outer loop** is **slower** than the **inner loop**;

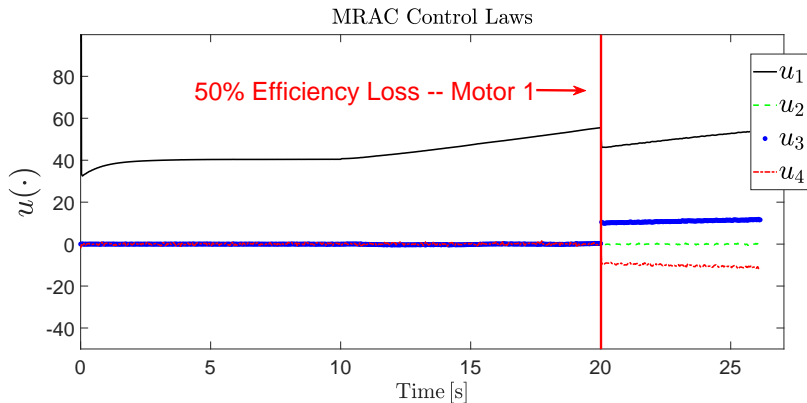
Tuning the Adaptive Rates

- **Start small**: $\Gamma \approx 1 - 10$;
- **Increase rates** until **oscillations** begin, then **decrease**;
- In our experience:
 - Γ_x is **most sensitive**;
 - **Higher Γ_r** = **better command tracking**;
 - **Higher Γ_x, Γ_Θ** = **better disturbance rejection**

MRAC vs. PID

- White quadrotor – MRAC;
- Black quadrotor – PID;
- **Mission**: Takeoff to $1m$, follow square trajectory $4m$ sides;
- Both experience **50% efficiency loss** in the front motor (**red arm**)

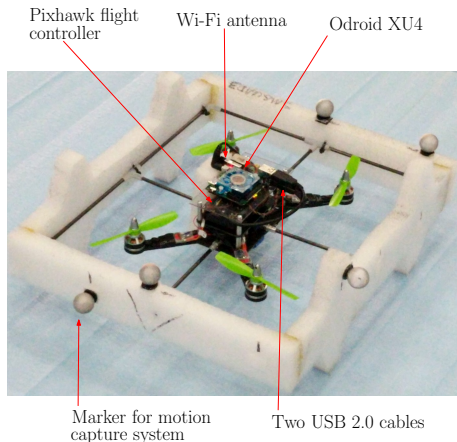
MRAC Controls Plot



MRAC controls change rapidly at $t = 20s$ to handle the **motor failure disturbance**

Quadcopter Architecture

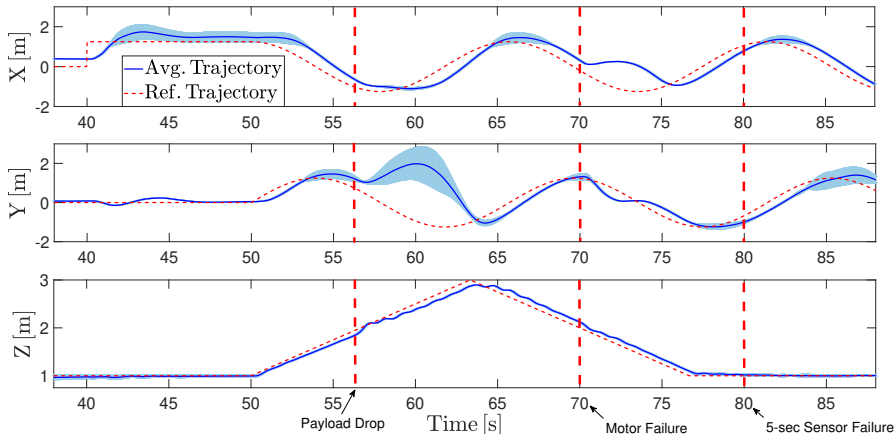
- Pixhawk flight controller used for inertial measurement unit and actuation
- ODroid XU4 companion computer to calculate controls
- QGroundControl to communicate with Pixhawk
- VICON motion capture system to deduce UAV position



Mission Scenario

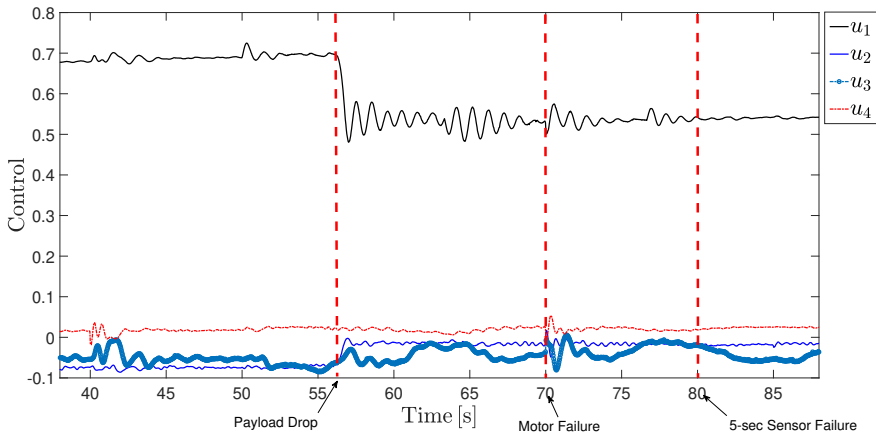
- Carry a 200g slung payload,
- Takeoff to 1m height,
- Move forward 1m in X ,
- Follow a spiral shaped trajectory,
- Slung payload is dropped at $t = 57\text{s}$,
- 25% motor failure of one motor at $t = 70\text{s}$,
- Roll angle frozen at $t = 80\text{s}$ for 5 seconds,
- Maintain $\psi(t) = 0, t \geq 0$,
- Land at takeoff point.

MRAC Trajectory Plot



Trajectory tracking deviates greatly when the system is **disturbed**, yet MRAC drives the disturbed system to the **desired trajectory**

MRAC Controls Plot



Controls drop to handle **payload drop** at $t = 57$ s.

Controls react to compensate for **motor failure** at time $t = 70$ s

Taguchi Method

Problem

- No known, systematic method to **tune nonlinear controllers**

Approach

- Input: **Independent variables** to test & **response variable** to measure
- Output: **Values of independent variables** that will **minimize the response variable**
- **Exploit** experimenter's **familiarity with the system** & **analysis of variation**

Taguchi Method Steps

- 1 Determine** number of **independent variables** I
 - Up to the experimenter's discretion
- 2 Determine** response variable $\bar{e}_i \triangleq \frac{1}{N} \sum_{l_i=1}^N e_i(t_{l_i})$
- 3 Determine** number of **level settings** k , per independent variable
- 4 Determine** minimum number of experiments
 $E \triangleq I(k - 1) + 1$
- 5 Select** orthogonal array
- 6 Run experiments** and **record** response variable
- 7 Perform analysis of variance** on **response variable** for each independent variable

Analysis of Variance (ANOVA)

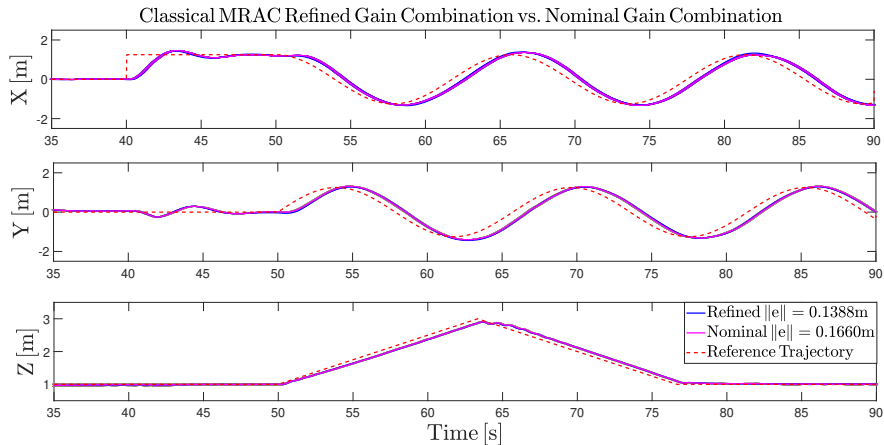
- Signal-to-noise ratio (S/N): quantify how response variable is affected by noise.
- S/N calculated for each independent variable & level setting
 - $S/N_{i,j} = -10 \log(\frac{1}{n} \sum_{m=1}^n \bar{e}_{m,j}^2)$, $i = 1, \dots, I$, $j = 1, \dots, k$,
where n is the number of times a level setting is tested
- Level setting with lowest S/N ratio selected for that independent variable
- Change in S/N indicates most sensitive independent variable

Results of Taguchi Method

- Classical MRAC & e -MRAC laws tuned using trial & error method
 - These gains dubbed the “nominal gains”
- **Result:** “refined gain combination” & most sensitive i. variables

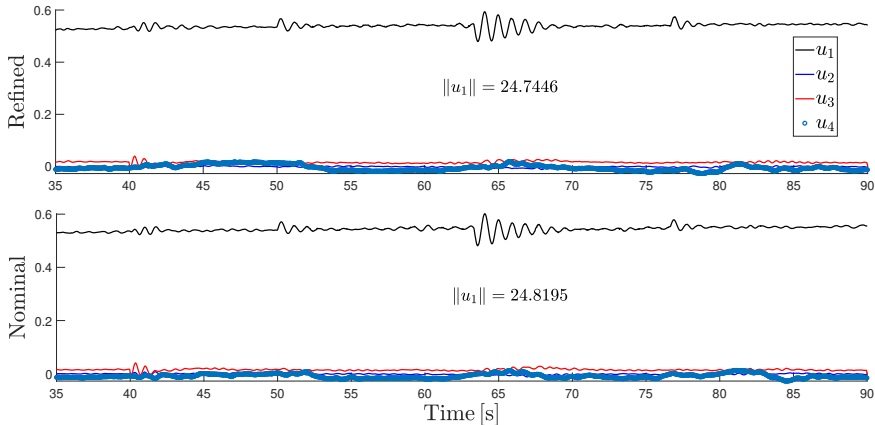
	Battery	MRAC	e -MRAC
Nominal \bar{e} (meters)	Fresh (16.5V)	0.1771	0.1958
	Med (15.9V)	0.1736	0.1812
	Low (15.1V)	0.1474	0.1650
	Average	0.1660	0.1807
Refined \bar{e} (meters)	Fresh (16.5V)	0.1397	0.1374
	Med (15.9V)	0.1383	0.1424
	Low (15.1V)	0.1385	0.1290
	Average	0.1388	0.1363
	% Change	-16.36	-24.58

MRAC Trajectory Plot



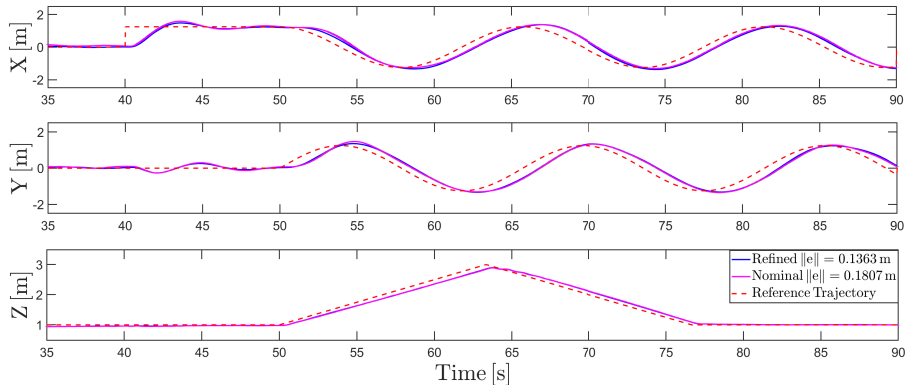
MRAC Controls Plot

Classical MRAC Refined Gain Combination vs. Nominal Gain Combination

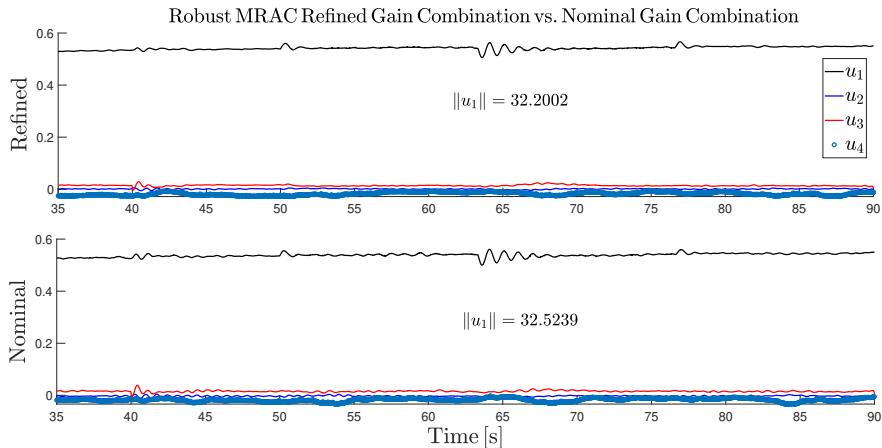


e-MRAC Trajectory Plot

Robust MRAC Refined Gain Combination vs. Nominal Gain Combination



e-MRAC Controls Plot



Questions?

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QUESTIONS?

