OPTIMAL CONTROL

JOSÉ C. GEROMEL

DSCE / School of Electrical and Computer Engineering UNICAMP, 13083-852, Campinas, SP, Brazil, geromel@dsce.fee.unicamp.br

Campinas, Brazil, July 2012

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Our first goal is to handle the following optimization problem

$$\min_{x \in \mathcal{X}} J(x)$$

where:

- t = 0 and t = T are the initial and final times. T > 0 is given.
- $x(t): [0, T] \to \mathbb{R}^n$ is a trajectory defined $\forall t \in [0, T]$.
- \mathcal{X} is the set of all feasible trajectories. It may include constraints on x(0) and x(T).
- $J(x): \mathcal{X} \to \mathbb{R}$ is an objective function of the general type

$$J(x) = \int_0^T f(x(t), \dot{x}(t)) dt$$

It is assumed that $f \in \mathbb{C}^1$. More sophisticated objective function will be addressed as well.

 To show how to obtain the optimality conditions for the previous problem we consider, in addition, the initial and final time constraints

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

• Assume a feasible trajectory $x^*(t)$ is optimal and is slightly perturbed

$$x(t) = x^*(t) + \epsilon \xi(t), \ \forall t \in [0, T]$$

where $\epsilon \in \mathbb{R}$ is an arbitrarily small parameter and $\xi(0)=\xi(T)=0$. From Taylor's series development we obtain

$$J(x) = J(x^*) + \epsilon \int_0^T \left\{ \frac{\partial f'}{\partial x} \xi + \frac{\partial f'}{\partial \dot{x}} \dot{\xi} \right\} dt + \mathcal{O}(\epsilon^2)$$

• The key observation is

$$\int_0^T \left\{ \frac{\partial f'}{\partial \dot{x}} \dot{\xi} \right\} dt = \left. \frac{\partial f'}{\partial \dot{x}} \dot{\xi} \right|_0^T - \int_0^T \left\{ \frac{d}{dt} \frac{\partial f'}{\partial \dot{x}} \dot{\xi} \right\} dt$$

which, imposing $\xi(0) = \xi(T) = 0$ yields

$$J(x) = J(x^*) + \epsilon \int_0^T \left\{ \frac{\partial f}{\partial x} - \frac{d}{dt} \frac{\partial f}{\partial \dot{x}} \right\}' \xi dt + \mathcal{O}(\epsilon^2)$$

• Since x(t) is optimal, we must have $J(x) \leq J(y)$ for all $\epsilon \in \mathbb{R}$ sufficiently small and all feasible trajectory $\xi(t)$. The consequence is that $x^*(t)$ must satisfy

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

• The next theorem provides a necessary condition for optimality

Theorem (6)

Let T > 0 be given. If x(t) satisfying $x(0) = x_0$ and $x(T) = x_T$ is a local minimum then it satisfies the so called Euler-Lagrange equation

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

- A trajectory candidate to be optimal can be determined by solving the Euler-Lagrange equation subject to the initial time $x(0) = x_0$ and final time $x(T) = x_T$ constraints.
- In general, the Euler-Lagrange equation requires a numerical procedure to be solved.

• Example 1 : Consider $x(t):[0,1]\to\mathbb{R}$ such that x(0)=0 and x(1)=1. We want to solve

$$\min_{\mathbf{x}\in\mathcal{X}}\int_0^1 \sqrt{1+\dot{\mathbf{x}}(t)^2}dt$$

The Euler-Lagrange equation is written as

$$\frac{d}{dt}\frac{\dot{x}}{\sqrt{1+\dot{x}^2}}=0$$

which reduces to

$$\ddot{x} = 0$$

Solving it under the previous constraints, we get the unique solution $x^*(t) = t$ which is the line segment connecting the points (0, x(0)) and (1, x(1)), as expected.

• **Example 2 :** In the plane (x,t) consider two points (1,0) and (0,1). Only under the gravity acceleration g, determine the curve x(t), $0 \le t \le 1$ such that a mass starting from the rest in the first point will reach the second one in minimum time. We have to solve

$$\min_{x \in \mathcal{X}} \int_0^1 \sqrt{\frac{1 + \dot{x}(t)^2}{2gx(t)}} dt$$

Applying the Euler-Lagrange equation we obtain

$$2x\ddot{x} + (\dot{x})^2 + 1 = 0$$

subject to x(0) = 1 and x(1) = 0. This is a nonlinear second order differential equation. How to solve it, analytically or numerically?

A more general problem

Our goal now is to handle the following optimization problem

$$\min_{x \in \mathcal{X}, T \ge 0} J(x, T)$$

where:

- t = 0 and t = T are the initial and final times. T > 0 is free.
- $x(t): [0, T] \to \mathbb{R}^n$ is a trajectory defined $\forall t \in [0, T]$.
- \mathcal{X} is the set of all feasible trajectories. It includes the constraints $x(0) = x_0$ and $x(T) = x_T$.
- $J(x,T): \mathcal{X} \times \mathbb{R} \to \mathbb{R}$ is an objective function of the general type

$$J(x,T) = \int_0^T f(x(t),\dot{x}(t))dt$$

It is assumed that $f \in \mathbb{C}^1$. The final time T is an optimization variable to be determined.

• Perturbing slightly the optimal trajectory as before and the optimal final time to $T = T^* + \epsilon \tau$ with some $\tau \in \mathbb{R}$, we have

$$J(x,T) = J(x,T^*) + \int_{T^*}^{T^* + \epsilon \tau} f(x(t),\dot{x}(t))dt$$

The first term has already been treated before, providing the first order approximation

$$J(x, T^*) = J(x^*, T^*) + \epsilon \int_0^{T^*} \left\{ \frac{\partial f}{\partial x} - \frac{d}{dt} \frac{\partial f}{\partial \dot{x}} \right\}' \xi dt + \epsilon \frac{\partial f'}{\partial \dot{x}} \xi \Big|_0^{T^*} + \mathcal{O}(\epsilon^2)$$

- Clearly $\xi(0) = 0$ since $x(0) = x^*(0) = 0$.
- However, the perturbed trajectory $x(t) = x^*(t) + \epsilon \xi(t)$, calculated at $t = T^*$ does not imply that $\xi(T^*) = 0$. Indeed, we can not impose $x(T^*) = x_T$ but we must impose that $x(T) = x(T^* + \epsilon \tau) = x_T$, that is

$$x(T^* + \epsilon \tau) = x(T^*) + \epsilon \dot{x}(T^*)\tau + \mathcal{O}(\epsilon^2)$$

= $x^*(T^*) + \epsilon(\xi(T^*) + \dot{x}^*(T^*)\tau) + \mathcal{O}(\epsilon^2)$

which from the fact that $x^*(T^*) = x_T$ requires that

$$\xi(T^*) = -\dot{x}^*(T^*)\tau$$

Putting these relations all together we have

$$J(x,T) = J(x^*,T^*) + \epsilon \int_0^{T^*} \left\{ \frac{\partial f}{\partial x} - \frac{d}{dt} \frac{\partial f}{\partial \dot{x}} \right\}' \xi dt +$$

$$+ \epsilon \left\{ f(x^*(T^*),\dot{x}^*(T^*)) - \frac{\partial f'}{\partial \dot{x}} \dot{x}^*(T^*) \right\} \tau + \mathcal{O}(\epsilon^2)$$

• Now, taking into account that the function $\xi(t)$, $\forall t \in [0, T^*]$ and the scalar $\tau \in \mathbb{R}$ are arbitrary, we obtain the next result by setting to zero the first order term of the development.

• The next theorem provides a necessary condition for optimality

Theorem (7)

A more general problem

If the pair (x(t), T) satisfying $x(0) = x_0$, $x(T) = x_T$ and $T \ge 0$ is a local minimum then it satisfies the Euler-Lagrange equation

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

and the so called transversality condition

$$f(x(T), \dot{x}(T)) - \frac{\partial f'}{\partial \dot{x}} \dot{x}(T) = 0$$

 Compared to Theorem (6), the present one has one more constraint. It is necessary since now the final time T > 0 is an additional variable to be determined.

• **Example 3**: We want to solve with x(0) = 0 and x(T) = 1

$$\min_{x \in \mathcal{X}, T \ge 0} \int_0^T (1 + x^2 + \dot{x}^2) dt$$

Applying the Euler-Lagrange equation we obtain $\ddot{x} - x = 0$ which yields

$$x(t) = \frac{\operatorname{senh}(t)}{\operatorname{senh}(T)}, \ \forall t \in [0, T]$$

while the transversality condition $\dot{x}(T) = \sqrt{2}$ allows us to determine the optimal final time as

$$\tanh(T) = \frac{1}{\sqrt{2}} \Longrightarrow T \approx 0.9$$

Discussion

• In the present case we have to handle trajectories $x(t):[0,T]\to\mathbb{R}^n$. For x(t) and $\xi(t)$ given we can evaluate the first order variation of $J(x + \epsilon \xi)$ in the general form

$$\delta J(x,\xi) = \lim_{\epsilon \to 0} \frac{J(x+\epsilon\xi) - J(x)}{\epsilon}$$

Hence, an extremal trajectory x^* satisfying the necessary condition for optimality is obtained from

$$\delta J(x^*,\xi) = 0, \ \forall \xi(t), \forall t \in [0,T]$$

This was the strategy used previously.

Discussion

• Once a candidate for minimum x^* has been obtained, we can go further to determine

$$\delta^2 J(x^*, \xi) = \lim_{\epsilon \to 0} \frac{J(x^* + \epsilon \xi) - J(x^*)}{\epsilon^2}$$

which characterizes a local minimum whenever

$$\delta^2 J(x^*, \xi) > 0, \ \forall \xi(t) \neq 0, \forall t \in [0, T]$$

This sufficient optimality test is accomplished by Taylor's series development. In general, the determination of $\delta^2 J(x^*, \xi)$ is not difficult but it is very time consuming.

Suggested problems

• **Problem 1:** Provide the necessary conditions for optimality of the problem

$$\min_{x \in \mathcal{X}} \int_0^T f(x(t), \dot{x}(t)) dt + g(x(T))$$

where $x(0) = x_0$, x(T) is free and T > 0 is given.

• **Problem** 2: Solve the problem

$$\min_{x \in \mathcal{X}} \int_0^1 \left(x(t)^2 + \dot{x}(t)^2 \right) dt$$

subject to x(0) = 1 and x(1) = 0.

Suggested problems

• Problem 3: Solve the problem

$$\min_{x \in \mathcal{X}} \int_0^1 (x(t)^2 + \dot{x}(t)^2) dt + 2x(1)$$

subject to x(0) = 1.

Problem 4: Solve the problem

$$\min_{x \in \mathcal{X}} \int_0^1 (x(t)^2 + \dot{x}(t)^2) dt + x(1)^2$$

subject to x(0) = 1.

Suggested problems

• Problem 5: Solve the problem

$$\min_{x \in \mathcal{X}} \int_0^T \left(x(t)^2 + \dot{x}(t)^2 \right) dt$$

subject to x(0) = 1, x(T) = 1 and T > 0 free.

• **Problem** 6: Solve the problem

$$\min_{x \in \mathcal{X}} \int_0^T \left(x(t)^2 - \dot{x}(t)^2 \right) dt$$

subject to x(0) = 1, x(T) = 1 and T > 0 free.