Cooperative control of a gripped load by a team of quadrotors

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Abstract. In this paper, an output tracking controller is proposed for cooperative transport of a gripped load by a team of quadrotors. The proposed control law requires the measurement of only four state variables: position and yaw angle of the system. Moreover, the controller provides rejection of step and ramp external force disturbances. In addition, the control basis vectors derived via optimization facilitate the real-time determination of quadrotors' control inputs. Numerical simulations show the effectiveness of the proposed control scheme and its superiority over formerly designed controllers for such systems.

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1. Introduction

Unmanned Aerial Vehicles (UAV) have widely attracted engineers and scientists in the past decade. These platforms can be used in military operations, search and rescue operations, urbanization, etc. To extend capabilities of UAVs, they should be capable of transporting payloads individually or cooperatively. Therefore, scientists have investigated the load transport problem by UAVs in many research articles. Studies in this field can be divided into two general categories: studies about cable-suspended loads [1-19] and researches about gripped loads [11,20-25]. Furthermore, in some of these studies, load is transported by an individual quadrotor [1-8,10,12-17,19,23,24], while, in some other researches [9,11,18,20-22,25], cooperative transport of payloads by UAV teams has been investigated. Different control techniques have been implemented to control UAVs and payloads in these studies such as adaptive control [2,3,14], reinforcement learning [4,5,12], geometric control [7-9,15,18], model predictive control [22], LQR [19,22], gain scheduling [29], PID control [23,24], and sliding mode control [25].

In this study, cooperative transport of a gripped load by a team of quadrotors is addressed. Output tracking control scheme is implemented to design a trajectory tracking controller such that only the measurement of position and yaw angle of the system is required. Furthermore, step and ramp external force disturbances are rejected by the designed controller. Moreover, to calculate the control inputs of the crowded network of the quadrotors in real-time, proper control basis vectors are derived by optimization of a cost function.

This study is organized as follows. Section 2 describes the modeling of the system in detail. The controller scheme is proposed in Section 3. In Section 4, numerical simulations are performed to examine the designed control laws in a cooperative transport of a gripped load. Conclusions are stated in Section 5.

2. Dynamic model

In this section, configuration, acting forces and mo-
ments, and equations of motion of the system are presented. Moreover, practical assumptions about modeling of the system are introduced.

2.1. System configuration
In this study, a team of identical quadrotors is implemented to grip and transport the load. To represent the rotational motion of the system with respect to the inertial reference frame, a body fixed frame $B_L$ is attached to the center of mass of the load. Moreover, a body fixed frame is attached to each quadrotor $Q_i$. Quadrotors grip the load such that $Z_{Q_i}$ for each quadrotor is parallel to $Z_{B_L}$ (Figure 1).

Furthermore, moment of inertia matrix of each quadrotor in its body-fixed frame $J_Q$ is diagonal and $J_{Q_c} = J_{Q_w}$. Moment of inertia tensor of the load in its body-fixed frame, $J_L$, is diagonal, too. Coordinates of the center of mass of each quadrotor in the body-fixed frame of the load are denoted by $x_i^{B_L}$, $y_i^{B_L}$, and $z_i^{B_L}$. In addition, the number of quadrotors is denoted by $n$. Quadrotors should grip the load such that:

$$\sum_{i=1}^{n} x_i^{B_L} = 0, \quad \sum_{i=1}^{n} y_i^{B_L} = 0, \quad \sum_{i=1}^{n} x_i^{B_L} y_i^{B_L} = 0.$$  

Consequently, coordinates of the center of mass of the entire system in $B_L$ frame denoted by $x_L^{B_L}$, $y_L^{B_L}$, and $z_L^{B_L}$ satisfy $x_L^{B_L} = y_L^{B_L} = 0$. Furthermore, axes of the body-fixed frame of the entire system, $B$, are chosen parallel to those of $B_L$ so that:

$$\sum_{i=1}^{n} x_i^B = 0, \quad \sum_{i=1}^{n} y_i^B = 0, \quad \sum_{i=1}^{n} x_i^B y_i^B = 0,$$  

in which the coordinates of the center of mass of each quadrotor in $B$ frame are denoted by $x_i^B$, $y_i^B$, and $z_i^B$. Hence, the moment of inertia matrix of the entire system, $J$, will be diagonal, too. This matrix can be evaluated by the following equation accordingly:

$$J_{xx} = J_{L_{xx}} + n J_{Q_{xx}} + m_Q \sum_{i=1}^{n} (y_i^B)^2 + (z_i^B - z_c^B)^2 + m_L (z_c^B)^2,$$

$$J_{yy} = J_{L_{yy}} + n J_{Q_{yy}} + m_Q \sum_{i=1}^{n} (x_i^B)^2 + (z_i^B - z_c^B)^2 + m_L (z_c^B)^2,$$

$$J_{zz} = J_{L_{zz}} + n J_{Q_{zz}} + m_Q \sum_{i=1}^{n} (y_i^B)^2 + (x_i^B)^2,$$

$$J = diag(J_{xx}, J_{yy}, J_{zz}).$$  

(1)

In Eq. (1), $m_Q$ and $m_L$ denote mass of each quadrotor and mass of the load, respectively. Moreover, $z_c^B$ represents $Z$ coordinate of the center of mass of the load in $B$ frame.

Obviously, total mass of the system can be evaluated as follows:

$$m = nm_Q + m_L.$$  

(2)

2.2. Forces and moments
Thrust force, $F$, and drag moment, $M_z$, produced by a rotor with an angular speed, $\omega$, can be approximated by:

$$F = K_F \omega^2,$$

$$M_z = K_M \omega^2.$$  

(3)

In Eq. (3), $K_F$ and $K_M$ are thrust and drag coefficients of the rotor. Moreover, drag moment acts in opposite direction of the rotor rotation direction. Each quadrotor has four rotors such that rotors number 1 to number 4 are on the positive $X_{Q_1}$, positive $Y_{Q_1}$, negative $X_{Q_1}$, and negative $Y_{Q_1}$, respectively. Furthermore, rotors number 1 and 3 rotate clockwise, while rotors number 2 and 4 rotate anti-clockwise. Distance from the axis of rotation of each rotor to the center of mass of the quadrotor is denoted by $L$. Therefore, net force and moments produced by rotors in $Q_i$ frame can be formulated as follows:
\[
\begin{bmatrix}
F_Q_i \\
M_{xQ_i} \\
M_{yQ_i} \\
M_{zQ_i}
\end{bmatrix} =
\begin{bmatrix}
K_F & K_F & K_F & K_F \\
0 & K_F L & 0 & -K_F L \\
-K_F L & 0 & K_F L & 0 \\
-K_M & K_M & -K_M & K_M
\end{bmatrix}\]

\[
\begin{bmatrix}
\omega_{1}^2 \\
\omega_{2}^2 \\
\omega_{3}^2 \\
\omega_{4}^2
\end{bmatrix}.
\]

(4)

Consequently, according to the system configuration, total force and moments acting on the system can be calculated as follows:

\[
\begin{bmatrix}
F_B \\
M_{xB} \\
M_{yB} \\
M_{zB}
\end{bmatrix} = \sum_{i=1}^{n} \begin{bmatrix}
1 & 0 & 0 & 0 \\
y_i^B & \cos(\psi_i) & -\sin(\psi_i) & 0 \\
-x_i^B & \sin(\psi_i) & \cos(\psi_i) & 0 \\
0 & 0 & 0 & 1
\end{bmatrix}
\begin{bmatrix}
F_{Q_i} \\
M_{xQ_i} \\
M_{yQ_i} \\
M_{zQ_i}
\end{bmatrix}.
\]

(5)

\(\psi_i\) is the yaw angle of the quadrotor \(Q_i\) with respect to \(B\).

2.3. Equations of motion

Translational equations of motion of the system in the inertial reference frame can be formulated as follows:

\[
\ddot{r} = \frac{F_B}{m}R_{e_3} - g_{e_3},
\]

in which \(r, g, e_3,\) and \(R\) are the position vector of the center of mass of the system, gravity acceleration, standard unit vector in \(z\) direction, and rotation matrix from the inertial frame to \(B\) frame.

Furthermore, equations of rotational motion of the system in \(B\) frame are represented as the following equation by denoting angular velocity of the system by \(\Omega\).

\[
J\dot{\Omega} = -\Omega \times J\Omega + \begin{bmatrix} M_{xB} \\
M_{yB} \\
M_{zB} \end{bmatrix}.
\]

(7)

3. Controller design

In this section, following the linearization of equations of motion with respect to the equilibrium point of the system, it is shown that translational equations of motion can be represented as a fourth-order linear system, while yaw dynamics of the system can be considered as a second-order linear system. Therefore, a fourth-order trajectory tracking controller and a second-order yaw controller are designed appropriately. Moreover, controller basis vectors are derived to determine control input of each quadrotor. Since the controller needs the measurement of position and yaw angle and, indeed, measurements made by motion capture systems are usually noisy, proper Kalman filters are introduced.

3.1. Linearized dynamic model

Rotation matrix \(R\) can be represented by successive Yaw-Pitch-Roll rotations as in Eq. (8), shown in Box I.

In Eq. (8), \(\phi, \theta,\) and \(\psi\) represent roll, pitch, and yaw angles, respectively. Therefore, equilibrium states of the system are obtained as follows:

\[
\begin{align*}
\phi &= \dot{\phi} = \theta = \dot{\theta} = \psi = 0, \\
\psi &= \psi_e.
\end{align*}
\]

(9)

According to the last line of Eq. (9), yaw angle of the system can be any desired value. Now, the linearized equation of motion with respect to the equilibrium point will be in the following form by assuming small angles approximation:

\[
\begin{align*}
\ddot{x} &= g\cos(\psi_e)\theta + g\sin(\psi_e)\dot{\phi}, \\
\ddot{y} &= g\sin(\psi_e)\theta - g\cos(\psi_e)\dot{\phi}, \\
\ddot{z} &= \frac{F_B}{m} - g,
\end{align*}
\]

(10)

By taking the second derivative of the translational equations of motion with respect to time and combining them with the equations of rotational motion, one can find the following equations:

\[
R = \begin{bmatrix}
\cos(\theta)\cos(\psi) & \cos(\psi)\sin(\phi)\sin(\theta) - \cos(\phi)\sin(\psi) \\
\cos(\theta)\sin(\psi) & \sin(\psi)\sin(\phi)\sin(\theta) + \cos(\phi)\cos(\psi) \\
-\sin(\theta) & \cos(\theta)\sin(\phi)
\end{bmatrix}
\begin{bmatrix}
\cos(\phi)\sin(\theta)\cos(\psi) + \sin(\phi)\sin(\psi) \\
\cos(\phi)\sin(\theta)\sin(\psi) - \sin(\phi)\cos(\psi) \\
\cos(\phi)\cos(\theta)
\end{bmatrix}.
\]

(8)

Box I
\[ x^{(4)} = g \cos(\psi_e) \frac{M_{yn}}{J_{yy}} + g \sin(\psi_e) \frac{M_{xn}}{J_{xx}}. \]
\[ y^{(4)} = g \sin(\psi_e) \frac{M_{yn}}{J_{yy}} - g \cos(\psi_e) \frac{M_{xn}}{J_{xx}}. \]
\[ z^{(4)} = \ddot{F}_B \frac{g}{m} - g. \]
\[ \ddot{\psi} = M_{zn} \frac{g}{J_{zz}}. \] 
\[ (11) \]

Consequently, equations of motion are simplified to:
\[ r^{(4)} = v, \]
\[ \ddot{\psi} = M_{zn} \frac{g}{J_{zz}}. \] 
\[ (12) \]

by following feedback linearization rule:
\[
\begin{bmatrix}
\ddot{F}_B \\
\dot{M}_{xn} \\
\dot{M}_{yn}
\end{bmatrix} =
\begin{bmatrix}
J_{xx} \sin(\psi_e)/g & -J_{xx} \cos(\psi_e)/g & 0 \\
J_{yy} \cos(\psi_e)/g & J_{yy} \sin(\psi_e)/g & 0 \\
0 & 0 & 1/m
\end{bmatrix}
\begin{bmatrix}
v_x \\
v_y \\
v_z
\end{bmatrix}
+ \begin{bmatrix}
0 \\
0 \\
g
\end{bmatrix}.
\] 
\[ (13) \]

3.2. Trajectory tracking controller design

According to Eq. (13), to track the desired trajectory denoted by \( r_T \), one can adopt \( v \) as:
\[ v = -K_p e_r - K_d \dot{e}_r - K_i \int e_r \, dt + r_T^{(4)} \]
\[ e_r = r - r_T, \] 
\[ (14) \]
in which \( K_p, K_d, K_i \), and \( K_j \) are positive constant gains. Therefore, dynamics of the trajectory tracking error for the linearized system will be as follows:
\[ e_r^{(4)} + K_p e_r^{(3)} + K_d \dot{e}_r + K_i \int e_r \, dt + K_j e_r = 0. \] 
\[ (15) \]

To track the desired trajectory, it is sufficient for the gains of the controller to be chosen such that polynomial of Eq. (15) be Hurwitz. By this selection of the gains, trajectory tracking error will converge to zero exponentially.

Moreover, the equation of translational motion for the system subjected to external force disturbances can be represented as follows:
\[ \ddot{r} = \frac{\ddot{F}_B}{m} \frac{g}{e_3} + \frac{d(t)}{m}, \] 
\[ (16) \]
where \( d(t) \) is the disturbance vector. Therefore, linearized equations of motion will be as follows:
\[ r^{(4)} = v + \frac{d(t)}{m}. \] 
\[ (17) \]

By Eq. (14), trajectory tracking dynamics will be as follows:
\[ e_r^{(4)} + K_p e_r^{(3)} + K_d \dot{e}_r + K_i \int e_r \, dt + \ddot{\psi} = \ddot{d}(t) / m. \]
\[ (18) \]

Therefore, steady-state tracking error is zero for step and ramp disturbances. In other words, step and ramp disturbances are rejected out by the controller properly.

3.3. Yaw tracking controller

To track the desired trajectory of yaw angle denoted by \( \psi_T \) exponentially, it is sufficient to choose \( M_{zn} \) as follows:
\[ M_{zn} = J_{zz} \left( -K_{p\psi} e_\psi - K_{d\psi} \dot{e}_\psi + \ddot{\psi}_T \right). \]
\[ (19) \]

In control law (Eq. (19)), \( K_{p\psi} \) and \( K_{d\psi} \) are positive constant gains.

3.4. Control basis vectors

Equation (5) defines four equations with 4n unknowns and can be rewritten as follows:
\[
\begin{bmatrix}
F_B \\
M_{xn} \\
M_{yn} \\
M_{zn}
\end{bmatrix} = A u.
\]
\[ (20) \]
in which \( A \in \mathbb{R}^{4 \times 4n} \) is a constant matrix and can be determined from Eq. (5) by considering \( u \) as:
\[
u = \begin{bmatrix}
F_{Q_1} M_{x_{Q_1}} M_{y_{Q_1}} M_{z_{Q_1}} \ldots F_{Q_n} M_{x_{Q_n}} M_{y_{Q_n}} M_{z_{Q_n}}
\end{bmatrix}^T.
\]
\[ (21) \]

For a system with more than one quadrotor, Eq. (20) remains undetermined. Therefore, one can choose 4n - 4 of inputs independent of the desired net force and moments. However, an optimal control input \( u^* \) that achieves the desired inputs while minimizing a cost function \( J_0 \) is adopted in this study. In other words,
\[
u^* = \arg \min_u \left\{ J_0 \left[ F_{B}^{des} \ M_{x_{n}}^{des} \ M_{y_{n}}^{des} \ M_{z_{n}}^{des} \right]^T \right\}.
\]
\[ (22) \]
where:
\[
J_0 = \sum_{i=1}^{n} W_{F_{Q_i}} F_{Q_i}^2 + W_{M_{x_{Q_i}}} M_{x_{Q_i}}^2 + W_{M_{y_{Q_i}}} M_{y_{Q_i}}^2 + W_{M_{z_{Q_i}}} M_{z_{Q_i}}^2
\]
\[ (23) \]
In other words, $J_0 = |Hq|^2$ in which:

$$H = \text{diag} \left( \sqrt{W_{F_{q_1}}}, \sqrt{W_{M_{\psi q_1}}}, \sqrt{W_{M_{\psi q_2}}}, \sqrt{W_{M_{\psi q_3}}} \right),$$

$$\quad \ldots, \sqrt{W_{F_{q_n}}}, \sqrt{W_{M_{\psi q_n}}}, \sqrt{W_{M_{\psi q_{n+1}}}}, \sqrt{W_{M_{\psi q_{n+2}}}} \right).$$

(24)

Hence, the optimal control input will be:

$$u^* = H^{-T} (AH^{-2} A^T)^{-1}$$

$$\left[ \begin{array}{c} P_{\text{des}}^B M_{\psi B} \ v_{\text{des}}^B M_{\gamma B} \end{array} \right]^T.$$ (25)

It is supposed that $W_{F_{q_1}} = W_F$ and $W_{M_{\psi q_1}} = W_{M_{\psi}}$ because quadrotors are identical. Moreover, roll and pitch moment weights are adopted as $W_{M_{\psi q_1}} = W_{M_{q_1}} = W_{M_{\psi}}$, because quadruplets are axially symmetric and roll and pitch moments are treated in the same way. After some algebraic manipulations, one can show that:

$$u^* = \left[ \begin{array}{c} U_F & U_{M_\psi} & U_{M_\psi} & U_{M_\psi} \end{array} \right]$$

$$\left[ \begin{array}{c} P_{\text{des}}^B m_{\text{des}}^B M_{\psi y_B} \ M_{\psi y_B} \end{array} \right]^T,$$ (26)

$U_F$, $U_{M_{\psi}}$, $U_{M_{\psi}}$, and $U_{M_{\psi}}$ are obtained by Eq. (27) as shown in Box II.

Therefore, control basis vectors can be simply calculated through Eq. (27).

3.5. Kalman filter

In the case of implementing a motion capture system, position and yaw angle of the load can be measured by the markers of such systems. However, measurements may be noisy, which should be filtered. Moreover, higher order derivatives of position and yaw angle for the designed feedback controllers should be estimated from the measurements. Thus, one can use Kalman filters to estimate these states, because system (13) is observable and measurements usually are disturbed by zero mean Gaussian white noises.

3.5.1. State transition and control-input models for the translational Kalman filter

To estimate first to third derivatives of the position of the load for feedback, state transition and control input models of the Kalman filter should be considered respectively as follows:

$$A_r = \begin{bmatrix} 1 & T_x & T_x^2 & T_x^3 \\ 0 & 1 & T_x & T_x^2 \\ 0 & 0 & 1 & T_x \\ 0 & 0 & 0 & 1 \end{bmatrix}, \quad B_r = \begin{bmatrix} T_x^4 \\ 0 \\ 0 \\ 0 \end{bmatrix}.$$ (28)

Sampling time is denoted in Eq. (28) by $T_s$ accordingly.

3.5.2. State transition and control-input models for yaw angle Kalman filter

State transition and control input models of the Kalman filter to estimate yaw angle and velocity from noisy measurements should be:

$$A_\psi = \begin{bmatrix} 1 & 0 & T_s \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}, \quad B_\psi = \begin{bmatrix} T_s^2 \\ T_s \\ 0 \end{bmatrix}.$$ (29)

4. Simulations

In this section, the performance of the designed control laws in position and yaw angle trajectory tracking of a rectangular cubic load (Figure 2) with a team of four identical quadrotors is examined. Moreover, orientation and location of the body-fixed frame $B_L$ are demonstrated in Figure 3. Moreover, physical and geometrical properties of the load and quadrotors are listed in Table 1.

The body-fixed frame of the entire system should be located at the center of the mass of the system parallel to $B_L$ as described formerly. To validate the designed controller fairly, the nonlinear model of the system represented by Eqs. (4) to (7) is used in simulations. This model has been verified experimentally in [21].

$$U_F = \frac{1}{n} \begin{bmatrix} 1 & 0 & 0 & 0 & \ldots & 1 & 0 & 0 & 0 \end{bmatrix}^T,$$

$$U_{M_\psi} = \frac{1}{n} \begin{bmatrix} 0 & 0 & 0 & 1 & \ldots & 0 & 0 & 0 & 1 \end{bmatrix}^T,$$

$$U_{M_\psi} = \frac{1}{\sum \left( y_i^B \right)^2 + n} \begin{bmatrix} W_{M_{\psi}} x_i^B \cos(\psi_1) - \sin(\psi_1) & \ldots & W_{M_{\psi}} y_i^B \cos(\psi_n) - \sin(\psi_n) & 0 \end{bmatrix}^T,$$

$$U_{M_\psi} = \frac{1}{\sum \left( x_i^B \right)^2 + n} \begin{bmatrix} W_{M_{\psi}} x_i^B \sin(\psi_1) \cos(\psi_1) & \ldots & - W_{M_{\psi}} x_i^B \sin(\psi_n) \cos(\psi_n) & 0 \end{bmatrix}^T.$$ (27)

Box II
4.1. Trajectory generation

In the considered mission for the quadrotors, they should transport the load from its initial position to a desired destination at a given time. In addition, they should rotate the load to a final desired yaw angle. Therefore, proper trajectories should be generated for position and yaw angle of the system. According to Eq. (13), snap of the position trajectory and acceleration of the yaw trajectory should be minimized to have a minimum input trajectory. Consequently, trajectory generation for position of the system can be formulated as follows:

\[
\min \int_{t_0}^{t_f} \left| r_T^{(4)}(t) \right|^2 dt
\]

subject to:
\[
\begin{align*}
    r_T(t_0) &= r_0 \\
    r_T^{(i)}(t_0) &= 0, \quad i = 1, 2, 3 \\
    r_T(t_f) &= r_f \\
    r_T^{(i)}(t_f) &= 0, \quad i = 1, 2, 3
\end{align*}
\]  \hspace{1cm} (30)

In this formulation, \( t_0, t_f, r_0 \), and \( r_f \) are the initial time, given final time, initial position, and final desired position, respectively. By introducing a variable as \( \tau = t - t_f \), trajectory generation can be rewritten as follows:

\[
\min \int_0^1 \left| r_T^{(4)}(\tau) \right|^2 d\tau
\]

subject to:
\[
\begin{align*}
    r_T(0) &= r_0 \\
    r_T^{(i)}(0) &= 0, \quad i = 1, 2, 3 \\
    r_T(1) &= r_f \\
    r_T^{(i)}(1) &= 0, \quad i = 1, 2, 3
\end{align*}
\]  \hspace{1cm} (31)

Now, trajectory generation can be formulated as a quadratic problem considering desired trajectory in each direction as a P-order polynomial of \( \tau \). For example, trajectory generation in \( X \) direction can be rewritten as the following quadratic problem:

\[
\min c_x^T H c_x.
\]

subject to \( A_x c_x = B_x \).  \hspace{1cm} (32)

for the desired trajectory defined as follows:

\[
x_T = c_{0x} + c_{1x}\tau + c_{2x}\tau^2 + \cdots + c_{px}\tau^p.
\]  \hspace{1cm} (33)

In the quadratic form of the trajectory generation problem, vector \( c_x \) is constructed as:

\[
c_x = [c_{0x} \quad c_{1x} \ldots \quad c_{px}]^T.
\]

Moreover, matrix \( H \) can be evaluated by calculating the following integral.

\[
H = \int_0^1 h(\tau) h(\tau) d\tau
\]

\[
h(\tau) = \begin{bmatrix} 0 & 0 & 0 & 24 & 120\tau & 360\tau^2 \\
& & & & & \vdots \end{bmatrix} \otimes I_{3\times 3}.
\]  \hspace{1cm} (34)

| Table 1. Physical and geometrical properties of the system. |
|---------------------------------------------------------------|
| \( m_L \) | 0.2 Kg |
| \( m_Q \) | 0.65 Kg |
| \( J_Q \) | \( \text{diag}(7.5, 7.5, 7.5, 13) \times 10^{-3} \text{ kg.m}^2 \) |
| \( a \) | 1.5 m |
| \( b \) | 0.8 m |
| \( c \) | 0.01 m |
| \( J_{xx} \) | 0.1845 kg.m² |
| \( J_{yy} \) | 0.6753 kg.m² |
| \( J_{zz} \) | 0.1071 kg.m² |
| \( x^B_L \) | -0.3 m |
| \( y^B_L \) | 0.3 m |
| \( z^B_L \) | 0 |
| \( \psi_1 \) | 0 |
| \( \psi_2 \) | 0.45 m |
| \( \psi_3 \) | 0.7 m |
| \( z^B_L, i = 1, 2, 3, 4 \) | 0.105 m |
Furthermore, matrices $A_x$ and $B_x$ are constructed as:

\[
A_x = \begin{bmatrix}
A_{0x} \\
A_{1x}
\end{bmatrix},
\]

\[
B_x = \begin{bmatrix}
B_{0x} \\
B_{1x}
\end{bmatrix},
\]

in which:

\[
A_{0x} = \begin{bmatrix}
1 & 0 & 0 & \cdots & 0 \\
0 & 1 & 0 & \cdots & 0 \\
0 & 0 & 2 & \cdots & 0 \\
0 & 0 & 0 & \cdots & 0
\end{bmatrix},
\]

\[
B_{0x} = \begin{bmatrix}
x_0 \\
0 \\
0
\end{bmatrix},
\]

\[
A_{1x} = \begin{bmatrix}
1 & 1 & 1 & \cdots & 1 \\
0 & 1 & 2 & \cdots & p \\
0 & 0 & 2 & \cdots & p(p-1) \\
0 & 0 & 0 & \cdots & p(p-1)(p-2)
\end{bmatrix},
\]

\[
B_{1x} = \begin{bmatrix}
x_f \\
0 \\
0
\end{bmatrix}.
\]

This procedure can be applied in other directions similarly. As mentioned before, the acceleration of the yaw trajectory should be minimized. Therefore, trajectory generation for yaw angle is an optimization problem as follows:

\[
\min_{x_0} \int_{t_0}^{t_f} \left| \psi_T^2(t) \right|^2 dt
\]

\[
\text{s.t.}
\]

\[
\begin{align*}
\psi_T(t_0) &= \psi_0 \\
\psi_T(0) &= 0 \\
\psi_T(t_f) &= \psi_f \\
\psi_T(0) &= 0
\end{align*}
\]

This optimization problem can be formulated as a quadratic problem by the same variable change and considering yaw trajectory as a polynomial function of $t$. Problem formulation for yaw angle trajectory generation is similar to the position trajectory planning procedure; however, matrices are different according to the optimization constraints.

4.3. Tracking performance of the controller

Controller gains are adopted by LQR design appropriately. Furthermore, position and yaw measurements are made noisy by white Gaussian zero-mean noises to examine the performance of the Kalman filters. Moreover, weighting parameters are adopted such that $\frac{W_{q}}{W_{p}} = \frac{1}{2}$. This ratio determines the importance of minimizing the thrust force or roll and yaw moments of each quadrotor. The performance of the proposed control scheme in tracking the desired trajectory of the load is demonstrated in Figure 4. Furthermore, yaw tracking performance of the controller is depicted in Figure 5. As demonstrated in these figures, although the controller designed by the linearized dynamic model of the system and the measurements is noisy, the desired trajectory has been tracked well by the proposed controller. Therefore, the designed controller along with Kalman filters can guarantee the stability of the system on the desired trajectories. Obviously, the estimation of the states by both of the filters is satisfactory because trajectory tracking aim is achieved. However, noisy measurements and filtered measurements are plotted in Figures 6 and 7 to

![Figure 4](image1.png)

**Figure 4.** Trajectory of the load versus the desired trajectory.

![Figure 5](image2.png)

**Figure 5.** Yaw trajectory of the load versus the desired yaw trajectory.
exhibit the performance of the position Kalman filter accordingly. Therefore, the measurement of position and yaw angle is sufficient for load trajectory tracking control by means of Kalman filtering.

4.4. Disturbance rejection performance of the controller

In the next simulations, the system is subjected to step and ramp force disturbances to examine disturbance rejection capability of the controller. At first, disturbance of the system is considered as \( d(t) = \begin{bmatrix} 5.6 & -2.8 & 8.4 \end{bmatrix} \). As depicted in Figure 8, tracking error converges to zero properly, and disturbance effect is attenuated by the proposed controller well. However, the controller of the study [21] has not rejected the disturbance, and the steady-state tracking error is about 1 m.

Moreover, the system is also subjected to a ramp disturbance \( d(t) = \frac{5.6t - 2.8t - 8.4t}{100} \) and performances of the controllers are compared accordingly. Figure 9 demonstrates that the proposed controller in this study has attenuated the disturbance properly while the controller of study [21] is unable to reject the disturbance.

Therefore, the proposed controller is more proper than formerly designed control laws to implement in real applications in which the considered system may be subjected to force disturbances such as wind.

5. Conclusion

In this paper, the performance of an output tracking controller is demonstrated on a challenging problem, i.e., cooperative transport of a gripped load by quadrotors. Simulations and analytical results illustrate that:
Figure 9. Trajectory tracking error in the presence of ramp disturbance by the proposed controller in this study and controller of study [21].

- The proposed controller provides a significant reduction in the disturbance influence on the tracking performance. Such capability of the controller is important in real applications where the effect of wind can cause problems;
- The designed control needs the measurement of only position and yaw angle of the system, which can be measured by motion capture systems such as Vicon in indoor applications;
- By applying the derived control basis vectors, real-time implementation of the control becomes possible.

The main drawback of the proposed control scheme is the required high bandwidth in the case of large networks. However, a similar centralized controller has been implemented to control gripped loads by four quadrotors in a study [21]. In future studies, the proposed controller will be implemented experimentally. In addition, a decentralized control scheme will be designed.

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