

EN530.603 Applied Optimal Control

Lecture 4: Trajectory Optimization Basics

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1 Variations of functions

We are interested in solving optimal control problems such as

轨迹
↓

$$\min J(x(\cdot), u(\cdot), t_f) = \int_{t_0}^{t_f} L(x(t), u(t), t) dt$$

subject to $\dot{x} = f(x, u, t)$ and other constraints. The cost J is called a *functional*, i.e. it is a function of functions since the trajectories $x(\cdot)$ and $u(\cdot)$ are functions of time. It is possible to optimize a functional in a similar way we optimize a regular function. In particular, there is a functional analog to the necessary conditions for a minimum of a function g given by $\nabla g = 0$. This analog is

differential of a function $\nabla g = 0 \Leftrightarrow$ variation of a functional $\delta J = 0$ 相当于对函数求导

Next, define the *change* in a functional, after *varying* $x(t)$ by $\delta x(t)$ at each t , by

$$\Delta J(x(\cdot), \delta x(\cdot)) = J(x(\cdot) + \delta x(\cdot)) - J(x(\cdot)).$$

disturbance 只在 $x(\cdot)$ 上

The *variation* $\delta J(x(\cdot), \delta x(\cdot))$ is a linear function of $\delta x(\cdot)$ defined by the following relationship

$$\Delta J(x(\cdot), \delta x(\cdot)) = \delta J(x(\cdot), \delta x(\cdot)) + o(\|\delta x\|),$$

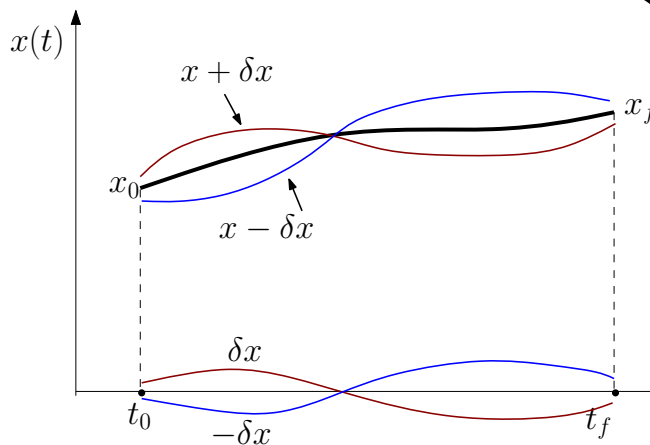
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where the *small-o* notation was used. For a positive integer p and a function $h : \mathbb{R}^n \rightarrow \mathbb{R}^m$ we have

$$h(x) = o(\|x\|^p),$$

if $\lim_{k \rightarrow \infty} \frac{h(x_k)}{\|x_k\|^p} = 0$ for all sequences x_k such that $x_k \rightarrow 0$ and $x_k \neq 0$ for all k .

In other words, roughly speaking, if $h(x)$ denotes the the second and higher order terms in ΔJ , then $h(x)$ has second and higher-order multiples of δx and thus $h(x)/\|\delta x\|$ is at least proportional to δx which tends to zero as $\delta x \rightarrow 0$.



$$\delta J = k \cdot \delta x$$

$x, \delta x$ 都是函数
 δx 是一个扰动

δx 只要连续
A $\delta x(t_0) = \delta x(t_f) = 0$
即可

不能对起点和终点有影响

$$\Delta J(x(\cdot), \delta x(\cdot)) = J(x(\cdot) + \delta x(\cdot)) - J(x(\cdot))$$

Similarly to standard function optimization, we can use the argument that at an optimum $x^*(\cdot)$ we have

$$\Delta J(x^*(\cdot), \delta x(\cdot)) \geq 0$$

since any change of the cost away from optimum must be positive. Taking $\delta x \rightarrow 0$ this implies that

$$\delta J(x^*(\cdot), \delta x(\cdot)) \geq 0.$$

But this must also hold for variations in the direction $-\delta x(\cdot)$, i.e.

$$\delta J(x^*(\cdot), -\delta x(\cdot)) \geq 0 \Rightarrow \delta J(x^*(\cdot), \delta x(\cdot)) \leq 0,$$

by the linearity of δJ . Both conditions can only be true when

$$\delta J(x^*(\cdot), \delta x(\cdot)) = 0,$$

for any $\delta x(\cdot)$.

当 x 是 optimal 时
任何关于 x 的改变都会导致 $\text{cost} \uparrow \Rightarrow \delta J \geq 0$
 x 为 optimal 时 δJ 是 0 类似于 $\nabla g(x^*) = 0$

2 The Euler-Lagrange Equations

Consider the cost function

$$J = \int_{t_0}^{t_f} g(x(t), \dot{x}(t)) dt.$$

Its variation becomes

$$\delta J = \int_{t_0}^{t_f} [g_x(x, \dot{x}) \delta x + g_{\dot{x}}(x, \dot{x}) \delta \dot{x}] dt,$$

since the variation corresponds to taking first-order terms (i.e. derivatives) of the functions at time t . Integrate by parts (recall $\int uv = uv - \int \dot{u}v$) to get

$$\delta J = \int_{t_0}^{t_f} \left[g_x(x, \dot{x}) \delta x - \frac{d}{dt} g_{\dot{x}}(x, \dot{x}) \delta x \right] dt + \underbrace{g_{\dot{x}}(x(t_f), \dot{x}(t_f)) \delta x(t_f)}_{=0}$$

Fixed boundary conditions. If $x(t_0)$ and $x(t_f)$ are given then $\delta x(0) = \delta x(t_f) = 0$ and

$$\delta J = \int_{t_0}^{t_f} \left[g_x(x, \dot{x}) - \frac{d}{dt} g_{\dot{x}}(x, \dot{x}) \right] \delta x(t) dt$$

Since $\delta x(t)$ are arbitrary and independent then $\delta J = 0$ only when

$$g_x(x, \dot{x}) - \frac{d}{dt} g_{\dot{x}}(x, \dot{x}) = 0,$$

which are called the Euler-Lagrange equations (EL).

Free boundary conditions. When $x(t_f)$ is not fixed, in addition to the EL equation, the following must hold

$$g_{\dot{x}}(x(t_f), \dot{x}(t_f)) = 0.$$

一般 $x(t_0)$ 是固定
起点已知

to make $\delta J = 0$



$$\text{E.L. eq.} \quad g_x(x, \dot{x}) - \frac{d}{dt}g_{\dot{x}}(x, \dot{x}) = 0,$$

2.1 Example: shortest path curve



Consider a one-dimensional problem with state $x(t)$ and the cost function

$$J = \int_{t_0}^{t_f} \sqrt{1 + \dot{x}^2} dt,$$

which in fact corresponds to the length of a curve in the (x, t) plane. The goal is to compute the shortest length curve between two given points (x_0, t_0) and (x_f, t_f) .

Applying the EL equations we get

fixed boundary

$$\frac{d}{dt} \frac{\dot{x}}{(1 + \dot{x}^2)^{1/2}} = 0 \quad \Leftrightarrow \quad \frac{\ddot{x}}{(1 + \dot{x}^2)^{3/2}} = 0,$$

which is satisfied when $\ddot{x} = 0$ or when

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

in this form, to get c_1 and c_0 .

need more info. like $x_0 = c_0$ $x_f = c_1$

i.e. when $x(t)$ is a straight line. Let $t_0 = 0$ and $t_f = 1$. It is easy to see that $c_0 = x_0$ and $c_1 = x_f - x_0$.

2.2 Particle in 3-D

$x(t_0)$ $x(t_f)$ are fixed

Let $g(x, \dot{x}, t) = \frac{1}{2}m\|\dot{x}\|^2 - V(x)$, where m denotes the mass and V denotes the potential energy of a particle with position $x \in \mathbb{R}^3$. The function g is actually the Lagrangian of the particle and the EL equations lead to

$$m\ddot{x} = -\nabla_x V,$$

能量方程

which is simply Newton's law. This is one of the simplest examples that illustrates that Lagrangian mechanics can be considered as a special case of optimal control.

3 Free final-time

原

$$\delta J = \int_{t_0}^{t_f} \left[g_x(x, \dot{x}) \delta x - \frac{d}{dt} g_{\dot{x}}(x, \dot{x}) \delta x \right] dt + g_{\dot{x}}(x(t_f), \dot{x}(t_f)) \delta x(t_f)$$

When the final time t_f is allowed to vary, the variation of J is expressed as

$$\begin{aligned} \delta J = & \int_{t_0}^{t_f} \left[g_x(x^*, \dot{x}^*, t) \delta x - \frac{d}{dt} g_{\dot{x}}(x^*, \dot{x}^*, t) \delta x \right] dt \\ & + g_{\dot{x}}(x^*(t_f), \dot{x}^*(t_f), t) \delta x(t_f) + \underline{g(x^*(t_f), \dot{x}^*(t_f), t_f) \delta t_f}, \end{aligned}$$

w.r.t t_f

where we employ the spatially-optimized trajectory $x^*(t)$ instead of just $x(t)$ to signify that after finding the optimal path without considering δt_f , then variations of t_f infinitesimally contribute to the cost by the term

$$g(x^*(t_f), \dot{x}^*(t_f), t_f) \delta t_f$$

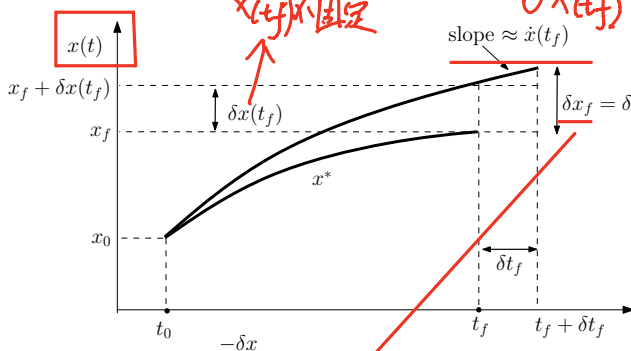
which is simply the cost at the last points multiplied by the time variations: think of it as a first-order approximation to the integral $\int_{t_f}^{t_f + \delta t_f} g(x^*, \dot{x}^*, t) dt$ which is what must be added to the cost when varying time.

↑
最后一段, 一直在增加

注意 y 轴只是 $x(t)$ 并不固定.

δt_f
 $\delta x(t_f)$

在 t_f 就已经有 disturb 了



上边: final time
= $t_f + \delta t_f$

$\delta x(t_f)$ 不足以表示变化

Define the total space-time variation δx_f by

x_f 是 x 最终的值

$x(t_f)$ 是 x 在 t_f 时的值 t_f 有扰动 δt_f

i.e. this variation combines the effects of varying the final state by keeping time t_f fixed, and then by varying t_f (See figure above).

We have

$$\delta J = \int_{t_0}^{t_f} \left[g_x(x^*, \dot{x}^*, t) \delta x - \frac{d}{dt} g_{\dot{x}}(x^*, \dot{x}^*, t) \delta x \right] dt + g_{\dot{x}}(x^*(t_f), \dot{x}^*(t_f), t) \delta x(t_f) + g(x^*(t_f), \dot{x}^*(t_f), t_f) \delta t_f,$$

展开
1st order approx

Unrelated t_f and $x(t_f)$.

We next consider the case when t_f and $x(t_f)$ are unrelated in which we have

$$g_{\dot{x}}(t_f) = 0, \quad [g(t_f) - g_{\dot{x}}(t_f) \dot{x}] \delta t_f = 0 \Rightarrow g(t_f) = 0$$

Function $\Theta(t_f)$. We next consider the case when the boundary constraint is given by

related t_f and $x(t_f)$

$$x(t_f) = \Theta(t_f)$$

boundary 是个关于 t_f 的函数

Variations δt_f and δx_f are related by

$$\delta x_f = \delta x(t_f) + \dot{x}(t_f) \delta t_f,$$

x 的方程是 fixed

$$\delta x(t_f) = 0$$

$$\delta x_f = \frac{d\Theta}{dt} \delta t_f,$$

where the total variation does not contain a spatial component $\delta x(t_f)$ since the final state is completely determined by time only. The necessary conditions become

所以只用管 \int 外的项

$$g_{\dot{x}}(t_f) \left[\frac{d\Theta}{dt} - \dot{x}^* \right]_{t=t_f} + g(t_f) = 0,$$

which are called transversality conditions.

$$g_{\dot{x}}(x^*, \dot{x}^*, t_f) \delta x_f + [g(x^*, \dot{x}^*, t_f) - g_{\dot{x}}(x^*, \dot{x}^*, t_f) \dot{x}] \delta t_f = 0$$

4 Differential Constraints

Consider the optimization of

$$J = \int_{t_0}^{t_f} g(x, \dot{x}, t) dt$$

$x(t_f)$ & t_f are fixed

where $x \in \mathbb{R}^{n+m}$ subject to

$$f(x(t), \dot{x}(t), t) = 0,$$

where $f = (f_1, \dots, f_n)$ are n constraints, and $x(t_f)$ and t_f are fixed. To obtain the necessary conditions, we define the Lagrangian multipliers $\lambda: [t_0, t_f] \rightarrow \mathbb{R}^n$ and the augmented cost

$$J_a = \int_{t_0}^{t_f} \{g(x, \dot{x}, t) + \lambda^T f(x, \dot{x}, t)\} dt$$

↑ constraint optimization 做法 t_f fixed

Taking variations

$$\delta J_a = \int_{t_0}^{t_f} \{[g_x + \lambda^T f_x] \delta x(t) + [g_{\dot{x}} + \lambda^T f_{\dot{x}}] \delta \dot{x} + \delta \lambda^T f\} dt$$

$J_a(x, \dot{x}, \lambda)$ λ is variable

Integrating by parts we get

$$\delta J_a = \int_{t_0}^{t_f} \left\{ [g_x + \lambda^T f_x - \frac{d}{dt} [g_{\dot{x}} + \lambda^T f_{\dot{x}}] \delta x + \delta \lambda^T f \right\} dt.$$

$$\delta J = \int_{t_0}^{t_f} \left[g_x(x, \dot{x}) - \frac{d}{dt} g_{\dot{x}}(x, \dot{x}) \right] \delta x(t) dt$$

If we define the augmented cost g_a by

$$g_a = g + \lambda^T f$$

the Euler-Lagrange equations become

$$\frac{\partial g_a}{\partial x} - \frac{d}{dt} \frac{\partial g_a}{\partial \dot{x}} = 0$$

which along with the constraint

$$f(x, \dot{x}, t) = 0$$

constitute the necessary conditions.

5 General Boundary Constraints

Let $x \in \mathbb{R}^n$ and consider the optimization of

terminal cost

$$J(x(\cdot), t_f) = \varphi(x(t_f), t_f) + \int_{t_0}^{t_f} g(x, \dot{x}, t) dt$$

subject to the free final time t_f and general boundary conditions

$$\psi(x(t_f), t_f) = 0, \quad \leftarrow \text{terminal constraint.}$$

where ψ is a vector of m functions. To obtain the necessary conditions define the augmented cost

$$J_a = \varphi(x(t_f), t_f) + \nu^T \psi(x(t_f), t_f) + \int_{t_0}^{t_f} g(x, \dot{x}, t) dt,$$

↑ another set of multipliers

where $\nu \in \mathbb{R}^m$. Let

$$w(x(t_f), \nu, t_f) = \varphi(x(t_f), t_f) + \nu^T \psi(x(t_f), t_f)$$

$Q(x_f, t_f)$

这里是 x_{final_time} 用 x_f 更准确

5

但不能用 x_f

$\delta x_f = \delta x(t_f)$
 $+ \dot{x}(t_f) \delta t_f$
 依旧适用

no dynamic constraints now

x_f free terminal!

$J_a(x(t_f), t_f, \nu, \dot{x}, \ddot{x})$

Taking variations

$$\delta J_a = w_x(t_f)\delta x_f + w_{t_f}(t_f)\delta t_f + \underline{\delta \nu^T \psi(t_f)} + \int_{t_0}^{t_f} [g_x \cdot \delta x(t) + g_{\dot{x}} \cdot \delta \dot{x}(t)] dt + g(t_f)\delta t_f,$$

where the last term is due to cost accrued from final time variations. Using integration by parts as well as the total variation definition $\delta x_f = \delta x(t_f) + \dot{x}(t_f)\delta t_f$ we have

$$\begin{aligned} \int_{t_0}^{t_f} g_{\dot{x}} \cdot \delta \dot{x}(t) dt &= g_{\dot{x}}(t_f)\delta x(t_f) - \int_{t_0}^{t_f} \frac{d}{dt} g_{\dot{x}} \cdot \delta x(t) \\ &= g_{\dot{x}}(t_f)(\delta x_f - \dot{x}(t_f)\delta t_f) - \int_{t_0}^{t_f} \frac{d}{dt} g_{\dot{x}} \cdot \delta x(t) \end{aligned}$$

which results in

$$\delta J_a = [w_x(t_f) + g_{\dot{x}}]\delta x_{t_f} + [w_{t_f}(t_f) + g(t_f) - g_{\dot{x}}(t_f)\dot{x}(t_f)]\delta t_f + \delta \nu^T \psi(t_f) + \int_{t_0}^{t_f} [g_x - \frac{d}{dt} g_{\dot{x}}]\delta x(t) dt,$$

The necessary conditions require that $\delta J_a = 0$ for arbitrary $\delta x(t), \delta \nu$ which is only possible if the following necessary conditions hold:

$$\nabla_x w(x(t_f), \nu, t_f) + \nabla_{\dot{x}} g(x(t_f), \dot{x}(t_f), t_f) = 0, \quad (1)$$

$$\frac{\partial}{\partial t_f} w(x(t_f), \nu, t_f) + g(x(t_f), \dot{x}(t_f), t_f) - \nabla_{\dot{x}} g(x(t_f), \dot{x}(t_f), t_f)^T \dot{x}(t_f) = 0, \quad (2)$$

$$\psi(x(t_f), t_f) = 0 \quad (3)$$

$$\nabla_x g(x(t), \dot{x}(t), t) - \frac{d}{dt} \nabla_{\dot{x}} g(x(t), \dot{x}(t), t) = 0, \quad t \in (t_0, t_f). \quad (4)$$