

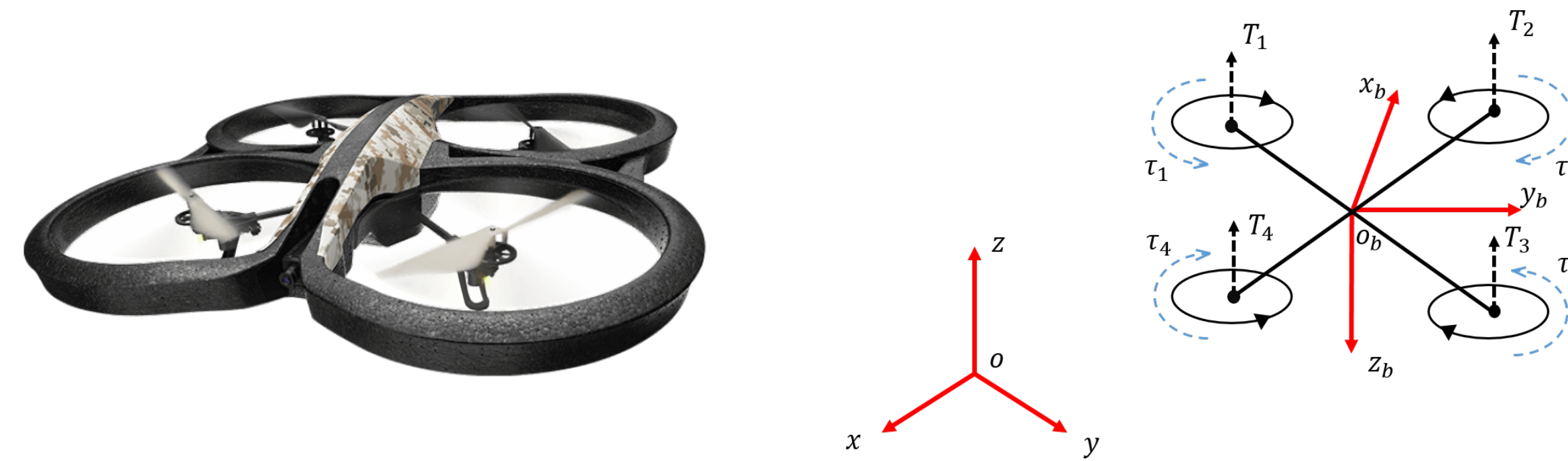
Modeling and diagnosis of UAVs

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Summary

Since the dynamic model of a quadcopter is too complex for estimation purposes, submodels are considered. First, a lateral dynamic model is used to estimate the mass changes of the quadcopter using only measurements from onboard sensors, which leads to a sensor-to-sensor problem. Second, a challenge of a general sensor-to-sensor problem is that it is sometimes not obvious to select which signals to use as input or output of a SISO system. It is shown that an Instrumental Variable (IV) approach gives identical results when estimating the forward and inverse models of a SISO system. Third, the IV method can also provide accurate estimates of the parameters of a Hammerstein model of the vertical dynamics in the closed-loop setting.

Quadcopter dynamic



In the body-fixed frame, the dynamic equations of a rigid body quadcopter using the Newton-Euler equations are given by

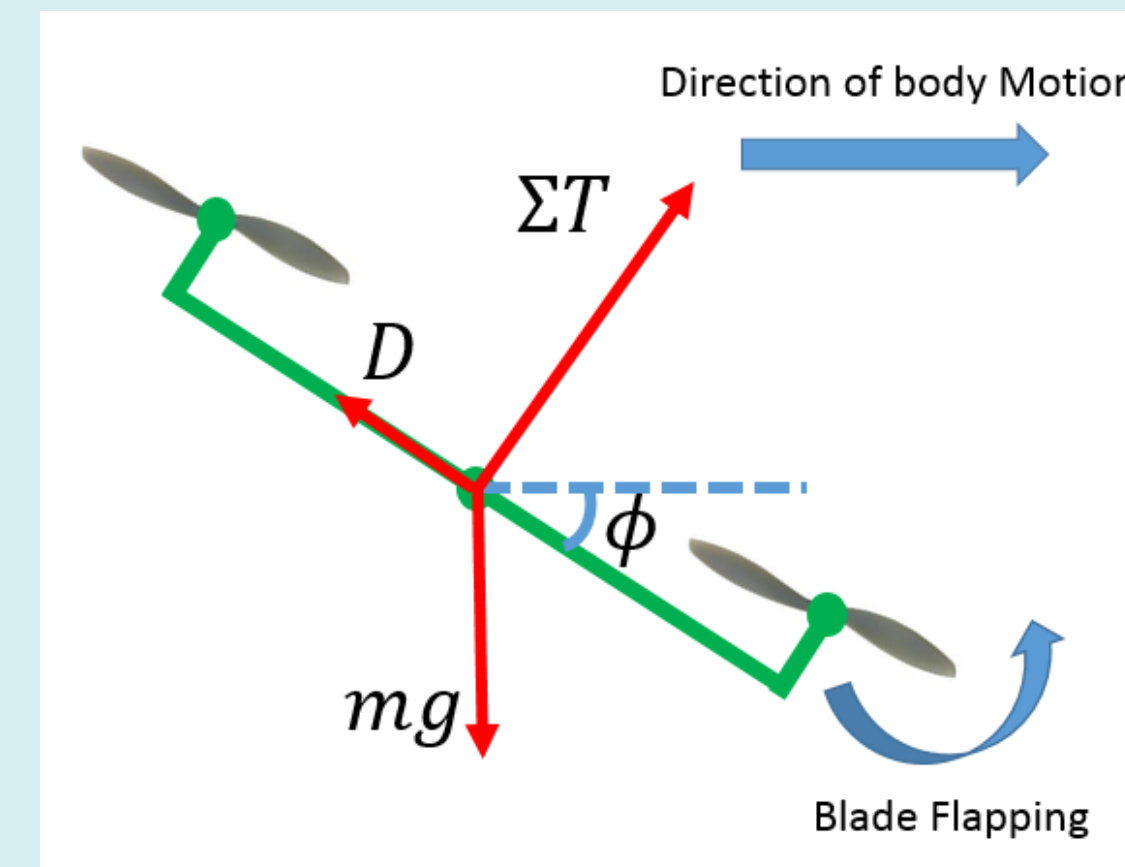
$$\begin{aligned} m\dot{V}_B + m\nu \times V_b &= mR^T g + E_B^F(\Omega) + D_B^F(V_b) \\ I\dot{\nu} + \nu \times (I\nu) &= O_B^T(\nu, \Omega) + E_B^T(\Omega) \end{aligned} \quad (1)$$

where m is the mass of the quadcopter.

Submodels of a quadcopter

- The drag effect makes the lateral acceleration linearly dependent on the lateral velocity in the body-fixed frame.
- The vertical thrust equation contains a linear term due to the induced velocity of the air flow interacting with the propellers.

Sensor-to-sensor problem



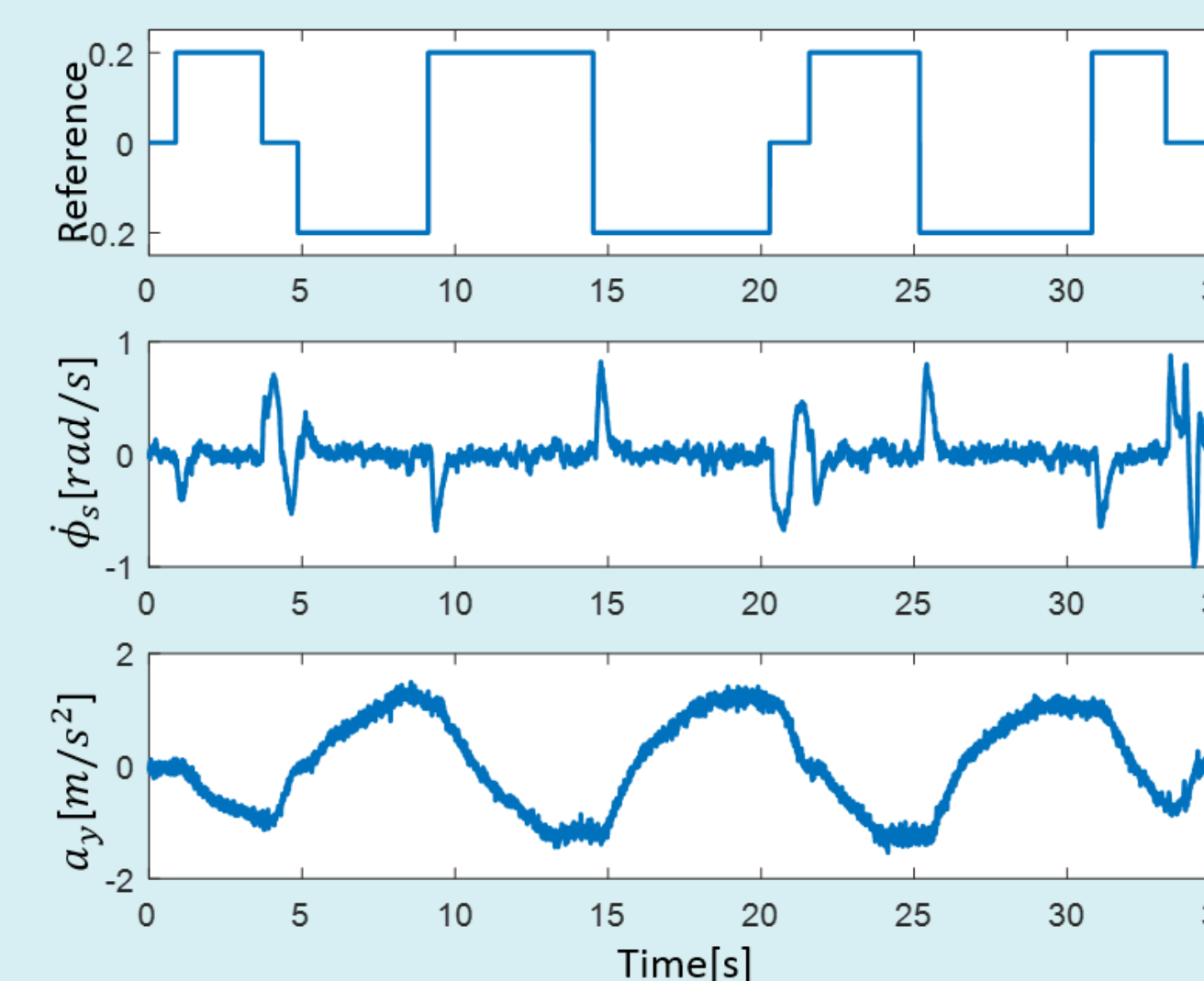
The effect of the relative speed of the blades with respect to free air divides the operating region of a propeller into two areas: a retreating and an advancing blade. The advancing blade has a higher relative velocity than the retreating one, which creates a force imbalance between the two areas.

Projecting (1) onto the lateral plane in the body-fixed frame

$$\dot{v} = g \cos(\theta) \sin(\phi) - \frac{\lambda_1}{m} v \quad (2a)$$

Using measurements from an IMU

$$a_y = \frac{\lambda_1}{m} v + e_{a_y} \quad \dot{\phi}_m = \dot{\phi} + e_{\dot{\phi}} \quad (3)$$

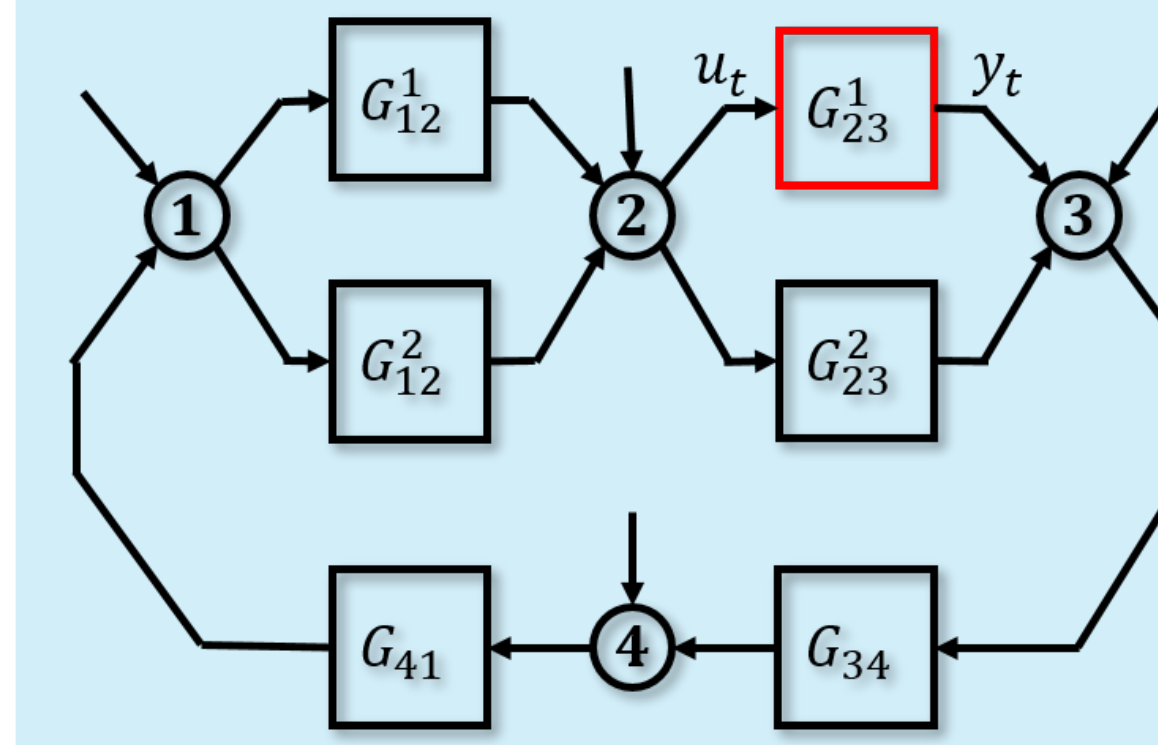


Three different mass datasets are collected.

m_{ref}	m_c	\hat{m}_c (LS)	\hat{m}_c (EKF)	\hat{m}_c (IV)
455 g	510 g	1362.5 ± 54.9 g	505.6 ± 258.8 g	504.1 ± 3.9 g
582 g	455 g	2126.2 ± 78.9 g	384.4 ± 161.2 g	580.9 ± 3.8 g
510 g	582 g	170.3 ± 6.9 g	458.9 ± 234.8 g	460.3 ± 3.4 g
582 g	455 g	795.8 ± 25.7 g	387.3 ± 187.3 g	587.5 ± 3.2 g
510 g	582 g	124.5 ± 4.6 g	689.7 ± 289.6 g	456.1 ± 3.0 g
582 g	455 g	373.1 ± 12.1 g	766.4 ± 370.7 g	505.2 ± 2.8 g

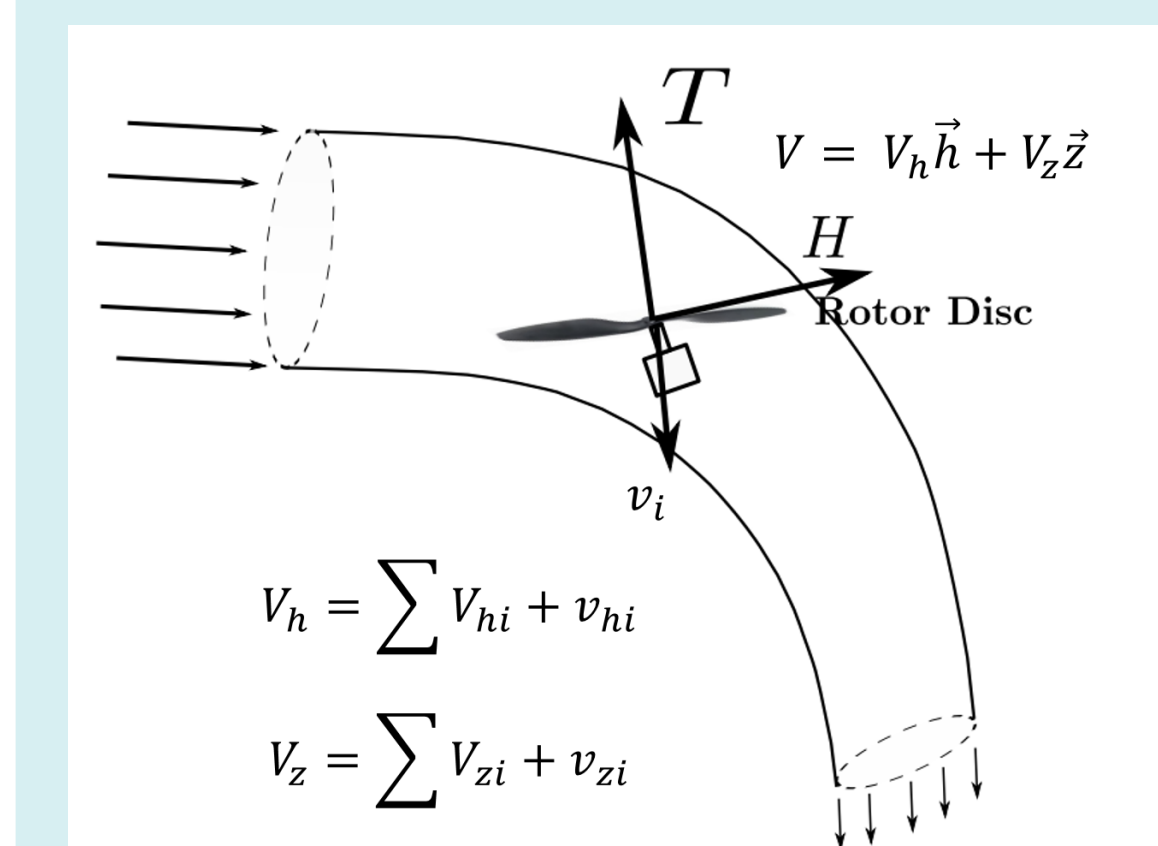
The least-squares and EKF methods give unreliable results while the IV method can detect the changes of mass accurately.

Errors-in-variables estimation



The IV method gives identical estimates of the forward model G_{23}^1 and the inverse model G_{32}^1 , regardless input-output selection, using finite data.

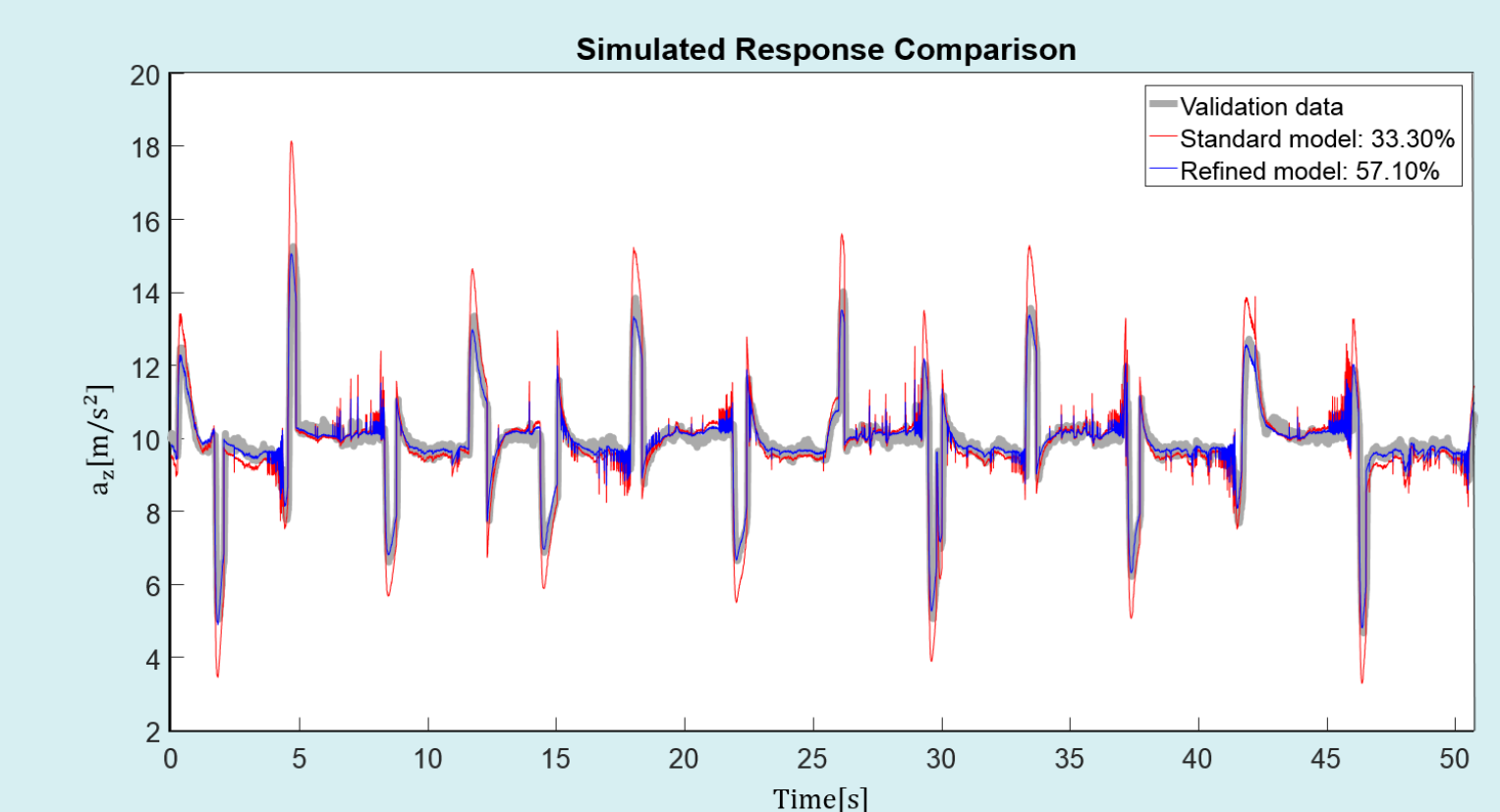
Nonlinear Hammerstein model



V_{hi} and V_{zi} are the horizontal and vertical velocities of the i^{th} rotor in the body-fixed frame, and v_{hi} and v_{zi} are the horizontal and vertical induced velocities of the air stream through the i^{th} rotor.

Projecting (1) onto the z_b axis yields

$$\dot{w} = \underbrace{\frac{k_1 u_t^2}{m}}_{\text{standard thrust}} - \frac{k_2 u_t}{m} - \frac{k_w}{m} w + g \cos \theta \cos \phi \quad (4)$$



The refined model (57.10% model fit) gives a more accurate estimate of the vertical dynamics of the quadcopter than the standard model (33.30% model fit).

Future work

1. Fault detection and isolation algorithms.
2. Other nonlinear block-oriented models.

