



An Aerodynamic Model-aided State Estimator for Multirotor UAVs



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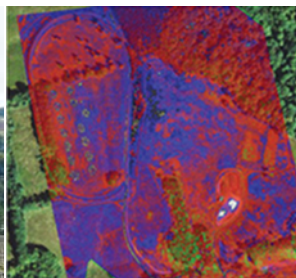
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Multi-rotor UAVs



Agriculture



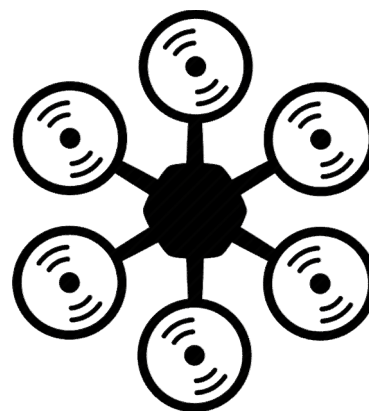
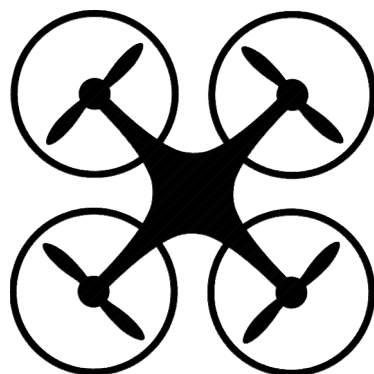
Entertainment



Service robots

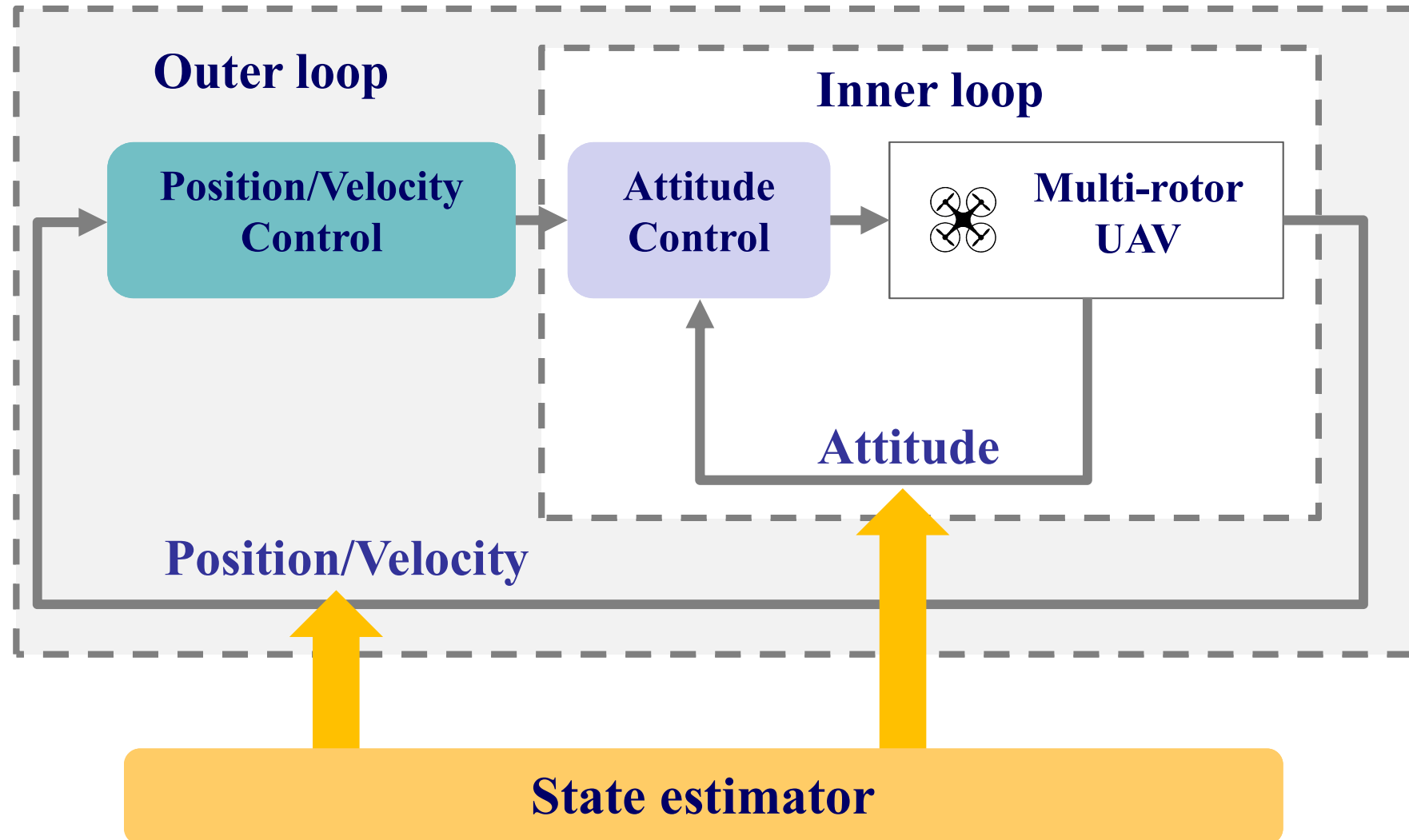


Photography



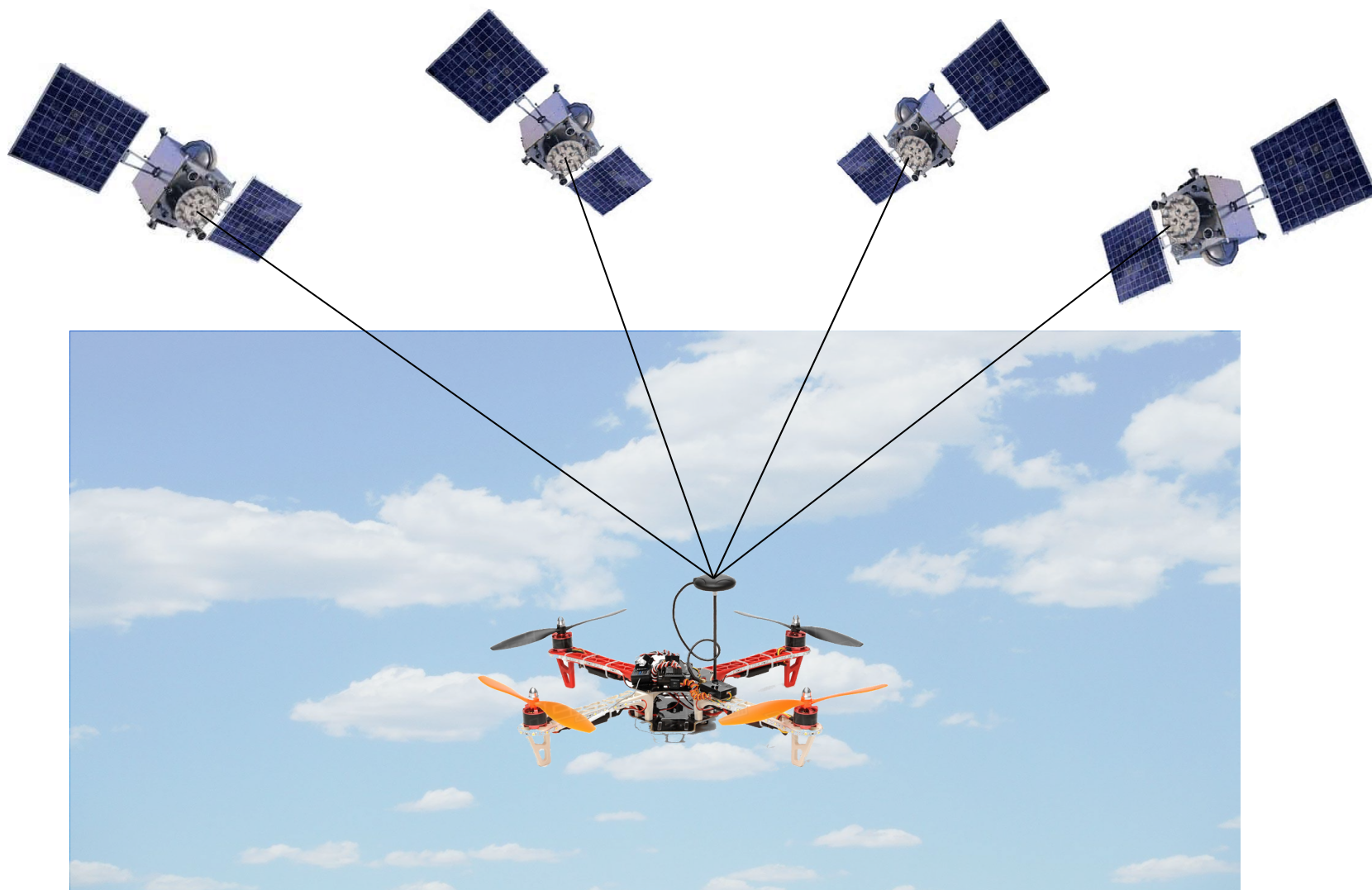


A typical control loop of a multi-rotor UAV





Open area





GPS-denied environments

Vision-based State Estimator:

Optical flow + Ultrasonic sensor

Visual odometry/SLAM

Visual-inertial odometry/SLAM



PX4flow



VI sensor



DJI Guidance

Texture-less
Dark scenes
Over-exposure

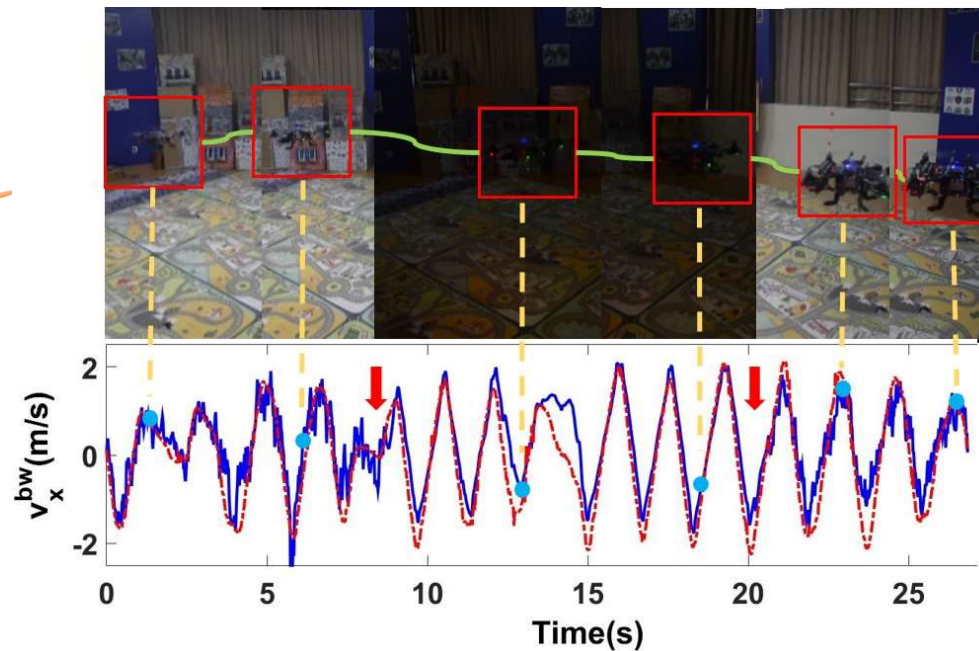
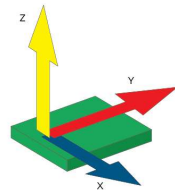




Aerodynamic model-aided state estimator

Special aerodynamic characteristic of multi-rotor UAVs:

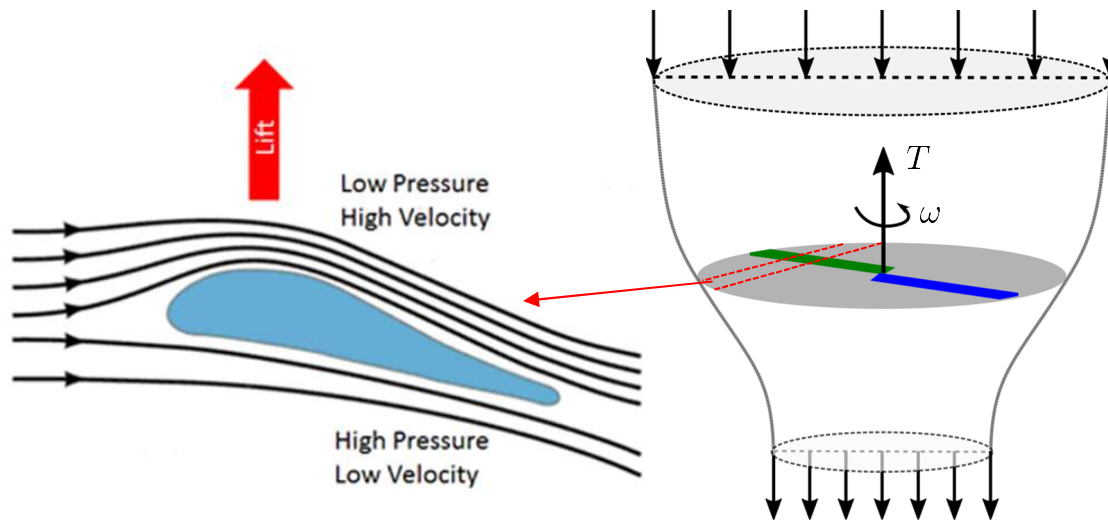
The horizontal velocity of the vehicle is **observable** from the accelerometer measurements.





Rotor Aerodynamics

Force in hovering state



$$T = c_T \omega^2$$

The thrust force T

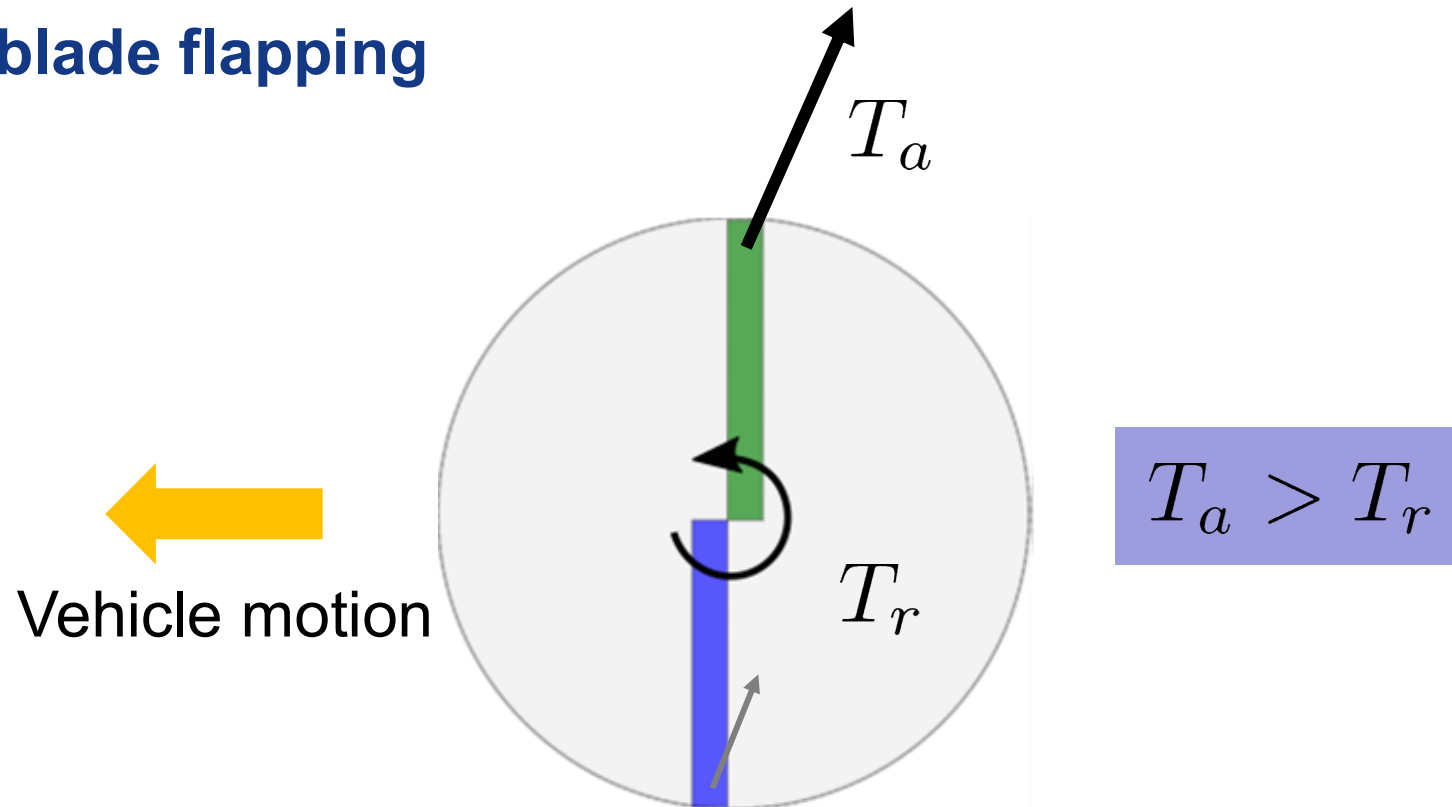
- 1) is perpendicular to the rotor plane
- 2) is proportional to the square of the spinning speed of the rotor



Rotor Aerodynamics

Forces in translation movement :

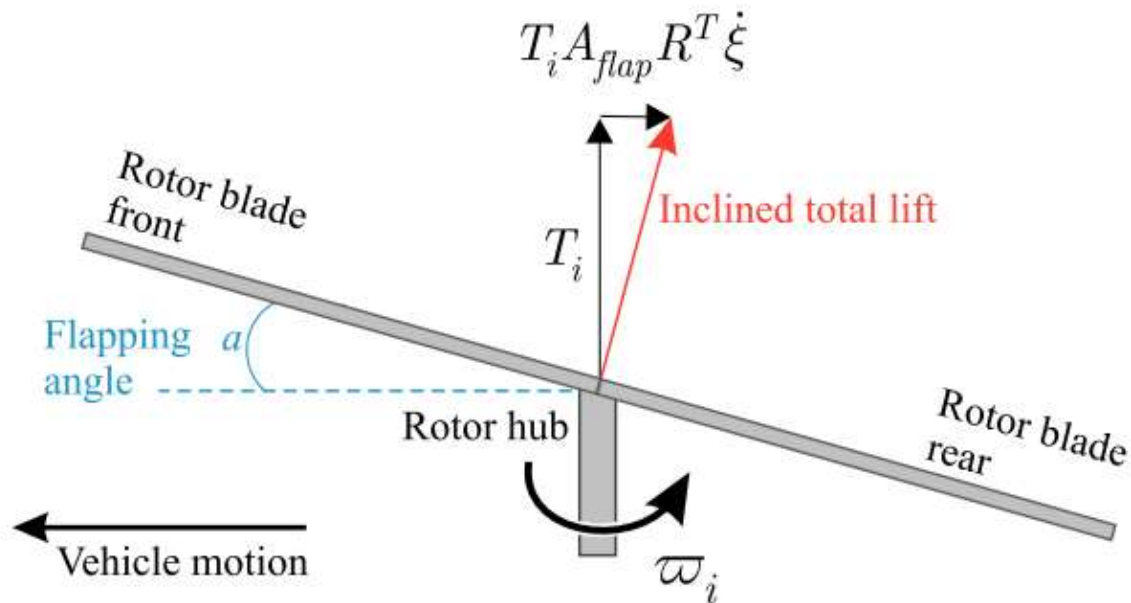
- blade flapping





Rotor Aerodynamics

- Forces in translation movement :
 - blade flapping

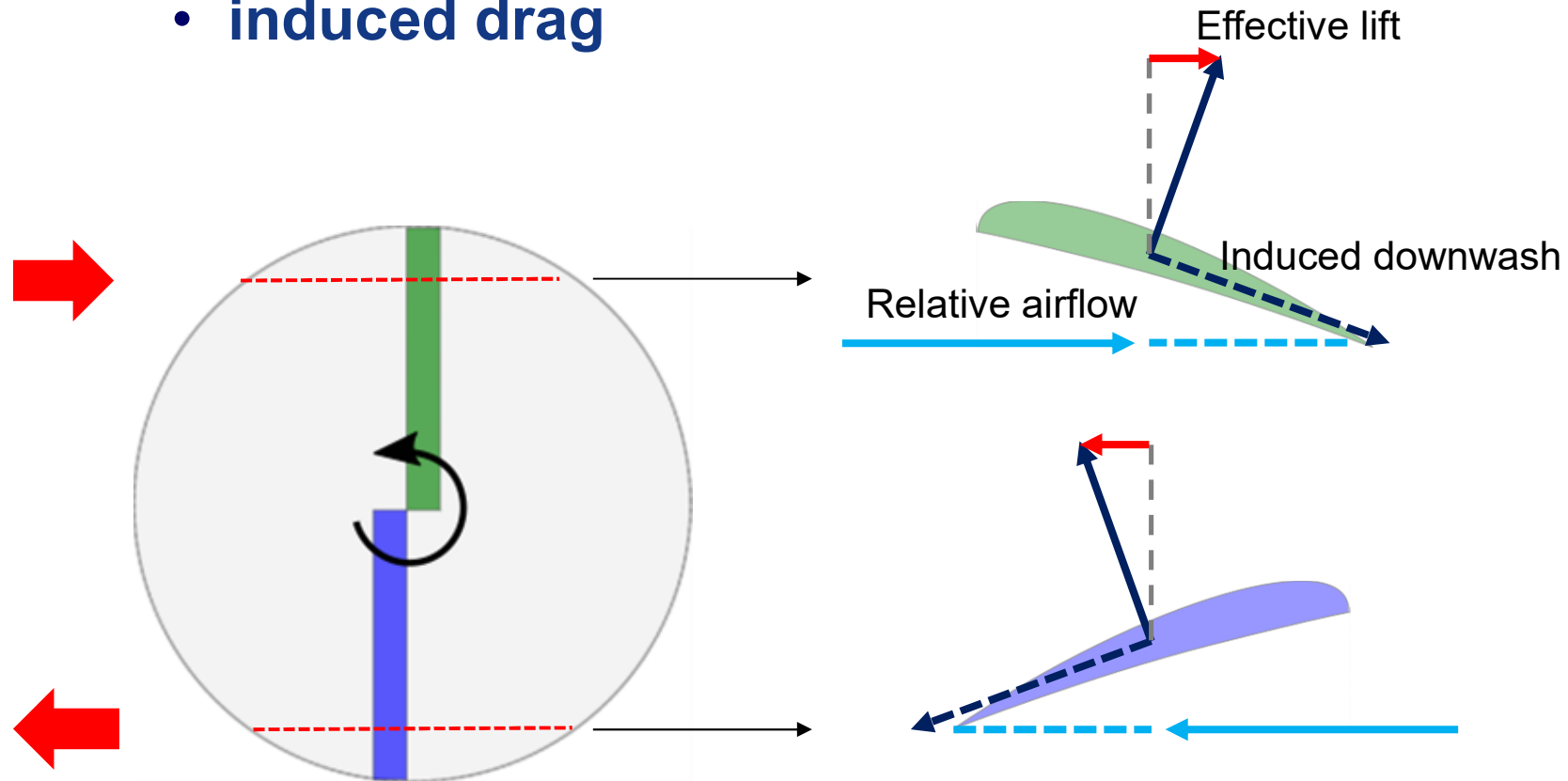




Rotor Aerodynamics

Induced drag in hovering :

- induced drag

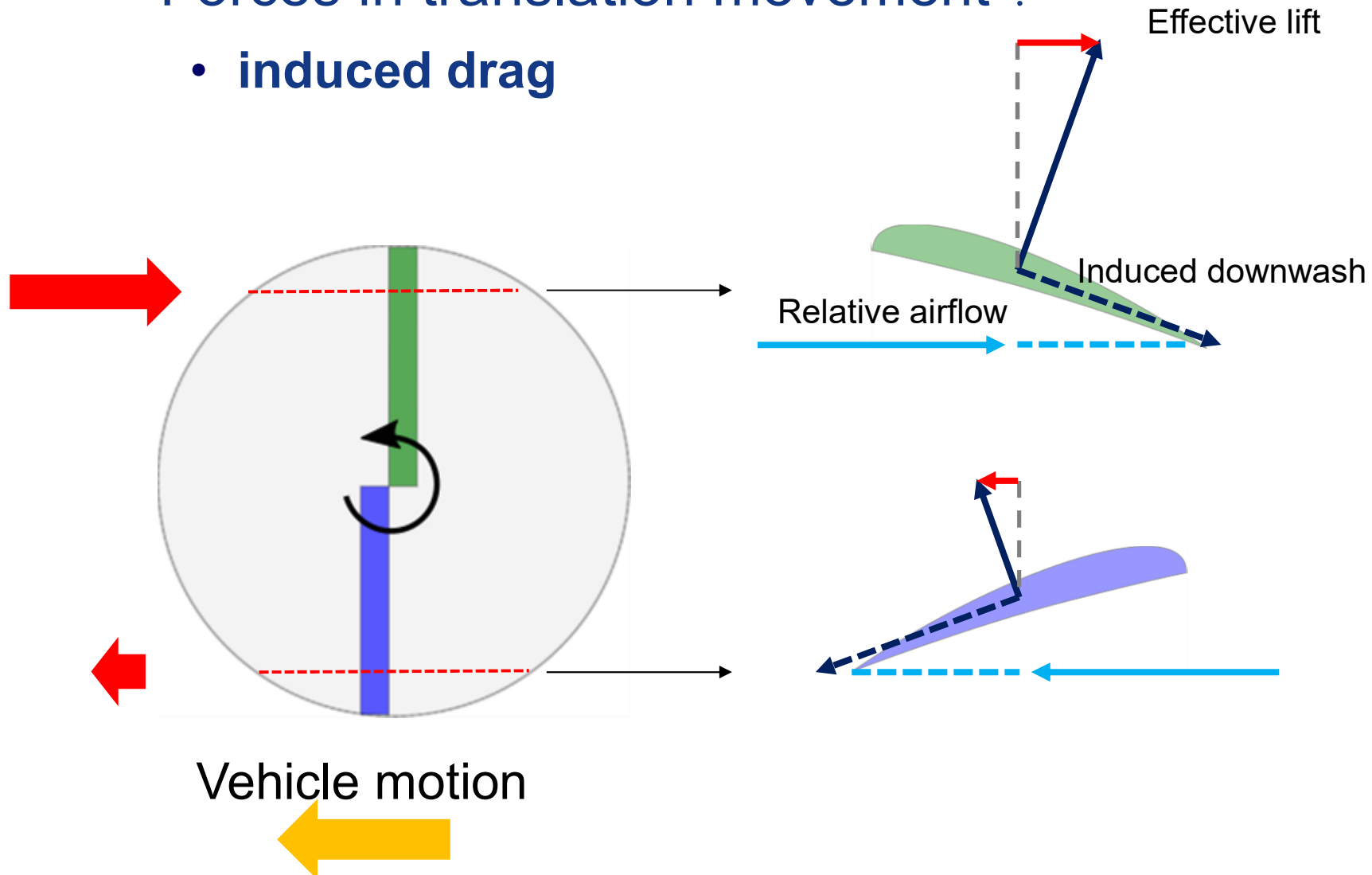




Rotor Aerodynamics

Forces in translation movement :

- induced drag





Rotor Aerodynamics

- When the UAV is in translation flight, it receives an horizontal force that is a combination of blade-flapping force and the induced drag,

$$H_i \approx \lambda \omega_i V_C^\perp$$

constant

Horizontal velocity in body frame

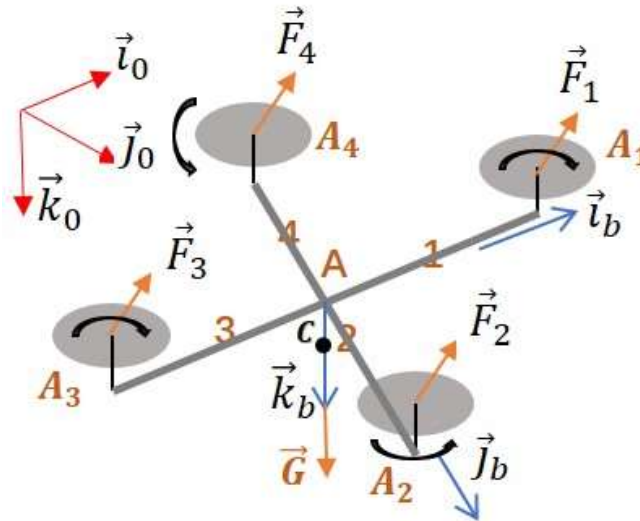
Spinning speed of the rotor

Principles of Helicopter Aerodynamics, Section 3.5



Multi-rotor Aerodynamics

- The total force generated by the rotors:

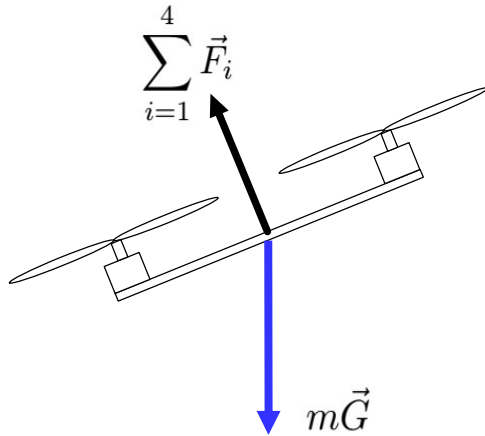


$$\sum_{i=1}^4 \vec{F}_i \approx -a \left(\sum_{i=1}^4 \omega_i^2 \right) \vec{k}_b - \lambda_1 \left(\sum_{i=1}^4 \omega_i \right) \vec{V}_C^\perp$$

Martin, Philippe, and Erwan Salaün. "The true role of accelerometer feedback in quadrotor control." Robotics and Automation (ICRA), 2010 IEEE International Conference on. IEEE, 2010.



Multi-rotor Aerodynamics



1) Accelerometer measures the special force (exclude the gravity) :

$$\sum_{i=1}^4 \vec{F}_i \approx -\alpha \left(\sum_{i=1}^4 \omega_i^2 \right) \vec{k}_b - \lambda \left(\sum_{i=1}^4 \omega_i \right) \vec{V}_C^\perp$$

2) Dynamic model from Newton's Second law:

$$\vec{F}_C = m\dot{\vec{V}}_C = m\vec{G} - \alpha \left(\sum_{i=1}^4 \omega_i^2 \right) \vec{k}_b - \lambda \left(\sum_{i=1}^4 \omega_i \right) \vec{V}_C^\perp$$

➡

$$\begin{aligned} \dot{\vec{V}}_C^b = & (-G \sin \theta - \mu \cdot v_x^b - (\omega_y^b v_z^b - \omega_z^b v_y^b)) \vec{i}_b \\ & + (G \cos \theta \sin \phi - \mu \cdot v_y^b - (\omega_z^b v_x^b - \omega_x^b v_z^b)) \vec{j}_b \\ & + (G \cos \theta \cos \phi - a_z - (\omega_x^b v_y^b - \omega_y^b v_x^b)) \vec{k}_b \end{aligned}$$



Multi-rotor Aerodynamics

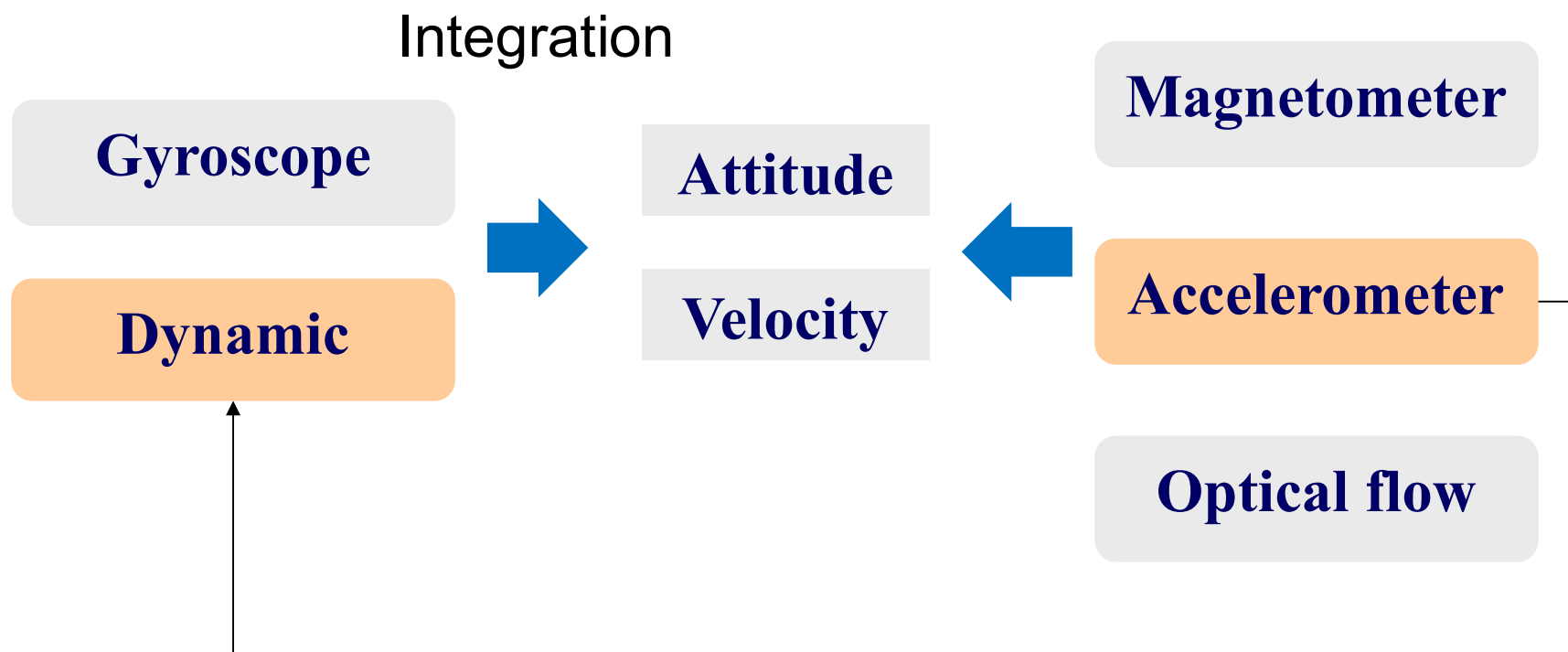


Two cues of multi-rotor aerodynamics:

- Body velocities in the rotor plane is observable
- Roll/pitch angle is observable*



Aerodynamic model-aided state estimator





Estimator Design

State definition & propagation:

$$\mathbf{x} = [\phi \ \theta \ \psi \ \beta_{g_x} \ \beta_{g_y} \ \beta_{g_z} \ v_x^b \ v_y^b \ v_z^b \ h \ \mu_x \ \mu_y]^T \quad \} \text{ state vector}$$

$$\left\{ \begin{array}{l} \dot{\phi} = \hat{\omega}_x^b + \hat{\omega}_y^b \tan \theta \sin \phi + \hat{\omega}_z^b \tan \theta \cos \phi \\ \dot{\theta} = \hat{\omega}_y^b \cos \phi - \hat{\omega}_z^b \sin \phi \\ \dot{\psi} = \hat{\omega}_y^b \sin \phi / \cos \theta + \hat{\omega}_z^b \cos \phi / \cos \theta \\ \dot{\beta}_{g_x}^b = -\frac{1}{\tau_{g_x}^b} \beta_{g_x}^b + \delta \beta_{g_x} \\ \dot{\beta}_{g_y}^b = -\frac{1}{\tau_{g_y}^b} \beta_{g_y}^b + \delta \beta_{g_y} \\ \dot{\beta}_{g_z}^b = -\frac{1}{\tau_{g_z}^b} \beta_{g_z}^b + \delta \beta_{g_z} \\ \dot{v}_x^b = -g \sin \theta + \mu_x v_x^b - (\hat{\omega}_y^b v_z^b - \hat{\omega}_z^b v_y^b) \\ \dot{v}_y^b = g \cos \theta \sin \phi + \mu_y v_y^b - (\hat{\omega}_z^b v_x^b - \hat{\omega}_x^b v_z^b) \\ \dot{v}_z^b = g \cos \theta \cos \phi + a_z - (\hat{\omega}_x^b v_y^b - \hat{\omega}_y^b v_x^b) \\ \dot{h} = -\sin \theta v_x^b + \cos \theta \sin \phi v_y^b + \cos \theta \cos \phi v_z^b \\ \dot{\mu}_x = \delta \mu_x \\ \dot{\mu}_y = \delta \mu_y, \end{array} \right. \quad \left. \begin{array}{l} \} \text{ attitude} \\ \} \text{ imu bias} \\ \} \text{ velocity} \\ \} \text{ height} \\ \} \text{ aerodynamic coefficients} \end{array} \right.$$



Estimator Design

Measurement Model

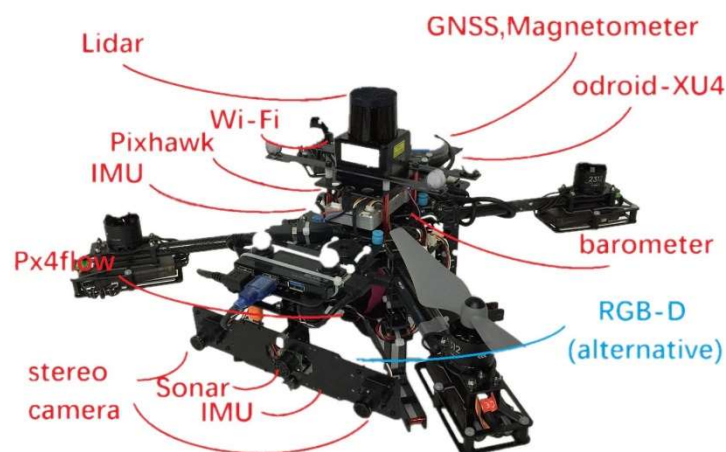
$$\mathbf{x} = [\phi \ \theta \ \psi \ \beta_{g_x} \ \beta_{g_y} \ \beta_{g_z} \ v_x^b \ v_y^b \ v_z^b \ h \ \mu_x \ \mu_y]^T \quad \} \text{ state vector}$$

$$\left\{ \begin{array}{l} Acc_x = \mu_x \cdot v_x^b + \delta_{a_x^b} \\ Acc_y = \mu_y \cdot v_y^b + \delta_{a_y^b} \end{array} \right. \quad \} \text{ accelerometer}$$
$$\left\{ \begin{array}{l} \hat{v}_x^b = v_x^b + \delta_{v_x^b} \\ \hat{v}_y^b = v_y^b + \delta_{v_y^b} \end{array} \right. \quad \} \text{ optical flow}$$
$$\hat{\psi} = \psi + \delta_{\psi} \quad \} \text{ magnetometer}$$
$$\hat{h} = h + \delta_h \quad \} \text{ sonar}$$



Experiment

- We test our estimator in real scenes with real platforms:



Our home-made quadcopter
Onboard computation



Parrot's ARDrone 2.0
Off board computation

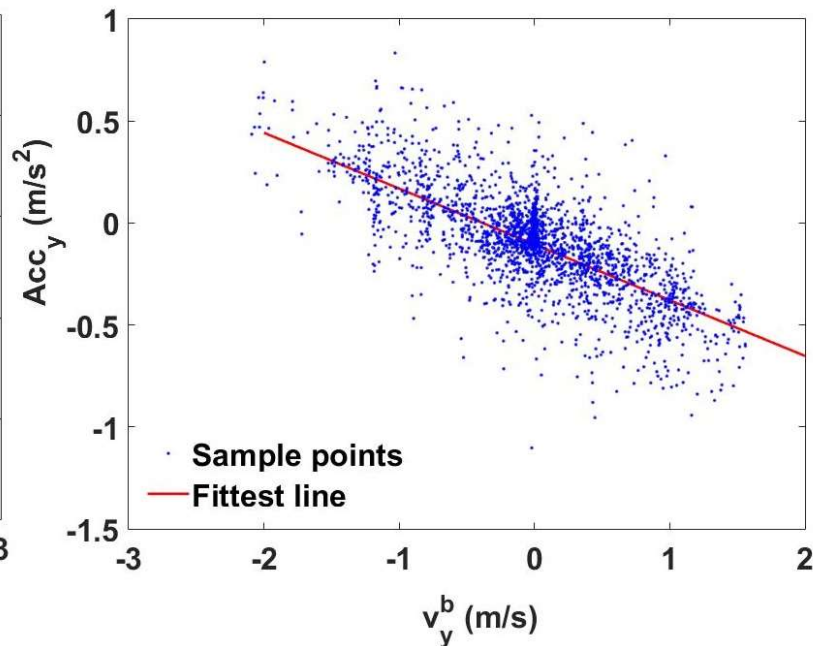
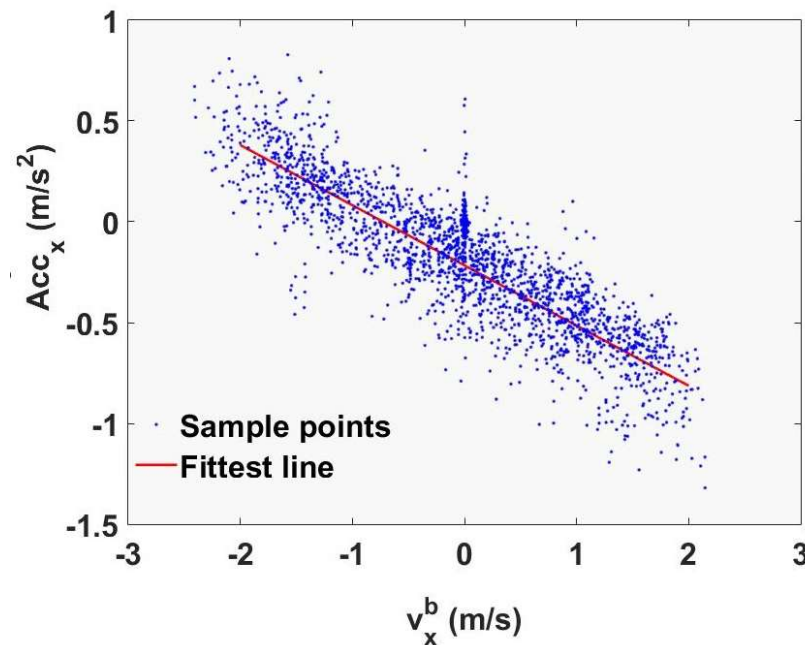


Experiment

Drag Coefficients Initialization by line fitting

$$\tilde{\mu}_x = \min_{\mu_x} (Acc_x - \mu_x v_x^b)^2$$

$$\tilde{\mu}_y = \min_{\mu_y} (Acc_y - \mu_y v_y^b)^2$$

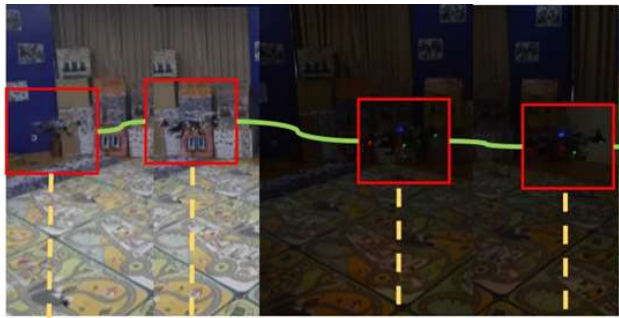


Left: line fitting of μ_x , **Right:** line fitting of μ_y



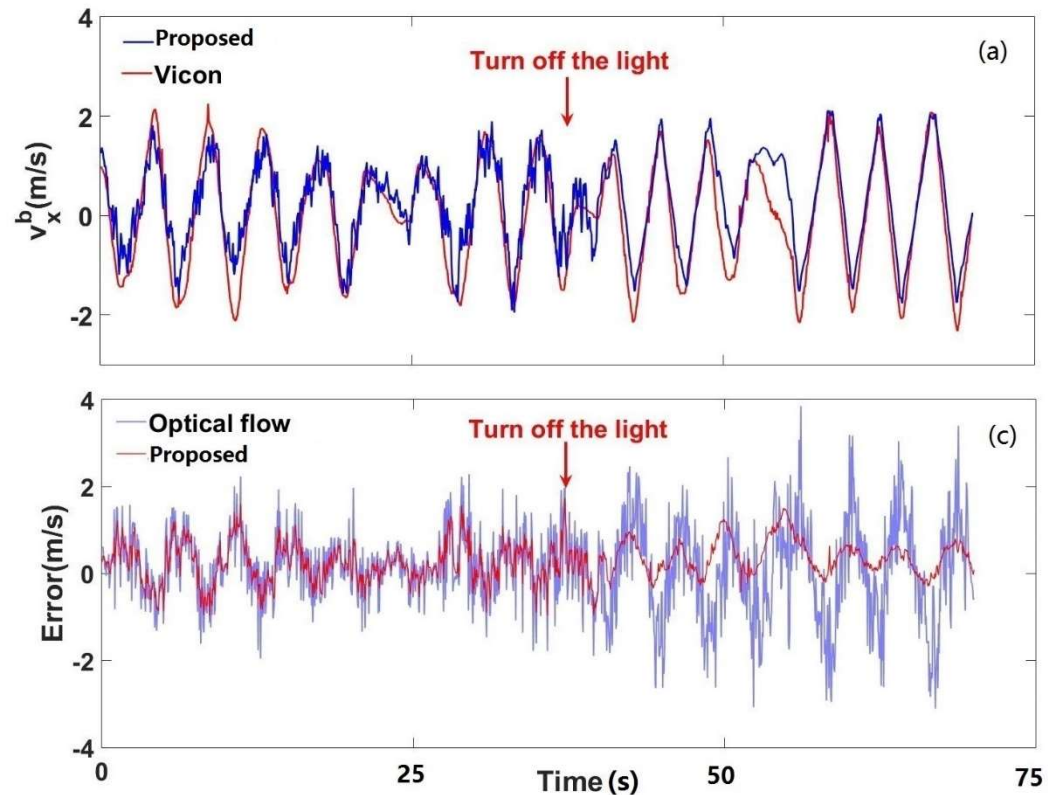
Experiment

Manual flight under controlled lighting condition



| | Light on | Light off |
|------|-----------------|-----------------|
| O.F | 0.355m/s | 0.893/ms |
| D.E. | 0.322m/s | 0.349m/s |

Average estimation error (m/s)

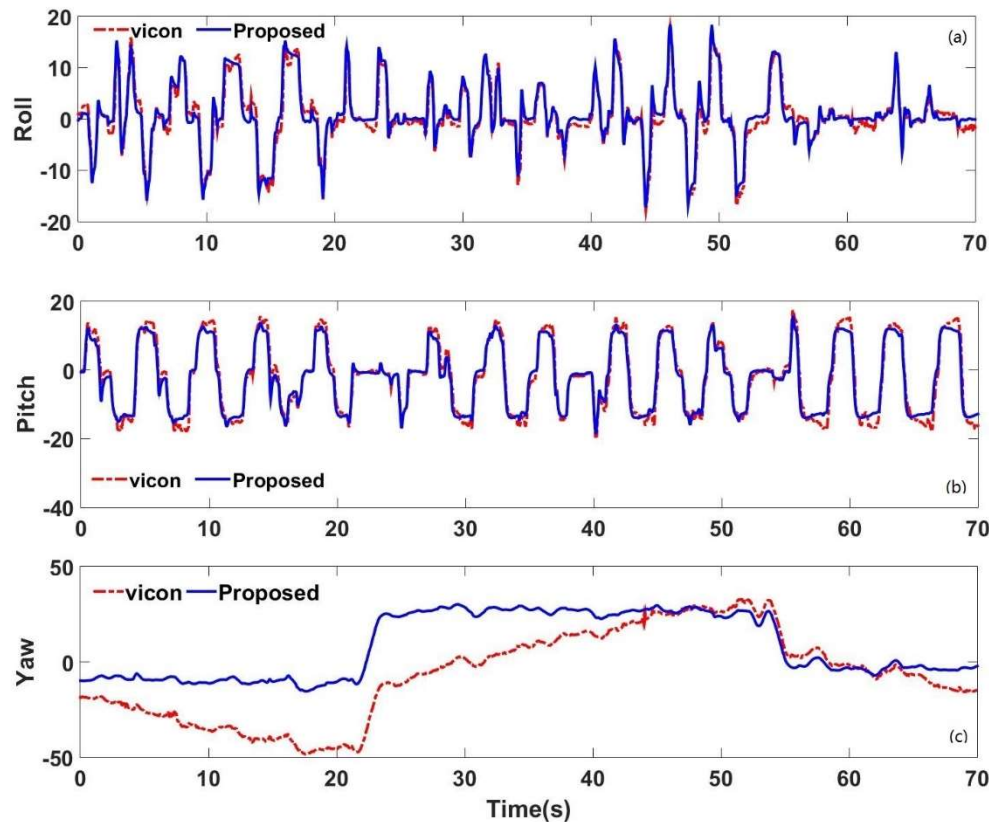




Experiment



Autonomous flight in complete darkness (Attitude estimation)



| | mean | STD |
|-------|--------|--------|
| Roll | 0.235° | 1.359° |
| Pitch | 0.172° | 2.024° |
| Yaw | 14.79° | 13.70° |

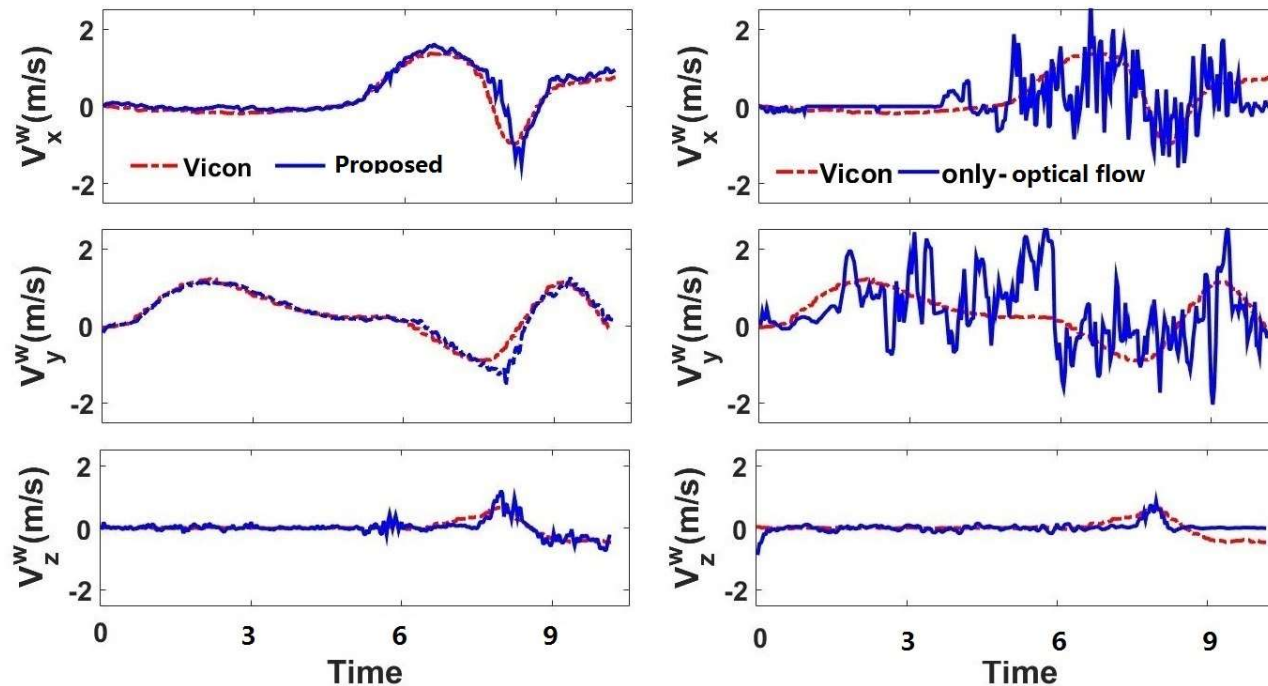
Estimation error (°)

**Yaw is not observable*



Experiment

- Autonomous flight in complete darkness (Velocity estimation)



**Optical flow is not usable, but the proposed estimator works well*

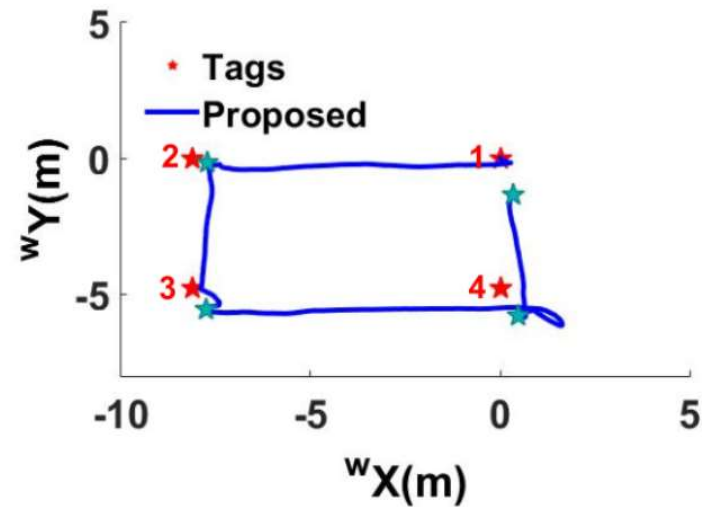
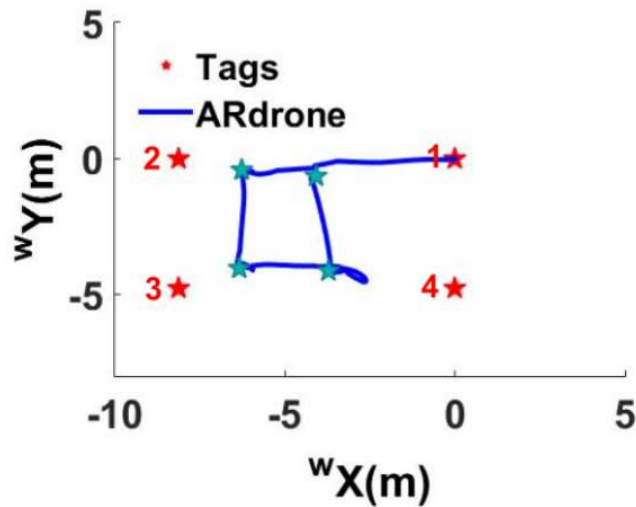


Experiment

Manual flight in a large hall at night



| | Estimated Distance (m) | Loop Error |
|----------|------------------------|------------|
| ARDrone | 16.00 | 4.13 |
| Proposed | 25.81 | 1.37 |





Conclusion

- Special aerodynamics of multi-rotor UAVs renders the body velocity and attitude observable.
- By incorporating this observation, we can design a fast and easily implemented state estimator to remedy the shortcoming of vision-based approach.
- The results of real-world experiments show the effectiveness and robustness of the proposed approach.



Q&A

