Trajectory Generation and Control for Quadrotors

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Introduction

- Quadrotors are popular research because of their agility and mobility in 3D space
- For a quadrotor maneuvering in tightly constraint environment, it is necessary to develop optimal trajectories to ensure safe passage through corridors and constraints [1].
- Develop flight plans that leverage the dynamics of system [1].



Objective

- Review of control and trajectory generation for quadrotors [1].
- Understand and implement differential flatness for trajectory generation
- Examples code given at [2, 3].
- Implement minimum snap trajectory and control for quadrotors based on flatness [1, 3].

^[1] Mellinger, Daniel, and Vijay Kumar. "Minimum snap trajectory generation and control for quadrotors." Robotics and Automation (ICRA), 2011 IEEE International Conference on. IEEE, 2011.

^[2] https://github.com/WenjinTao/Robotics--Coursera/tree/master/Robotics-Aerial Robotics

^[3] https://www.coursera.org/learn/robotics-flight

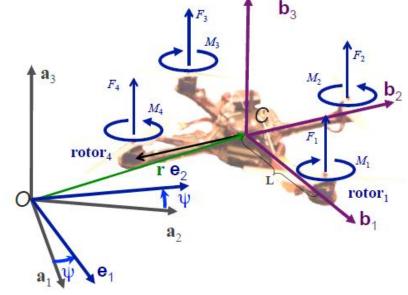
Dynamic Model

$$q = \begin{bmatrix} x \\ y \\ z \end{bmatrix}$$

$$q = \begin{bmatrix} \varphi \\ \theta \\ \psi \end{bmatrix}$$
Roll
Pitch
Yaw

Angular velocity in body frame
$$\begin{bmatrix} p \\ q \\ r \end{bmatrix} = \begin{bmatrix} c\theta & 0 & -c\phi s\theta \\ 0 & 1 & s\phi \\ s\theta & 0 & c\phi c\theta \end{bmatrix} \begin{bmatrix} \dot{\phi} \\ \dot{\theta} \\ \dot{\psi} \end{bmatrix}$$

$$egin{bmatrix} \dot{\phi} \ \dot{ heta} \ \dot{\psi} \end{bmatrix}$$



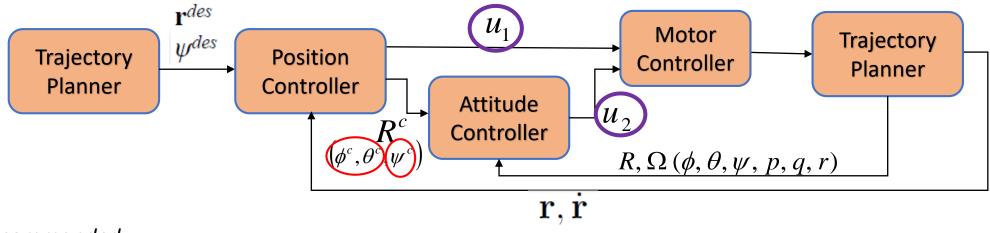
$$m\ddot{r} = \begin{bmatrix} 0 \\ 0 \\ -mg \end{bmatrix} + R_{BA} \begin{bmatrix} 0 \\ 0 \\ F_1 + F_2 + F_3 + F_4 \end{bmatrix}$$
 Newton's Equation of motion

$$I\begin{bmatrix} \dot{q} \\ \dot{q} \\ \dot{r} \end{bmatrix} = \begin{bmatrix} L(F_2 - F_4) \\ L(F_3 - F_1) \\ M_1 - M_2 + M_3 - M_4 \end{bmatrix} - \begin{bmatrix} p \\ q \\ r \end{bmatrix} \times I\begin{bmatrix} p \\ q \\ r \end{bmatrix} = \begin{bmatrix} c \psi c \theta - s \phi s \psi s \theta & -c \phi s \psi & c \psi s \theta + c \theta s \phi s \psi \\ -c \theta s \psi + c \psi s \phi s \theta & c \phi c \psi & s \psi s \theta - c \psi c \theta s \phi \\ -c \phi s \theta & s \phi & c \phi c \theta \end{bmatrix}$$
Source: N. Michael, D. Mellinger, Q. Lindsey, and V. Kumar, The GRASP Multiple Micro-UAV Testbed," IEEE Robotics and Automation Magazine, 4

$$-\begin{bmatrix} p \\ q \\ r \end{bmatrix} \times I \begin{bmatrix} p \\ q \\ r \end{bmatrix}$$

$$= \begin{bmatrix} c \psi c \theta - s \phi s \psi s \theta & -c \phi s \psi & c \psi s \theta + c \theta s \phi s \psi \\ c \theta s \psi + c \psi s \phi s \theta & c \phi c \psi & s \psi s \theta - c \psi c \theta s \phi \\ -c \phi s \theta & s \phi & c \phi c \theta \end{bmatrix}$$

Control



commanded
$$(\ddot{r}_{i,des} - \ddot{r}_{i,c}) + k_{d,i} (\dot{r}_{i,des} - \dot{r}_{i}) + k_{p,i} (r_{i,des} - \ddot{r}_{i}) = 0$$

$$\uparrow \qquad \uparrow \qquad \qquad \uparrow \qquad \qquad \downarrow$$
 specified

$$u_1 = m(g + \ddot{r}_{3,c})$$

$$\phi_c = \frac{1}{g} \left(\ddot{r}_{1,c} \sin \psi_{des} - \ddot{r}_{2,c} \cos \psi_{des} \right)$$

$$\theta_c = \frac{1}{g} \left(\ddot{r}_{1,c} \cos \psi_{des} - \ddot{r}_{2,c} \sin \psi_{des} \right)$$

$$u_2 = \begin{bmatrix} k_{p,\phi}(\phi_c - \phi) + k_{d,\phi}(p_c - p) \\ k_{p,\theta}(\theta_c - \theta) + k_{d,\theta}(q_c - q) \\ k_{p,\psi}(\psi_c - \psi) + k_{d,\psi}(r_c - r) \end{bmatrix}$$

Differential Flatness

- **Definition:** A non linear system is *differentially flat* if we can find a set of outputs such that we can express all states and inputs in terms of those outputs and their derivatives [1].
- Example: Kinematic car

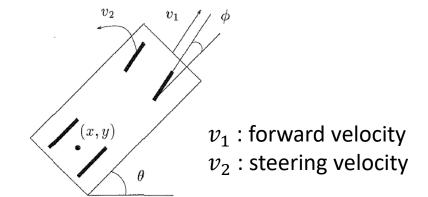
Equations of Motion

$$\dot{x} = \cos \theta \cos \phi \ v_1$$

$$\dot{y} = \sin \theta \cos \phi \ v_2$$

$$\dot{\theta} = \frac{1}{l} \sin \phi \ v_1$$

$$\dot{\phi} = v_2$$



• If output is (x, y), ϕ can be expressed in terms of flat outputs as

$$\phi = \arctan(\frac{1}{l}(\dot{x}^2 + \dot{y}^2)^{\frac{3}{2}}, \dot{x}\ddot{x} - \dot{y}\ddot{y})$$

3D Trajectory Generation

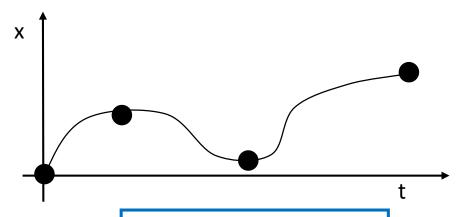
• Optimal Trajectory:

$$x^*(t) = \underset{x(t)}{\operatorname{arg\,min}} \int_{0}^{T} L(\dot{x}, x, t) dt$$
 L: functional

• Smooth Trajectories:

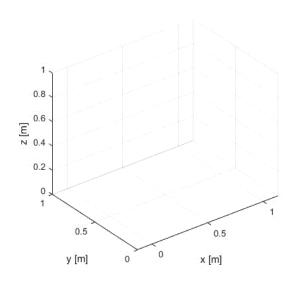
$$x^*(t) = \operatorname*{arg\,min}_{x(t)} \int_0^T (x^{(n)})^2 dt$$

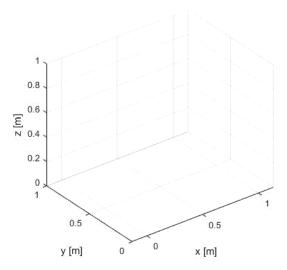
- n = 1, shortest distance (velocity)
- \rightarrow n = 2, minimum acceleration
- \rightarrow n = 3, minimum jerk
- \rightarrow n = 4, minimum snap
- Constraints:
 - ➤ Must pass through way points:
 - > Velocity, Acceleration, Jerk are zero at the end points
 - ➤ Velocity, acceleration, 3rd to 6th derivatives are continuous
 - ightharpoonup Convert the equation into matrix for $A\alpha=b$ and solve for a

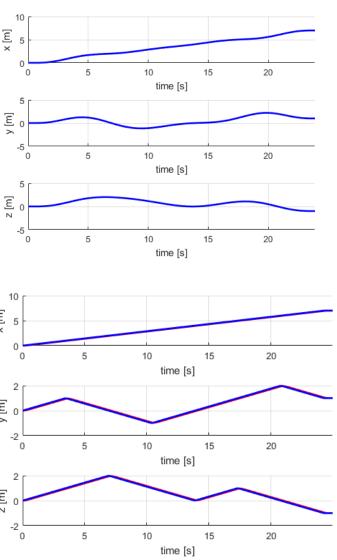


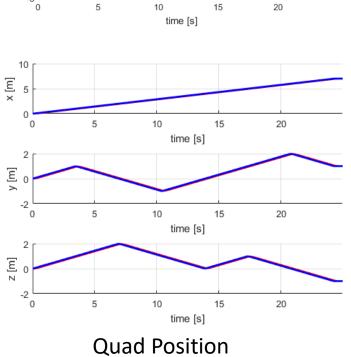
$$x(t) = C_0 + C_1 t + C_2 t^2 + C t^3 + C_4 t^4 + C_5 t^5 + C_6 t^5 + C_7 t^7$$

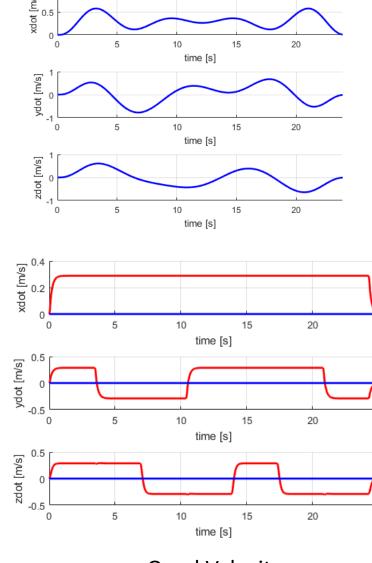
Simulation Results











Conclusion & Future Work

- Implemented minimum snap trajectory and control for quadrotors
- The trajectories are optimal
- Can navigate based on ways points with smooth trajectories
- Can be extended to motion planning for precise and aggressive maneuvers
- Implement in experimental platform
- If possible extend to multi-agents

Questions?