

Lecture 6.2 - Differential Flatness

Fri 13 Feb 07
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Trajectory generation: find $\dot{x}_d(t)$, $u_d(t)$ that satisfy

$$(*) \quad \dot{x}_d = f(x_d, u_d) \quad x_d(0) = x_0 \quad x_d(T) = x_f$$

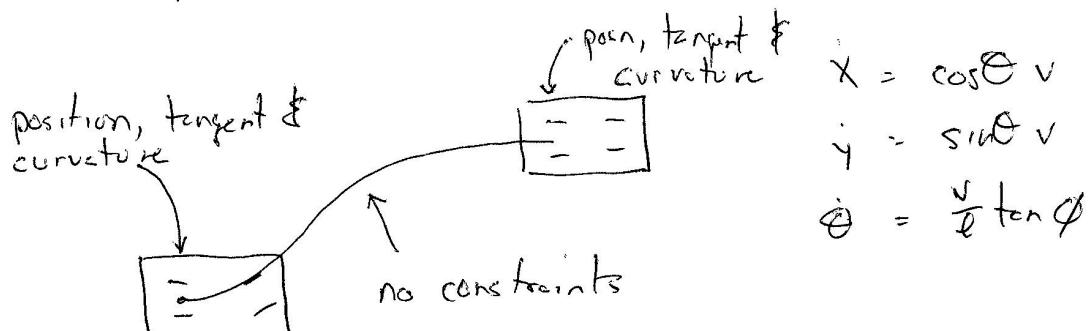
In addition, we may wish to satisfy additional constraints

- Input saturation: $|u(t)| < M$
- State constraints: $g(x) \leq 0$
- Tracking: $h(x) = r(t)$
- Optimization: $\min \int_0^T L(x, u) dt + V(x(T), u(T))$

Special case: $L(x, u) = \|h(x) - r(t)\|^2 \Rightarrow$ tracking

In principle, we can use maximum principle to solve this, but often difficult to work out explicit solutions for x_d , u_d from two point boundary value problem \Rightarrow look for other solutions.

Motivating example: kinematic car



point, tangent &
curvature

$$\begin{aligned} x &= \cos\theta v \\ y &= \sin\theta v \\ \dot{\theta} &= \frac{v}{l} \tan\phi \end{aligned}$$

Given x, y , we can solve for $\theta = \arctan(\dot{y}/\dot{x})$, $v = \dot{x}/\cos\theta$, $\phi = \arctan(\frac{l}{v\dot{\theta}}) \Rightarrow$ get trajectory for full state. Initial &

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Differential Flatness

Defn A nonlinear system (*) is differentially flat, if there exists α functions $\alpha \in \mathbb{R}$ such that

$$z = \alpha(x, u, \dot{u}, \dots, u^{(q)})$$

and we can write the solutions of the differential equation as functions of z and a finite number of derivatives

$$\begin{aligned} x &= \beta(z, \dot{z}, \dots, z^{(p)}) \\ u &= \gamma(z, \dot{z}, \dots, z^{(q)}) \end{aligned}$$

Example: for kinematic car $z = \alpha(\dot{x}) = (x, y)$

Remarks

1. For a differentially flat system, the flat outputs, z , completely define the feasible trajectories of the system
2. The number of flat outputs = number of system inputs
3. General theory for determining if a system is flat is hard; usually guess & check
4. General classes of systems that are flat:
 - Reachable linear systems
 - Mechanical systems with m configuration variables and m inputs
 - Feedback linearizable ~~systems~~ nonlinear systems

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Using flatness to plan trajectories

Suppose we wish to generate a feasible trajectory for NL system

$$\dot{x} = f(x, u) \quad x(0) = x_0 \quad x(T) = x_f$$

If system is differential flat then

$$x(0) = \beta(z(0), \dot{z}(0), \dots, z^{(p)}(0)) = x_0$$

$$x(T) = \gamma(z(T), \dot{z}(T), \dots, z^{(p)}(T)) = x_f$$

Find any $z(0), \dot{z}(0), \dots, z(T), \dot{z}(T), \dots$ that satisfy these constraints (often unique). Choose

$$z(t) = \sum_{i=1}^N \alpha_i \psi_i^*(t) \quad \begin{array}{l} \alpha_i \in \mathbb{R} \\ \psi_i \text{ smooth basis func} \end{array}$$

$$\dot{z}(t) = \sum_{i=1}^N \alpha_i \dot{\psi}_i^*(t)$$

$$\vdots$$

$$z^{(p)}(t) = \sum_{i=1}^N \alpha_i \psi_i^{*(p)}(t)$$

$$\left[\begin{array}{cccc} \psi_1^*(0) & \psi_2^*(0) & \dots & \psi_N^*(0) \\ \dot{\psi}_1^*(0) & \dot{\psi}_2^*(0) & \dots & \dot{\psi}_N^*(0) \\ \vdots & & & \\ \psi_1^{*(p)}(0) & \psi_2^{*(p)}(0) & \dots & \psi_N^{*(p)}(0) \\ \psi_1(T) & \psi_2(T) & \dots & \psi_N(T) \\ \vdots & & & \\ \psi_1^{*(p)}(T) & \psi_2^{*(p)}(T) & \dots & \psi_N^{*(p)}(T) \end{array} \right] \left[\begin{array}{c} \alpha_1 \\ \vdots \\ \alpha_N \end{array} \right] = \left[\begin{array}{c} z(0) \\ \dot{z}(0) \\ \vdots \\ z^{(p)}(0) \\ z(T) \\ \vdots \\ z^{(p)}(T) \end{array} \right]$$

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Example Nonholonomic integrator

$$\dot{x}_1 = u_1$$

Differentially flat w/ $z = (x_2, x_3)$

$$\dot{x}_2 = u_2$$

$$\dot{x}_3 = \dot{x}_1 / u_1 \quad \dot{x}_1 = z_1 \quad \dot{x}_2 = \dot{x}_3 / \dot{x}_1 = \dot{z}_2 / \dot{z}_1$$

$$\dot{x}_3 = x_2 u_1$$

$$x_3 = z_2 \quad u_1 = \dot{z}_1$$

$$u_2 = \dot{x}_2 = \frac{d}{dt} \left(\frac{\dot{z}_2}{\dot{z}_1} \right) = \frac{\ddot{z}_2}{\dot{z}_1} - \frac{\dot{z}_2 \ddot{z}_1}{\dot{z}_1^2}$$

Note: intuition says taking derivatives is "noisy". Here we are taking derivatives of basis functions \Rightarrow noise isn't an issue.

$$z_1(t) = \sum a_{1,i} \psi_{1,i}(t) \quad z_2(t) = \sum a_{2,i} \psi_{2,i}(t)$$

Generate a trajectory from origin to a point $x = (x_{1f}, x_{2f}, x_{3f})$

Choose basis functions as simple polynomials

$$\begin{aligned} \psi_{1,1}(t) &= 1 & \psi_{1,2}(t) &= t & \psi_{1,3}(t) &= t^2 & \psi_{1,4}(t) &= t^3 \\ \psi_{2,1}(t) &= 1 & \psi_{2,2}(t) &= t & \psi_{2,3}(t) &= t^2 & \psi_{2,4}(t) &= t^3 \end{aligned}$$

$$\begin{matrix} z_1 \\ z_2 \\ \vdots \\ \dot{z}_1 \\ \dot{z}_2 \\ \vdots \end{matrix} = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 1 & T & T^2 & T^3 & 0 & 0 \\ 0 & 1 & 2T & 3T^2 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & T \\ 0 & 0 & 0 & 0 & 0 & 2T \end{bmatrix} \begin{bmatrix} a_{11} \\ a_{12} \\ a_{13} \\ a_{21} \\ a_{22} \\ a_{23} \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \\ 0 \\ 0 \\ x_{1f} \\ 1 \\ x_{3f} \\ x_{2f} \end{bmatrix}$$

8x8 matrix

Invertible for most T

Doesn't depend on x_0, x_f

Some freedom left

Remarks

- There are several remaining degrees of freedom ($T, 1, \text{etc}$) that have to be specified \Rightarrow solutions are not unique
- Want to parameterize and solve least squares

Using flatness for constrained, sub-optimal trajectory generation

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Return to full problem

$$\mathcal{L} = \mathcal{L}(x, u), \quad \min \int_0^T L(x, u) dt + V(x(T), u(T))$$

subject to

$$\dot{x} = f(x, u) \quad g(x, u) \leq 0 \quad \leftarrow \text{state + input constraints}$$

If system is differentially flat, can write $x = \sum a_i \psi_i(t)$

$$\begin{aligned} x &= \beta(z, \dot{z}, \dots, z^{(n)}) = \beta(a, t) \\ u &= \gamma(\quad\quad\quad) - \gamma(a, t) \end{aligned}$$

Can rewrite entire problem in terms of α

$$\min \int_0^T L(\beta(\alpha, t), \gamma(a, t)) dt + V(\beta(\alpha, T), \gamma(a, T))$$

$$g(\beta(\alpha, t), \gamma(a, t)) \leq 0 \quad \text{No dynamics}$$

This is an optimization problem in fixed set of parameters:
 given a , can compute all quantities. Very good numerical tools available.

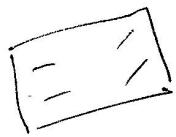
Remarks

1. $g(\beta(\alpha, t), \gamma(a, t))$ give constraint for each $t \Rightarrow$ infinite # of constraints \Rightarrow hard. Can replace with $g(\beta(a, t_i), \gamma(a, t_i))$ for $t_1, t_2, \dots, t_m \Rightarrow$ finite approx.
2. Can extend to non-flat, etc (more next week)
3. Sub-optimal, but fast

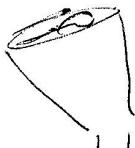
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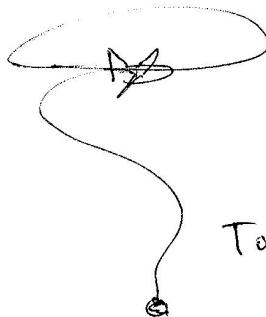
Example of flat systems



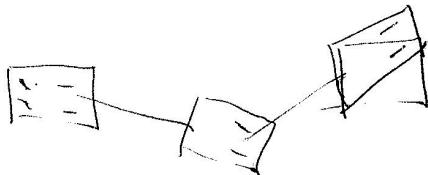
Kin car



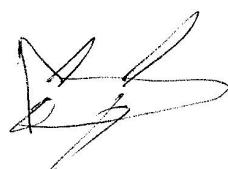
Ducted fan



Towed cable



N-trailer



Lift (elevator) \ll Lift (wings)