

Wind tolerant controller and robust INS/GPS sensor fusion architecture for multirotor UAV

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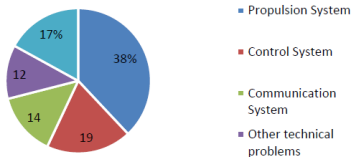
Motivations

UAVs Reliability



UAV crash

Failure Modes of Multirotors

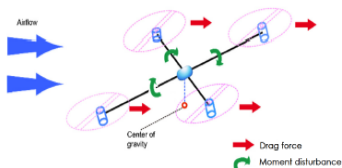


Failure rate on a conventional aircraft

[1] S. Reimann, J. Amos, E. Bergquist, J. Cole, J. Phillips, S. Shuster, UAV for Reliability, AEM 4331 - Aerospace Vehicle Design, December 2013.

Problem formulation

Wind perturbations on system



- Additional uncertainties
- Induced disturbance forces and moments
- Loss of stability

Problem formulation

Onboard sensors vulnerable to hardware faults



IMU



Barometer



GPS unit



Compass

Aim of Our Work

- Maintaining system performance and stability in the presence of model uncertainties and external perturbations.
- Developing a strategy to cope with sensor faults.

Main Contributions

- Proposition of a nonlinear observer based on super-twisting theory to estimate the wind forces.
- Proposition of new EKF based GPS/INS fusion architecture to detect and isolate faulty sensors and software issues.

Platform

Quadrotor S500 frame



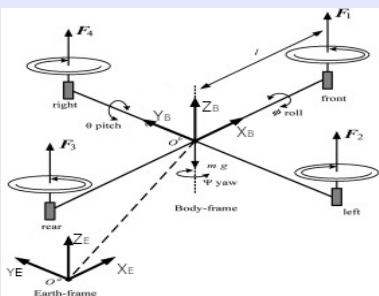
System's Parameters

Mass m	1.1 kg
Inertia I_{xx}, I_{yy}	$2.2 * 10^{-2} \text{Kg.m}^2$
Inertia I_{zz}	$5.5 * 10^{-2} \text{Kg.m}^2$
Thrust factor K_f	$2.75 * 10^{-5} \text{Ns}^2/\text{rad}^2$
Drag factor K_t	$3.6 * 10^{-7} \text{Nm}/\text{rad}^2$
Length of the arm l	0.22 m
Rotor's Inertia J_r	negligible

The thrust and torque coefficients are provided by the manufacturer (www.dji.com).

State variables and Control inputs

Schematic representation



$O_B: \{O_B, X_B, Y_B, Z_B\}$ Body-fixed frame

$O_E: \{O_E, X_E, Y_E, Z_E\}$ Earth-fixed frame

State Variables

position	x, y
altitude	z
roll	ϕ
pitch	θ
yaw	ψ
roll velocity	p
pitch velocity	q
yaw velocity	r

Virtual control inputs

Total thrust	u_t
roll torque	τ_ϕ
pitch torque	τ_θ
yaw torque	τ_ψ

Real inputs

motors speeds $\omega_i \quad i = 1, \dots, 4$

NonLinear Model

The Quadrotor's dynamics are written as [5]:

$$\left\{ \begin{array}{l} \ddot{x} = (\cos \phi \sin \theta \cos \psi + \sin \phi \sin \psi) * \frac{u_t}{m} \\ \ddot{y} = (\cos \phi \sin \theta \sin \psi - \sin \phi \cos \psi) * \frac{u_t}{m} \\ \ddot{z} = (\cos \phi \cos \theta) \frac{u_t}{m} - g \\ \ddot{\phi} = \frac{I_{yy} - I_{zz}}{I_{xx}} \dot{\theta} \dot{\psi} - \frac{J_r}{I_{xx}} \dot{\theta} \Omega + \frac{1}{I_{xx}} \tau_{\phi} \\ \ddot{\theta} = \frac{I_{zz} - I_{xx}}{I_{yy}} \dot{\phi} \dot{\psi} - \frac{J_r}{I_{yy}} \dot{\phi} \Omega + \frac{1}{I_{yy}} \tau_{\theta} \\ \ddot{\psi} = \frac{I_{xx} - I_{yy}}{I_{zz}} \dot{\phi} \dot{\theta} + \frac{1}{I_{zz}} \tau_{\psi} \end{array} \right. \quad (1)$$

The relation between the virtual inputs and the motors speeds:

$$\left\{ \begin{array}{l} u_t = F_1 + F_2 + F_3 + F_4 \\ \tau_{\phi} = (F_4 - F_2) * l \\ \tau_{\theta} = (F_3 - F_1) * l \\ \tau_{\psi} = (\tau_1 + \tau_3) - (\tau_2 + \tau_4) \end{array} \right. \quad (2)$$

Generalities

nonlinear system general form

$$\underline{x}^{(n)} = f(\underline{x}, \dot{\underline{x}}, \ddot{\underline{x}}, \dots, \underline{x}^{(n-1)}, t) + g(\underline{x}, \dot{\underline{x}}, \ddot{\underline{x}}, \dots, \underline{x}^{(n-1)}, t)u + w(t) \quad (3)$$

- $\underline{x}^{(n)}$: state vector
- u : virtual input vector
- f : modeled dynamics function
- g : control function
- $w(t)$: unmodeled dynamics

Uncertain functions Boundaries

$$\begin{aligned} |f(\underline{x}, t) - \hat{f}(\underline{x}, t)| &\leq F(\underline{x}, t) \\ |g(\underline{x}, t) - \hat{g}(\underline{x}, t)| &\leq G(\underline{x}, t) \\ |w(\underline{x}, t) - \hat{w}(\underline{x}, t)| &\leq W(\underline{x}, t) \\ |\dot{w}(\underline{x}, t) - \dot{\hat{w}}(\underline{x}, t)| &\leq \delta(\underline{x}, t) \end{aligned} \quad (4)$$

Sliding variable

- Defining the tracking error as :

$$\tilde{x} = x - x_d \quad (5)$$

- Introducing sliding variable :

$$s = \dot{\tilde{x}} + \lambda \tilde{x} \quad (6)$$

Sliding variable properties

$$s \begin{cases} \dot{s} \text{ contains } u \\ s \mapsto 0 \text{ when } t \mapsto +\infty \Rightarrow \tilde{x} \mapsto 0 \end{cases} \quad (7)$$

Lyapunov function

- Positive definite Lyapunov function :

$$V(t) = \frac{1}{2}s^2 \quad (8)$$

- Control law :

$$\begin{aligned} u &= \frac{1}{\hat{g}}(\hat{u} - k * \text{sign}(s)) \\ \hat{u} &= -\hat{f} + \ddot{x}_d + \lambda \tilde{x} \end{aligned} \quad (9)$$

Lyapunov condition

$$\dot{V}(t) = s\dot{s} = s(f - \hat{f}) + w(t)s - k|s| \leq -\eta|s| < 0 \quad (10)$$

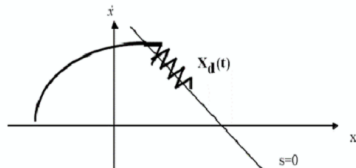
- To ensure this condition:

Gain

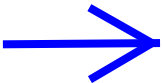
$$k = F + W + \eta \quad (11)$$

Problems

1) Chattering



2) Over-estimation of F & W



Motor saturation & chattering

3) Under-estimation of F & W



Possibility of Lyapunov instability

Solutions

- Chattering

Super-twisting algorithm

$$u(t) = u_1 + u_2 \begin{cases} u_1 = -\alpha_1 |s|^\tau \text{sign}(s), & \tau \in]0, 0.5] \\ \dot{u}_2 = -\alpha_2 \text{sign}(s) \end{cases} \quad (12)$$

- Over/under-estimation of F :

Observer-based controller

$$\hat{u} = -\hat{f} - \hat{f}_{wind} + \ddot{x}_d + \lambda \tilde{x} \quad (13)$$

$$\dot{V}(t) = s\dot{s} = s(f - \hat{f}) + s(f_{wind} - \hat{f}_{wind}) + w(t) - k|s| \leq -\eta|s| \quad (14)$$

Wind estimation

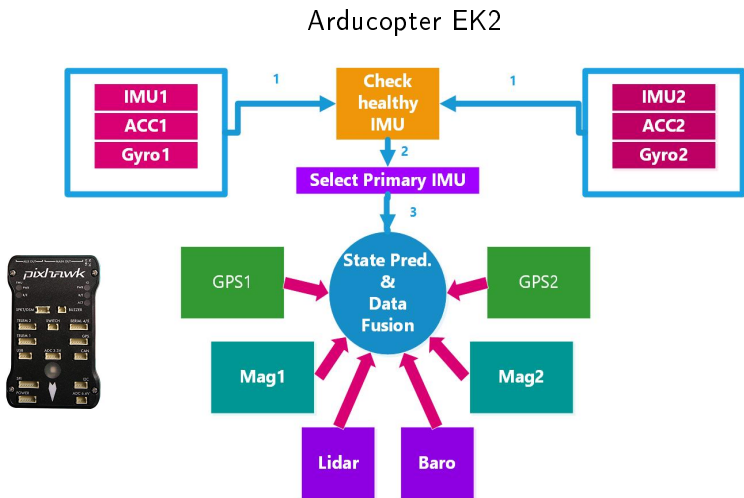
Nonlinear Super-twisting Observer

$$\left\{ \begin{array}{l} \dot{z}_0 = v_0 + \frac{u_t}{m} u_x - K_p(\dot{x} - \dot{x}_d) \\ v_0 = -\alpha_{0x}|z_0 - \sigma|^{2/3} \text{sgn}(z_0 - \sigma) + z_1 \\ \dot{z}_1 = v_1, \quad v_1 = -\alpha_{1x}|z_1 - v_0|^{1/2} \text{sgn}(z_1 - v_0) + z_2 \\ \dot{z}_2 = -\alpha_{2x} \text{sgn}(z_2 - v_1) \\ \hat{F}_x = z_1 \end{array} \right. \quad (15)$$

[1] J. Davila, L. Fridman, and A. Levant, "Second-order sliding-mode observer for mechanical systems," IEEE Transactions on Automatic Control, vol. 50, no. 11, Novembre 2005.

[2] Y. B. Shtessel, I. A. Shkolnikov, and A. Levant, "Smooth second-order sliding modes: Missile guidance application," Automatica, vol. 43, no.8, 2007.

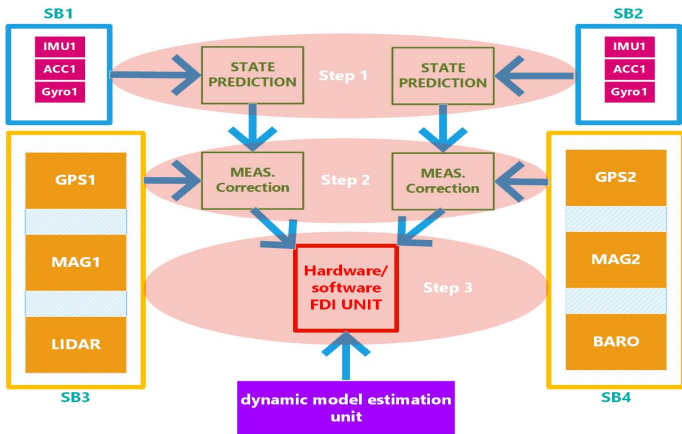
Multi-sensor data fusion architecture



Riseborough, Paul. "Application of data fusion to aerial robotics." Proc. of Embedded Linux Conference.

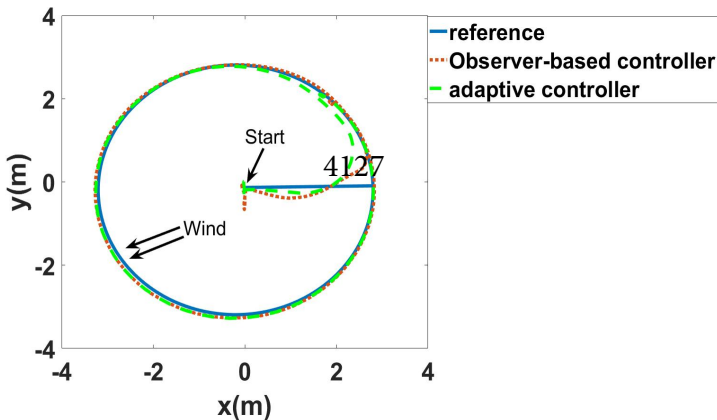
Multi-sensor data fusion architecture

Proposed architecture



Simulation - Adaptive Vs observer-based controller

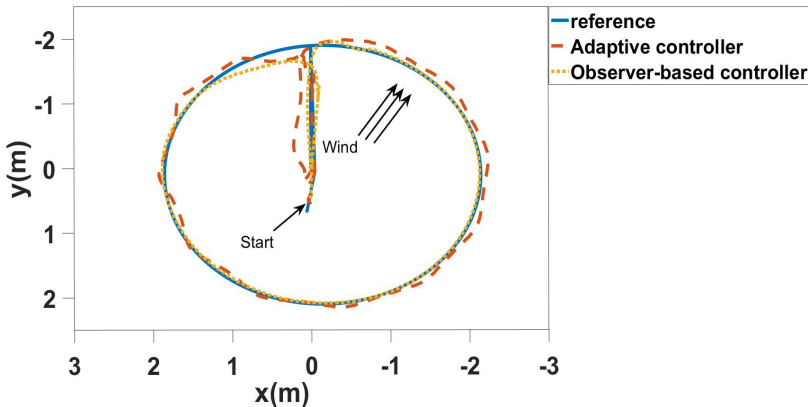
Small wind forces < 1 N



S. Rajappa, C. Masone, and P. Stegagno, "Adaptive super twisting controller for a quadrotor uav," *IEEE International Conference on Robotics and Automation (ICRA)*, pp. 2971–2977, 2016.

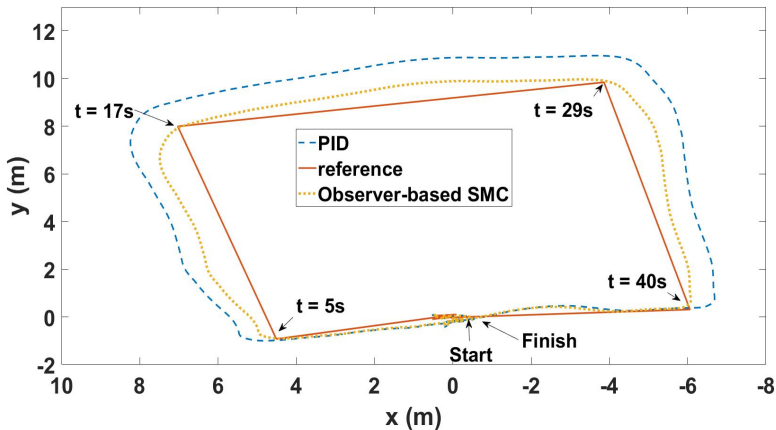
Simulation - Adaptive Vs observer-based controller

Wind forces $\sim 5-6$ N

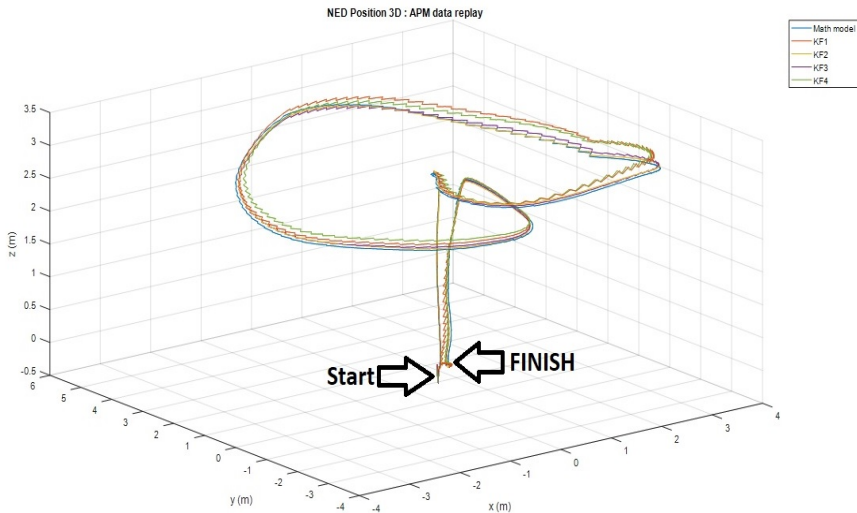


Experiment - PID Vs observer-based controller

Average wind forces ~ 0.13 N



Simulation - Data fusion architecture



Video

Real flight Test of Observer-based STA Controller Robust to Wind Perturbation for Multirotor UAV



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Thank You

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