Dynamic Modeling, Control and Simulation of an Autonomous Quadrotor

MEAM 620 Project 1, Phase 1

Due: Tuesday, Feb 7th 11:59 PM

1 Introduction

In the last seven years, advances in materials, electronics, sensors and batteries have fueled a growth in the development of Micro Unmanned Aerial Vehicles (MAVs) that are between 0.1-0.5 m in length and 0.1-0.5 kg in mass [1]. A few groups have built and analyzed MAVs in the 10 cm range [2, 3]. One of the smallest is the Picoflyer with a 60 mm propeller diameter and a mass of 3.3 g [4]. Platforms in the 50 cm range are more prevalent with several groups having built and flown systems of this size [5–7]. In fact, there are several commercially available RC helicopters and research grade helicopters in this size range [8]. A review of the challenges in develop autonomous micro UAVs is presented in [9].

In this project, we introduce the modeling of quad rotors used in the GRASP Multiple MAV testbed to support research on coordinated, dynamic flight of MAVs. This document is derived from the complete reference [10] which describes the approach to modeling, control and integrating off-the-shelf quad rotors.

2 System Description

2.1 Quadrotor Platform

For this project we will be using the CrazyFlie 2.0 platform made by Bitcraze, shown in Fig. 1. The CrazyFlie has a motor to motor (diagonal) distance of 92 mm, and a mass of 30 g, including a battery. An onboard microcontroller allows low-level control and estimation to be done onboard the robot. An onboard IMU provides feedback of angular velocities and accelerations.

2.2 VICON Motion Capture System

The VICON Motion Capture System provides a state estimate for the quadrotor, which is nearly ground truth [11]. The VICON system offers three important features. First, *it is fast*: the system can be run at or below 375 Hz. Second, *it is precise*: experimental tests show that the deviations of position estimates for single static markers are on the order of $50 \,\mu$ m which is well beyond the requirements for flight. Lastly, *it is robust*: the VICON *Tracker* software assumes that markers in models are rigidly attached which enables the system to maintain tracking even if all but one camera are occluded. Using this software, tracking of quadrotors is rarely lost, even during extreme situations such as fast maneuvers (speeds of $3.5 \,\text{m/s}$, accelerations of $15 \,\text{m/s}^2$ and angular speeds of $1000 \,^{\circ}$ /s).

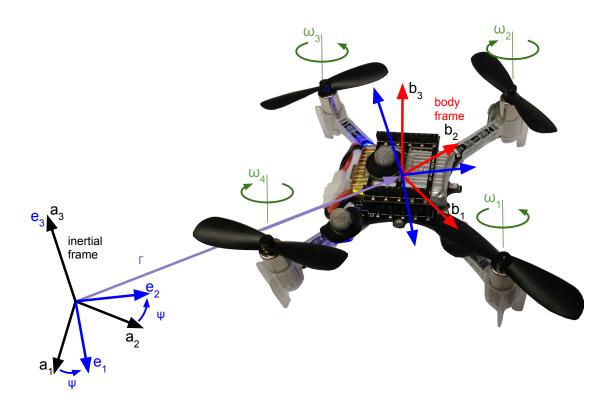


Figure 1: The CrazyFlie 2.0 robot. Note the reflective motion capture markers attached. A pair of motors spins counter clockwise while the other pair spins clockwise, such that when all propellers spin at the same speed, the net torque in the yaw direction is zero. The pitches on the corresponding propellers are reversed so that the thrust is always pointing in the $\mathbf{b_3}$ direction for all propellers. Shown also is the transformation from the inertial frame to the body-fixed frame. First a rotation by ψ around the $\mathbf{a_3}$ axis (leading to coordinate system $\mathbf{e_1}, \mathbf{e_2}, \mathbf{e_3}$) is performed, followed by a translation \mathbf{r} to the center of mass C of the robot. Subsequent rotations by ϕ and θ generate the final body-fixed coordinate system \mathcal{B} , where the axes $\mathbf{b_1}$ and $\mathbf{b_2}$ are aligned with the arms, and $\mathbf{b_3}$ is perpendicular to them.

2.3 Software and Integration

Position control and other high level commands are computed in Matlab at 100 Hz and sent to the robot via the CrazyRadio (2.4GHz). Attitude control is performed onboard using the microcontroller

3 Modeling

3.1 Coordinate Systems and Reference frames

The coordinate systems and free body diagram for the quadrotor are shown in Fig. 1. The inertial frame, A, is defined by the triad a_1 , a_2 , and a_3 with a_3 pointing upward. The body frame, B, is attached to the center of mass of the quadrotor with b_1 coinciding with the preferred forward direction and b_3 perpendicular to the plane of the rotors pointing vertically up during perfect hover (see Fig. 1). These vectors are parallel to the principal axes. The center of mass is C. Rotor 1 is a distance L away along b, 2 is L away along b_2 , while 3 and 4 are similarly L away along the negative b_1 and b_2 respectively.

3.2 Inertial Properties

Since \mathbf{b}_i are principal axes, the inertia matrix referenced to the center of mass along the \mathbf{b}_i reference triad, I, is a diagonal matrix. In practice, the three moments of inertia can be estimated by weighing individual components of the quadrotor and building a physically accurate model in SolidWorks. The key parameters for the rigid body dynamics for the CrazyFlie platform are as follows:

- (a) mass: $m = 0.030 \,\mathrm{kg}$;
- (b) the distance from the center of mass to the axis of a motor: $L = 0.046 \,\mathrm{m}$; and
- (c) the components of the inertia dyadic using b_i as the SRT:

$$[I_C]^{\mathbf{b}_i} = \begin{bmatrix} 1.43 \times 10^{-5} & 0 & 0\\ 0 & 1.43 \times 10^{-5} & 0\\ 0 & 0 & 2.89 \times 10^{-5} \end{bmatrix}.$$

3.3 Motor Model

Each rotor has an angular speed ω_i and produces a vertical force F_i according to

$$F_i = k_F \omega_i^2. \tag{1}$$

Experimentation with a fixed rotor at steady-state shows that $k_F \approx 6.11 \times 10^{-8} \, \text{N/rpm}^2$. The rotors also produce a moment according to

$$M_i = k_M \omega_i^2. (2)$$

The constant, k_M , is determined to be about $1.5 \times 10^{-9} \,\mathrm{Nm/rpm^2}$ by matching the performance of the simulation to the real system.

Data obtain from system identification experiments suggest that the rotor speed is related to the commanded speed by a first-order differential equation

$$\dot{\omega}_i = k_m(\omega_i^{\text{des}} - \omega_i).$$

This motor gain, k_m , is found to be about $20\,\mathrm{s}^{-1}$ by matching the performance of the simulation to the real system. The desired angular velocities, ω_i^{des} , are limited to a minimum and maximum value determined through experimentation.

However, as a first approximation, we can assume the motor controllers to be perfect and the time constant k_m associated with the motor response to be arbitrarily small. In other words, we can assume the actual motor velocities ω_i are equal to the commanded motor velocities, ω_i^{des} .

3.4 Rigid Body Dynamics

Kinematics We will use Z - X - Y Euler angles to model the rotation of the quadrotor in the world frame. To get from \mathcal{A} to \mathcal{B} , we first rotate about \mathbf{a}_3 through the the yaw angle, ψ , to get the triad \mathbf{e}_i . A rotation about the \mathbf{e}_1 through the roll angle, ϕ gets us to the triad \mathbf{f}_i (not shown in the figure). A third pitch rotation about \mathbf{f}_2 through θ results in the body-fixed triad \mathbf{b}_i . You will need the rotation matrix for transforming components of vectors in \mathcal{B} to components of vectors in \mathcal{A} :

$${}^{\mathcal{A}}[R]_{\mathcal{B}} = \begin{bmatrix} c\psi c\theta - s\phi s\psi s\theta & -c\phi s\psi & c\psi s\theta + c\theta s\phi s\psi \\ c\theta s\psi + c\psi s\phi s\theta & c\phi c\psi & s\psi s\theta - c\psi c\theta s\phi \\ -c\phi s\theta & s\phi & c\phi c\theta \end{bmatrix}. \tag{3}$$

We will denote the components of angular velocity of the robot in the body frame by p, q, and r:

$${}^{\mathcal{A}}\omega_{\mathcal{B}} = p\mathbf{b}_1 + q\mathbf{b}_2 + r\mathbf{b}_3.$$

These values are related to the derivatives of the roll, pitch, and yaw angles according to

Newton's Equations of Motion Let \mathbf{r} denote the position vector of C in A. The forces on the system are gravity, in the $-\mathbf{a}_3$ direction, and the forces from each of the rotors, F_i , in the \mathbf{b}_3 direction. The equations governing the acceleration of the center of mass are

$$m\ddot{\mathbf{r}} = \begin{bmatrix} 0\\0\\-mg \end{bmatrix} + R \begin{bmatrix} 0\\0\\F_1 + F_2 + F_3 + F_4 \end{bmatrix}.$$
 (5)

We will define our first input u_1 to be

$$u_1 = \sum_{i=1}^4 F_i.$$

Euler's Equations of Motion In addition to forces, each rotor produces a moment perpendicular to the plane of rotation of the blade, M_i . Rotors 1 and 3 rotate in the $-\mathbf{b_3}$ direction while 2 and 4 rotate in the $+\mathbf{b_3}$ direction. Since the moment produced on the quadrotor is opposite to the direction of rotation of the blades, M_1 and M_3 act in the $\mathbf{b_3}$ direction while M_2 and M_4 act in the $-\mathbf{b_3}$ direction. L is the distance from the axis of rotation of the rotors to the center of mass of the quadrotor.

The angular acceleration determined by the Euler equations is

$$I\begin{bmatrix} \dot{p} \\ \dot{q} \\ \dot{r} \end{bmatrix} = \begin{bmatrix} L(F_2 - F_4) \\ L(F_3 - F_1) \\ M_1 - M_2 + M_3 - M_4 \end{bmatrix} - \begin{bmatrix} p \\ q \\ r \end{bmatrix} \times I \begin{bmatrix} p \\ q \\ r \end{bmatrix}. \tag{6}$$

We can rewrite this as:

$$I\begin{bmatrix} \dot{p} \\ \dot{q} \\ \dot{r} \end{bmatrix} = \begin{bmatrix} 0 & L & 0 & -L \\ -L & 0 & L & 0 \\ \gamma & -\gamma & \gamma & -\gamma \end{bmatrix} \begin{bmatrix} F_1 \\ F_2 \\ F_3 \\ F_4 \end{bmatrix} - \begin{bmatrix} p \\ q \\ r \end{bmatrix} \times I \begin{bmatrix} p \\ q \\ r \end{bmatrix}. \tag{7}$$

where $\gamma = \frac{k_M}{k_F}$ is the relationship between lift and drag given by Equations (1-2). Accordingly, we will define our second set of inputs to be the vector of moments \mathbf{u}_2 given by:

$$\mathbf{u}_2 = \begin{bmatrix} 0 & L & 0 & -L \\ -L & 0 & L & 0 \\ \gamma & -\gamma & \gamma & -\gamma \end{bmatrix} \begin{bmatrix} F_1 \\ F_2 \\ F_3 \\ F_4 \end{bmatrix}.$$

4 Robot Controllers

4.1 The Nominal State

Our controllers are derived by linearizing the equations of motion and motor models (5-2) at an operating point that corresponds to the nominal hover state, $\mathbf{r} = \mathbf{r}_0$, $\theta = \phi = 0$, $\psi = \psi_0$, $\dot{\mathbf{r}} = 0$, and $\dot{\phi} = \dot{\theta} = \dot{\psi} = 0$, where the roll and pitch angles are small $(c\phi \approx 1, c\theta \approx 1, s\phi \approx \phi, \text{ and } s\theta \approx \theta)$. At this state the lift from the propellers is given by:

$$F_{i,0} = \frac{mg}{4},$$

The nominal values for the inputs at hover are $u_{1,0} = mg$, $\mathbf{u}_{2,0} = \mathbf{0}$.

Linearizing (5), we get:

$$\ddot{r}_{1} = g(\Delta\theta \cos\psi_{0} + \Delta\phi \sin\psi_{0})$$

$$\ddot{r}_{2} = g(\Delta\theta \sin\psi_{0} - \Delta\phi \cos\psi_{0})$$

$$\ddot{r}_{3} = \frac{1}{m}u_{1} - g$$
(8)

Linearizing (6), we get:

$$\begin{bmatrix} \dot{p} \\ \dot{q} \\ \dot{r} \end{bmatrix} = I^{-1} \begin{bmatrix} 0 & L & 0 & -L \\ -L & 0 & L & 0 \\ \gamma & -\gamma & \gamma & -\gamma \end{bmatrix} \begin{bmatrix} F_1 \\ F_2 \\ F_3 \\ F_4 \end{bmatrix}$$
(9)

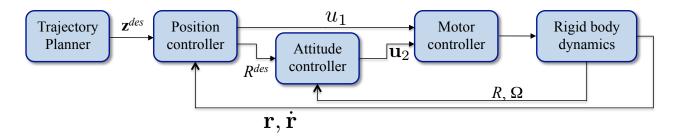


Figure 2: The position and attitude control loops.

If we assume the rotor craft is symmetric so $I_{xx} = I_{yy}$, we get:

$$\begin{split} \dot{p} &= \frac{u_{2,x}}{I_{xx}} = \frac{L}{I_{xx}} (F_2 - F_4) \\ \dot{q} &= \frac{u_{2,y}}{I_{yy}} = \frac{L}{I_{yy}} (F_3 - F_1), \\ \dot{r} &= \frac{u_{2,z}}{I_{zz}} = \frac{\gamma}{I_{zz}} (F_1 - F_2 + F_3 - F_4). \end{split}$$

In other words, the equations of motion are decoupled in terms of angular accelerations. Each component of angular acceleration depends only on the appropriate component of \mathbf{u}_2 .

4.2 Position and Attitude Control

The control problem is to determine the four inputs, $\{u_1, \mathbf{u}_2\}$ required to hover or to follow a desired trajectory, \mathbf{z}^{des} . As shown in Figure 2, we will use errors in the robot's position to drive a position controller from (8) which directly determines u_1 . The model in (8) also allows us to derive a desired orientation. The attitude controller for this orientation is derived from the model in (9).

The transformation of the inputs into $\{u_1, \mathbf{u}_2\}$ is straightforward and is described in [10]. The inner attitude control loop uses onboard accelerometers and gyros to control the roll, pitch, and yaw and runs at approximately 500 Hz, while the outer position control loop uses estimates of position and velocity of the center of mass to control the trajectory in three dimensions at 100-200 Hz.

Attitude Control We now present a proportional plus derivative (PD) attitude controller to track a trajectory in SO(3) specified in terms of a desired roll, pitch and yaw angle. Since our development of the controller will be based on linearized equations of motion, the attitude must be close to the nominal hover state where the roll and pitch angles are small.

Near the nominal hover state the proportional plus derivative control laws take the form:

$$\mathbf{u}_{2} = I \begin{bmatrix} k_{p,\phi}(\phi^{\text{des}} - \phi) + k_{d,\phi}(p^{\text{des}} - p) \\ k_{p,\theta}(\theta^{\text{des}} - \theta) + k_{d,\theta}(q^{\text{des}} - q) \\ k_{p,\psi}(\psi^{\text{des}} - \psi) + k_{d,\psi}(r^{\text{des}} - r) \end{bmatrix}$$
(10)

Position Control In the next subsection, we will present two position control methods that use the roll and pitch angles as inputs to drive the position of the quadrotor. In both methods, the position control algorithm will determine the desired roll and pitch angles, θ^{des} and ϕ^{des} , which can be used to compute the desired speeds from (10). The

first, a hover controller, is used for station-keeping or maintaining the position at a desired position vector, \mathbf{r}_0 . The second tracks a specified trajectory, $\mathbf{r}_T(t)$, in three dimensions. In both cases, the desired yaw angle, is specified independently. It can either be a constant, ψ_0 or a time varying quantity, $\psi_T(t)$. We will assume that the desired trajectory:

$$\mathbf{z}^{\mathrm des} = \begin{bmatrix} \mathbf{r}_T(t) \\ \psi_T(t) \end{bmatrix},$$

will be provided by a trajectory planner as an input to specify the trajectory of the position vector and the yaw angle we are trying to track.

4.3 Hover Controller

For hovering, $\mathbf{r}_T(t) = \mathbf{r}_0$ and $\psi_T(t) = \psi_0$. The command accelerations, \ddot{r}_i^{des} , are calculated from a proportional plus derivative (PD) controller. Define the position error in terms of components using the standard reference triad \mathbf{a}_i by:

$$e_i = (r_{i,T} - r_i).$$

In order to guarantee that this error goes exponentially to zero, we require

$$(\ddot{r}_{i,T} - \ddot{r}_i^{\text{des}}) + k_{d,i}(\dot{r}_{i,T} - \dot{r}_i) + k_{p,i}(r_{i,T} - r_i) = 0, \tag{11}$$

where $\dot{r}_{i,T} = \ddot{r}_{i,T} = 0$ for hover.

From (8) we can obtain the relationship between the desired accelerations and roll and pitch angles. Given that $\Delta\theta = \theta - \theta_0 = \theta$ and $\Delta\phi = \phi - \phi_0 = \phi$, we can write

$$\ddot{r}_1^{\text{des}} = g(\theta^{\text{des}} \cos \psi_T + \phi^{\text{des}} \sin \psi_T)$$
$$\ddot{r}_2^{\text{des}} = g(\theta^{\text{des}} \sin \psi_T - \phi^{\text{des}} \cos \psi_T)$$
$$\ddot{r}_3^{\text{des}} = \frac{1}{m} u_1 - g.$$

For hover, the last equation yields:

$$u_1 = mg + m\ddot{r}_3^{\text{des}} = mg - m\left(k_{d,3}\dot{r}_3 + k_{p,3}(r_3 - r_{3,0})\right) \tag{13}$$

The other two equations can be used to compute the desired roll and pitch angles for the attitude controller:

$$\phi^{\text{des}} = \frac{1}{g} (\ddot{r}_1^{\text{des}} \sin \psi_T - \ddot{r}_2^{\text{des}} \cos \psi_T)$$
 (14a)

$$\theta^{\text{des}} = \frac{1}{g} (\ddot{r}_1^{\text{des}} \cos \psi_T + \ddot{r}_2^{\text{des}} \sin \psi_T)$$
 (14b)

The desired roll and pitch velocities are taken to be zero.

$$p^{\text{des}} = 0 \tag{15a}$$

$$q^{\text{des}} = 0 \tag{15b}$$

Since the yaw, $\psi_T(t)$ is prescribed independently by the trajectory generator, we get:

$$\psi^{\text{des}} = \psi_T(t) \tag{16a}$$

$$r^{\rm des} = \dot{\psi}_T(t) \tag{16b}$$

These equations provide the setpoints for the attitude controller in (10).

Note that the attitude controller must run an order of magnitude faster than the position control loop in order for this to work. In practice as discussed in [10], the position controller runs at $100\,\mathrm{Hz}$, while the inner attitude control loop runs at $500\,\mathrm{Hz}$.

4.4 3D Trajectory Control

The 3-D Trajectory Controller is used to follow three-dimensional trajectories with modest accelerations so the near-hover assumptions hold. To derive this controller we follow the same steps as in (11) except that $\dot{r}_{i,T}$ and $\ddot{r}_{i,T}$ are no longer zero but are obtained from the specification of the trajectory. If the near-hover assumption holds and the dynamics are linear with no saturation on the inputs, a controller that generates the desired acceleration \ddot{r}_i^{des} according to (11) is guaranteed to drive the error exponentially to zero. However, it is possible that the commanded trajectory has twists and turns that are too hard to follow perfectly because of errors in the model or limitations on the input thrusts. We suggest a modification that allows the robot to follow the path even if it may not be able to follow the time-parameterized trajectory.

Let \mathbf{r}_T denote the closest point on the desired trajectory to the the current position \mathbf{r} , and let the desired velocity and acceleration obtained by differentiating the specified trajectory be given by $\dot{\mathbf{r}}_T$ and $\ddot{\mathbf{r}}_T$ respectively. Let the unit tangent vector of the trajectory (unit vector along $\dot{\mathbf{r}}_T$) be $\dot{\mathbf{t}}$. The unit normal to the trajectory, $\hat{\mathbf{n}}$, is derived by differentiating the tangent vector with respect to time or arc length, and finally the unit binormal vector is the cross product $\hat{\mathbf{b}} = \hat{\mathbf{t}} \times \hat{\mathbf{n}}$. We define the position and velocity errors according to the following equations:

$$\mathbf{e}_p = ((\mathbf{r}_T - \mathbf{r}) \cdot \hat{\mathbf{n}})\hat{\mathbf{n}} + ((\mathbf{r}_T - \mathbf{r}) \cdot \hat{\mathbf{b}})\hat{\mathbf{b}}$$

and

$$\mathbf{e}_v = \dot{\mathbf{r}}_T - \dot{\mathbf{r}}.$$

Note that here we ignore position error in the tangential direction and only considering position error in the plane that is normal to the curve at the closest point. Once again we calculate the commanded acceleration, \ddot{r}_i^{des} , from the PD feedback as shown in (11). In vector notation, this is:

$$(\ddot{\mathbf{r}}_T - \ddot{\mathbf{r}}^{\text{des}}) + \mathbf{k}_d \mathbf{e}_v + \mathbf{k}_p \mathbf{e}_p = 0, \tag{17}$$

Finally we use (14 - 16) to compute the desired roll and pitch angles. Again we refer the readers to [10] for more details including experimental results obtained from these controllers.

5 Controller Design

The controller design in the previous sections was based on the decoupling of the position and attitude control subsystems as motivated in Fig. 2. This is only valid when the attitude control loop is much faster than the position control loop permitting us to use, what is called in the control theory literature (see [12]), a *backstepping* approach to controller design. This is roughly the approach that justifies the development in Sections 4.3-4.4.

It is also possible to consider the coupled system with positions and orientations concurrently. To see this, we consider a planar model of the quadrotor which can be obtained by restricting the motion of the quadrotor to the y-z plane with zero roll (θ) and yaw (ψ) . In such a case, (5-6) reduce to:

$$m \begin{bmatrix} \ddot{y} \\ \ddot{z} \\ \ddot{\phi} \end{bmatrix} = \begin{bmatrix} 0 \\ -mg \\ 0 \end{bmatrix} + \begin{bmatrix} -\frac{1}{m}\sin\phi & 0 \\ \frac{1}{m}\cos\phi & 0 \\ 0 & \frac{1}{I_{xx}} \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix}$$
 (18)

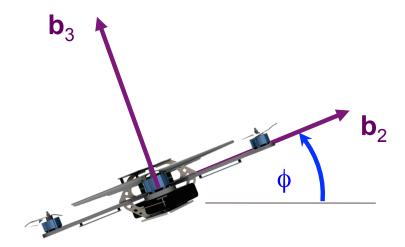


Figure 3: A planar model with zero roll and yaw in the y-z plane. .

We can write this in state space notation in the usual form:

$$\dot{\mathbf{x}} = \mathbf{f}(\mathbf{x}, \mathbf{u}) \tag{19}$$

where

$$\mathbf{x} = \left[y, z, \phi, \dot{y}, \dot{z}, \dot{\phi} \right]^T, \ \mathbf{u} = \left[u_1, u_2 \right]^T.$$

5.1 Control of the linearized system

As before we linearize the system about the hover state. The hover state, $(\mathbf{x}_0, \mathbf{u}_0)$, can be written as (why?):

$$\mathbf{x}_0 = egin{bmatrix} y_0 \ z_0 \ 0 \ 0 \ 0 \end{bmatrix}, \ \mathbf{u}_0 = egin{bmatrix} mg \ 0 \end{bmatrix}.$$

We now linearize the system about this point to get:

$$\begin{split} \dot{\mathbf{x}} &= & \frac{\partial}{\partial \mathbf{x}} \mathbf{f}(\mathbf{x}, \mathbf{u})|_{(\mathbf{x}_0, \mathbf{u}_0)} (\mathbf{x} - \mathbf{x}_0) + \frac{\partial}{\partial \mathbf{u}} \mathbf{f}(\mathbf{x}, \mathbf{u})|_{(\mathbf{x}_0, \mathbf{u}_0)} (\mathbf{u} - \mathbf{u}_0) \\ &= & \mathbf{A}(\mathbf{x} - \mathbf{x}_0) + \mathbf{B}(\mathbf{u} - \mathbf{u}_0) \end{split}$$

where

$$A = \dots, B = \dots$$

The controller takes the form of a linear feedback law:

$$(\mathbf{u} - \mathbf{u}_0) = \mathbf{K}(\mathbf{x} - \mathbf{x}_0) \tag{20}$$

where the 2×6 gain matrix, **K**, must be synthesized to optimize the performance of the system. For a linear system, this is generally posed as a *linear-quadratic feedback regulator* (LQR) design problem where **K** is chosen to minimize the cost functional:

$$\int_0^\infty (\mathbf{x} - \mathbf{x}_0)^T \mathbf{Q} (\mathbf{x} - \mathbf{x}_0) + (\mathbf{u} - \mathbf{u}_0)^T \mathbf{R} (\mathbf{u} - \mathbf{u}_0) dt$$
 (21)

which weights excursions from the error position and magnitudes of the input effort over the trajectory via suitably chosen matrices \mathbf{Q} and \mathbf{R} . See matlab function lqr for the continuous LQR design problem and its close relative dlqr for the discrete controller design problem.

5.2 Control of the nonlinear system

In Equation (18), the nonlinearity arises from the sine and cosine functions of the pitch ϕ . Returning to the discussion in Section 4.3, is it possible to design a *nonlinear* controller that can guarantee performance over a wider range of conditions including large excursions from hover? The answer is yes and the approach presented here to address this is based closely on [13, 14].

We start with (11) and write:

$$\begin{bmatrix} \ddot{y}^{\text{des}} \\ \ddot{z}^{\text{des}} \end{bmatrix} = \begin{bmatrix} \ddot{y}_T \\ \ddot{z}_T \end{bmatrix} + \mathbf{k}_d \begin{bmatrix} \dot{y}_T - \dot{y} \\ \dot{z}_T - \dot{x} \end{bmatrix} + \mathbf{k}_p \begin{bmatrix} y_T - y \\ z_T - x \end{bmatrix}. \tag{22}$$

Because it is impossible to directly produce accelerations along the b_2 direction, we do the next best thing by projecting the desired acceleration vector, \mathbf{k} , along the b_3 direction and compute the inputs required to produce this acceleration:

$$u_{1} = mg + \left[-\sin\phi \cos\phi\right] \begin{bmatrix} \begin{bmatrix} \ddot{y}_{T} \\ \ddot{z}_{T} \end{bmatrix} + \mathbf{k}_{d} \begin{bmatrix} \dot{y}_{T} - \dot{y} \\ \dot{z}_{T} - \dot{x} \end{bmatrix} + \mathbf{k}_{p} \begin{bmatrix} y_{T} - y \\ z_{T} - x \end{bmatrix} \end{bmatrix}. \tag{23}$$

We then compute the input that is required to align the b_3 vector along the desired acceleration. This is derived from the error computed as follows:

$$e_{rot} = \frac{\mathbf{b}_3 \times \mathbf{k}}{\|\mathbf{k}\|} \cdot \mathbf{b}_1, \tag{24}$$

where \mathbf{k} is the desired acceleration vector in (22). This is effectively the sine of the angle between the desired acceleration vector and the vector along which the robot can produce an acceleration. Thus the second control input must drive this error to zero. Accordingly we write:

$$u_2 = I_{xx}(-k_r e_{rot} - k_w \dot{e}_{rot}) \tag{25}$$

Equations (23, 25) with any positive definite \mathbf{k}_p and \mathbf{k}_d and positive k_r and k_ω guarantee global convergence. See [13] for a discussion of convergence for this nonlinear controller.

6 LQR Design for the linearized model

LQR is a design technique from optimal control theory [15] to design controllers for linear systems to minimize a quadratic cost function subject to the dynamic constraint $\dot{\mathbf{x}} = \mathbf{A}\mathbf{x} + \mathbf{B}\mathbf{u}$, resulting in a linear state-feedback controller.

Theorem 1 If the linear system $\dot{\mathbf{x}} = \mathbf{A}\mathbf{x} + \mathbf{B}\mathbf{u}$ is controllable and if \mathbf{Q} is a positive semi-definite matrix and \mathbf{R} is positive definite matrix, then there exists a solution to the optimization problem

$$\min_{u(t)} \int_0^T \left[\mathbf{x}^T(t) \mathbf{Q} \mathbf{x}(t) + \mathbf{u}^T(t) \mathbf{R} \mathbf{u}(t) \right] dt, \tag{26}$$

and the solution is given by

- $\mathbf{u}^*(t) = -\mathbf{K}(t)\mathbf{x}(t)$, where
- $\mathbf{K}(t) = \mathbf{R}^{-1}\mathbf{B}^T\mathbf{S}(t)$, and $\mathbf{S}(t)$ is the solution of the Ricatti equation:
- $\bullet \ -\dot{\mathbf{S}}(t) = \mathbf{Q} + \mathbf{S}(t)\mathbf{A} + \mathbf{A}^T\mathbf{S}(t) \mathbf{S}(t)\mathbf{B}\mathbf{R}^{-1}\mathbf{B}^T\mathbf{S}(t).$

The solution of the infinite-horizon $(T \to \infty)$ optimization problem,

$$\min_{u(t)} \int_0^\infty \left[\mathbf{x}^T(t) \mathbf{Q} \mathbf{x}(t) + \mathbf{u}^T(t) \mathbf{R} \mathbf{u}(t) \right] dt, \tag{27}$$

also exists and is given by

- $\mathbf{u}^*(t) = -\mathbf{K}\mathbf{x}(t)$, where
- $\mathbf{K} = \mathbf{R}^{-1}\mathbf{B}^T\mathbf{S}$, and \mathbf{S} is the solution of the Algebraic Ricatti Equation (ARE):
- $\bullet \mathbf{Q} + \mathbf{S}\mathbf{A} + \mathbf{A}^T\mathbf{S} \mathbf{S}\mathbf{B}\mathbf{R}^{-1}\mathbf{B}^T\mathbf{S} = 0.$
- Matlab Command: [K S] = lgr(A, B, Q, R).

Remark 1 There is a trade-off between speed of response and the magnitude of control signals. Penalizing the controls more heavily than the states (R large in comparison to Q), tends to result in slower response, while penalizing the states more heavily than the controls (Q large in comparison to R) tends to result in faster response but at the expense of larger control signals.

Remark 2 A few choices of Q and R for an example system $\dot{x} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} x + \begin{bmatrix} 0 \\ 1 \end{bmatrix} u$ are below:

- $\mathbf{Q} = 0.1 \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$, $\mathbf{R} = 1$. (This would result in slow convergence but with small control signals.)
- $\mathbf{Q} = 100 \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$, $\mathbf{R} = 1$. (This results in fast convergence but at the expense large control signals)
- $\mathbf{Q} = \begin{bmatrix} \frac{1}{1^2} & 0 \\ 0 & \frac{1}{0.2^2} \end{bmatrix}$, $\mathbf{R} = \frac{1}{1^2}$. (This requests $\max |x_1| \le 1$, $\max |x_2| \le 0.2$ and $\max |u| \le 1$, which is respected by the lqr controller)

Remark 3 To stabilize to a final goal state, (27) is used, while (26) is used for stabilizing to a time-dependent state along the desired trajectory rather than to a single goal state.

7 Project Work

7.1 Tasks

You will simulate the quadrotor dynamics and control using the matlab simulator posted on the meam620 wiki. The simulator relies on the numerical solver ode45, details about this solver can be easily found online. You don't need to be an expert in numerical methods, but it would be beneficial if you know the basics of how ode solvers work. Your tasks include:

1. Trajectory Generator

You will first implement two trajectory generators that generate a circle and a diamond trajectory. In order to specify the trajectory in the simulator, you will need to change the variable trajhandle. trajhandle is the name of a function that takes in current time t and current quadrotor number qn and returns a struct of desired state as a function of time. The struct desired_state has 5 fields, which are pos, vel, acc, yaw and yawdot. For a proper trajectory, you need to specify pos, vel and acc, whereas yaw and yawdot can be leave as 0.

- (a) circle.m
 - A helix in the xy plane of radius 5 m centered about the point (0,0,0) starting at the point (5,0,0). The z coordinate should start at 0 and end at 2.5. The helix should go counter-clockwise.
- (b) diamond.m A "diamond helix" with corners at (0,0,0), $(0,\sqrt{2},\sqrt{2})$, $(0,0,2\sqrt{2})$, and $(0,-\sqrt{2},\sqrt{2})$ when projected into the yz plane, and an x coordinate starting at 0 and ending at 1. The quadrotor should start at (0,0,0) and end at (1,0,0).

2. Controller

You will then implement a controller in file controller.m that makes sure that your quadrotor follows the desired trajectory. This controller will be used again in the following phases. The controller takes in a cell struct qd, current time t, current quadrotor number qn and quadrotor parameters params and outputs thrust F, moments M, desired thrust, roll, pitch and yaw trpy and derivatives of desired roll, pitch and yaw drpy.

In this phase, since we are doing simulation, you will only need to output thrust F and moments M and leave trpy and drpy untouched. qd is a cell array contains structs for each quadrotor. This means $qd\{qn\}$ contains all the information (position, velocity, euler angles and etc.) about quadrotor number qn. Those information can be accessed via $qd\{qn\}$. pos for example. The reason for using qd here is simply to be compatible with KMel's code that you will use in the final phase of this project to fly the quadrotor.

3. Simulation

Lastly, you need to fill in the appropriate variables for trajhandle and controlhandle in the script runsim.m for simulating your result.

7.2 Simulator

The quadrotor simulator comes with the student code. Before implementing your own functions, you should first try running runsim.m in your matlab. If you see a quadrotor falling from position (0,0,0) then the simulator works on your computer and you may continue with other tasks. This is because the outputs of controller.m are all

zeros, thus no thrust generated.

When you have the basic version of your trajectory generator and controller done, you will be able to see the quadrotor flying in the space with proper roll, pitch, and yaw,leaving trails of desired and actual position behind. The desired position is color-coded blue and the actual position is red. After the simulation finishes, two plots of position and velocity will be generated to give you an overview of how well your trajectory generator and controller are doing. Note that you will not be graded on these plots but by the automatic grader.

7.3 Submission

When you are finished you may submit your code via turnin. The project name for this assignment is titled "projlphasel" so the command to submit should be

```
\# turnin -c meam620 -p proj1phase1 -v *
```

Your turnin submission should contain:

- 1. A README file detailing anything we should be aware of.
- 2. Files crazyflie.m, controller.m, diamond.m, circle.m, as well as any other Matlab files you need to run your code. (That being said, we recommend you to only submit those 4 files)

Shortly after submitting you should receive an e-mail from meam620@seas.upenn.edu stating that your submission has been received. You can check on the status of your submission at https://alliance.seas.upenn.edu/~meam620/monitor/. Once the automated tests finish running, you should receive another e-mail containing your test results. This report will only tell you whether you passed or failed tests; it will not tell you what the test inputs were or why your code failed. Your code will be given at most 5 minutes to complete all the tests. There is no limit on the number of times you can submit your code.

Please do not attempt to determine what the automated tests are or otherwise try to game the automated test system. Any attempt to do so will be considered cheating, resulting in a 0 on this assignment and possible disciplinary action.

You will be graded on successful completion of the code and how quickly and accurately your quadrotor follows the circle and diamond paths and two other trajectories which will not be released.

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