



On-board Cable Attitude Measurement and Controller for Outdoor Aerial Transportation

Pratik Prajapati ¹, Sagar Parekh ¹ and Vineet Vashista* ¹

¹Human-Centered Robotics Lab, IIT Gandhinagar, Ahmedabad, INDIA.

*Corresponding author. E-mail: vineet.vashista@iitgn.ac.in.

Received: xx xxx xxx; **Revised:** xx xxx xxx; **Accepted:** xx xxx xxx

Keywords: Human-quadcopter interaction, Aerial transportation, Cable-suspended payload

Abstract

Deploying quadcopters for aerial transportation can be cost-effective in impromptu material handling applications. However, such applications are limited mainly due to the requirement of onboard localization sensors and associated computation. The current work presents a human-controlled modality to successfully execute spontaneous outdoor flight of a quadcopter with a cable-suspended payload. Stable and smooth flights are achieved through an onboard integration of a custom-built sensor system and a controller to minimize payload oscillations. The feasibility of the proposed modality is demonstrated by conducting outdoor experiments and a case study in an unstructured environment.

1. Introduction

Many field applications, such as automotive and packaging industries, agriculture and warehouse applications, construction sites, etc., involve repetitive transportation of materials/objects from one place to another. Deploying robotics devices to perform these repetitive activities can be cost-effective and enhance overall productivity by reducing material handling time. Various kinds of robotic devices have been used to perform specific types of applications [1], the majority of which are programmed to execute the tasks autonomously and often in well-structured environments. The reachable workspace is an important aspect of such robotic applications. Notably, aerial robots can be deployed in dynamic and unstructured outdoor environments with a relatively larger workspace, and their use for the task of transportation has been proposed for applications in payload delivery [2]. Extending quadcopters for

impromptu payload transportation in outdoor settings can help reduce infrastructure costs and improve productivity.

In the literature, various methodologies have been proposed for aerial transportation using quadcopters. The payloads are mainly transported by being directly attached to the quadcopter's chassis [3–7] or grabbed using the mechanical arms/grippers [8,9] or suspended using the cables. The authors in [10] presented a case where quadcopters autonomously pick and place rectangle-shaped modular elements to create special cubic structures, and the authors in [11] demonstrated cooperative construction of 7 m height architecture using rectangle foam elements. The authors in the [12,13] also presented a case where quadcopters cooperatively constructed a tensile structure using ropes to create a temporary bridge. The use of mechanical arms/grippers or direct attachment of payloads affects the flight dynamics of the quadcopter and gets limited by payloads' size and shape. In contrast, transporting a payload through a cable attached at the quadcopter's center of mass (COM) can be advantageous as the quadcopter's attitude dynamics remain unaffected, which provides agility. Further, cable suspension can be easily extended for cooperative payload transportation. However, cable suspended payloads are prone to oscillations during transportation, which requires control strategies to eliminate oscillations.

Various methods have been presented in the literature for a quadcopter with a cable-suspended payload system to track complex trajectories with minimal payload oscillations, such as differential flatness-based controllers [14–18], optimal controllers [19–24], and reinforcement learning-based controllers [25,26]. In addition, the use of multiple aerial robots to lift a payload cooperatively can increase the load-carrying capacity and has been the focus of many studies. The authors in [27–29] demonstrated such a modality in an indoor environment where required feedback of the states was estimated using a motion capture system. Using image processing techniques in [30–32] and rotary encoders in [33], the authors demonstrated estimation of the cable attitude to stabilize the payload swings during autonomous outdoor transportation. The authors in [34,35] demonstrated collaborative transportation using multiple quadcopters by utilizing force and admittance controller.

Considering the under-actuated, non-linear and coupled dynamics of a quadcopter with cable-suspended payload system, the existing works [14–26, 32] have been focused on the development of agile, autonomous, and accurate trajectory tracking to perform specific tasks. Accordingly, for localization in indoor settings motion capture system is widely used; however, it restricts outdoor applications. For localization and obstacle avoidance in unstructured outdoor environments, external sensors such

as LIDARS, cameras, visual-inertial odometry, etc., become necessary to install. Also, enabling fully autonomous flight in variable and changing outdoor environments is a computationally costly and challenging task. Alternatively, some of the human-in-the-loop modalities presented in [36–41] have demonstrated the involvement of human operators in various capacities to show potential benefits in high-level logical planning and established successful teleoperation of a complex and coupled dynamical system.

The current work present a preliminary design and control approach that enables stable and smooth human-controlled aerial transportation of cable-suspended payload using a quadcopter in outdoor environment. The main components of this approach include on-board integration of a portable sensor device to accurately measure the cable states and a controller to minimize the payload oscillations using the cable state feedback during outdoor flight. The current work further demonstrates the feasibility of this approach through two outdoor flight modalities, namely 1) Semi-autonomous flight, and 2) Human-controlled flight. Finally, a preliminary case study is presented to demonstrate the feasibility of the proposed modality in outdoor applications such as impromptu aerial transportation at a construction site.

2. METHODS

2.1. Dynamical Model

The quadcopter with a cable-suspended payload system is shown in the Fig. 1. The inertial reference frame and body-fixed reference frame of the quadcopter are denoted as three orthogonal unit vectors $\{\mathbf{e}_1, \mathbf{e}_2, \mathbf{e}_3\}$ and $\{\mathbf{b}_1, \mathbf{b}_2, \mathbf{b}_3\}$ respectively. The third inertial frame axis, \mathbf{e}_3 , is taken as the vertically upward direction and the third body-fixed axis of the quadcopter, \mathbf{b}_3 , is taken perpendicular to the plane of the quadcopter, pointing upwards. The position of the quadcopter and the payload in frame $\{I\}$ are denoted by $X = [x, y, z]^T \in \mathbb{R}^3$ and $X_p = [x_p, y_p, z_p]^T \in \mathbb{R}^3$ respectively. The attitude of the quadcopter is represented by standard ZXY Euler angle representation, rotation matrix $\mathbf{R} \in SO(3)$, such that, (ϕ, θ, ψ) define the quadcopter's roll, pitch, and yaw angle respectively [42]. The angular velocity of the quadcopter in frame $\{B\}$ is denoted as $\boldsymbol{\omega} \in \mathbb{R}^3$. The attitude of the cable is represented as the angular position of the cable along \mathbf{e}_1 and \mathbf{e}_2 axes, i.e., ϕ_p and θ_p respectively. $m \in \mathbb{R}$ and $\mathbf{J} \in \mathbb{R}^{3 \times 3}$ denote quadcopter's mass and moment of inertia in frame $\{B\}$. $m_p \in \mathbb{R}$ and $l \in \mathbb{R}$ denote payload's

Nomenclature

$\{I\}$	Inertial frame of reference with three orthogonal unit vectors $\{e_1, e_2, e_3\}$
$\{B\}$	Body frame of reference of the quadcopter with three orthogonal unit vectors $\{b_1, b_2, b_3\}$
$g \in \mathbb{R}$	Acceleration due to gravity
$m \in \mathbb{R}$	Mass of the quadcopter
$J \in \mathbb{R}^{3 \times 3}$	Moment of inertia of the quadcopter in frame $\{B\}$
ϕ, θ, ψ	Roll, pitch, and yaw angles of the quadcopter
$R \in SO(3)$	Attitude of the quadcopter
$\omega \in \mathbb{R}^3$	Body frame angular velocity of the quadcopter
$m_p \in \mathbb{R}$	Mass of the payload
$l \in \mathbb{R}$	Length of the cable
ϕ_p, θ_p	Cable attitude about e_1 and e_2 axes
$\phi_{p,CAM}, \theta_{p,CAM}$	Cable attitude measured by CAM device about b_1 and b_2 axes
ϕ_c, θ_c	Cable Attitude Controller commands along roll and pitch angle
$(\phi_h, \theta_h, \dot{\psi}_h, \dot{z}_h)$	Higher-level command for roll angle, pitch angle, desired yaw rate, and desired velocity along e_3 axis of the quadcopter
$X = [x, y, z]^T \in \mathbb{R}^3$	Position of the quadcopter in the inertial frame
$X_p = [x_p, y_p, z_p]^T \in \mathbb{R}^3$	Position of the payload in the inertial frame
$F \in \mathbb{R}$	Total thrust generated by the quadcopter
$M = [M_1, M_2, M_3]^T \in \mathbb{R}^3$	Moment generated by the quadcopter about the b_1, b_2 and b_3 axes

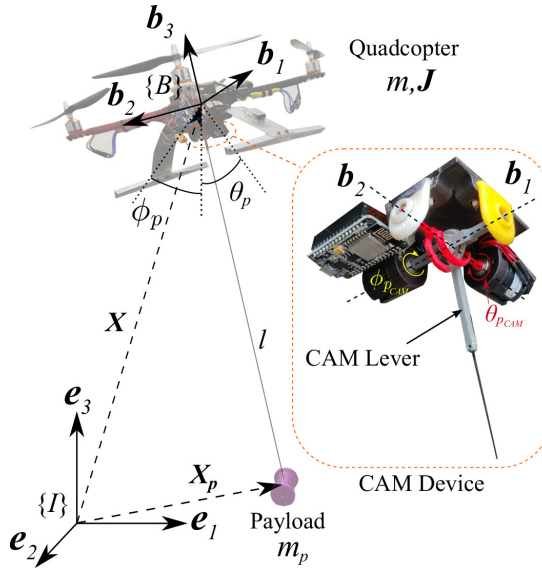


Figure 1. Experimental setup of a quadcopter and a cable-suspended payload system. The payload is suspended using a cable through the CAM device, which is mounted underneath quadcopter's chassis. X and X_p denote the position vectors of the quadcopter and payload in the inertial frame, $\{I\}$, respectively. ϕ_p and θ_p are the angular position of the cable about e_1 and e_2 axes respectively. $\{B\}$ is the body-fixed reference frame attached to quadcopter's principal axis. $\phi_{p,CAM}$ and $\theta_{p,CAM}$ are the cable attitude measured by CAM device about b_1 and b_2 axes respectively.

mass and cable length respectively. The cable is considered to be massless and when it remains taut, the position of the payload in frame $\{I\}$ is given by Eqs. (1 & 2), where, $c(\cdot) = \cos(\cdot)$ and $s(\cdot) = \sin(\cdot)$.

$$\mathbf{X}_p = \mathbf{X} + \mathbf{R}_x(\phi_p) \mathbf{R}_y(\theta_p) \begin{bmatrix} 0 & 0 & -l \end{bmatrix}^T \quad (1)$$

$$\mathbf{R}_x(\phi_p) = \begin{bmatrix} 1 & 0 & 0 \\ 0 & c\phi_p & -s\phi_p \\ 0 & s\phi_p & c\phi_p \end{bmatrix}, \quad \mathbf{R}_y(\theta_p) = \begin{bmatrix} c\theta_p & 0 & s\theta_p \\ 0 & 1 & 0 \\ -s\theta_p & 0 & c\theta_p \end{bmatrix} \quad (2)$$

The total thrust force and moment generated by four motors of the quadcopter are denoted as $F \in \mathbb{R}$ and $\mathbf{M} = [M_1, M_2, M_3]^T \in \mathbb{R}^3$ respectively. Using Euler-Lagrange's equation, the equations of motion for the quadcopter-payload system with cable attachment at the quadcopter's center are given in Eqs. (3 & 4). A vector, s , is used to define quadcopter's position and cable attitude, $s = [X^T, \phi_p, \theta_p]^T$. The matrices $\mathbf{M}(s) \in \mathbb{R}^{5 \times 5}$, $\mathbf{C}(s, \dot{s}) \in \mathbb{R}^{5 \times 5}$, and $\mathbf{G}(s) \in \mathbb{R}^{5 \times 1}$ are inertia, coriolis & centrifugal, and gravity terms and their expression are given in the Appendix.

$$\mathbf{M}(s)\ddot{s} + \mathbf{C}(s, \dot{s})\dot{s} + \mathbf{G}(s) = \begin{bmatrix} F\mathbf{R}e_3 \\ 0_{2 \times 1} \end{bmatrix} \quad (3)$$

$$\mathbf{J}\dot{\omega} + \omega \times \mathbf{J}\omega = \mathbf{M} \quad (4)$$

2.2. CAM Device

In the current work, a custom built sensor setup, referred to as Cable Attitude Measurement (CAM) device, is mounted under the quadcopter's chassis to suspend the payload. Similar to a joystick functioning, CAM device allows and measures the rotation of its lever about two orthogonal axes, \mathbf{b}_1 and \mathbf{b}_2 , as shown in Fig. 1. Similar approaches have been used in prior works to estimate cable attitude in a cable driven parallel robot [43] and for transporting suspended payload using helicopter [33]. The current work achieves a portable design with sensing and communication capabilities for a quadcopter to measure cable attitude. Consequently, enabling the implementation of an on-board controller that minimizes payload oscillations for outdoor flights. In brief, CAM device uses two 12-bit magnetic encoders from Broadcom Inc. that are aligned with \mathbf{b}_1 and \mathbf{b}_2 axes. To extract the readable data from the encoders,

NodeMCU ESP8266 WiFi module is used. The device has a 3D printed base made of polylactic acid plastic with a size of 6 cm \times 6 cm \times 3 cm. The mass of the device is 80g and can measure angles in the range $[-65^\circ \ 65^\circ]$ about the two axes. The cable is attached to an extended rod, labeled as CAM lever, which passes through the slotted \mathbf{b}_2 axis to be attached to \mathbf{b}_1 axis as shown in Fig. 1. Accordingly, CAM device measures cable attitude in frame $\{B\}$, i.e., $\phi_{p,CAM}$ and $\theta_{p,CAM}$ about \mathbf{b}_1 and \mathbf{b}_2 axes respectively at a rate of 760 Hz.

In this work, a custom-made PX4 autopilot based quadcopter with an X-configuration frame and arm lengths of 23 cm is used. Consumer grade motors, electronic speed controllers (ESCs), and propellers are used that allowed a maximum thrust-to-weight ratio of 3 : 1. The physical parameters of the experimental platform are mentioned in Table 1. The CAM device is rigidly mounted below the quadcopter's chassis, such that the chassis geometrical center and CAM lever pivot point are aligned in the chassis plane and have a small offset of 2 cm along \mathbf{b}_3 axis.

Description	Notation	Value
Mass of quadcopter	m	1.2 kg
Moment of Inertia	\mathbf{J}	diag(0.016, 0.017, 0.032) kg m ²
Mass of payload	m_p	82 g
Length of cable	l	1 m

Table 1. Physical parameters of the experimental setup.

With the cable attitude values in frame $\{B\}$, $\phi_{p,CAM}$ and $\theta_{p,CAM}$, the payload position in frame $\{I\}$ is written in Eq. (5), where quadcopter's attitude, \mathbf{R} , from an on-board IMU is utilized. For simplicity, the effect of offset along \mathbf{b}_3 axis is ignored. Accordingly, cable attitude in frame $\{I\}$, i.e., ϕ_p and θ_p , are evaluated by equating Eqs. (1 & 5).

$$\mathbf{X}_p = \mathbf{X} + \mathbf{R} \mathbf{R}_x(\phi_{p,CAM}) \mathbf{R}_y(\theta_{p,CAM}) \begin{bmatrix} 0 & 0 & -l \end{bmatrix}^T \quad (5)$$

Validation: Cable attitude measurements from the CAM device are verified using a motion capture system from Vicon. In Fig. 2, two sets of experimental data are presented, (A) Static trial, where the suspended payload was perturbed with the system rigidly mounted on a stationary frame, and (B) Quadcopter movement trial, where the quadcopter's chassis was moved manually in the workspace to induce payload oscillations.

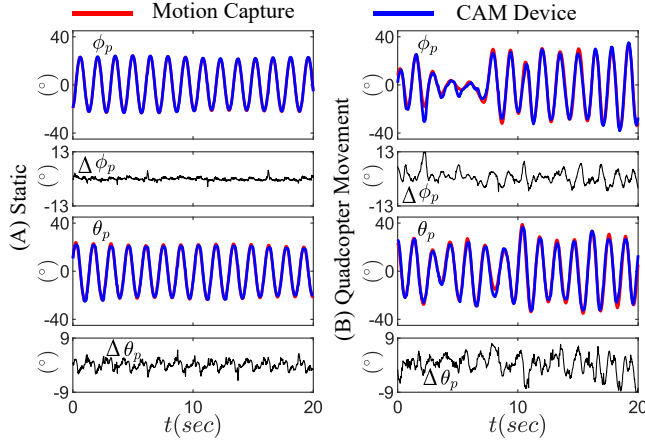


Figure 2. The experimental results of the CAM device validation. (A) Static Trial: where the suspended payload was perturbed with the system rigidly mounted on a stationary frame. (B) Quadcopter movement trial: where the quadcopter’s chassis was moved manually in the workspace to induce payload oscillations. $\Delta\phi_p$ and $\Delta\theta_p$ are the difference between motion capture data to CAM device data.

A close match is observed between the cable attitude values, represented in frame $\{I\}$, between the two data sources for both the trials. Small differences ($\Delta\phi_p, \Delta\theta_p$) are also observed particularly at payload’s extreme positions, which can be due to the cumulative effect of swinging payload inertia and CAM device friction. Relatively small root mean square error in (ϕ_p, θ_p) values are observed, ($1.5^\circ, 1.8^\circ$) for trial (A) and ($3.16^\circ, 3.12^\circ$) for trial (B). With these relatively small errors, CAM device provides a reasonable feedback of the cable attitude in real-time that is used for CAC to minimize the payload oscillations during the flight.

2.3. On-board Controller Design

The overall objective of the controller is to enable stable and smoother outdoor aerial transportation of a cable-suspended payload using a quadcopter. In this context, an on-board control strategy that establishes a three-fold interaction between a human operator, CAM device, and PX4 autopilot controller is implemented. In particular, the controller is designed to take higher-level commands in the form of quadcopter’s roll, ϕ_h , pitch, θ_h , yaw rate, $\dot{\psi}_h$, and velocity along e_3 axis, \dot{z}_h , to attain a specific quadcopter attitude and altitude respectively. These values over time decide the desired quadcopter’s path to be followed, say $\mathbf{X}_d = [x_d, y_d, z_d]^T$. Furthermore, based on the cable attitude feedback from the

CAM device, the controller administers counter measures to minimize payload oscillations during payload transportation while executing the commanded maneuver \mathbf{X}_d . Accordingly, the total thrust force and moment, F and \mathbf{M} , are generated to fly the quadcopter-payload system.

To formulate the control architecture, a linearized model of the system in Eqs. (3 & 4) is developed about the hover equilibrium configuration, marked by zero quadcopter's translational position and velocity, ($\mathbf{X} = \dot{\mathbf{X}} = [0, 0, 0]^T$), zero quadcopter's roll-pitch-yaw angles, ($\phi = \theta = \psi = 0$), zero quadcopter's angular rates, ($\dot{\phi} = \dot{\theta} = \dot{\psi} = 0$), and zero cable attitude and its rates, ($\phi_p = \theta_p = 0, \dot{\phi}_p = \dot{\theta}_p = 0$). Further, the thrust force and moment generated by the quadcopter at the hover equilibrium configuration are $F = (m + m_p)g$ and $\mathbf{M} = [0, 0, 0]^T$ respectively. The linearized model is represented in Eqs. (6-8), where, $m_t = (m + m_p)$. Noting that the quadcopter's attitude dynamics, Eq. (7), is not affected by its translational dynamics, Eq. (6), and payload's rotational dynamics, Eq. (8). Hence, quadcopter's attitude controller can be designed separately. In this work, we utilized PX4 autopilot's attitude controller which generates required moment \mathbf{M} to track the desired attitude of the quadcopter, i.e., (ϕ_d, θ_d, ψ_d). This attitude controller is based on the standard PID controller, as described in [44, 45].

$$\ddot{x} = \frac{m_t g}{m} \theta - \frac{m_p g}{m} \theta_p, \quad \ddot{y} = -\frac{m_t g}{m} \phi + \frac{m_p g}{m} \phi_p, \quad \ddot{z} = \frac{F}{m_t} - g \quad (6)$$

$$\ddot{\phi} = \frac{M_1}{J_1}, \quad \ddot{\theta} = \frac{M_2}{J_2}, \quad \ddot{\psi} = \frac{M_3}{J_3} \quad (7)$$

$$\ddot{\phi}_p = \frac{m_t g}{m l} \phi - \frac{m_t g}{m l} \phi_p, \quad \ddot{\theta}_p = \frac{m_t g}{m l} \theta - \frac{m_t g}{m l} \theta_p \quad (8)$$

From the linearized system's dynamics, Eqs. (6-8), it is noted that the quadcopter's attitude dynamics in ϕ and θ can regulate the quadcopter's translational dynamics, specifically along e_1 and e_2 axes, and also the payload's rotational dynamics, which are also coupled with each other. Accordingly, the values of angles (ϕ_d, θ_d) in the attitude controller are planned to navigate the system along desired course with reduced payload oscillations. In particular, from the coupled terms of Eqs. (6) and (8), quadcopter's roll and pitch, (ϕ, θ), are extracted as in Eq. (9). Error dynamics in quadcopter's position and payload attitude is defined in Eq. (10), where $e_{(\cdot)} = (\cdot) - (\cdot)_d$, and $(\cdot) = \{x, y, z, \phi_p, \theta_p\}$. Further, $K_{(\cdot)}$ and $K_{(\cdot)}$ are like proportional and derivative gains and can be tuned.

$$\phi = -\frac{\ddot{y}}{g} - \frac{m_p l}{m_t g} \ddot{\phi}_p, \quad \theta = \frac{\ddot{x}}{g} - \frac{m_p l}{m_t g} \ddot{\theta}_p \quad (9)$$

$$\ddot{e}_{(\cdot)} + K_{(\cdot)} \dot{e}_{(\cdot)} + K_{(\cdot)} e_{(\cdot)} = 0 \quad (10)$$

$$\phi_d = \underbrace{\frac{-1}{g} \left[\ddot{y}_d - K_{\dot{y}} \dot{e}_y - K_y e_y \right]}_{\phi_h} + \underbrace{\frac{-m_p l}{m_t g} \left[\ddot{\phi}_{p_d} - K_{\dot{\phi}_p} \dot{e}_{\phi_p} - K_{\phi_p} e_{\phi_p} \right]}_{\phi_c} \quad (11)$$

$$\theta_d = \underbrace{\frac{1}{g} \left[\ddot{x}_d - K_{\dot{x}} \dot{e}_x - K_x e_x \right]}_{\theta_h} + \underbrace{\frac{-m_p l}{m_t g} \left[\ddot{\theta}_{p_d} - K_{\dot{\theta}_p} \dot{e}_{\theta_p} - K_{\theta_p} e_{\theta_p} \right]}_{\theta_c} \quad (12)$$

Using (y, ϕ_p) and (x, θ_p) error dynamics in ϕ and θ respectively from Eq. (9), the desired roll angle, ϕ_d , and desired pitch angle, θ_d , are calculated as in Eqs. (11) and (12) respectively. Further, desired roll and pitch angles are divided into two components, i) (ϕ_h, θ_h) which corresponds to track the desired translational position along e_2 and e_1 direction respectively, and ii) (ϕ_c, θ_c) which corresponds to track desired angular position along ϕ_p and θ_p direction respectively. In the current work, the objective is to minimize the payload oscillations while transportation. Hence, the desired cable attitude is kept zero. i.e., $\phi_{p_d} = \theta_{p_d} = 0$. Accordingly, the terms (ϕ_c, θ_c) are referred to as Cable Attitude Controller, CAC, commands. Noting the uncoupled dynamics of ψ from Eqs. (6-8), the yaw rate, $\dot{\psi}_h$, is used to decide quadcopter's heading angle, i.e., desired yaw angle $\psi_d = \int \dot{\psi}_h dt$. Similarly, as dynamics is uncoupled, velocity along e_3 axis, \dot{z}_h , decides the quadcopter's altitude, i.e., desired altitude $z_d = \int \dot{z}_h dt$. Further, z_d value is used in z error dynamics to control the altitude of the system as per Eq. (13).

$$F = m_t (\ddot{z}_d - K_{\dot{z}} \dot{e}_z - K_z e_z) + m_t g \quad (13)$$

The overall control architecture of the system, as implemented in this work, is shown in Fig. 3. The controller is designed based on the linearized dynamics of the system with the assumption of cable remains taut. Hence, for stable and smooth payload transportation and avoid aggressive maneuvers,

bounds on the higher-level commands are imposed as: $\phi_h, \theta_h \in [-45^\circ \ 45^\circ]$, $\dot{\psi}_h \in [-45^\circ \ 45^\circ]/\text{sec}$, $\dot{z}_h \in [-250 \ 250] \text{ cm/sec}$. Further, as PX4 autopilot attitude controller is defined based on linear PID controller its inputs, (ϕ_d, θ_d) are bounded in the region $[-45^\circ \ 45^\circ]$ for stability point of view. The higher-level commands $(\phi_h, \theta_h, \dot{\psi}_h, \dot{z}_h)$ can be planned either using on-board sensors for localization and obstacle avoidance to enable fully autonomous flight, or can be supplied directly from the human operator using RC remote control joystick, or can be a combination of these two modalities to present a semi-autonomous flight. The CAC commands (ϕ_c, θ_c) are calculated on-board PX4 autopilot at 100 Hz where CAM device provides the real-time cable attitude feedback.

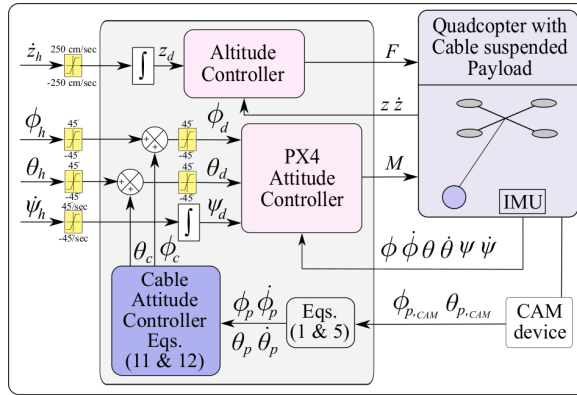


Figure 3. Overall control architecture of the system, where $(\phi_h, \theta_h, \dot{\psi}_h, \dot{z}_h)$ are higher-level commands. Using cable attitude measured by CAM device in frame $\{B\}$, i.e., $\phi_{p,CAM}, \theta_{p,CAM}$ and IMU data, the cable attitude and its rate in frame $\{I\}$, i.e., $(\phi_p, \dot{\phi}_p, \theta_p, \dot{\theta}_p)$ are estimated using Eqs. (1 & 5) in PX4 autopilot. Cable Attitude Controller calculates ϕ_c and θ_c using Eqs. (11 & 12) and these commands are added to ϕ_h and θ_h respectively. The desired yaw angle is calculated by integrating the yaw rate command, $\dot{\psi}_h$. On-board PX4 autopilot attitude controller generates required moment, \mathbf{M} , that tracks the desired attitude of the quadcopter. The desired altitude, z_d , is calculated by integrating \dot{z}_h and using altitude controller Eq. (13), thrust force, F , is applied to the system.

3. OUTDOOR EXPERIMENTS

Two sets of outdoor experiments with the quadcopter-payload system are presented in this section. In the first set of experiments, out of four higher-level commands, (ϕ_h, θ_h) commands are computed using on-board GPS sensor and other two commands $(\dot{\psi}_h, \dot{z}_h)$ are taken from the human operator using RC remote control joystick. In second set of experiments all the four higher-level commands are taken from the human operator using RC remote control joystick. As per the proposed control modality, out of

the four commands, $(\phi_h, \theta_h, \dot{\psi}_h, \dot{z}_h)$, CAC commands, (ϕ_c, θ_c) , are only supplemented to (ϕ_h, θ_h) to plan the desired values of (ϕ_d, θ_d) . Accordingly, within each set of experiments, a comparison is presented between the cases when CAC is not used, i.e., $\phi_c = \theta_c = 0$, and when it is used to control the cable attitude. To make comparison between these two cases, 3D trajectory traced by the quadcopter is estimated using a GPS sensor, quadcopter's roll and pitch angles (ϕ, θ) are estimated using PX4 autopilot's IMU, and cable attitude (ϕ_p, θ_p) are estimated using CAM device. Further, higher-level commands (ϕ_h, θ_h) are estimated using GPS sensor by Eqs. (11-12) in first set of experiments and using RC remote control joystick in second set of experiments. All the experimental data were logged on the PX4 autopilot.

3.1. Semi-autonomous Flight

In this experiment, the commands (ϕ_h, θ_h) are computed using an on-board GPS sensor that provides the quadcopter's latitude and longitude information in real-time. Using standard ECEF (Earth-Centered, Earth Fixed) coordinate system, quadcopter's translational positions, (x, y) , and velocities, (\dot{x}, \dot{y}) are calculated. The task involved in this experiment was to navigate the system through four set-points located at the corners of a square path separated by 10 m at a height of 10 m from the ground. This is shown by points 1-2-3-4-1 in Figs. 4 (A.1, B.1). The tuned gains of the on-board set-point navigation control allowed a maximum velocity of 2.5 m/sec along two directions. During the experiment, when the system hovered at an altitude between 10 ± 3 m, the pre-programmed on-board set-point navigation control mode was activated remotely by RC remote control joystick. The human operator supplied \dot{z}_h commands to maintain the height of the system at 10 m. Further, the yaw rate command, $\dot{\psi}_h$, was not required for executing this task and was kept at zero.

The experimental results are shown for without CAC case in Fig. 4 (A) and for with CAC case in Fig. 4 (B). The output of the set-point navigation control, (ϕ_h, θ_h) commands, and quadcopter roll and pitch, (ϕ, θ) , are shown in Figs. 4 (A.2, A.3) and (B.2, B.3). It is observed that (ϕ, θ) values followed (ϕ_h, θ_h) commands very closely for without CAC case compared to with CAC case. This is because, in the CAC case, the quadcopter response is as per (ϕ_h, θ_h) and non-zero CAC, (ϕ_c, θ_c) , commands. From Figs. 4 (A.1) and (B.1), it is observed that the overall quadcopter maneuver was closer to the desired square trajectory in the CAC case compared to without CAC case. Notably, the variations in the payload

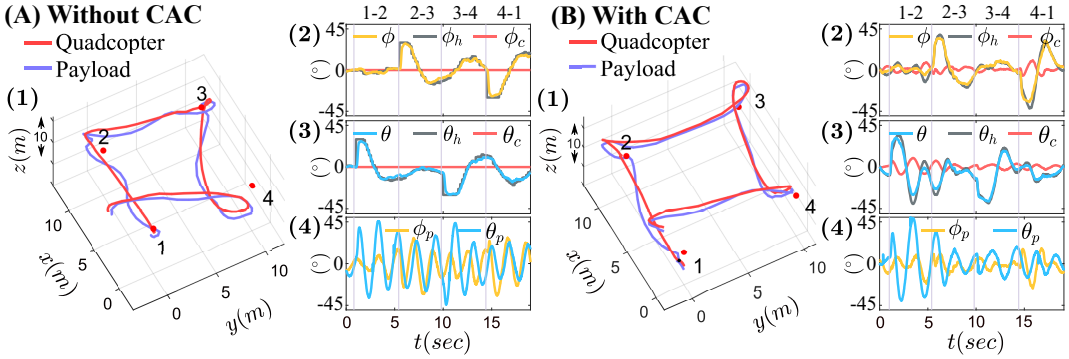


Figure 4. The experimental results for semi-autonomous flight. Plots (A) show when CAC is not implemented and plots (B) show when CAC is implemented. (1) 3D trajectory traced by the quadcopter and payload. The system started its motion from point 1 and followed a path 1-2-3-4-1. The distance between each plots are 10 m. (2) show quadcopter roll angle, ϕ , human roll command, ϕ_h , CAC roll command, ϕ_c . (3) show quadcopter pitch angle, θ , human pitch command, θ_h , and CAC pitch command, θ_c . (4) show the cable attitude, ϕ_p, θ_p .

position relative to quadcopter were higher for without CAC case. This implies that the quadcopter-payload system deviates from its intended trajectory when navigation commands (ϕ_h, θ_h) are planned on-board without incorporating the countermeasures to minimize the payload oscillations.

From the logged data of the flight the relative position of the payload from the quadcopter during the flight along (e_1, e_2, e_3) axes are computed as $(-2.66 \pm 24.29 \text{ cm}, -0.97 \pm 34.15 \text{ cm}, -90.49 \pm 6.84 \text{ cm})$ for without CAC case and $(-3.68 \pm 13.16 \text{ cm}, -1.59 \pm 27.62 \text{ cm}, -94.92 \pm 6.84 \text{ cm})$ for with CAC case. In particular, it is noted that the percentage reductions in the standard deviation of the payload's relative position along (e_1, e_2) axes are about (46%, 20%) for the CAC case compared to without CAC case. Moreover, the payload was on an average of 94.92 cm from the quadcopter along e_3 axis when CAC was used, which is much closer to the length of $l = 100 \text{ cm}$ used for suspending the payload during the experiment. From Figs. 4 (A.4) and (B.4), a notable reduction is observed in the cable attitude values, (ϕ_p, θ_p), for with CAC case compared to without CAC case. In general, the mean and standard deviation in (ϕ_p, θ_p) values are around $(-1.54 \pm 14.35^\circ, -0.65 \pm 21.07^\circ)$ for without CAC case and $(-2.26 \pm 7.66^\circ, -0.95 \pm 16.86^\circ)$ for with CAC case. Using Fast Fourier Transform (FFT) analysis, the dominant frequency component of the payload altitude, (ϕ_p, θ_p), are (0.55 Hz, 0.51 Hz) for without CAC case and (0.1 Hz, 0.17 Hz) for with CAC case respectively. These observations indicate a significant reduction in the payload oscillations when CAC commands were supplemented to the commands of the on-board set-point navigation control.

Overall, the semi-autonomous flight experiments successfully demonstrate the use of CAM device to provide on-board cable attitude feedback and CAC implementation to minimize the payload oscillations during the flight.

3.2. Human-controlled Flight

In this experiment, a human operator fully controlled the quadcopter-payload system by applying four commands, $(\phi_h, \theta_h, \dot{\psi}_h, \dot{z}_h)$ remotely using RC remote control joystick. The task for this experiment was to fly the system along a square path in the horizontal plane with 30 m sides as indicated by points 1-2-3-4-1 in Figs. 5 (A.1) and (B.1). Indicators for these points were provided on the ground as a reference of the square path to the human operator during the experiment. Considering the task in the horizontal plane, the operator mainly applied (ϕ_h, θ_h) commands. The flight was performed at an appropriate altitude between 10 ± 3 m from the ground, which if required was maintained using \dot{z}_h command. Further, as the yaw rate, $\dot{\psi}_h$, command particularly alters the quadcopter's heading angle, it was not required for executing this task.

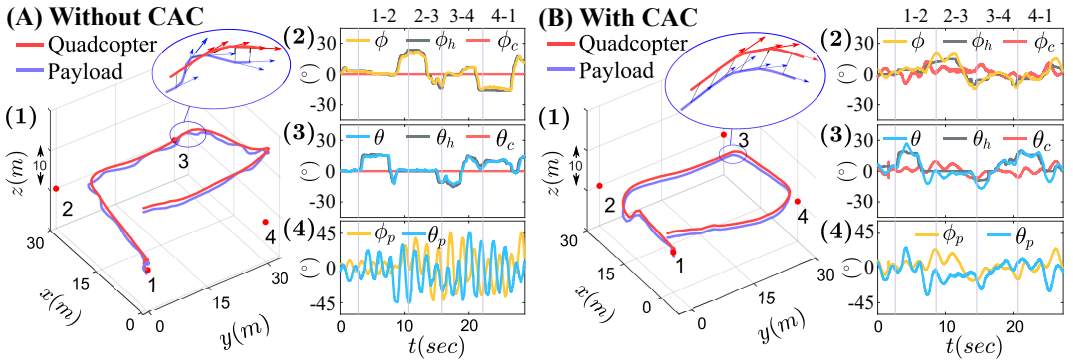


Figure 5. The experimental results for human-controlled flight. Plots (A) shows when CAC is not implemented and plots (B) show when CAC is implemented. (1) 3D trajectory traced by the quadcopter and payload. Point 1 is marked at which human operator starts the task. Points 2, 3, 4 are marked at the corners of the 30 m \times 30 m square path at the same altitude as point 1. (2) Quadcopter's roll angle, ϕ , human roll command, ϕ_h , and CAC roll command, ϕ_c . (3) Quadcopter's pitch angle, θ , human pitch command, θ_h , and CAC pitch command, θ_c . (4) Cable attitude, (ϕ_p, θ_p) .

The experimental result are shown for without CAC case in Fig. 5 (A) and for with CAC case in Fig. 5 (B). To follow a square path, the human operator specified the roll and pitch commands, (ϕ_h, θ_h) , simultaneously to turn the system at the corners as shown in Figs. 5 (A.2, A.3) and (B.2, B.3). For the

human-controlled flight, even though, the exact tracking of the desired square path with reference on the ground is quite challenging, it is observed that the human operator flew the system closer to the path in both cases, as in Figs. 5 (A.1) and (B.1). However, the variations in the payload position relative to the quadcopter were higher for without CAC case, which also affected the overall flight maneuver. This implies that applying countermeasures to minimize payload oscillations can be effective for human-controlled flights of a quadcopter-payload system.

It is observed from Figs. 5 (A) and (B) that with the use of CAM device and CAC, the overall human-controlled flight performance and experience improved. In particular, quadcopter followed the human commands (ϕ_h, θ_h) very closely in without CAC case, as seen in Figs. 5 (A.2, A.3). In contrast, in with CAC case, CAC commands (ϕ_c, θ_c) minimize the payload oscillations during the flight, quadcopter's rolling and pitching motion (ϕ, θ) varied about (ϕ_h, θ_h) , as in Figs. 5 (B.2, B.3). Further, to execute the quadcopter-payload flight in without CAC case, human operator supplied commands instantaneously and of large amplitudes seen as discrete steps in Fig. 5 (A.2, A.3), which were to change the flight direction and to compensate for payload oscillations. Due to the coupled dynamics of the system, abrupt quadcopter motion induced payload oscillations, which in turn require human operator to apply compensatory responses. In contrast, the implementation of CAC in with CAC case supplied (ϕ_c, θ_c) commands to reduce the payload oscillation along with the human commands. This implementation resulted in smoother quadcopter and payload motion and elicited relatively continuous human commands of lower magnitude during the flight as seen in Fig. 5 (B.2, B.3).

From the logged data of the flight the relative position of the payload from the quadcopter along $(\mathbf{e}_1, \mathbf{e}_2, \mathbf{e}_3)$ axes are $(1.27 \pm 30.87 \text{ cm}, 8.66 \pm 29.31 \text{ cm}, 89.58 \pm 9.88 \text{ cm})$ for without CAC case and $(3.44 \pm 16.16 \text{ cm}, 9.08 \pm 19.25 \text{ cm}, 96.24 \pm 3.95 \text{ cm})$ for with CAC case. Compared to without CAC case, the percentage reductions in the standard deviation of payload's relative position along $(\mathbf{e}_1, \mathbf{e}_2, \mathbf{e}_3)$ axes in with CAC case are about (47%, 34%, 60%) during the flight. Moreover, the vertical distance between the quadcopter and payload was on an average 96 cm in with CAC case, which is closer to the cable length of $l = 100 \text{ cm}$, as compared to 89 cm in without CAC case. Essentially, these data show that the payload remains quite underneath the quadcopter when CAC is implemented.

Notable change can also be observed in the payload oscillations from the variations of (ϕ_p, θ_p) in Figs. 5 (A.4) and (B.4) during the flight. The mean and standard deviation values in (ϕ_p, θ_p) are around $(0.69 \pm 18.95^\circ, -5.6 \pm 19.03^\circ)$ in without CAC case and $(1.98 \pm 9.3^\circ, -5.43 \pm 11.39^\circ)$ in with

CAC case. Thus, the percentage reduction in the standard deviation is about (50.6%, 40.1%) in (ϕ_p, θ_p) respectively when CAC is implemented. Moreover, using Fast Fourier Transform (FFT) analysis, the dominant frequency component of the payload altitude, (ϕ_p, θ_p) , are (0.56 Hz, 0.6 Hz) for without CAC case and (0.11 Hz, 0.17 Hz) for with CAC case respectively. It essentially indicates that with the use of CAC the payload oscillates at a lower frequency. These observations indicate attenuated payload oscillations when CAC commands were supplemented with human commands.

Essentially, with the implementation of CAC, the payload consistently maintains its position underneath the quadcopter, which can be a simpler approach to execute impromptu payload transportation tasks in outdoor settings. This is further highlighted in the enlarged view of the quadcopter-payload trajectory by the translational velocity vectors immediately before and after the turn at Point 3, refer Fig. 5 (A.1) for without CAC case and Fig. 5 (B.1) for with CAC case. Payload's velocity vectors point in different directions and vary abruptly with respect to quadcopter's vectors in without CAC case compared to with CAC case. As payload oscillates at high frequency and with a large amplitudes in without CAC case, considerable change in its velocity, and consequently momentum, is expected at the corners of the square trajectory or whenever quadcopter changes directions. As a result, a large reactive force acts on the quadcopter resisting its motion, which requires the operator to apply large and sudden inputs to maintain a path. In contrast, the use of CAC helps in reducing oscillations magnitude and frequency, which imply comparatively smaller reactive force on the quadcopter. Thus, the operator is mainly required to send commands to maneuver along a path, and a comparatively smoother flight is observed.

Overall, the experimental results of human-controlled flight successfully demonstrate the capability of the proposed modality of using an on-board CAM device to estimate the cable attitude, on-board implementation of CAC to minimize the payload oscillations, and incorporating humans' cognitive ability to navigate a quadcopter-payload system in the outdoor settings.

3.3. Case Study

A case study is conducted at a real construction site to demonstrate the feasibility of the proposed modality in impromptu object transportation. The task considered in the case study is to transport a brick-sized object (mass 82 gm) from the ground to the roof of a 40 ft tall building. For simplicity, a snap hook is used to attach and detach the payload with the quadcopter via a cable originating at the CAM device's lever.

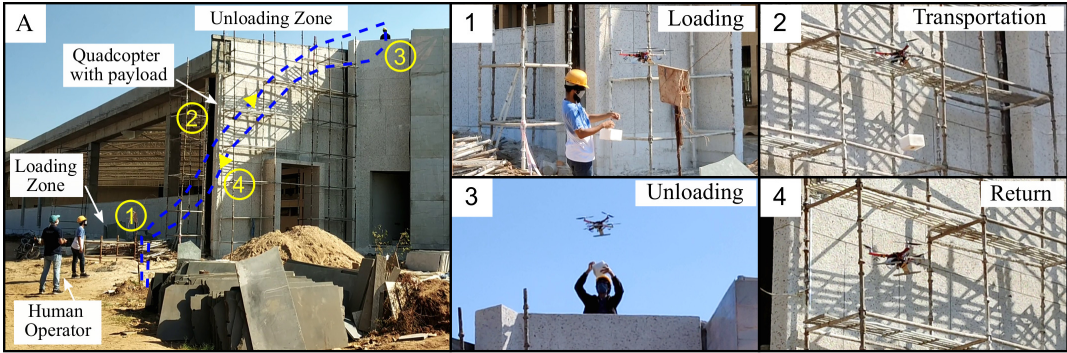


Figure 6. (A) Overall steps of aerial transportation of object at a real construction site, Snapshots: (1) Worker loads the object, (2) Human operator transport the payload, (3) Worker at roof unloads the object, and (4) Human operator returns the quadcopter at loading zone.

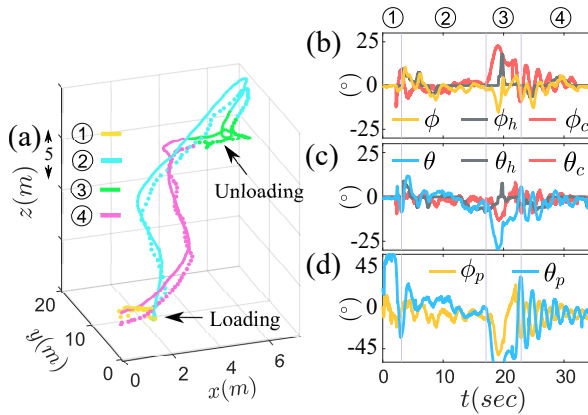


Figure 7. Experimental results of conducted case study. (a) 3D trajectory followed by the quadcopter and payload. Quadcopter's path is shown as solid line and payload's path is shown as dotted line. (b) Quadcopter's roll angle, ϕ , human roll command, ϕ_h , and CAC roll command, ϕ_c . (c) Quadcopter's pitch angle, θ , human pitch command, θ_h , and CAC pitch command, θ_c . (d) Cable attitude, (ϕ_p, θ_p) .

The overall transportation process is shown in Fig. 6 (A). An operator first flies the quadcopter to the loading zone, where a construction worker attaches the object by the hook as shown in Fig. 6 (1). The operator then commands the system to transport the object to the unloading zone as shown in Fig. 6 (2). During the flight, any oscillations produced in the object are attenuated by the implemented CAC, providing a smoother flying experience to the operator. At the unloading zone, another construction worker unloads the object as shown in Fig. 6 (3) after which the operator flies the quadcopter back to the loading zone to transport more objects.

Corresponding experimental results of the presented case study are shown in Fig. 7. The 3D trajectory followed by the quadcopter and payload with indication of four processes are shown in Fig. 7 (a) and

commanded roll and pitch commands, (ϕ_h, θ_h) are shown in the Fig. 7 (b) and (c). During loading and unloading, the worker grabs the hook to attach or detach the payload, leading to a large payload attitude as can be observed in Fig. 7 (d) for the processes (1) and (3). Consequently, CAC supplied appropriate (ϕ_c, θ_c) commands, Figs. 7 (b) and (c), to result in quadcopter rolling and pitching movements to attenuate the payload oscillations. Accordingly, as observed in Fig. 7 (d), the oscillations of the payload are rapidly reduced.

3.4. Discussion and Future Work

The presented experimental results and case study show the performance of CAC and CAM device in effectively reducing the payload oscillations and consequently, minimizing the operator's effort required for maneuvering the system. Thus, the proposed modality is easily deployable to perform impromptu object transportation tasks safely and smoothly.

In particular, the image processing techniques proposed in literature [30–32] to estimate the cable attitude are subjected to higher computation cost with a limited rate of about 50 Hz. Also, the accuracy of these techniques depend on the outdoor lighting conditions, and necessity of using cues and tags for payload detection. In contrast, CAM device is a mechanical sensor and its performance is not limited by similar constraints. Moreover, CAM device being portable achieves on-board sensing and cable attitude estimation at a rate of 760 Hz. The current work demonstrates that the CAM device can reliably be used to compute payload oscillation measure for outdoor flight.

Various control modalities such as geometric controller [14, 18, 27], input shaping [23], reinforcement learning [25, 26], admittance controller [34, 35] have been adopted in the literature to attenuate the payload oscillations. Although the presented CAC is based on a proportional-derivative controller, the experimental results highlight that CAC successfully minimizing the cable oscillations and keeps it underneath the quadcopter, leading to a smooth flight. However, it is noted that the performance of the CAC can be further improved to enable the practical use of the proposed modality. In particular, non-linear and adaptive control schemes can be implemented that guaranteed stability under external disturbances, changes in the mass of the payload and length of the cable for better performance. Additionally, the problems related to the cable slackness will be resolved in future works by detecting cable slacking in real-time, as presented in [19].

Human supervision plays an important role in the applications such as aerial surveillance, search and rescue operations for applying higher-level commands as mentioned in [40, 41]. In this context, we presented a human-in-the-loop control strategy that successfully demonstrates smooth outdoor aerial transportation of suspended payload using CAM device and CAC. In the future, for a thorough evaluation of the presented modality, studies will be conducted with multiple subjects with rigorous experiment protocols that emulate real-world scenarios.

4. CONCLUSION

The current work demonstrates human-controlled aerial transportation of a cable-suspended payload using a quadcopter in outdoor settings. To achieve this, a state feedback controller named Cable Attitude Controller (CAC) is successfully implemented to minimize the payload oscillations for stable and smooth transportation. Further, on-board portable device, Cable Attitude Measurement (CAM) device is developed to measure the cable attitude in real-time, enabling on-board implementation of CAC. Two sets of outdoor experiments, semi-autonomous flight and human-controlled flight, and a case study at a real construction site successfully illustrate the feasibility of the presented modality for outdoor aerial transportation applications.

5. ACKNOWLEDGEMENT

This work is supported by Core Research Grant (CRG/2020/004990) from SERB India.

APPENDIX

The expressions for Matrices $\mathcal{M}(s) \in \mathbb{R}^{5 \times 5}$, $\mathcal{C}(s, \dot{s}) \in \mathbb{R}^{5 \times 5}$, and $\mathcal{G}(s) \in \mathbb{R}^{5 \times 1}$ are given below,

$$\mathcal{M}(s) = \begin{bmatrix} m_t & 0 & 0 & 0 & -m_p \gamma \\ 0 & m_t & 0 & & \\ 0 & 0 & m_t & & M_t^T \\ 0 & & & m_p \gamma^2 & 0 \\ -m_p \gamma & M_t & & 0 & m_p l^2 \end{bmatrix}$$

where, $m_t = m + m_p$, $\gamma = lc\theta_p$, and $\mathbf{M}_t = \begin{bmatrix} m_p lc\phi_p c\theta_p & m_p ls\phi_p c\theta_p \\ -m_p ls\phi_p s\theta_p & m_p lc\phi_p s\theta_p \end{bmatrix}_{2 \times 2}$.

$$\mathbf{C}(s, \dot{s}) = \begin{bmatrix} \vdots & 0 & m_p ls\phi_p \dot{\theta}_p \\ & C_1 & -m_p ls\phi_p c\theta_p \dot{\theta}_p \\ \mathbf{0}_{5 \times 3} & m_p lc\theta_p c\phi_p \dot{\phi}_p & C_2 \\ & 0 & -m_p l^2 s(2\theta_p) \dot{\phi}_p \\ \vdots & m_p l^2 c\theta_p s\theta_p \dot{\phi}_p & 0 \end{bmatrix}$$

where, $C_1 = -m_p ls\phi_p c\theta_p \dot{\phi}_p - 2m_p lc\phi_p s\theta_p \dot{\theta}_p$, and $C_2 = -2m_p ls\phi_p s\theta_p \dot{\phi}_p + m_p lc\phi_p c\theta_p \dot{\theta}_p$

$$\mathbf{G}(s) = \begin{bmatrix} \mathbf{0}_{2 \times 1} \\ m_t g \\ m_p g ls\phi_p c\theta_p \\ m_p g lc\phi_p s\theta_p \end{bmatrix}$$

References

- [1] Sam Cubero. *Industrial robotics: theory, modelling and control*. Pro Literatur Verlag, 2006.
- [2] Miae Kim and Eric T Matson. A cost-optimization model in multi-agent system routing for drone delivery. In *International Conference on Practical Applications of Agents and Multi-Agent Systems*, pages 40–51. Springer, 2017.
- [3] Justin Thomas, Giuseppe Loianno, Joseph Polin, Koushil Sreenath, and Vijay Kumar. Toward autonomous avian-inspired grasping for micro aerial vehicles. *Bioinspiration & biomimetics*, 9(2):025010, 2014.
- [4] Daniel Mellinger, Quentin Lindsey, Michael Shomin, and Vijay Kumar. Design, modeling, estimation and control for aerial grasping and manipulation. In *2011 IEEE/RSJ International Conference on Intelligent Robots and Systems*, pages 2668–2673. IEEE, 2011.
- [5] Daniel Mellinger, Michael Shomin, Nathan Michael, and Vijay Kumar. Cooperative grasping and transport using multiple quadrotors. In *Distributed autonomous robotic systems*, pages 545–558. Springer, 2013.

- [6] Giuseppe Loianno and Vijay Kumar. Cooperative transportation using small quadrotors using monocular vision and inertial sensing. *IEEE Robotics and Automation Letters*, 3(2):680–687, 2017.
- [7] Z. Wang, S. Singh, M. Pavone, and M. Schwager. Cooperative object transport in 3D with multiple quadrotors using no peer communication. In *2018 IEEE International Conference on Robotics and Automation (ICRA)*, pages 1064–1071, May 2018.
- [8] Suseong Kim, Seungwon Choi, and H Jin Kim. Aerial manipulation using a quadrotor with a two dof robotic arm. In *2013 IEEE/RSJ International Conference on Intelligent Robots and Systems*, pages 4990–4995. IEEE, 2013.
- [9] P. O. Pereira, R. Zanella, and D. V. Dimarogonas. Decoupled design of controllers for aerial manipulation with quadrotors. In *2016 IEEE/RSJ International Conference on Intelligent Robots and Systems (IROS)*, pages 4849–4855, 2016.
- [10] Quentin Lindsey, Daniel Mellinger, and Vijay Kumar. Construction of cubic structures with quadrotor teams. *Proc. Robotics: Science & Systems VII*, 2011.
- [11] Federico Augugliaro, Sergei Lupashin, Michael Hamer, Cason Male, Markus Hehn, Mark W Mueller, Jan Sebastian Willmann, Fabio Gramazio, Matthias Kohler, and Raffaello D’Andrea. The flight assembled architecture installation: Cooperative construction with flying machines. *IEEE Control Systems Magazine*, 34(4):46–64, 2014.
- [12] Federico Augugliaro, Ammar Mirjan, Fabio Gramazio, Matthias Kohler, and Raffaello D’Andrea. Building tensile structures with flying machines. In *2013 IEEE/RSJ International Conference on Intelligent Robots and Systems*, pages 3487–3492. IEEE, 2013.
- [13] Ammar Mirjan, Fabio Gramazio, Matthias Kohler, Federico Augugliaro, and Raffaello D’Andrea. Architectural fabrication of tensile structures with flying machines. *Green Design, Materials and Manufacturing Processes*, pages 513–518, 2013.
- [14] Koushil Sreenath, Taeyoung Lee, and Vijay Kumar. Geometric control and differential flatness of a quadrotor uav with a cable-suspended load. In *52nd IEEE Conference on Decision and Control*, pages 2269–2274. IEEE, 2013.
- [15] Koushil Sreenath, Nathan Michael, and Vijay Kumar. Trajectory generation and control of a quadrotor with a cable-suspended load—a differentially-flat hybrid system. In *2013 IEEE International Conference on Robotics and Automation*, pages 4888–4895. IEEE, 2013.

- [16] P. Kotaru, G. Wu, and K. Sreenath. Differential-flatness and control of quadrotor(s) with a payload suspended through flexible cable(s). In *2018 Indian Control Conference (ICC)*, pages 352–357, 2018.
- [17] Jun Zeng, Prasanth Kotaru, Mark W Mueller, and Koushil Sreenath. Differential flatness based path planning with direct collocation on hybrid modes for a quadrotor with a cable-suspended payload. *IEEE Robotics and Automation Letters*, 2020.
- [18] J. Zeng, P. Kotaru, and K. Sreenath. Geometric control and differential flatness of a quadrotor uav with load suspended from a pulley. In *2019 American Control Conference (ACC)*, pages 2420–2427, 2019.
- [19] Sarah Tang and Vijay Kumar. Mixed integer quadratic program trajectory generation for a quadrotor with a cable-suspended payload. In *2015 IEEE International Conference on Robotics and Automation (ICRA)*, pages 2216–2222. IEEE, 2015.
- [20] Philipp Foehn, Davide Falanga, Naveen Kuppuswamy, Russ Tedrake, and Davide Scaramuzza. Fast trajectory optimization for agile quadrotor maneuvers with a cable-suspended payload. In *Robotics: Science and Systems*, pages 1–10, 2017.
- [21] Gerardo De La Torre, Evangelos Theodorou, and Eric N Johnson. Autonomous suspended load operations via trajectory optimization and variational integrators. *Journal of Guidance, Control, and Dynamics*, 40(2):278–291, 2017.
- [22] Shicong Dai, Taeyoung Lee, and Dennis S Bernstein. Adaptive control of a quadrotor UAV transporting a cable-suspended load with unknown mass. In *53rd IEEE Conference on Decision and Control*, pages 6149–6154. IEEE, 2014.
- [23] D. Guo and K. K. Leang. Image-based estimation, planning, and control of a cable-suspended payload for package delivery. *IEEE Robotics and Automation Letters*, 5(2):2698–2705, 2020.
- [24] M Eusebia Guerrero-Sánchez, D Alberto Mercado-Ravell, Rogelio Lozano, and C Daniel García-Beltrán. Swing-attenuation for a quadrotor transporting a cable-suspended payload. *ISA transactions*, 68:433–449, 2017.
- [25] Aleksandra Faust, Ivana Palunko, Patricio Cruz, Rafael Fierro, and Lydia Tapia. Learning swing-free trajectories for uavs with a suspended load. In *2013 IEEE International Conference on Robotics and Automation*, pages 4902–4909. IEEE, 2013.
- [26] Aleksandra Faust, Ivana Palunko, Patricio Cruz, Rafael Fierro, and Lydia Tapia. Automated aerial

- suspended cargo delivery through reinforcement learning. *Artificial Intelligence*, 247:381–398, 2017.
- [27] Taeyoung Lee, Koushil Sreenath, and Vijay Kumar. Geometric control of cooperating multiple quadrotor uavs with a suspended payload. In *52nd IEEE conference on decision and control*, pages 5510–5515. IEEE, 2013.
- [28] Farhad A Goodarzi and Taeyoung Lee. Stabilization of a rigid body payload with multiple cooperative quadrotors. *Journal of Dynamic Systems, Measurement, and Control*, 138(12):121001, 2016.
- [29] K. Mohammadi, S. Sirouspour, and A. Grivani. Control of multiple quad-copters with a cable-suspended payload subject to disturbances. *IEEE/ASME Transactions on Mechatronics*, 25(4):1709–1718, 2020.
- [30] Michael Gassner, Titus Cieslewski, and Davide Scaramuzza. Dynamic collaboration without communication: Vision-based cable-suspended load transport with two quadrotors. In *2017 IEEE International Conference on Robotics and Automation (ICRA)*, pages 5196–5202. IEEE, 2017.
- [31] Morten Bisgaard, Anders la Cour-Harbo, Eric N Johnson, and Jan Dimon Bendtsen. Vision aided state estimator for helicopter slung load system. *IFAC Proceedings Volumes*, 40(7):425–430, 2007.
- [32] S. Tang, V. Wüest, and V. Kumar. Aggressive flight with suspended payloads using vision-based control. *IEEE Robotics and Automation Letters*, 3(2):1152–1159, 2018.
- [33] M. Bernard and K. Kondak. Generic slung load transportation system using small size helicopters. In *2009 IEEE International Conference on Robotics and Automation*, pages 3258–3264, 2009.
- [34] Andrea Tagliabue, Mina Kamel, Sebastian Verling, Roland Siegwart, and Juan Nieto. Collaborative transportation using mavs via passive force control. In *2017 IEEE International Conference on Robotics and Automation (ICRA)*, pages 5766–5773. IEEE, 2017.
- [35] Andrea Tagliabue, Mina Kamel, Roland Siegwart, and Juan Nieto. Robust collaborative object transportation using multiple mavs. *The International Journal of Robotics Research*, page 0278364919854131, 2017.
- [36] Dongjun Lee, Antonio Franchi, Hyoung Il Son, ChangSu Ha, Heinrich H Bühlhoff, and Paolo Robuffo Giordano. Semiautonomous haptic teleoperation control architecture of multiple unmanned aerial vehicles. *IEEE/ASME Transactions on Mechatronics*, 18(4):1334–1345, 2013.

- [37] C. Masone, H. H. Bühlhoff, and P. Stegagno. Cooperative transportation of a payload using quadrotors: A reconfigurable cable-driven parallel robot. In *2016 IEEE/RSJ International Conference on Intelligent Robots and Systems (IROS)*, pages 1623–1630, 2016.
- [38] P. Prajapati, S. Parekh, and V. Vashista. On the human control of a multiple quadcopters with a cable-suspended payload system. In *2020 IEEE International Conference on Robotics and Automation (ICRA)*, pages 2253–2258, 2020.
- [39] Pratik Prajapati, Sagar Parekh, and Vineet Vashista. Collaborative transportation of cable-suspended payload using two quadcopters with human in the loop. In *2019 28th IEEE International Conference on Robot and Human Interactive Communication (RO-MAN)*, pages 1–6. IEEE, 2019.
- [40] Bas Vergouw, Huub Nagel, Geert Bondt, and Bart Custers. Drone technology: Types, payloads, applications, frequency spectrum issues and future developments. In *The future of drone use*, pages 21–45. Springer, 2016.
- [41] FJ Perez-Grau, R Ragel, F Caballero, A Viguria, and A Ollero. Semi-autonomous teleoperation of uavs in search and rescue scenarios. In *2017 International Conference on Unmanned Aircraft Systems (ICUAS)*, pages 1066–1074. IEEE, 2017.
- [42] N. Michael, D. Mellinger, Q. Lindsey, and V. Kumar. The grasp multiple micro-uav testbed. *IEEE Robotics Automation Magazine*, 17(3):56–65, 2010.
- [43] A. Fortin-Côté, P. Cardou, and A. Campeau-Lecours. Improving cable driven parallel robot accuracy through angular position sensors. In *2016 IEEE/RSJ International Conference on Intelligent Robots and Systems (IROS)*, pages 4350–4355, 2016.
- [44] Lorenz Meier, Petri Tanskanen, Friedrich Fraundorfer, and Marc Pollefeys. Pixhawk: A system for autonomous flight using onboard computer vision. In *2011 IEEE International Conference on Robotics and Automation*, pages 2992–2997, 2011.
- [45] PIXHAWK: Copter Attitude Controller. <https://ardupilot.org/dev/docs/apmcopter-programming-attitude-control-2.html>.