



# An Aerodynamic Model-aided State Estimator for Multirotor UAVs



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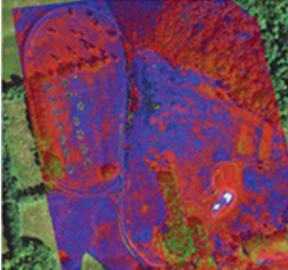
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# Multi-rotor UAVs



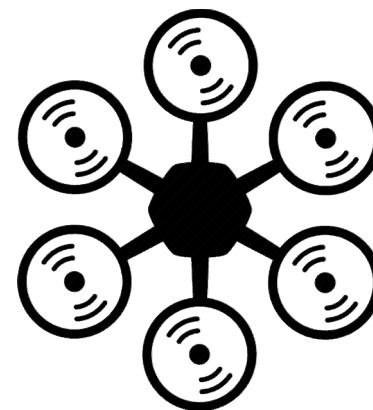
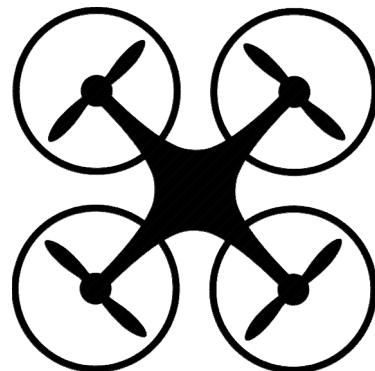
Agriculture

Entertainment



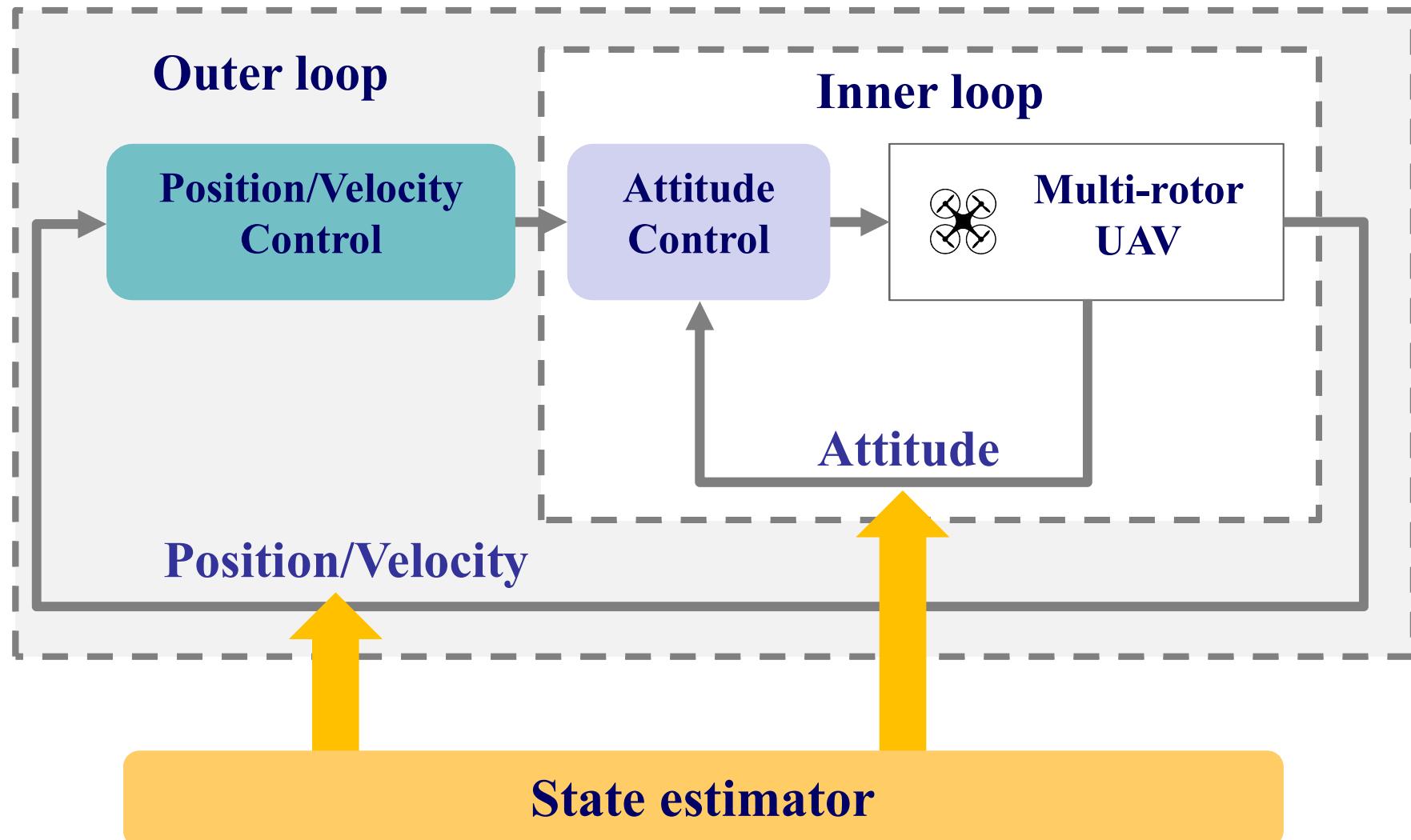
Service robots

Photography



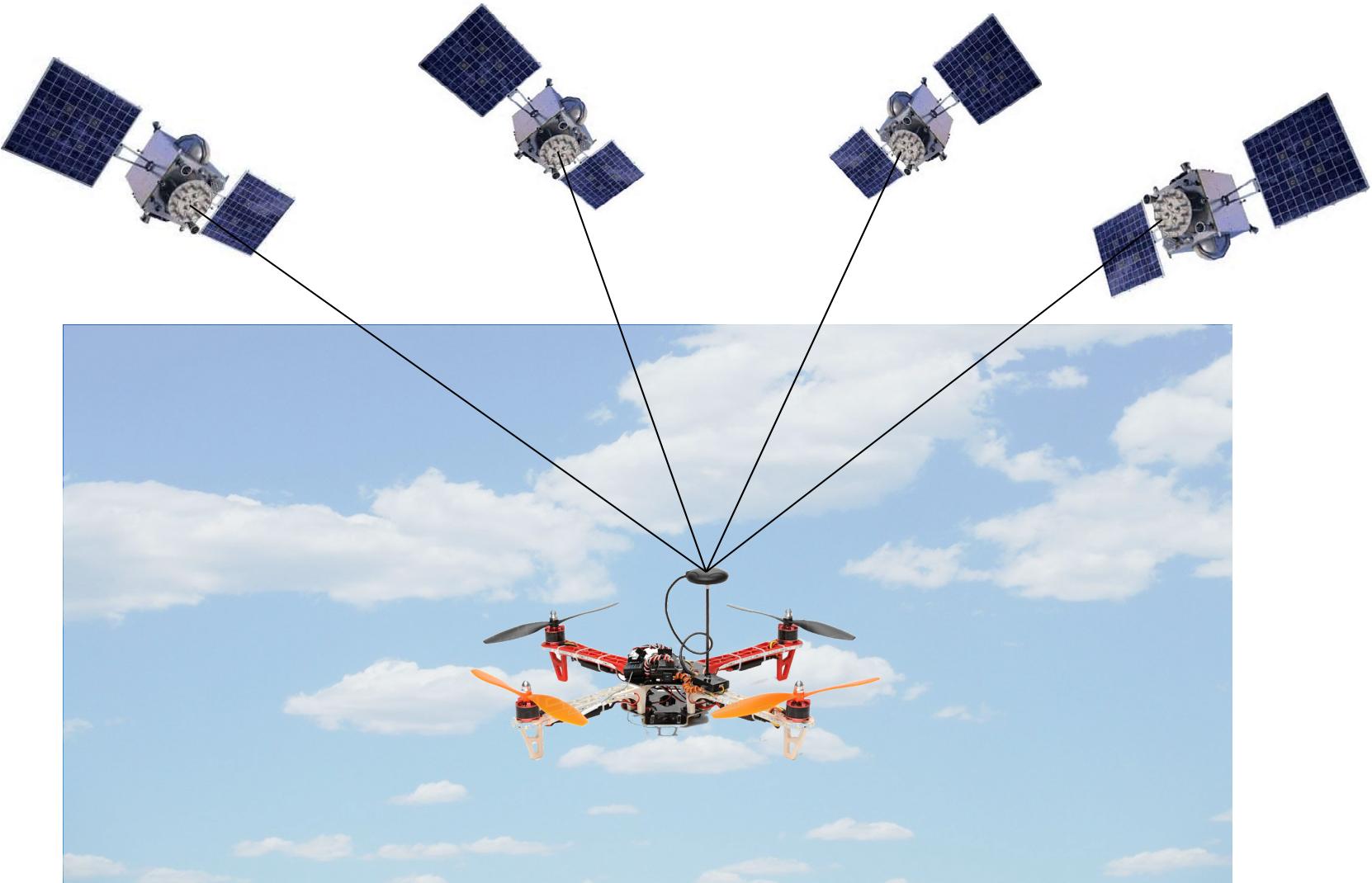


# A typical control loop of a multi-rotor UAV





# 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**

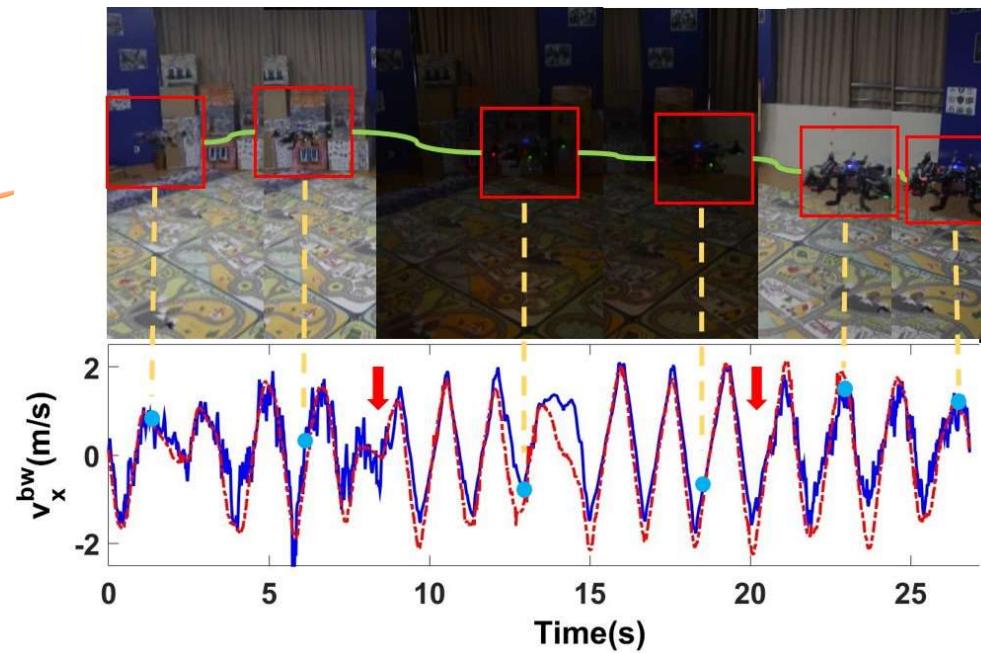
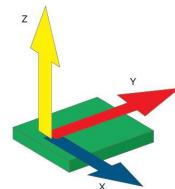




# Aerodynamic model-aided state estimator

Special aerodynamic characteristic of multi-rotor UAVs:

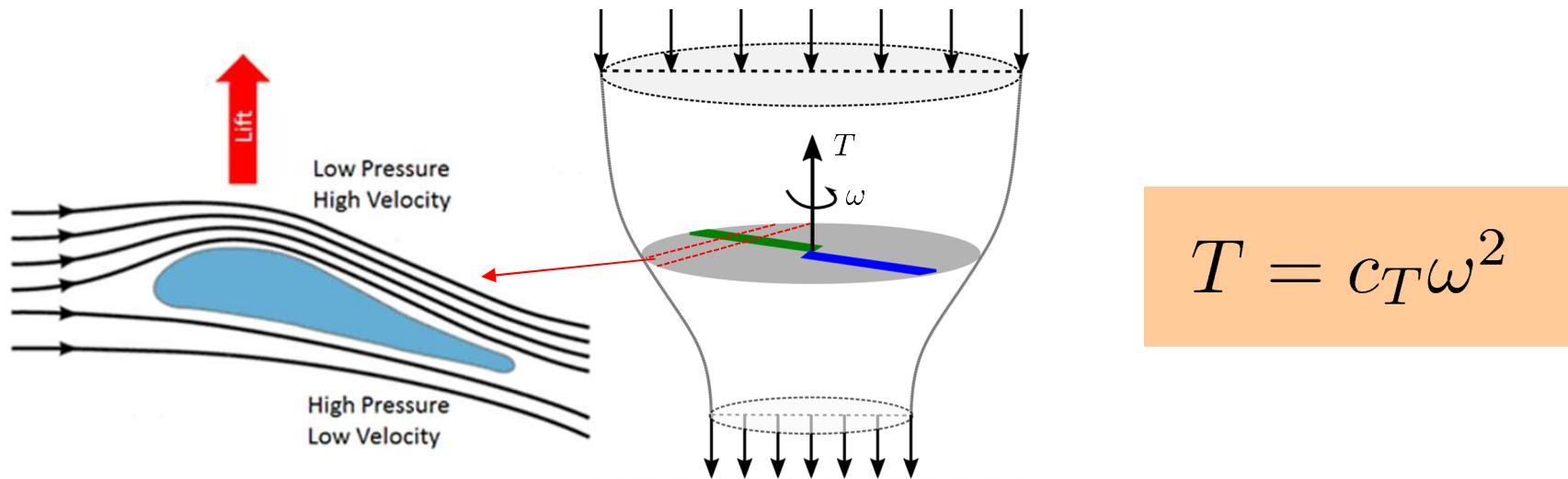
The horizontal velocity of the vehicle is **observable** from the accelerometer measurements.





# Rotor Aerodynamics

## Force in hovering state



$$T = c_T \omega^2$$

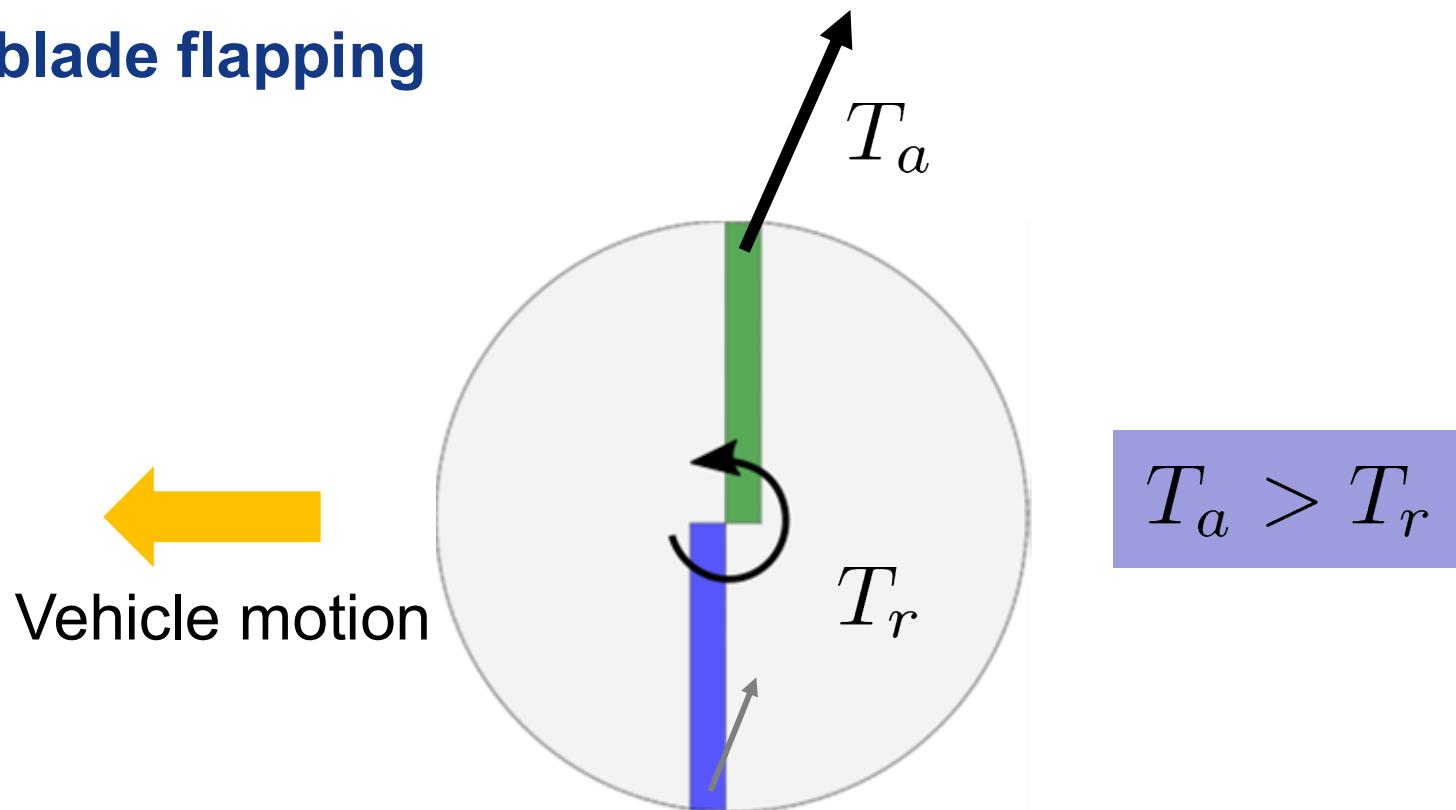
The thrust force  $T$

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



# Rotor Aerodynamics

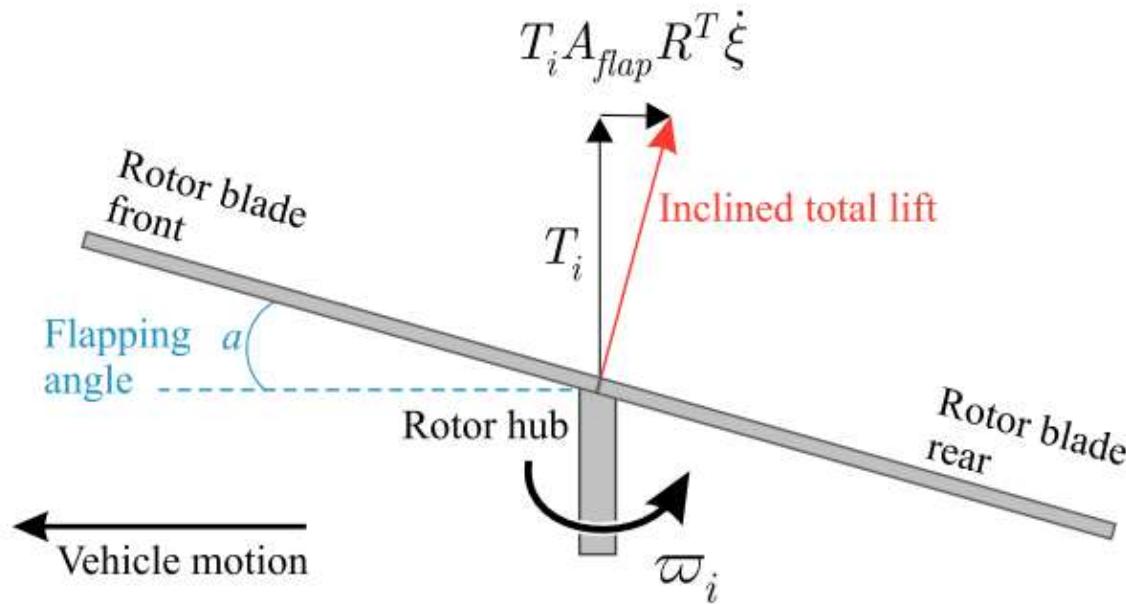
- Forces in translation movement :
  - blade flapping





# Rotor Aerodynamics

- Forces in translation movement :
  - blade flapping

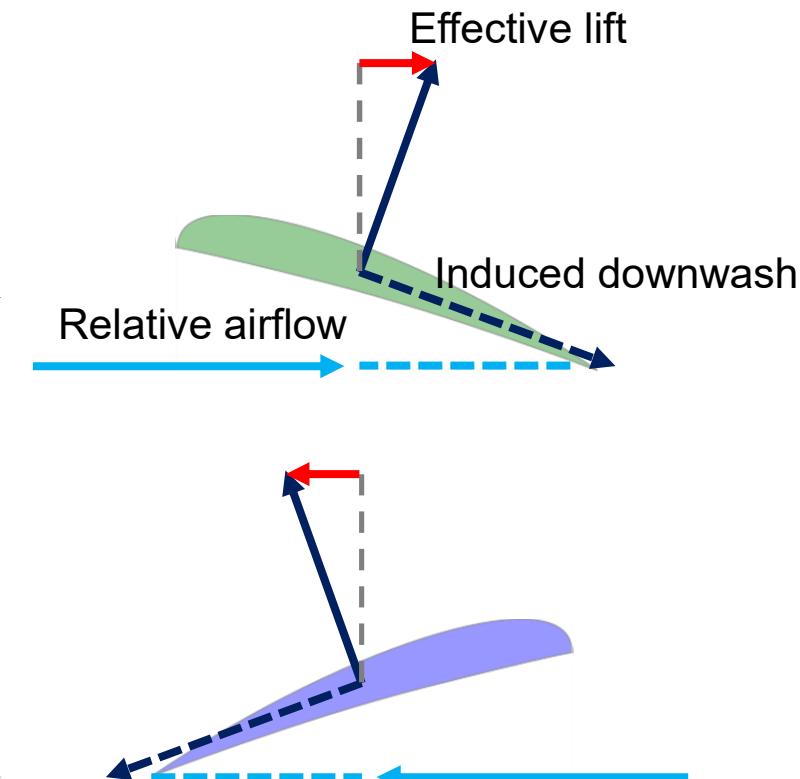
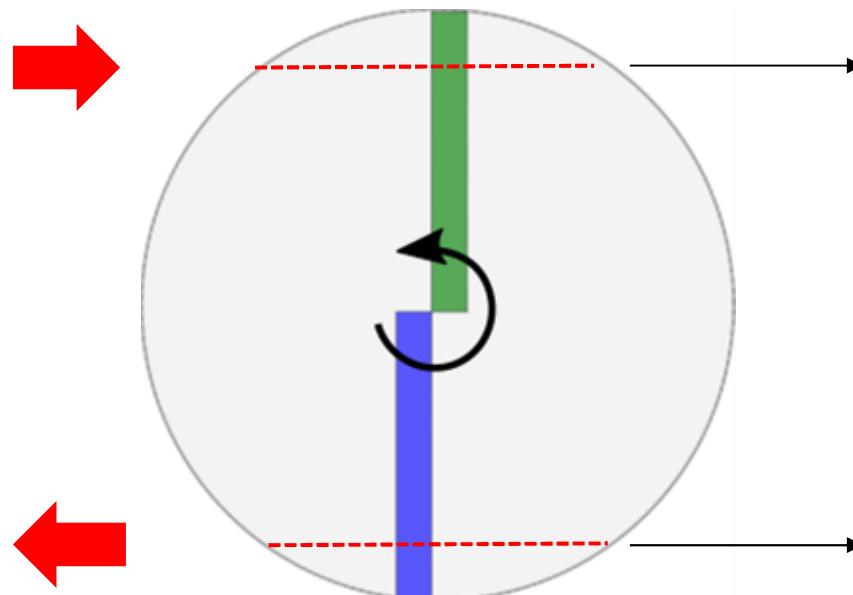




# Rotor Aerodynamics

- Induced drag in hovering :

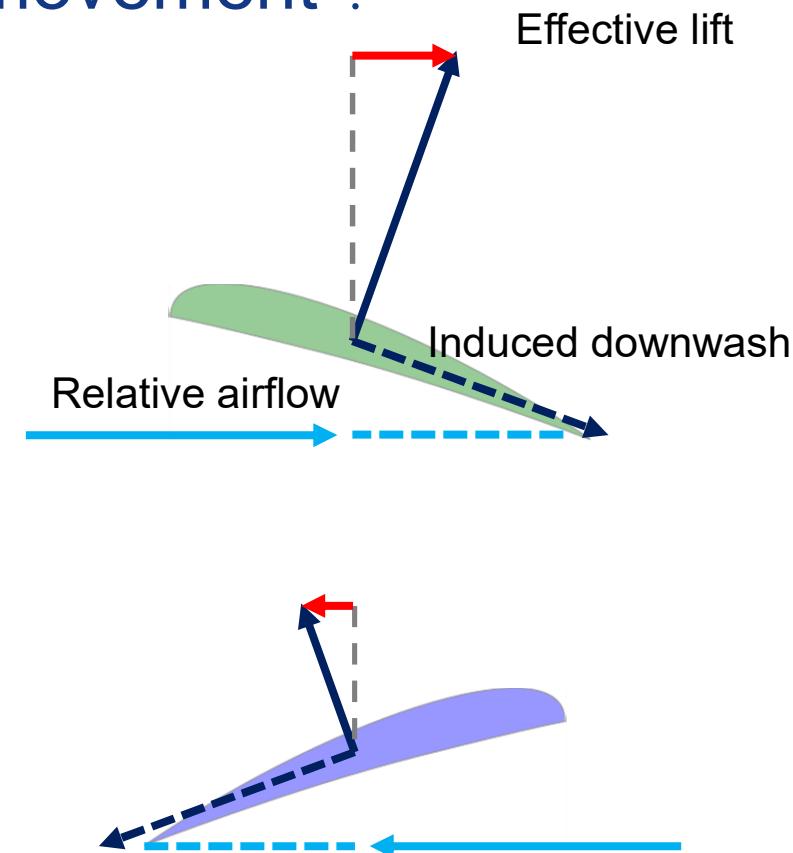
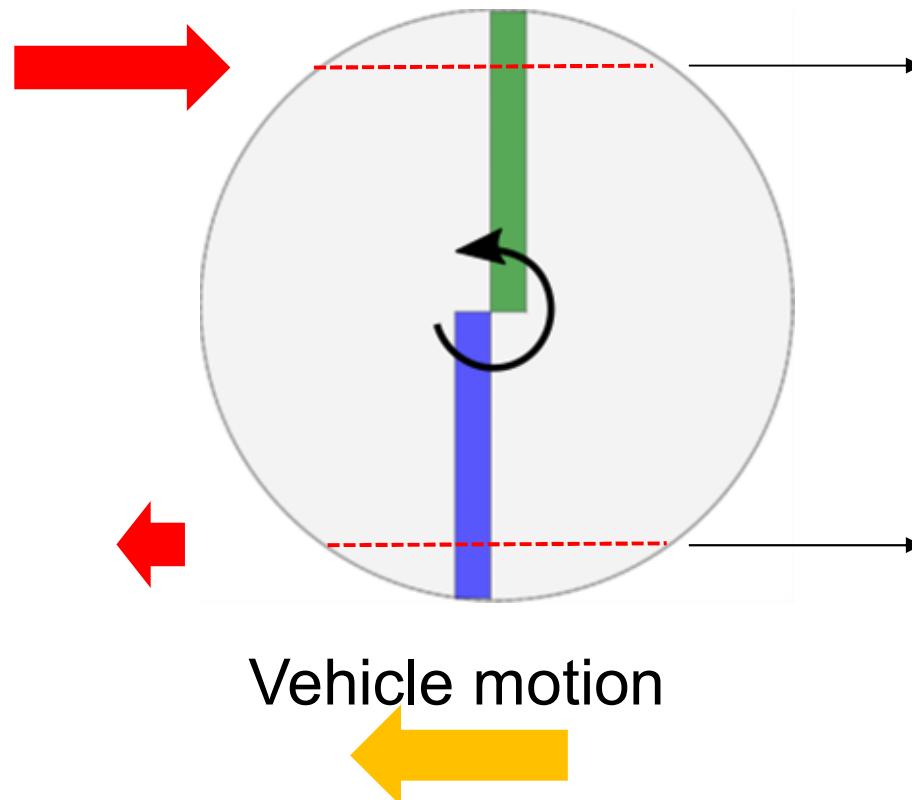
- induced drag





# 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,

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

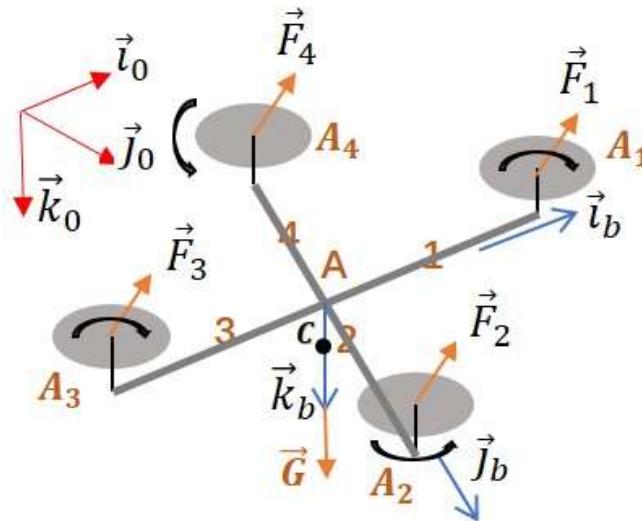
↓      ↓  
constant      Horizontal velocity in body frame  
↓  
Spinning speed of the rotor

*Principles of Helicopter Aerodynamics*, Section 3.5



# Multi-rotor Aerodynamics

- The total force generated by the rotors:

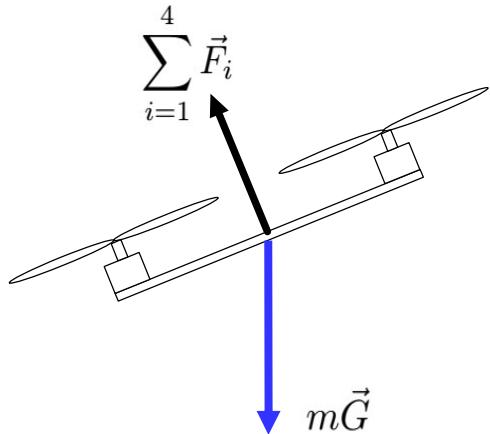


$$\sum_{i=1}^4 \vec{F}_i \approx -a \left( \sum_{i=1}^4 \omega_i^2 \right) \vec{k}_b - \lambda_1 \left( \sum_{i=1}^4 \omega_i \right) \vec{V}_C^\perp$$

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:

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

→

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

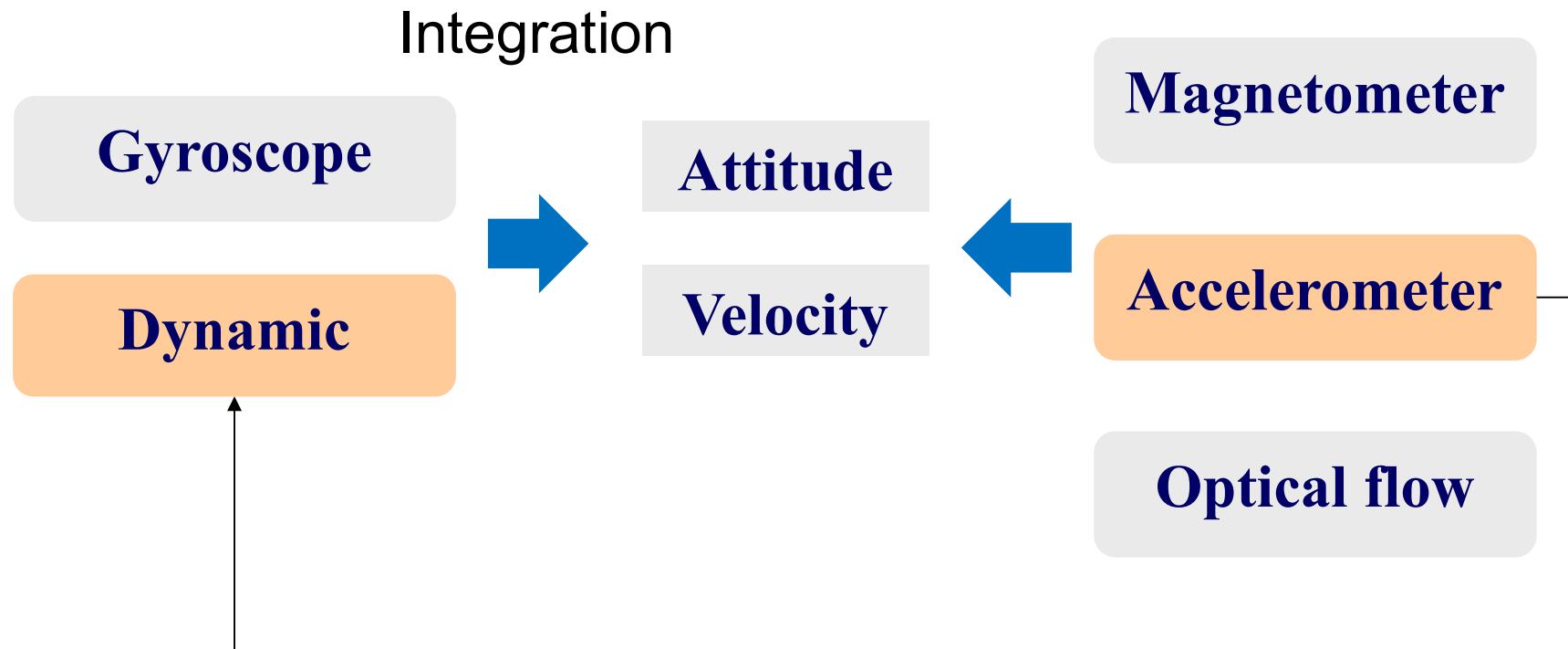


# Multi-rotor Aerodynamics

- ❶ **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:

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

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



# Estimator Design

## Measurement Model

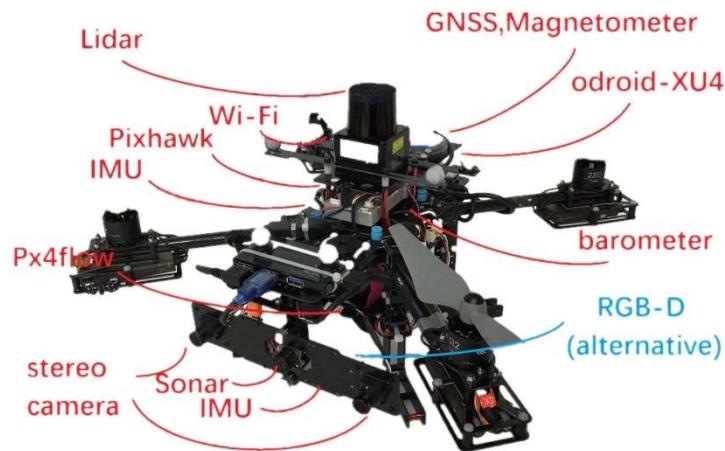
$$\mathbf{x} = [\phi \ \theta \ \psi \ \beta_{g_x} \ \beta_{g_y} \ \beta_{g_z} \ v_x^b \ v_y^b \ v_z^b \ h \ \mu_x \ \mu_y]^T \quad \text{state vector}$$

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



# 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

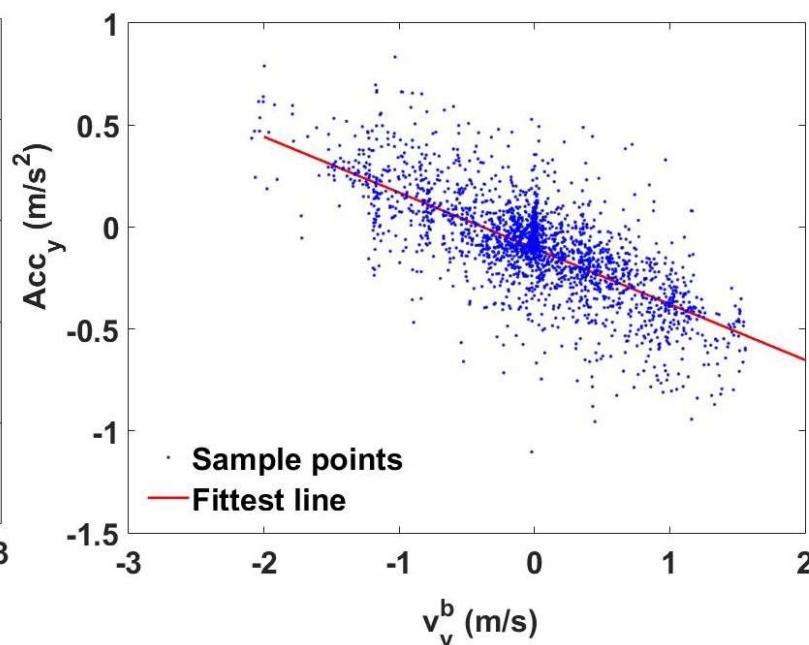
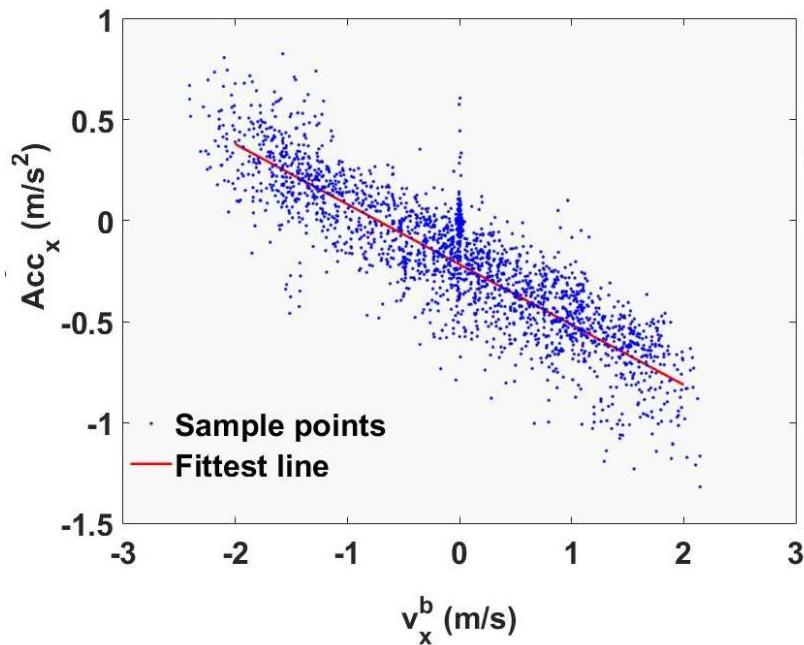


# Experiment

- Drag Coefficients Initialization by line fitting

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

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

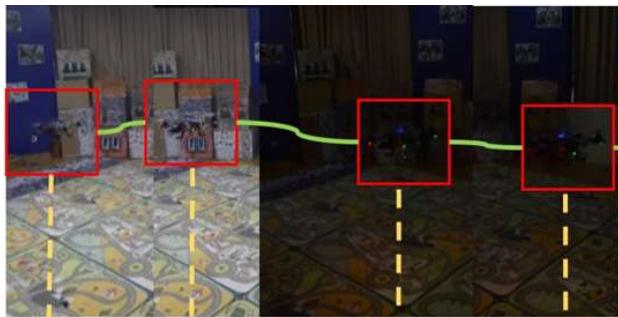


**Left:** line fitting of  $\mu_x$ ,    **Right:** line fitting of  $\mu_y$



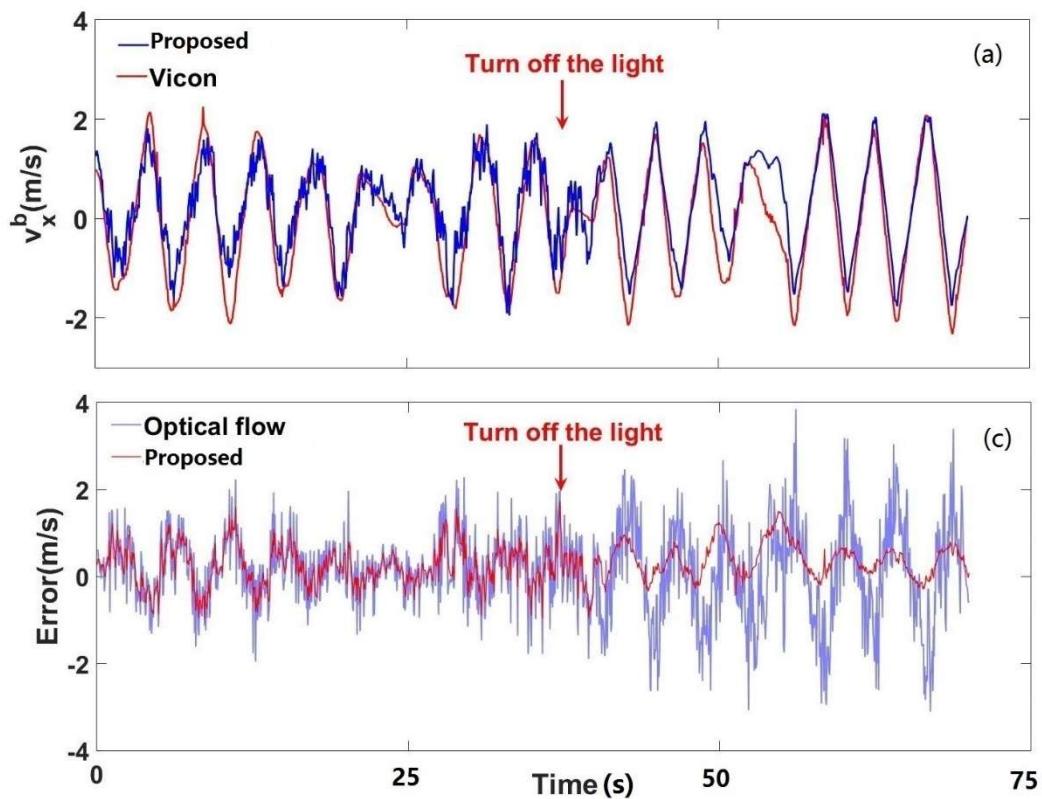
# Experiment

- Manual flight under controlled lighting condition



	Light on	Light off
O.F	0.355m/s	0.893/ms
D.E.	<b>0.322m/s</b>	<b>0.349m/s</b>

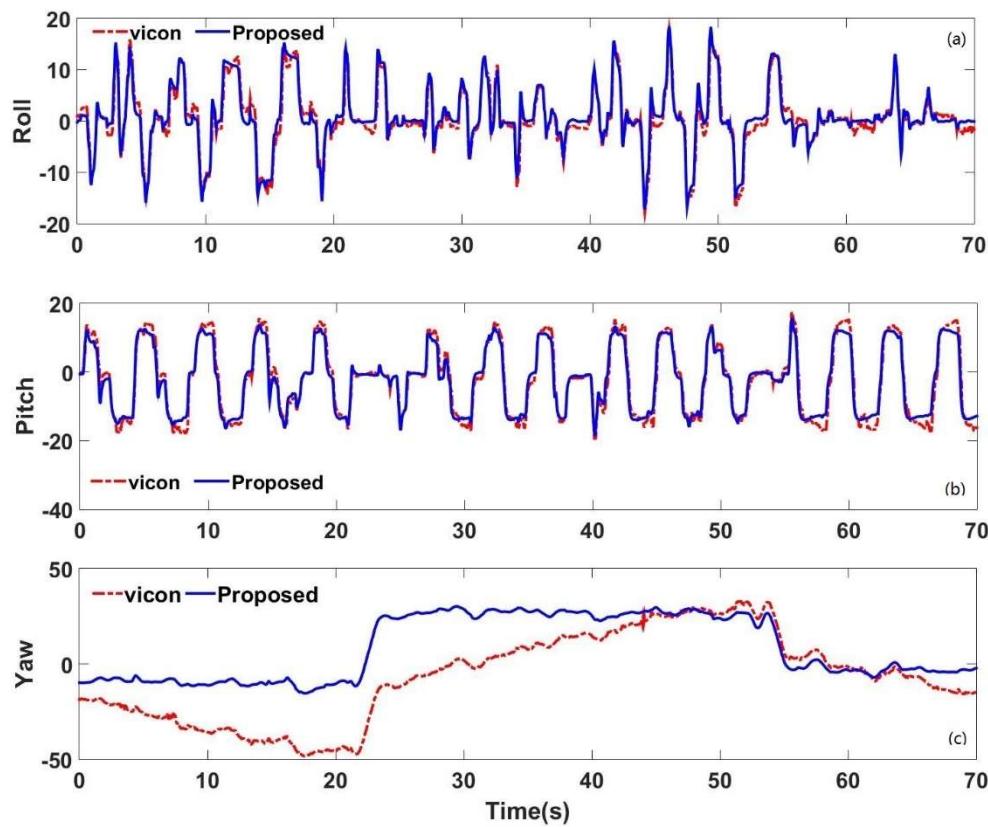
Average estimation error (m/s)





# Experiment

- Autonomous flight in complete darkness  
(Attitude estimation)



	mean	STD
Roll	$0.235^\circ$	$1.359^\circ$
Pitch	$0.172^\circ$	$2.024^\circ$
Yaw	$14.79^\circ$	$13.70^\circ$

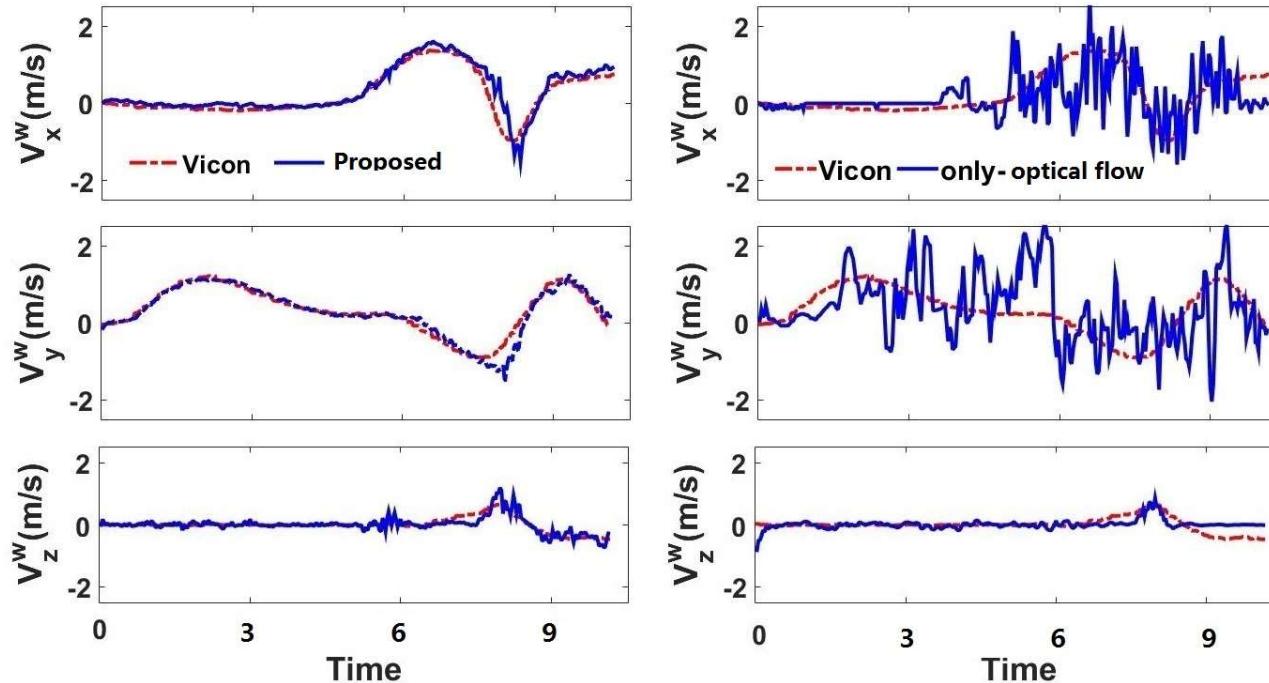
Estimation error ( $^\circ$ )

\*Yaw is not observable



# Experiment

- Autonomous flight in complete darkness  
(Velocity estimation)



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

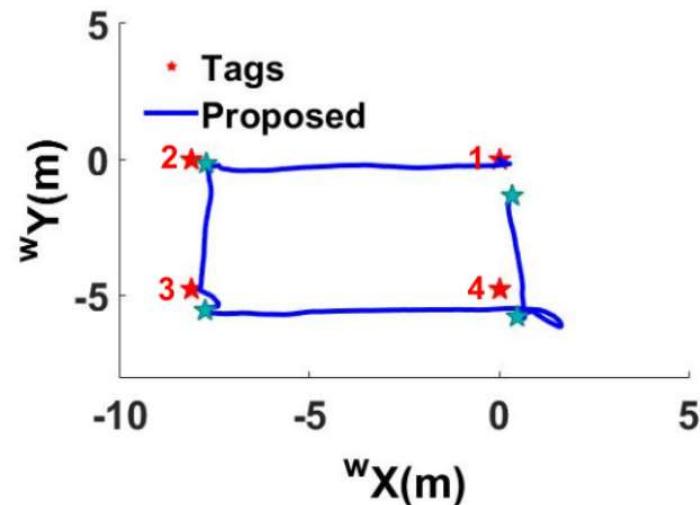
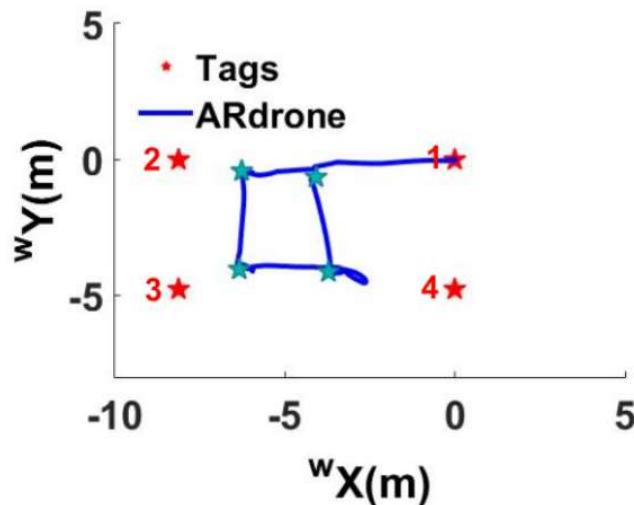


# Experiment

- 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

Experiment #1

Velocity estimation under large illumination change

