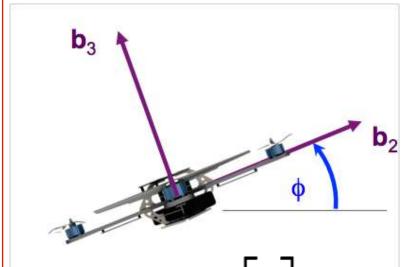
Planar Quadrotor



Planar Quadrotor Model



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

$$x = egin{bmatrix} x_1 \ x_2 \end{bmatrix} = egin{bmatrix} z \ \phi \ \dot{y} \ \dot{z} \ \dot{\phi} \end{bmatrix}$$

Linearized Dynamic Model

Equations of motion

$$\ddot{y} = -\frac{u_1}{m}\sin(\phi)$$

$$\ddot{z} = -g + \frac{u_1}{m}\cos(\phi)$$

$$\ddot{\phi} = \frac{u_2}{I_{xx}}$$
Dynamics are nonlinear

Equilibrium hover configuration

$$y_0, z_0, \phi_0 = 0, u_{1,0} = mg, u_{2,0} = 0,$$

Linearized dynamics

$$\ddot{y} = -g\phi$$

$$\ddot{z} = -g + \frac{u_1}{m}$$

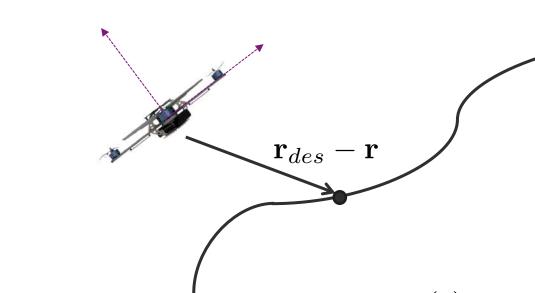
$$\ddot{\phi} = \frac{u_2}{I_{xx}}$$



Trajectory Tracking

Given $\mathbf{r}_T(t), \mathbf{\dot{r}}_T(t), \mathbf{\ddot{r}}_T(t)$

$$\mathbf{r}_T(t) = egin{bmatrix} y(t) \ z(t) \end{bmatrix}$$



desired trajectory (position, velocity, acceleration)

$$e_p = \mathbf{r}_T(t) - \mathbf{r}$$

$$e_v = \mathbf{\dot{r}}_T(t) - \mathbf{\dot{r}}$$

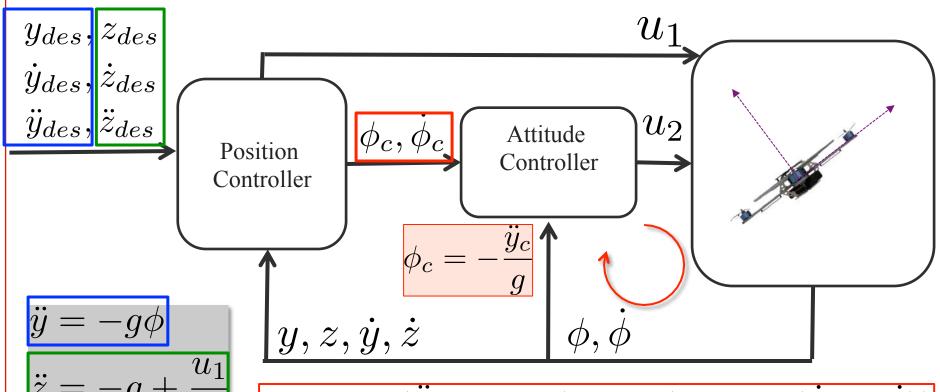
$$(\ddot{\mathbf{r}}_T(t) - \ddot{\mathbf{r}}_c) + k_{d,x}e_v + k_{p,x}e_p = 0$$



Commanded acceleration, calculated by the controller

Nested Control Structure

$$u_1 = m(g + \ddot{z}_{des} + k_{d,z}(\dot{z}_{des} - \dot{z}) + k_{p,z}(z_{des} - z))$$



$$\ddot{\ddot{z}} = -g + \frac{u_1}{m}$$

$$\ddot{\ddot{\phi}} = \frac{u_2}{\tau}$$

$$u_2 = I_{xx}(\ddot{\phi}_c + k_{p,\phi}(\phi_c - \phi) + k_{d,\phi}(\dot{\phi}_c - \dot{\phi}))$$



Control Equations

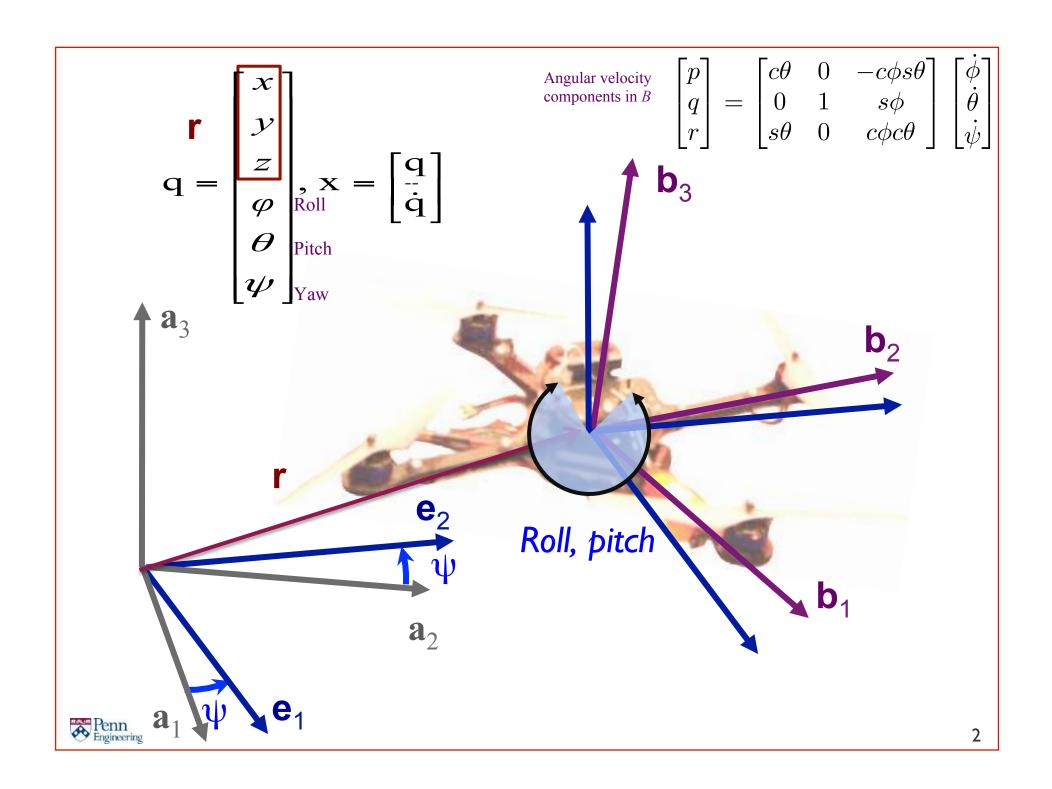
$$u_{1} = m(g + \ddot{z}_{des} + (k_{d,z})(\dot{z}_{des} - \dot{z}) + (k_{p,z})(z_{des} - z))$$

$$u_{2} = (k_{p,\phi})(\phi_{c} - \phi) + (k_{d,\phi})(\dot{\phi}_{c} - \dot{\phi})$$

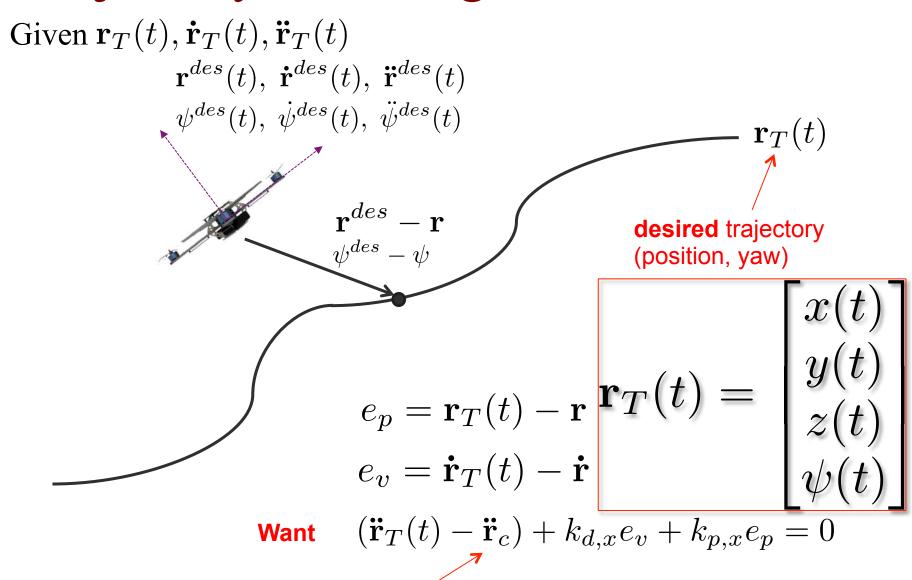
$$\phi_{c} = -\frac{1}{g}(\ddot{y}_{des} + (k_{d,y})(\dot{y}_{des} - \dot{y}) + (k_{p,y})(y_{des} - y))$$

3-D Quadrotor



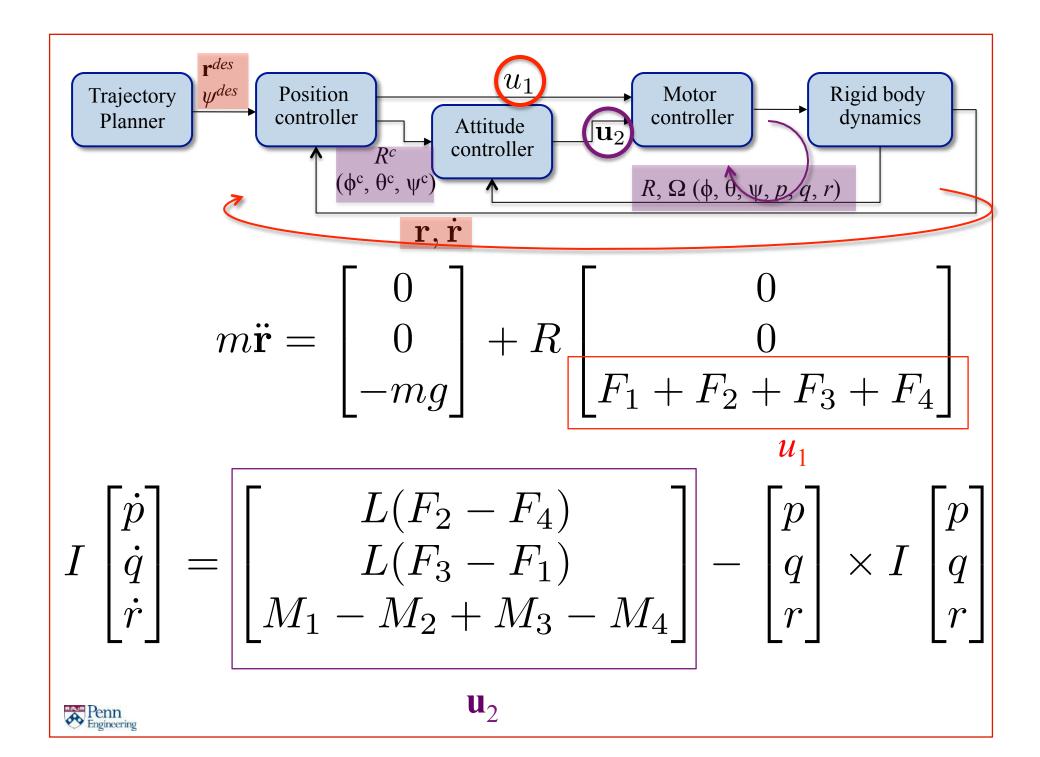


Trajectory Tracking in 3 Dimensions

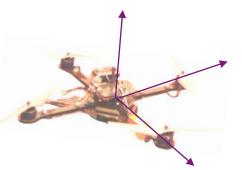




Commanded acceleration, calculated by the controller



Control for Hovering



Linearize the dynamics at the hover configuration

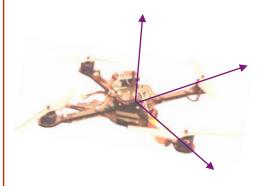
$$(u_1 \sim mg, \theta \sim 0, \phi \sim 0, \psi \sim \psi_0)$$
$$(u_2 \sim 0, p \sim 0, q \sim 0, r \sim 0)$$

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

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



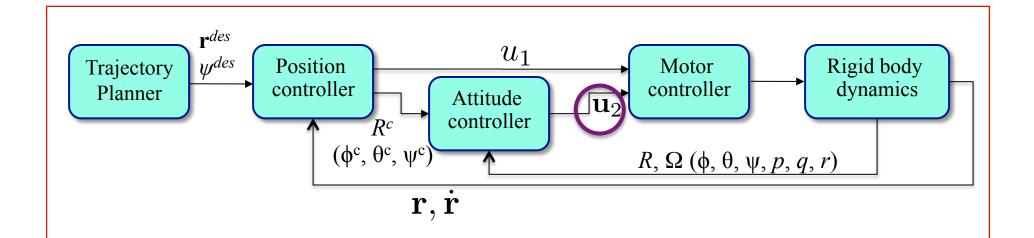
Control for Hovering



$$(u_2 \sim 0, p \sim 0, q \sim 0, r \sim 0)$$

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

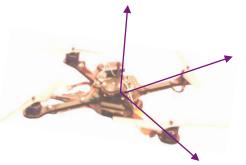




$$\mathbf{u}_{2} = \begin{bmatrix} k_{p,\phi}(\phi_{c} - \phi) + k_{d,\phi}(p_{c} - p) \\ k_{p,\theta}(\theta_{c} - \theta) + k_{d,\theta}(q_{c} - q) \\ k_{p,\psi}(\psi_{c} - \psi) + k_{d,\psi}(r_{c} - r) \end{bmatrix}$$



Control for Hovering

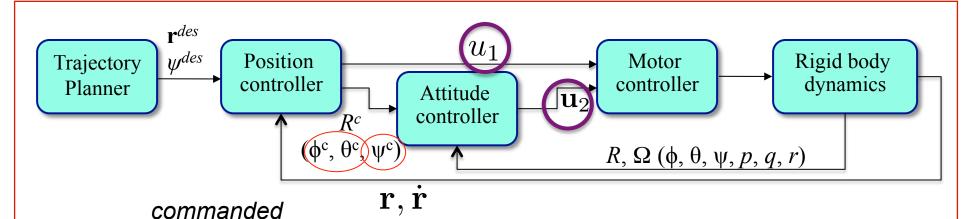


$$m\ddot{\mathbf{r}} = \begin{bmatrix} 0\\0\\-mg \end{bmatrix} + R \begin{bmatrix} 0\\0\\F_1 + F_2 + F_3 + F_4 \end{bmatrix}$$
 ation
$$u_1$$

Linearization

$$(u_1 \sim mg, \theta \sim 0, \phi \sim 0, \psi \sim \psi_0)$$
$$\ddot{r}_1 = \ddot{x} = g(\theta \cos \psi + \phi \sin \psi)$$
$$\ddot{r}_2 = \ddot{y} = g(\theta \sin \psi - \phi \cos \psi)$$





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

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

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

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

$$\psi_c = \psi^{des}$$

specified

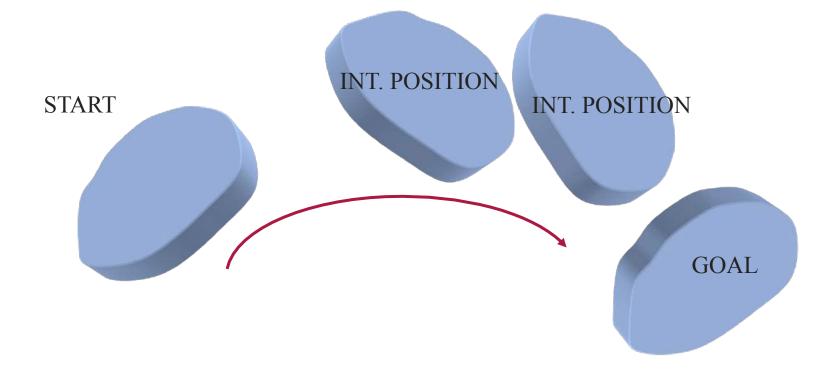
$$\mathbf{u}_{2} = \begin{bmatrix} k_{p,\phi}(\phi_{c} - \phi) + k_{d,\phi}(p_{c} - p) \\ k_{p,\theta}(\theta_{c} - \theta) + k_{d,\theta}(q_{c} - q) \\ k_{p,\psi}(\psi_{c} - \psi) + k_{d,\psi}(r_{c} - r) \end{bmatrix}$$



Time, Motion and Trajectories



Smooth three dimensional trajectories



Applications

- Trajectory generation in robotics
- Planning trajectories for quad rotors



General Set up

- Start, goal positions (orientations)
- Waypoint positions (orientations)
- Smoothness criterion
 Generally translates to minimizing rate of change of "input"
- Order of the system (n)
 Order of the system determines the input
 Boundary conditions on (n-1)th order and lower derivatives



Calculus of Variations

$$x^{\star}(t) = \underset{x(t)}{\operatorname{argmin}} \int_{0}^{T} \mathcal{L}(\dot{x}, x, t) dt$$
function

function

Examples

• Shortest distance path (geometry) $x^*(t) = \arg\min_{x(t)} \int_0^T \dot{x}^2 dt$

$$x^{\star}(t) = \arg\min_{x(t)} \int_0^T \dot{x}^2 dt$$

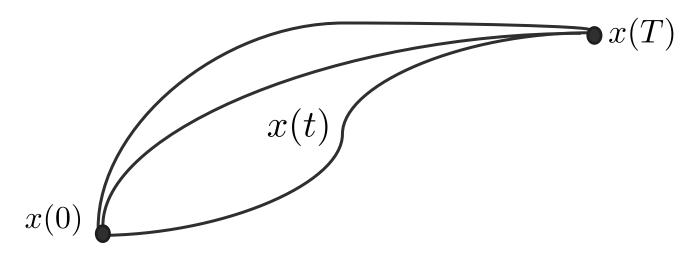
• Fermat's principle (optics)
$$x^{\star}(t) = \operatorname*{argmin}_{x(t)} \int_{0}^{T} 1 dt$$

• Principle of least action (mechanics) $x^*(t) = \underset{x(t)}{\operatorname{argmin}} \int_0^T T(\dot{x}, x, t) - V(x, t) dt$

Calculus of Variations

$$x^{\star}(t) = \underset{x(t)}{\operatorname{argmin}} \int_{0}^{T} \mathcal{L}(\dot{x}, x, t) dt$$

Consider the set of all differentiable curves, x(t), with a given x(0) and x(T).





Calculus of Variations

$$x^{\star}(t) = \underset{x(t)}{\operatorname{argmin}} \int_{0}^{T} \mathcal{L}(\dot{x}, x, t) dt$$

Euler Lagrange Equation

Necessary condition satisfied by the "optimal" function x(t)

$$\frac{d}{dt} \left(\frac{\partial \mathcal{L}}{\partial \dot{x}} \right) - \frac{\partial \mathcal{L}}{\partial x} = 0$$



Smooth trajectories (n=1)

$$x^{*}(t) = \arg\min_{x(t)} \int_{0}^{T} \dot{x}^{2} dt$$

$$x(0) = x_{0}, \ x(T) = x_{T} \quad \underset{u = \dot{x}}{\overset{input}{u = \dot{x}}}$$

Smooth trajectories (n=1)

$$x^{\star}(t) = \arg\min_{x(t)} \int_{0}^{T} \dot{x}^{2} dt$$

Euler Lagrange Equation

$$\frac{d}{dt} \left(\frac{\partial \mathcal{L}}{\partial \dot{x}} \right) - \frac{\partial \mathcal{L}}{\partial x} = 0$$

$$\mathcal{L}(\dot{x}, x, t) = (\dot{x})^2 \quad \Longrightarrow \quad \ddot{x} = 0$$

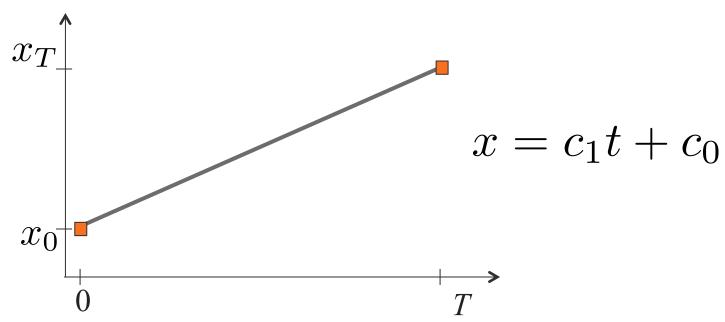
$$x = c_1 t + c_0$$



Smooth trajectories (n=1)

$$x^{\star}(t) = \arg\min_{x(t)} \int_{0}^{T} \dot{x}^{2} dt$$

$$x(0) = x_0, \ x(T) = x_T$$





Smooth trajectories (general *n*)

$$x^{\star}(t) = \underset{x(t)}{\operatorname{argmin}} \int_{0}^{T} \left(x^{(n)}\right)^{2} dt$$

$$\lim_{x \to \infty} u = x^{(n)}$$



Euler-Lagrange Equation

$$x^{\star}(t) = \underset{x(t)}{\operatorname{argmin}} \int_{0}^{T} \mathcal{L}\left(x^{(n)}, x^{(n-1)}, \dots, \dot{x}, x, t\right) dt$$

Euler Lagrange Equation

Necessary condition satisfied by the "optimal" function

$$\frac{\partial \mathcal{L}}{\partial x} - \frac{d}{dt} \left(\frac{\partial \mathcal{L}}{\partial \dot{x}} \right) + \frac{d^2}{dt^2} \left(\frac{\partial \mathcal{L}}{\partial \ddot{x}} \right) + \dots + (-1)^n \frac{d^n}{dt^n} \left(\frac{\partial \mathcal{L}}{\partial x^{(n)}} \right) = 0$$



Smooth Trajectories

$$x^{\star}(t) = \underset{x(t)}{\operatorname{argmin}} \int_{0}^{T} \left(x^{(n)}\right)^{2} dt$$

 \bullet n=1, shortest distance

velocity

- \bullet n=2, minimum acceleration
- \bullet n=3, minimum jerk
- *n*=4, minimum snap

n – order of system n^{th} derivative is input



Smooth Trajectories

$$x^{\star}(t) = \underset{x(t)}{\operatorname{argmin}} \int_{0}^{T} \left(x^{(n)}\right)^{2} dt$$

 \bullet n=1, shortest distance

velocity

- \bullet n=2, minimum acceleration
- \bullet n=3, minimum jerk
- *n*=4, minimum snap

Why is the minimum velocity curve also the shortest distance curve?



Minimum Jerk Trajectory

Design a trajectory x(t) such that x(0) = a, x(T) = b

$$x^{\star}(t) = \underset{x(t)}{\operatorname{argmin}} \int_{0}^{T} \mathcal{L}(\ddot{x}, \ddot{x}, \dot{x}, x, t) dt$$

$$\mathcal{L} = (\ddot{x})^2$$

Euler-Lagrange:

$$\frac{\partial \mathcal{L}}{\partial x} - \frac{d}{dt} \left(\frac{\partial \mathcal{L}}{\partial \dot{x}} \right) + \frac{d^2}{dt^2} \left(\frac{\partial \mathcal{L}}{\partial \ddot{x}} \right) - \frac{d^3}{dt^3} \left(\frac{\partial \mathcal{L}}{\partial x^{(3)}} \right) = 0$$

$$x^{(6)} = 0$$

$$x = c_5 t^5 + c_4 t^4 + c_3 t^3 + c_2 t^2 + c_1 t + c_0$$



Solving for Coefficients

$$x = c_5 t^5 + c_4 t^4 + c_3 t^3 + c_2 t^2 + c_1 t + c_0$$

Boundary conditions:

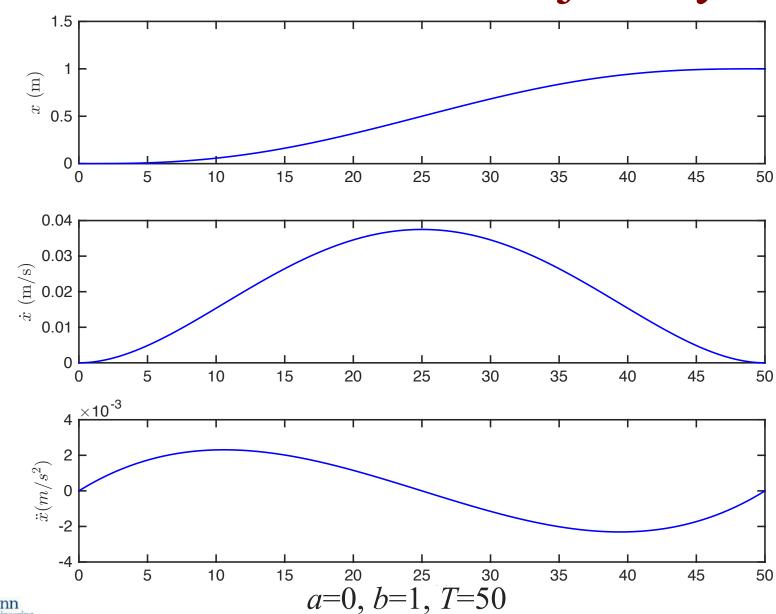
	Position	Velocity	Acceleration
t = 0	а	0	0
t = T	b	0	0

Solve:

$$\begin{bmatrix} a \\ b \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 & 0 & 0 & 0 & 0 & 1 \\ T^5 & T^4 & T^3 & T^2 & T & 1 \\ 0 & 0 & 0 & 0 & 1 & 0 \\ 5T^4 & 4T^3 & 3T^2 & 2T & 1 & 0 \\ 0 & 0 & 0 & 2 & 0 & 0 \\ 20T^3 & 12T^2 & 6T & 2 & 0 & 0 \end{bmatrix} \begin{bmatrix} c_5 \\ c_4 \\ c_3 \\ c_2 \\ c_1 \\ c_0 \end{bmatrix}$$



Minimum Jerk Trajectory





Extensions to multiple dimensions

$$(x^{\star}(t), y^{\star}(t)) = \arg\min_{x(t), y(t)} \int_{0}^{T} \mathcal{L}(\dot{x}, \dot{y}, x, y, t) dt$$

Euler Lagrange Equation

Necessary condition satisfied by the "optimal" function

$$\frac{d}{dt} \left(\frac{\partial \mathcal{L}}{\partial \dot{x}} \right) - \frac{\partial \mathcal{L}}{\partial x} = 0$$

$$\frac{d}{dt} \left(\frac{\partial \mathcal{L}}{\partial \dot{y}} \right) - \frac{\partial \mathcal{L}}{\partial y} = 0$$



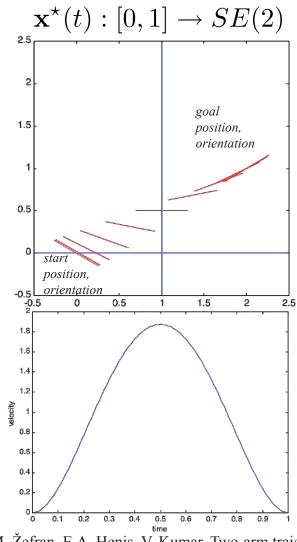
Minimum Jerk for Planar Motions

Minimum-jerk trajectory in (x, y, θ)

$$\min_{x(t),y(t),\theta(t)} \int_0^1 \left(\ddot{x}^2 + \ddot{y}^2 + \ddot{\theta}^2 \right) dt$$

Human two-handed manipulation tasks

- Noise in the neural control signal increases with size of the control signal
- Rate of change of muscle fiber lengths is critical in relaxed, voluntary motions



G.J. Garvin, M. Žefran, E.A. Henis, V. Kumar, Two-arm trajectory planning in a manipulation task, *Biological Cybernetics*, January 1997, Volume 76, Issue 1, pp 53-62

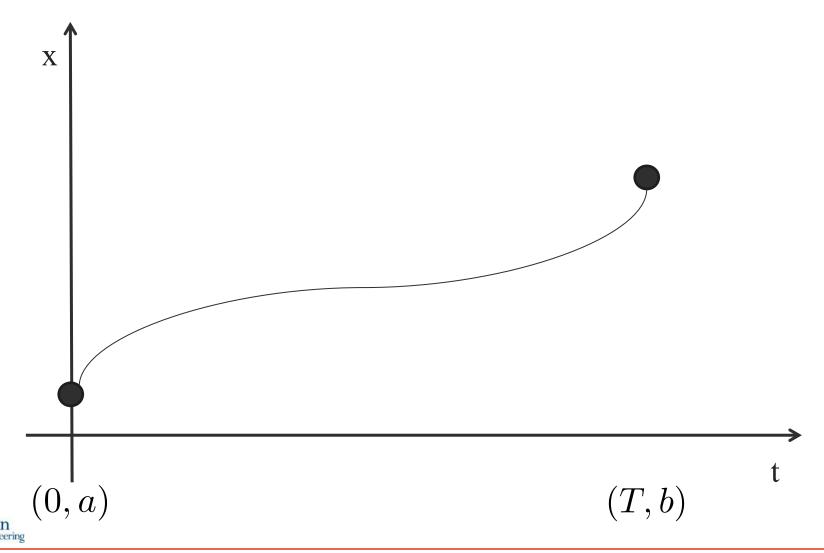


Waypoint Navigation



Smooth 1D Trajectories

Design a trajectory x(t) such that x(0) = a, x(T) = b

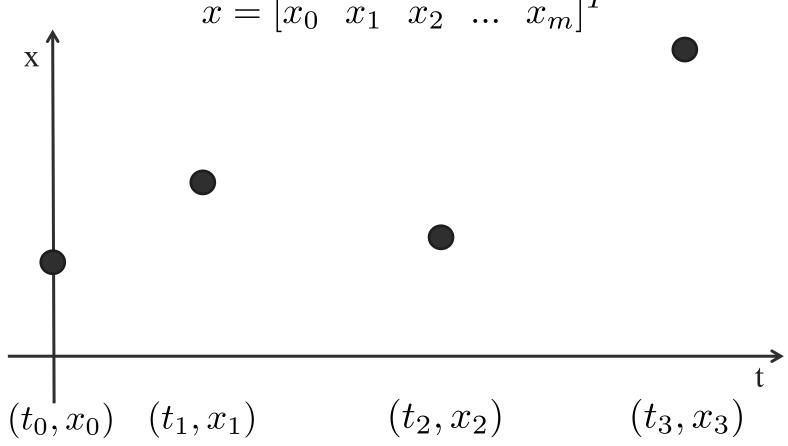


Multi-Segment 1D Trajectories

Design a trajectory x(t) such that:

$$t = [t_0 \ t_1 \ t_2 \ \dots \ t_m]^T$$

 $x = [x_0 \ x_1 \ x_2 \ \dots \ x_m]^T$



Penn Engineering

Multi-Segment 1D Trajectories

Design a trajectory x(t) such that:

$$t = [t_0 \ t_1 \ t_2 \ \dots \ t_m]^T$$

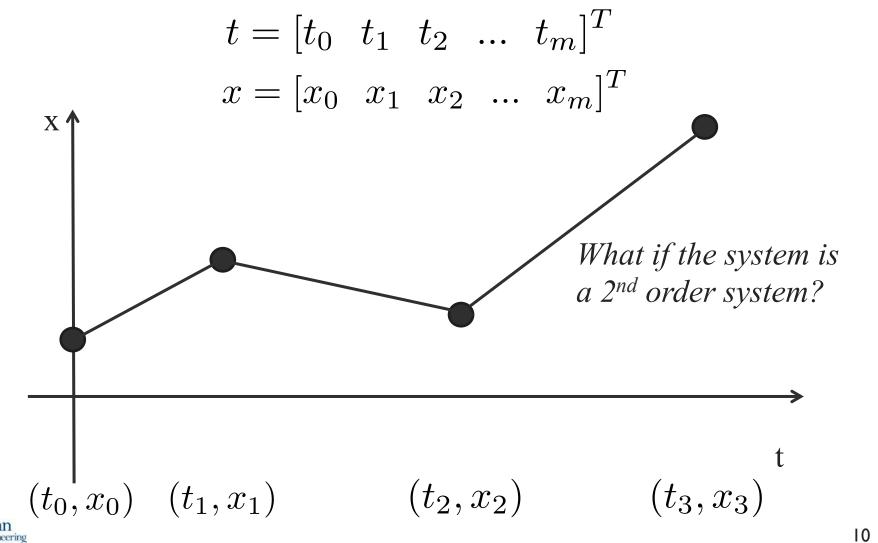
 $x = [x_0 \ x_1 \ x_2 \ \dots \ x_m]^T$

Define piecewise continuous trajectory:

$$x(t) = \begin{cases} x_1(t), & t_0 \le t < t_1 \\ x_2(t), & t_1 \le t < t_2 \\ \dots & \\ x_m(t), & t_{m-1} \le t < t_m \end{cases}$$



Continuous but not Differentiable



Minimum Acceleration Curve for 2nd Order Systems

$$t = \begin{bmatrix} t_0 & t_1 & t_2 & \dots & t_m \end{bmatrix}^T \\ x = \begin{bmatrix} x_0 & x_1 & x_2 & \dots & x_m \end{bmatrix}^T \\ & \underset{x(t)}{\text{min}} \left[\int_{t_0}^{t_1} (\ddot{x}^2) dt + \dots + \int_{t_{m-1}}^{t_m} (\ddot{x}^2) dt \right] \\ \bullet \\ & (t_0, x_0) & (t_1, x_1) & (t_2, x_2) & (t_3, x_3) \end{bmatrix}$$

Design a trajectory x(t) such that:

$$t = [t_0 \ t_1 \ t_2 \ \dots \ t_m]^T$$

 $x = [x_0 \ x_1 \ x_2 \ \dots \ x_m]^T$

$$\min_{x(t)} \left[\int_{t_0}^{t_1} (\ddot{x}^2) dt + \dots + \int_{t_{m-1}}^{t_m} (\ddot{x}^2) dt \right]$$

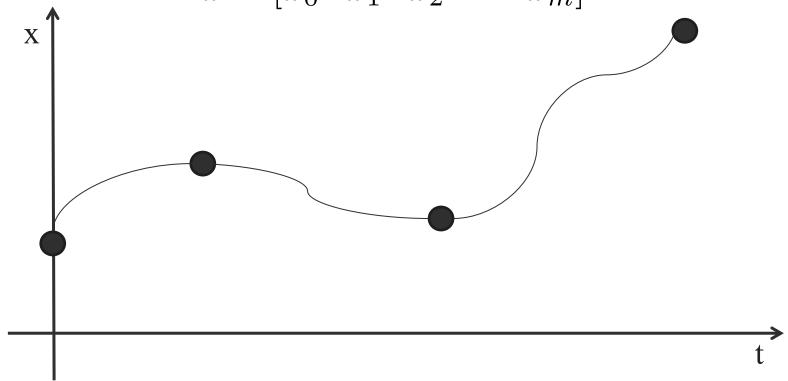
$$x(t) = \begin{cases} x_1(t) = c_{1,3}t^3 + c_{1,2}t^2 + c_{1,1}t + c_{1,0}, & t_0 \le t < t_1 \\ x_2(t) = c_{2,3}t^3 + c_{2,2}t^2 + c_{2,1}t + c_{2,0}, & t_1 \le t < t_2 \\ \dots & \vdots \\ x_m(t) = c_{m,3}t^3 + c_{m,2}t^2 + c_{m,1}t + c_{m,0}, & t_{m-1} \le t < t_m \end{cases}$$

4m degrees of freedom



$$t = [t_0 \ t_1 \ t_2 \ \dots \ t_m]^T$$

 $x = [x_0 \ x_1 \ x_2 \ \dots \ x_m]^T$





Design a trajectory x(t) such that:

$$t = \begin{bmatrix} t_0 & t_1 & t_2 & \dots & t_m \end{bmatrix}^T$$

$$x = \begin{bmatrix} x_0 & x_1 & x_2 & \dots & x_m \end{bmatrix}^T$$

$$x_1(t_1) = x_2(t_1) = x_1$$

$$x_2(t_2) = x_3(t_2) = x_2$$

$$x_1(t_0) = x_0$$

2m

$$t = \begin{bmatrix} t_0 & t_1 & t_2 & \dots & t_m \end{bmatrix}^T$$

$$x = \begin{bmatrix} x_0 & x_1 & x_2 & \dots & x_m \end{bmatrix}^T$$

$$x_1(t_1) = x_2(t_1) = x_1$$

$$\dot{x}_1(t_1) = \dot{x}_2(t_1)$$

$$\ddot{x}_1(t_1) = \ddot{x}_2(t_1)$$

$$\ddot{x}_1(t_1) = \ddot{x}_2(t_1)$$

$$x_2(t_2) = \ddot{x}_3(t_2) = x_2$$

$$\dot{x}_2(t_2) = \dot{x}_3(t_2)$$

$$\ddot{x}_2(t_2) = \ddot{x}_3(t_2)$$

$$t = \begin{bmatrix} t_0 & t_1 & t_2 & \dots & t_m \end{bmatrix}^T$$

$$x = \begin{bmatrix} x_0 & x_1 & x_2 & \dots & x_m \end{bmatrix}^T$$

$$x_1(t_1) = x_2(t_1) = x_1$$

$$\dot{x}_1(t_1) = \dot{x}_2(t_1)$$

$$\ddot{x}_1(t_1) = \ddot{x}_2(t_1)$$

$$\ddot{x}_1(t_1) = \ddot{x}_2(t_1)$$

$$\dot{x}_2(t_2) = \ddot{x}_3(t_2) = x_2$$

$$\dot{x}_2(t_2) = \dot{x}_3(t_2)$$

$$\dot{x}_1(t_0) = 0$$

$$\dot{x}_2(t_2) = \ddot{x}_3(t_2)$$

$$\dot{x}_2(t_2) = \ddot{x}_3(t_2)$$



Spline for nth order system

$$t = \begin{bmatrix} t_0 & t_1 & t_2 & \dots & t_m \end{bmatrix}^T$$

$$x = \begin{bmatrix} x_0 & x_1 & x_2 & \dots & x_m \end{bmatrix}^T$$

$$x_1(t_1) = x_2(t_1) = x_1$$

$$x_1^{2(n-1)}(t_1) = x_2^{2(n-1)}(t_1)$$

$$x_2(t_2) = x_3(t_2) = x_2$$

$$x_3(t_3) = x_3$$

$$x_$$

Spline for nth order system

$$t = \begin{bmatrix} t_0 & t_1 & t_2 & \dots & t_m \end{bmatrix}^T$$

$$x = \begin{bmatrix} x_0 & x_1 & x_2 & \dots & x_m \end{bmatrix}^T$$

$$x_1(t_1) = x_2(t_1) = x_1$$

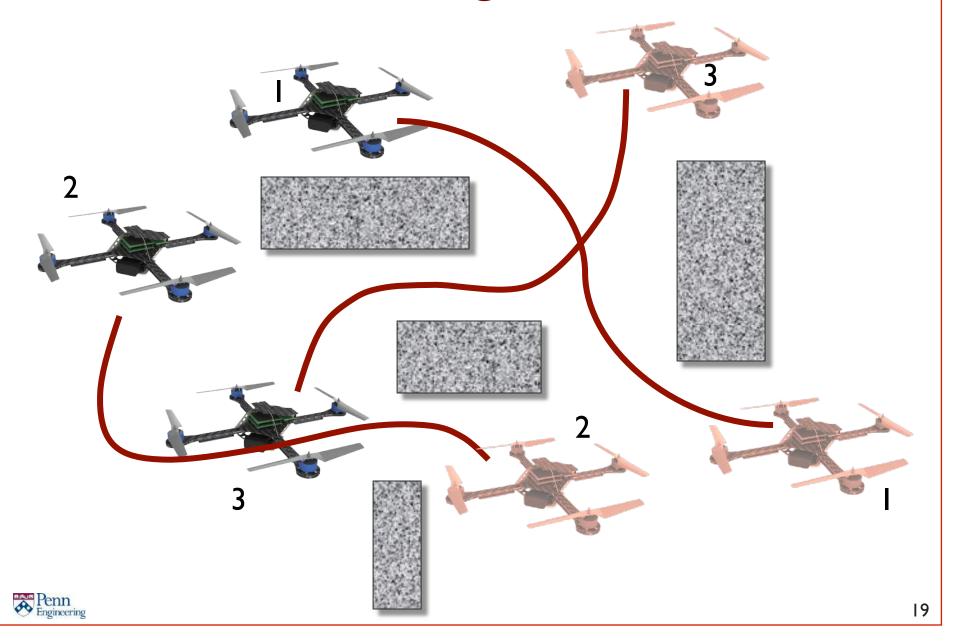
$$x_1^{2(n-1)}(t_1) = x_2^{2(n-1)}(t_1)$$

$$x_3(t_3) = x_3$$

$$x_3(t_3) = x_$$



Motion Planning of Quadrotors



Motion Planning for Quadrotors





$${}^{A}\mathbf{\omega}^{B} = p \mathbf{b}_{1} + q \mathbf{b}_{2} + r \mathbf{b}_{3}$$

 $m\ddot{\mathbf{r}} = egin{bmatrix} 0 \ 0 \ -mg \end{bmatrix}$

B

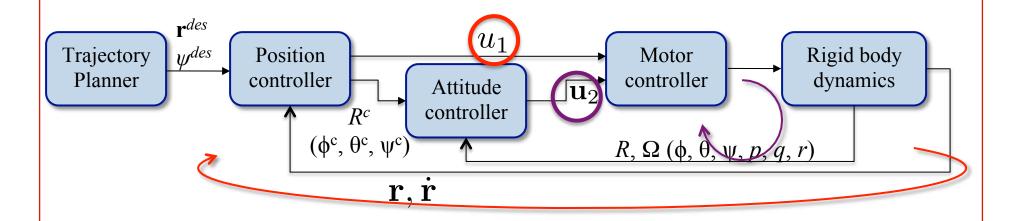
$$I\begin{bmatrix} \dot{p} \\ \dot{q} \\ \dot{r} \end{bmatrix} = \begin{bmatrix} L(F_2 - F_4) \\ L(F_3 - F_1) \\ M_1 - M_2 + M_3 - M_4 \end{bmatrix} -$$

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

Penn Engineering Components in the body frame along \mathbf{b}_1 , \mathbf{b}_2 , and \mathbf{b}_3 , the principal axes

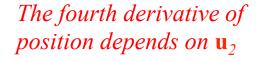
Rotation of thrust

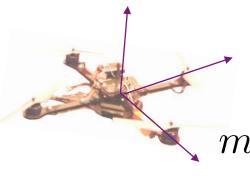
Position Control



Position control loop relies on an inner attitude control loop







$$n\ddot{\mathbf{r}} = \begin{bmatrix} 0 \\ 0 \\ -m \end{bmatrix}$$

$$\begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$F_1 + F_2 + F_3 + F_4$$

$$R(heta,\phi,\psi)$$

$$u_1$$

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

The second derivative of position depends on u_1

The second derivative of the rotation matrix depends on \mathbf{u}_2

$$I egin{bmatrix} \dot{p} \ \dot{q} \ \dot{r} \end{bmatrix} =$$

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$$\begin{bmatrix} L(F_2 - F_4) \\ L(F_3 - F_1) \\ M_1 - M_2 + M_3 - M_4 \end{bmatrix}$$

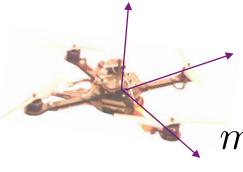
 $egin{bmatrix} p \ q \ x \end{bmatrix} imes I egin{bmatrix} p \ q \ r \end{bmatrix}$

4

Linearized Model

$$(\theta \sim 0, \phi \sim 0, \psi \sim 0)$$

$$(p \sim 0, q \sim 0, r \sim 0)$$



$$m\ddot{\mathbf{r}} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$-mg$$

$$I = 0$$

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

$$I(\theta,\phi,\psi)$$

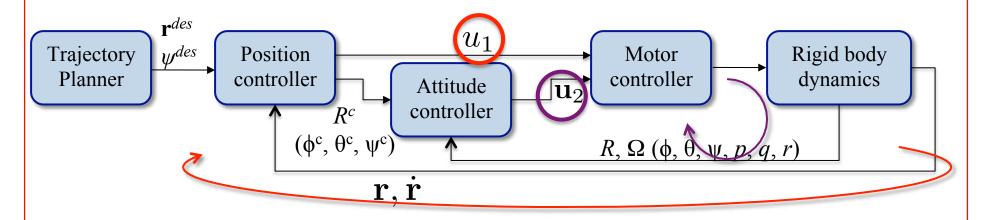
$$\begin{bmatrix} p \\ q \\ r \end{bmatrix} = \begin{bmatrix} c\theta & 0 & -c\phi s\theta \\ 0 & 1 & s\phi \\ s\theta & 0 & c\phi c\theta \end{bmatrix} \begin{bmatrix} e^{i\theta} & 0 & -c\phi s\theta \\ 0 & 1 & s\phi \\ 0 & c\phi c\theta \end{bmatrix}$$

$$I \begin{bmatrix} \dot{p} \\ \dot{q} \\ \dot{r} \end{bmatrix} =$$

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

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

Minimum Snap Trajectory



The position control system is a fourth order system

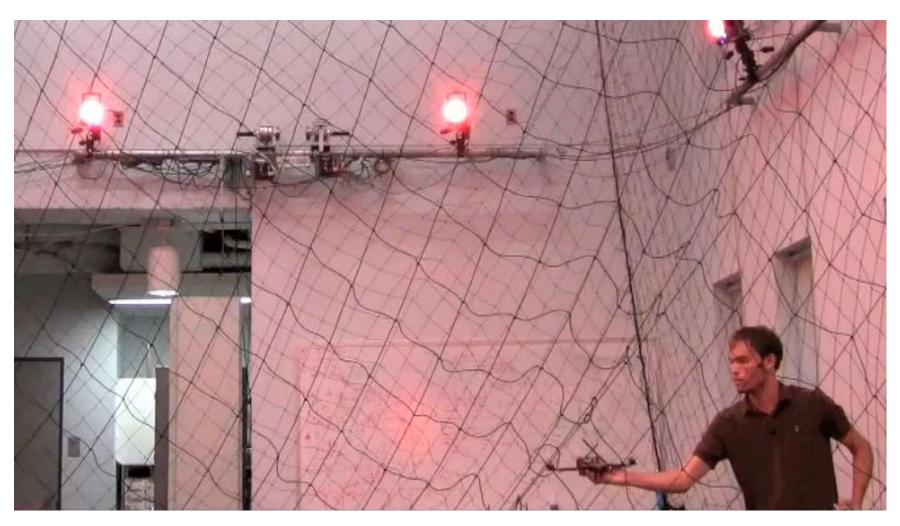
Want trajectories that can be differentiated four times

Minimum Snap Trajectory

$$x^{\star}(t) = \arg\min_{x(t)} \int_0^T \left(x^{(iv)}\right)^2 dt$$



Inner Attitude Control Loop



Daniel Mellinger, Nathan Michael, and Vijay Kumar. Trajectory Generation and Control for Precise Aggressive Maneuvers with Quadrotors. *International Journal of Robotics Research*, Apr. 2012.



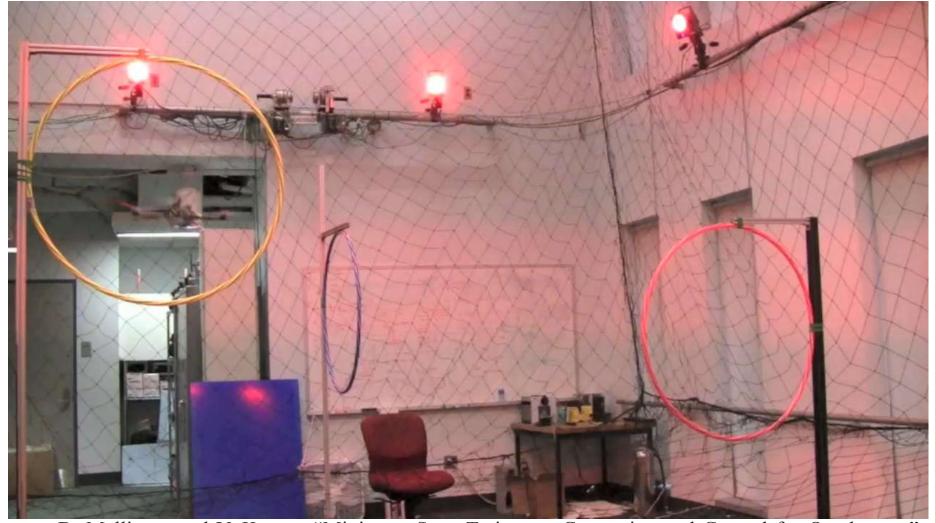
Minimum Snap Trajectories



D. Mellinger and V. Kumar, "Minimum Snap Trajectory Generation and Control for Quadrotors," *Proc. IEEE International Conference on Robotics and Automation*. Shanghai, China, May, 2011.



Automated Synthesis of Trajectories



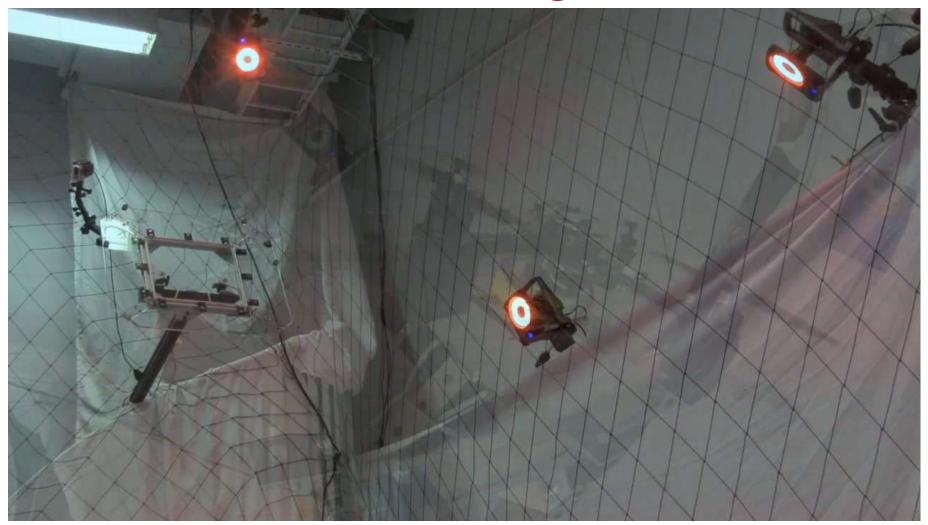
D. Mellinger and V. Kumar, "Minimum Snap Trajectory Generation and Control for Quadrotors," Proc. IEEE International Conference on Robotics and Automation. Shanghai, China, May, 2011.

Aerial Grasping and Manipulation



Justin Thomas, Joe Polin, Koushil Sreenath, and Vijay Kumar, "Avian-inspired grasping for quadrotor micro UAVs," *ASME International Design Engineering Technical Conference* (IDETC), Portland, Oregon, August 2013.

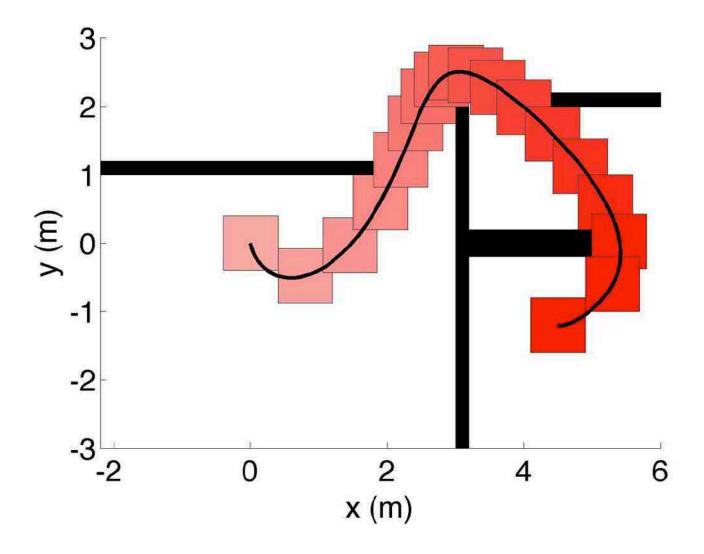
Perching



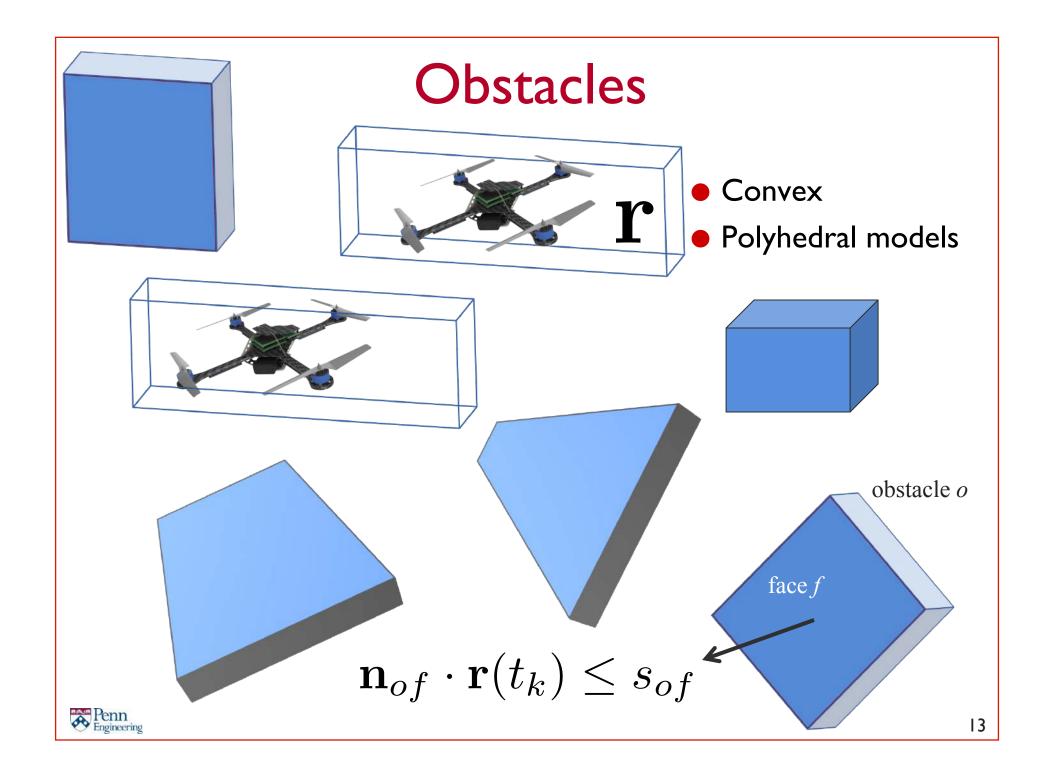
J. Thomas, G. Loianno, M. Pope, E. W. Hawkes, M. A. Estrada, H. Jiang, M. R. Cutkosky, and V. Kumar, "Planning and Control of Aggressive Maneuvers for Perching on Inclined and Vertical Surfaces," in *International Design Engineering Technical Conference & Computers and Information in Engineering Conference (IDETC/CIE)*, Boston MA, August 2015.



Min Snap Trajectory with Constraints







Integer Constraints for Obstacle Avoidance

$$\mathbf{n}_{of} \cdot \mathbf{r}(t_k) \le s_{of} + Mb_{ofk}, \ \forall f = 1, ..., n_f(o)$$

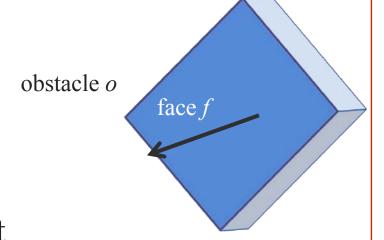
obstacle

 n_f number of faces

 t_k kth time instant

binary variable

M large positive constant



$$\mathbf{n}_{of} \cdot \mathbf{r}(t_k) \leq s_{of}$$

$$\sum_{f=1}^{n_f(o)} b_{ofk} \le n_f(o) - 1$$



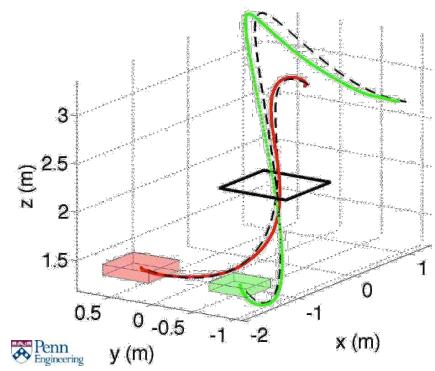
Transporting Suspended Payloads

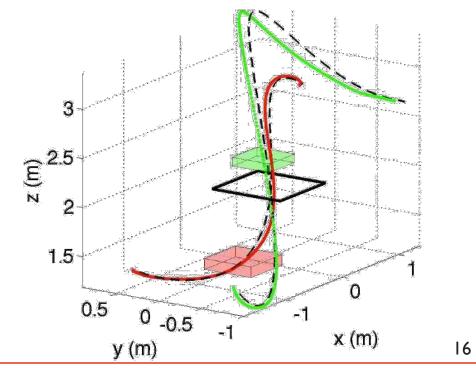


S. Tang and V. Kumar, "Mixed Integer Quadratic Program Trajectory Generation for a Quadrotor with a Cable-Suspended Payload," *in IEEE International Conference on Robotics and Automation*, May 2015.

Results









Aleksandr Kushleyev, Daniel Mellinger, Caitlin Powers, Vijay Kumar, "Towards a swarm of agile micro quadrotors," *Autonomous Robots*, Vol. 35, No. 4, Pg. 287-300, 2013.



Minimum Velocity Trajectories from the Euler-Lagrange Equations



Minimum Velocity Trajectory

Find the function x(t) such that:

$$x^{\star}(t) = \underset{x(t)}{\operatorname{argmin}} \int_{0}^{T} \mathcal{L}(\dot{x}, x, t) dt$$
$$= \underset{x(t)}{\operatorname{argmin}} \int_{0}^{T} \dot{x}^{2} dt$$

Euler-Lagrange equation:

$$\frac{d}{dt} \left(\frac{\partial \mathcal{L}}{\partial \dot{x}} \right) - \frac{\partial \mathcal{L}}{\partial x} = 0$$



Minimum Velocity Trajectory

Euler-Lagrange equation:

$$\frac{d}{dt} \left(\frac{\partial \mathcal{L}}{\partial \dot{x}} \right) + \frac{\partial \mathcal{L}}{\partial x} = 0$$

Cost function:

$$\mathcal{L}\left(\dot{x}, x, t\right) = (\dot{x})^2$$

Euler-Lagrange terms:

$$\left(\frac{\partial \mathcal{L}}{\partial x}\right) = 0$$
 \longleftarrow No x appears in \mathcal{L}

$$\left(\frac{\partial \mathcal{L}}{\partial \dot{x}}\right) = 2\dot{x}$$

$$\frac{d}{dt} \left(\frac{\partial \mathcal{L}}{\partial \dot{x}} \right) = \frac{d}{dt} (2\dot{x}) = 2\ddot{x}$$



Minimum Velocity Trajectory

Euler-Lagrange equation:

$$\frac{d}{dt} \left(\frac{\partial \mathcal{L}}{\partial \dot{x}} \right) - \frac{\partial \mathcal{L}}{\partial x} = 0$$

Euler-Lagrange terms: $2\ddot{x} - 0 = 0 \longrightarrow 2\ddot{x} = 0 \longrightarrow \ddot{x} = 0$

Integrate to get the velocity: $\dot{x} = c_1$

Integrate to get position: $x(t) = c_1 t + c_0$



Solving for Coefficients of Minimum Jerk Trajectories



Minimum Jerk Trajectory

$$x^{\star}(t) = \underset{x(t)}{\operatorname{argmin}} \int_{0}^{T} \mathcal{L}(\ddot{x}, \ddot{x}, \dot{x}, x, t) dt = \underset{x(t)}{\operatorname{argmin}} \int_{0}^{T} \ddot{x}^{2} dt$$

We can solve the Euler-Lagrange equation:

$$\frac{\partial \mathcal{L}}{\partial x} - \frac{d}{dt} \left(\frac{\partial \mathcal{L}}{\partial \dot{x}} \right) + \frac{d^2}{dt^2} \left(\frac{\partial \mathcal{L}}{\partial \ddot{x}} \right) - \frac{d^3}{dt^3} \left(\frac{\partial \mathcal{L}}{\partial x^{(3)}} \right) = 0$$

to get the condition:

$$x^{(6)} = 0$$

Thus, we want a trajectory of the form:

$$x(t) = c_5 t^5 + c_4 t^4 + c_3 t^3 + c_2 t^2 + c_1 t + c_0$$

Solving for Coefficients

Boundary conditions:

	Position	Velocity	Acceleration
t = 0	a	0	0
t = T	b	0	0

Position constraints: $x(t) = c_5 t^5 + c_4 t^4 + c_3 t^3 + c_2 t^2 + c_1 t + c_0$

$$x(0) = c_0 = a$$

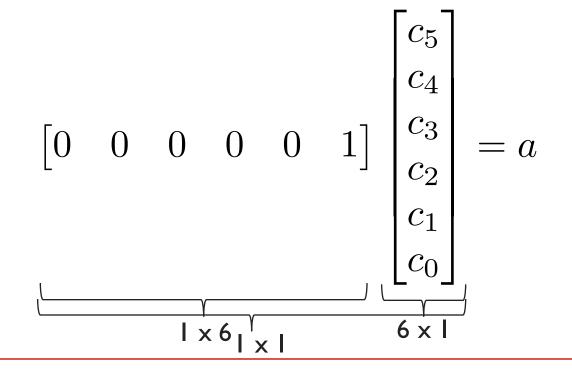
$$x(T) = c_5(T)^5 + c_4(T)^4 + c_3(T)^3 + c_2(T)^2 + c_4(T) + c_0 = b$$



Solving for Coefficients

Position constraints in matrix form:

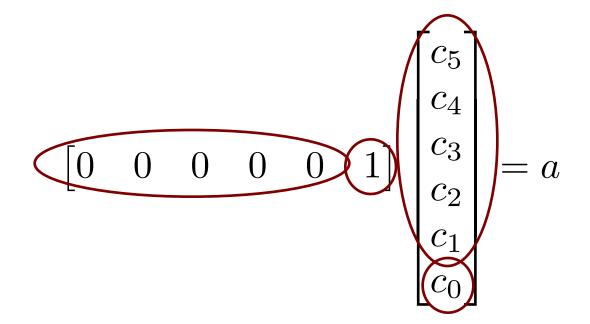
$$x(0) = c_0 = a$$





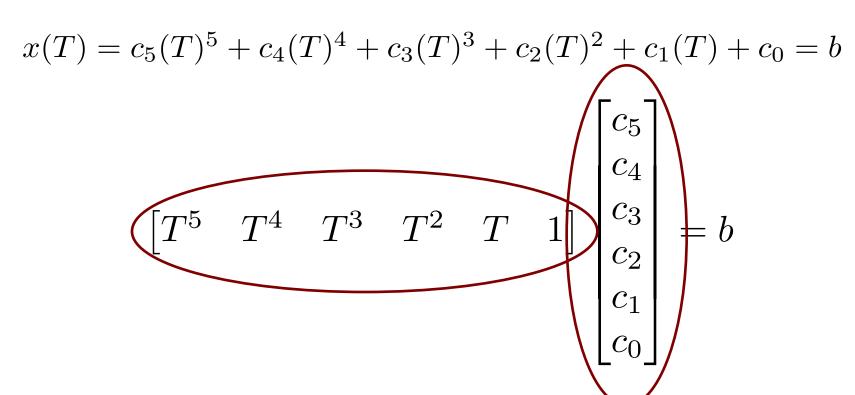
Position constraints in matrix form:

$$x(0) = c_0 = a$$





Position constraints in matrix form:





Boundary conditions:

	Position	Velocity	Acceleration
t = 0	a	0	0
t = T	b	0	0

Velocity constraints: $\dot{x}(t) = 5c_5t^4 + 4c_4t^3 + 3c_3t^2 + 2c_2t + c_1$

$$\dot{x}(0) = c_1 = 0$$

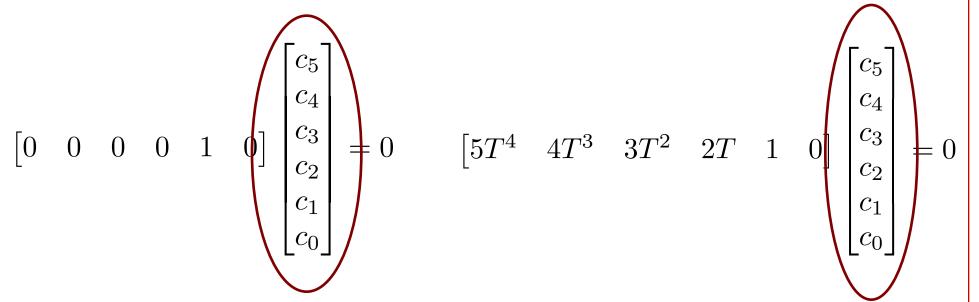
$$\dot{x}(T) = 5c_5(T)^4 + 4c_4(T)^3 + 3c_3(T)^2 + 2c_2(T) + c_1 = 0$$



Velocity constraints in matrix form:

$$\dot{x}(0) = c_1 = 0$$

$$\dot{x}(T) = 5c_5(T)^4 + 4c_4(T)^3 + 3c_3(T)^2 + 2c_2(T) + c_1 = 0$$



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Boundary conditions:

	Position	Velocity	Acceleration
t = 0	a	0	0
t = T	Ь	0	0

Acceleration constraints: $\ddot{x}(t) = 20c_5t^3 + 12c_4t^2 + 6c_3t^2 + 2c_2$

$$\ddot{x}(0) = 2c_2 = 0$$

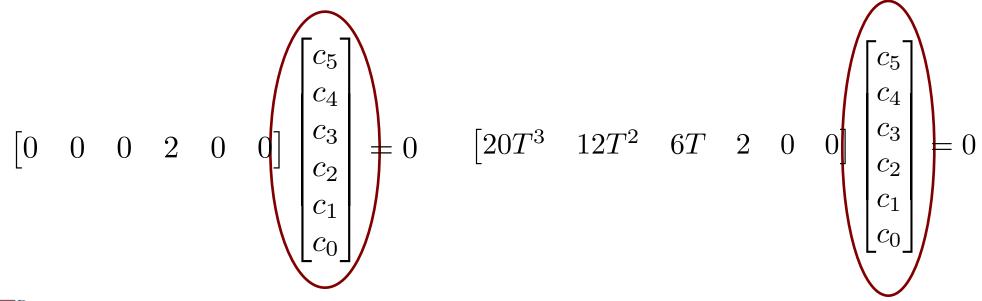
$$\ddot{x}(T) = 20c_5(T)^3 + 12c_4(T)^2 + 6c_3(T)^2 + 2c_2 = 0$$



Acceleration constraints in matrix form:

$$\ddot{x}(0) = 2c_2 = 0$$

$$\ddot{x}(T) = 20c_5(T)^3 + 12c_4(T)^2 + 6c_3(T)^2 + 2c_2 = 0$$



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Boundary conditions:

	Position	Velocity	Acceleration
t = 0	a	0	0
t = T	b	0	0

Combine constraints into one matrix expression:

$$\begin{bmatrix} 0 & 0 & 0 & 0 & 0 & 1 \\ T^5 & T^4 & T^3 & T^2 & T & 1 \\ 0 & 0 & 0 & 0 & 1 & 0 \\ 5T^4 & 4T^3 & 3T^2 & 2T & 1 & 0 \\ 0 & 0 & 0 & 2 & 0 & 0 \\ 20T^3 & 12T^2 & 6T & 2 & 0 & 0 \end{bmatrix} \begin{pmatrix} c_5 \\ c_4 \\ c_3 \\ c_2 \\ c_1 \\ c_0 \end{pmatrix} \neq \begin{bmatrix} a \\ b \\ 0 \\ 0 \\ 0 \end{bmatrix}$$



Find the minimum jerk trajectory with boundary conditions:

	Position	Velocity	Acceleration
t = 0	a = 0	0	0
t = T = 1	b = 5	0	0

$$Ax = b \\ x = A^{-1}b \\ A \longrightarrow \begin{bmatrix} 0 & 0 & 0 & 0 & 0 & 1 \\ 1 & 1 & 1 & 1 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 \\ 5 & 4 & 3 & 2 & 1 & 0 \\ 0 & 0 & 0 & 2 & 0 & 0 \\ 20 & 12 & 6 & 2 & 0 & 0 \end{bmatrix} \begin{bmatrix} c_5 \\ c_4 \\ c_3 \\ c_2 \\ c_1 \\ c_0 \end{bmatrix} = \begin{bmatrix} 0 \\ 5 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$



$$x = \begin{bmatrix} 30 \\ -75 \\ 50 \\ 0 \\ 0 \\ 0 \end{bmatrix} \longleftrightarrow \begin{bmatrix} c_5 \\ c_4 \\ c_3 \\ c_2 \\ c_1 \\ c_0 \end{bmatrix}$$

$$x(t) = 30t^5 - 75t^4 + 50t^3$$



We can verify that this trajectory does in fact satisfy all boundary conditions:

	Position	Velocity	Acceleration
t = 0	a = 0 🗸	0	0
t = T = I	b = 5 🗸	0	0

$$x(t) = 30t^5 - 75t^4 + 50t^3$$

$$x(0) = 0$$

$$x(1) = 30(1)^5 - 75(1)^4 + 50(1)^3 = 5$$



We can verify that this trajectory does in fact satisfy all boundary conditions:

	Position	Velocity	Acceleration
t = 0	a = 0 🗸	0 🗸	0
t = T = I	b = 5 🗸	0 🗸	0

$$\dot{x}(t) = 150t^4 - 300t^3 + 150t^2$$

$$\dot{x}(0) = 0$$

$$\dot{x}(1) = 150 - 300 + 150 = 0$$



We can verify that this trajectory does in fact satisfy all boundary conditions:

	Position	Velocity	Acceleration
t = 0	a = 0 🗸	0 🗸	0 🗸
t = T = 1	b = 5 🗸	0 🗸	0 🗸

$$\ddot{x}(t) = 600t^3 - 900t^2 + 300t$$

$$\ddot{x}(0) = 0$$

$$\ddot{x}(1) = 600 - 900 + 300 = 0$$



Find the Minimum Jerk Trajectory

The order of coefficients in the matrix of unknowns matters.

$$egin{bmatrix} T^5 & T^4 & T^3 & T^2 & T & 1 \end{bmatrix} egin{bmatrix} c_5 \ c_4 \ c_3 \ c_2 \ c_1 \ c_0 \end{bmatrix} = b$$



Find the Minimum Jerk Trajectory

The order of coefficients in the matrix of unknowns matters.

$$\begin{bmatrix} 1 & T & T^2 & T^3 & T^4 & T^5 \end{bmatrix} \begin{bmatrix} c_0 & c_1 & c_2 & c_3 & c_4 & c_5 & c_5 \end{bmatrix}$$



Find the Minimum Jerk Trajectory

The order of coefficients in the matrix of unknowns matters.

$$egin{bmatrix} c_4 & c_1 & c_2 & c_5 & c_3 & c_0 \ \end{bmatrix} egin{bmatrix} c_4 & c_1 & c_2 & c_2 & c_5 & c_5 \ c_3 & c_0 & c_0 \ \end{bmatrix}$$



Minimum Velocity Trajectories



Minimum Velocity Trajectory

Why is the minimum velocity curve also the shortest distance curve?

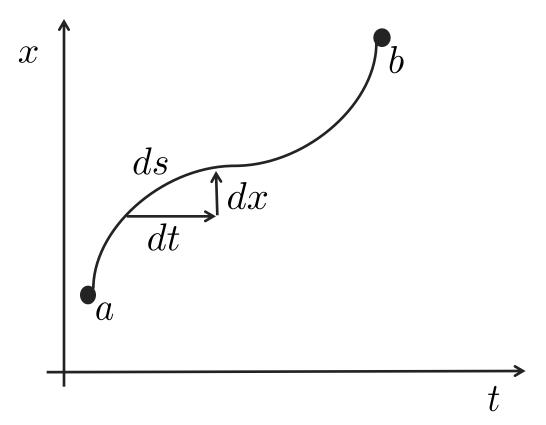
To get the minimum velocity trajectory, we solved:

$$x^{\star}(t) = \underset{x(t)}{\operatorname{argmin}} \int_{0}^{T} \dot{x}^{2} dt$$

From the Euler-Lagrange equations, the solution is:

$$x(t) = c_1 t + c_0$$

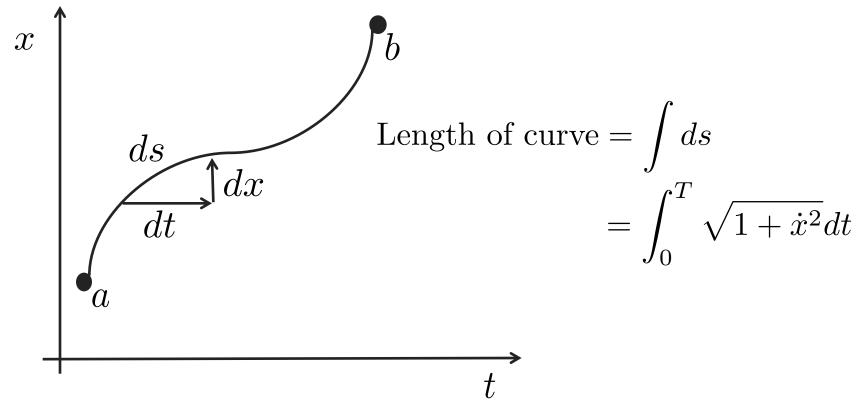




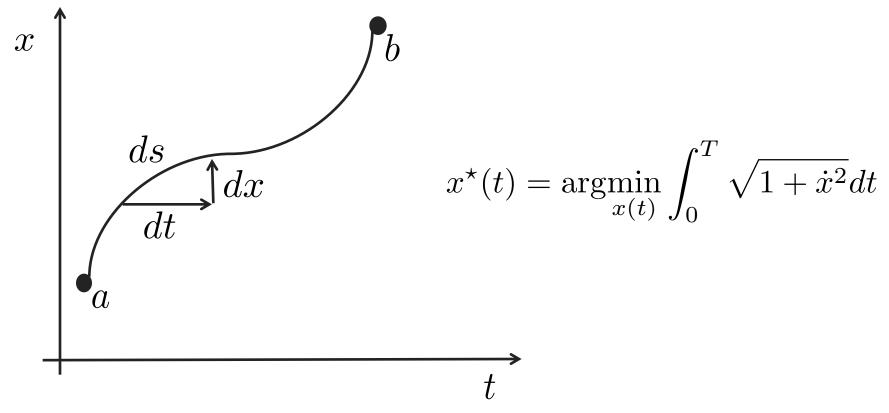
$$ds = \sqrt{dt^2 + dx^2}$$

$$= \sqrt{1 + \left(\frac{dx}{dt}\right)^2} dt$$

$$= \sqrt{1 + \dot{x}^2} dt$$









$$x^{\star}(t) = \operatorname*{argmin}_{x(t)} \int_{0}^{T} \sqrt{1 + \dot{x}^{2}} dt$$

$$\mathcal{L}(\dot{x}, x, t) = \sqrt{1 + \dot{x}^2}$$

Euler-Lagrange equation:

$$\frac{d}{dt} \left(\frac{\partial \mathcal{L}}{\partial \dot{x}} \right) - \frac{\partial \mathcal{L}}{\partial x} = 0$$



Euler-Lagrange equation:

$$\frac{d}{dt} \left(\frac{\partial \mathcal{L}}{\partial \dot{x}} \right) + \frac{\partial \mathcal{L}}{\partial x} = 0$$

Cost-function:
$$\mathcal{L}(\dot{x},x,t)=\sqrt{1+\dot{x}^2}$$

Euler-Lagrange terms:

$$\left(\frac{\partial \mathcal{L}}{\partial x}\right) = 0$$
 \longleftarrow No x appears in \mathcal{L}

$$\left(\frac{\partial \mathcal{L}}{\partial \dot{x}}\right) = \frac{\dot{x}}{\sqrt{1 + \dot{x}^2}}$$

$$\frac{d}{dt} \left(\frac{\partial \mathcal{L}}{\partial \dot{x}} \right) = \frac{d}{dt} \left(\frac{\dot{x}}{\sqrt{1 + \dot{x}^2}} \right)$$



Euler-Lagrange equation:

$$\frac{d}{dt} \left(\frac{\partial \mathcal{L}}{\partial \dot{x}} \right) - \frac{\partial \mathcal{L}}{\partial x} = 0$$

Euler-Lagrange terms: $\frac{d}{dt} \left(\frac{\dot{x}}{\sqrt{1 + \dot{x}^2}} \right) = 0$

Integrate to get velocity: $\frac{\dot{x}}{\sqrt{1+\dot{x}^2}}=K \longrightarrow \dot{x}=\sqrt{\frac{K^2}{1-K^2}}=c_1$

Integrate to get position: $x(t) = c_1 t + c_0$ \leftarrow Same as minimum velocity solution

Linearization of Quadrotor Equations of Motion



Quadrotor Equations of Motion

Linear momentum balance:

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

Angular momentum balance:

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



Quadrotor Equations of Motion

Linear momentum balance:

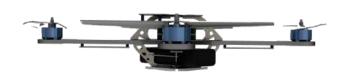
$$m\ddot{\mathbf{r}} = \begin{bmatrix} 0\\0\\-mg \end{bmatrix} + R \begin{bmatrix} 0\\0\\u_1 \end{bmatrix}$$

Angular momentum balance:

$$I \begin{bmatrix} \dot{p} \\ \dot{q} \\ \dot{r} \end{bmatrix} = \begin{bmatrix} u_{2x} \\ u_{2y} \\ u_{2z} \end{bmatrix} - \begin{bmatrix} p \\ q \\ r \end{bmatrix} \times I \begin{bmatrix} p \\ q \\ r \end{bmatrix}$$



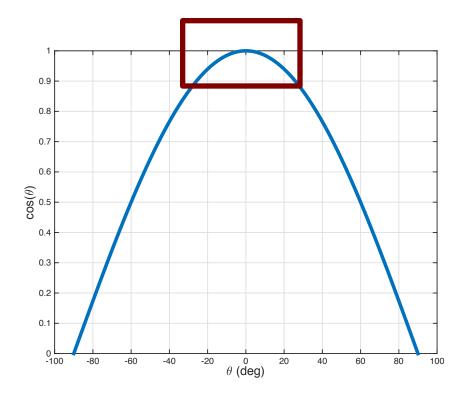
Equilibrium Hover Configuration



$$\mathbf{r} = \mathbf{r}_0, \theta = \phi = 0, \psi = \psi_0$$

$$\dot{\mathbf{r}} = 0, \dot{\theta} = \dot{\phi} = \dot{\psi} = 0$$

What is the value of $cos(\theta)$ near $\theta = 0$?



Can be approximated with the Taylor Series:

 $\cos(\theta)$

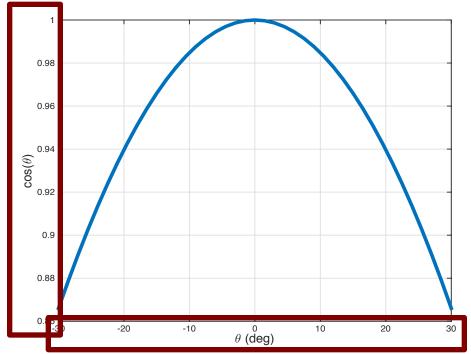
$$\approx \cos(\theta)|_{\theta=0} + \frac{d\cos(\theta)}{d\theta}|_{\theta=0}\theta$$

$$\approx 1 - \sin(\theta)|_{\theta=0}\theta$$

$$\approx 1$$



What is the value of $\cos(\theta)$ near $\theta = 0$?



Can be approximated with the Taylor Series:

 $\cos(\theta)$

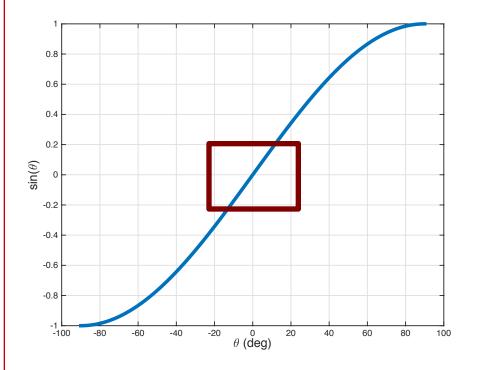
$$\approx \cos(\theta)|_{\theta=0} + \frac{d\cos(\theta)}{d\theta}|_{\theta=0}\theta$$

$$\approx 1 - \sin(\theta)|_{\theta=0}\theta$$

$$\approx 1$$



What is the value of $\sin(\theta)$ near $\theta = 0$?



Can be approximated with the Taylor Series:

 $\sin(\theta)$

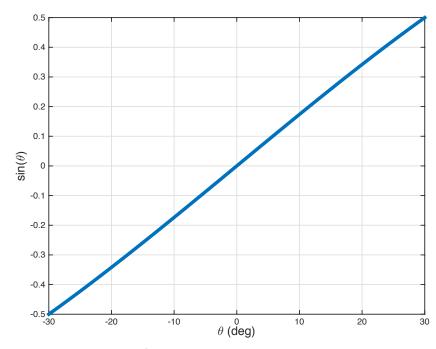
$$\approx \sin(\theta)|_{\theta=0} + \frac{d\sin(\theta)}{d\theta}|_{\theta=0}\theta$$

$$\approx 0 + \cos(\theta)|_{\theta=0}\theta$$

$$\approx \theta$$



What is the value of $\sin(\theta)$ near $\theta = 0$?



sine function looks linear around $\theta = 0$

Can be approximated with the Taylor Series:

 $\sin(\theta)$

$$\approx \sin(\theta)|_{\theta=0} + \frac{d\sin(\theta)}{d\theta}|_{\theta=0}\theta$$

$$\approx 0 + \cos(\theta)|_{\theta=0}\theta$$

$$\approx \theta$$



Linearized Equations of Motion

What are the equations of motion of the quadrotor when it is near the equilibrium hover configuration?

$$\mathbf{r} \approx \mathbf{r}_0, \theta \approx \phi \approx 0, \psi \approx \psi_0$$

$$\dot{\mathbf{r}} \approx 0, \dot{\theta} \approx \dot{\phi} \approx \dot{\psi} \approx 0$$



$$m\ddot{\mathbf{r}} = \begin{bmatrix} 0\\0\\-mg \end{bmatrix} + R \begin{bmatrix} 0\\0\\u_1 \end{bmatrix}$$



$$m\ddot{\mathbf{r}} = \begin{bmatrix} 0 \\ 0 \\ -mg \end{bmatrix} + \begin{bmatrix} c\psi c\theta - s\phi s\psi s\theta & -c\phi s\psi & c\psi s\theta + c\theta s\phi s\psi \\ c\theta s\psi + c\psi s\theta s\phi & c\phi c\psi & s\psi s\theta - c\psi c\theta s\phi \\ -c\phi s\theta & s\phi & c\phi c\theta \end{bmatrix} \begin{bmatrix} 0 \\ 0 \\ u_1 \end{bmatrix}$$



$$m\ddot{x} = (c\psi s\theta + c\theta s\phi s\psi) u_1$$

$$m\ddot{y} = (s\psi s\theta - c\psi c\theta s\phi) u_1$$

$$m\ddot{z} = -mg + (c\phi c\theta) u_1$$

Substituting in the approximation:

$$\sin(\theta) \approx \theta, \sin(\phi) \approx \phi, \cos(\theta) \approx \cos(\phi) \approx 1$$



$$m\ddot{x} = (\theta c\psi + \phi s\psi) u_1$$

$$m\ddot{y} = (\theta s\psi - \phi c\psi) u_1$$

$$m\ddot{z} = -mg + u_1$$

The second derivative of position is proportional to u_1 !



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

$$p = \dot{\phi}c\theta - \dot{\psi}c\phi s\theta$$
$$q = \dot{\theta} + \dot{\psi}s\phi$$
$$r = \dot{\phi}s\theta + \dot{\psi}c\phi c\theta$$

Substituting in the approximation:

$$\sin(\theta) \approx \theta, \sin(\phi) \approx \phi, \cos(\theta) \approx \cos(\phi) \approx 1$$



$$p = \dot{\phi} - \dot{\psi}\theta$$
$$q = \dot{\theta} + \dot{\psi}\phi$$
$$r = \dot{\phi}\theta + \dot{\psi}$$

Substituting in the approximation:

$$\dot{\psi}\theta \approx \dot{\psi}\phi \approx \dot{\phi}\theta \approx 0$$

Higher order terms: Product of two terms around 0 is approximately 0.



$$p = \dot{\phi}$$

$$q = \dot{\theta}$$

$$r = \dot{\psi}$$



$$I\begin{bmatrix} \dot{p} \\ \dot{q} \\ \dot{r} \end{bmatrix} = \begin{bmatrix} u_{2x} \\ u_{2y} \\ u_{2z} \end{bmatrix} - \begin{bmatrix} p \\ q \\ r \end{bmatrix} \times I\begin{bmatrix} p \\ q \\ r \end{bmatrix}$$

Substituting in the approximation:

$$I_{xy} \approx I_{yx} \approx I_{xz} \approx I_{zx} \approx I_{yz} \approx I_{zy} \approx 0$$



$$\begin{bmatrix} I_{xx} & 0 & 0 \\ 0 & I_{yy} & 0 \\ 0 & 0 & I_{zz} \end{bmatrix} \begin{bmatrix} \dot{p} \\ \dot{q} \\ \dot{r} \end{bmatrix} = \begin{bmatrix} u_{2x} \\ u_{2y} \\ u_{2z} \end{bmatrix} - \begin{bmatrix} 0 & r & -q \\ -r & 0 & p \\ q & -p & 0 \end{bmatrix} \begin{bmatrix} I_{xx} & 0 & 0 \\ 0 & I_{yy} & 0 \\ 0 & 0 & I_{zz} \end{bmatrix} \begin{bmatrix} p \\ q \\ r \end{bmatrix}$$



$$I_{xx}\dot{p} = u_{2x} - I_{yy}qr + I_{zz}qr$$

$$I_{yy}\dot{q} = u_{2y} + I_{xx}pr - I_{zz}pr$$

$$I_{zz}\dot{r} = u_{2z} - I_{xx}pq + I_{yy}pq$$

Substituting in the approximation:

$$qr \approx pr \approx pq$$

 $\approx \dot{\theta}\dot{\psi} \approx \dot{\phi}\dot{\psi} \approx \dot{\phi}\dot{\theta} \approx 0$

Higher order terms: Product of two terms around 0 is approximately 0.



$$I_{xx}\dot{p} = u_{2x}$$

$$I_{yy}\dot{q} = u_{2y}$$

$$I_{zz}\dot{r} = u_{2z}$$

Substituting in the approximation:

$$p \approx \dot{\phi}$$

$$q \approx \dot{\theta}$$

$$r \approx \dot{\psi}$$



$$\ddot{\phi} = \frac{u_{2x}}{I_{xx}}$$

$$\ddot{\theta} = \frac{u_{2y}}{I_{yy}}$$

$$\ddot{\psi} = \frac{u_{2z}}{I_{zz}}$$



Equations of Motion

Recall the linearized linear momentum equation:

$$m\ddot{x} = (\theta c\psi + \phi s\psi) u_1$$

Differentiating the equation:

$$m\ddot{x} = (\theta c\psi + \phi s\psi)\dot{u}_1 + (\dot{\theta}c\psi - \theta s\psi\dot{\psi} + \dot{\phi}s\psi + \phi c\psi\dot{\psi})u_1$$

Differentiating again:

$$m\ddot{x} = (\theta c\psi + \phi s\psi) \ddot{u}_1 + 2\left(\dot{\theta}c\psi - \theta s\psi\dot{\psi} + \dot{\phi}s\psi + \phi c\psi\dot{\psi}\right)\dot{u}_1 + \left(\ddot{\theta}c\psi - \dot{\theta}s\psi\dot{\psi} - \theta s\psi\ddot{\psi} - \theta c\psi\dot{\psi}^2 + \ddot{\phi}s\psi + \dot{\phi}c\psi\dot{\psi} + \phi c\psi\ddot{\psi} - \phi c\psi\dot{\psi}^2\right)u_1$$



Equations of Motion

Substituting in the approximation:

$$\ddot{\phi} = \frac{u_{2x}}{I_{xx}}, \ddot{\theta} = \frac{u_{2y}}{I_{yy}}, \ddot{\psi} = \frac{u_{2z}}{I_{zz}}$$

The linear momentum equation becomes:

$$m\ddot{x} = \dots + \left(\frac{u_{2y}}{I_{yy}}c\psi + \frac{u_{2z}}{I_{zz}}\theta(c\psi - s\psi) + \frac{u_{2x}}{I_{xx}}s\psi\right)u_1$$

The fourth derivative of position is proportional to u_2 !

