

Lecture 20 *

December 5, 2012

1 Pontryagin Maximum Principal (First Version)

Theorem: Pontryagin Maximum Principal-First Version suppose it is desired to steer the state of

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

so as to maximize (minimize)

$$J = \int_0^T l(x, u) dt + \psi(x(T))$$

If this defines a normal problem then there exists a *costate* vector $\lambda(t)$ such that λ , together with the optimizing values of x, u jointly satisfy

$$\dot{x} = \frac{\delta H}{\delta \lambda}$$

$$\lambda(T) = \frac{\delta \psi}{\delta x}(x(T))$$

$$\frac{\delta H}{\delta u} = 0$$

Where the Hamiltonian H is defined as

$$H(\lambda, X, u) = \lambda^T F(x, u) + l(x, u)$$

Consider the same problem, optimization, in which the terminal value of $x(T)$ is (partially) specified

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

*This work is being done by various members of the class of 2012

and we wish to optimize

$$J = \psi(x(T)) + \int_0^T l(x(t), u(t)) dt$$

with specific terminal values $X_1(T), \dots, X_r(T)$ ($r \leq n$)

The approach for the solution again involves infinite dimensional Lagrange Multipliers

$$\hat{J} = \psi(x, t) + \int_0^T H(\lambda(t), x(t), u(t)) - \lambda(t)^T \dot{x}(t) dt$$

with

$$H(\lambda, x, u) = \lambda^T f(x, u) + l(x, u).$$

suppose

$$x(T) = \begin{pmatrix} x_1(T) \\ \cdot \\ \cdot \\ \cdot \\ x_r(T) \\ x_n(T) \end{pmatrix}$$

with $X_1(T), \dots, X_r(T)$ specified.

In order to first orderize J

$$\delta \bar{J} = \frac{\delta \psi}{\delta x} (x^*(T) - \lambda(T))^T \delta x(T) + \int_0^T \frac{\delta H}{\delta x} (\lambda, x^*, u^*) + \dot{\lambda}(t)^T \delta x(t) dt + \int_0^T \frac{\delta H}{\delta x} (\lambda, x^*, u^*) \delta u(t) dt$$

As in the last lecture we choose $\lambda(t)$ to satisfy $\dot{\lambda}(t) = \frac{-\delta H}{\delta x}$ making the 2nd term in the above equation vanish since $\delta_{x_j}(T) = 0$ for $i = 1, \dots, r$. The corresponding costate component $\lambda_j(T)$ are not constrained so now the P.M.P. becomes

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

$$\dot{\lambda} = \frac{-\delta f^T}{\delta x} \lambda(t) - \frac{\delta l}{\delta x}$$

$$\lambda_{r+1}(T) = \frac{\delta \psi}{\delta x} (x(T))_{r+1}$$

.....

$$\lambda_n(T) = \frac{\delta \psi}{\delta x} (x(T))_n$$

$$\frac{\delta H}{\delta U} = 0,$$

$$X_1(T) = x_1$$

$$X_2(T) = X_2$$

↓

$$X_r(T) = X_r$$

Example Finite Dimensional LTI system

We would like to steer

$$\dot{x} = Ax + Bu$$

from $x(0) = x_0$ to $x(T) = x_1$ so as to minimize

$$J = \frac{1}{2} \int_0^T \|u(t)\|^2 dt$$

To solve, first let us write down the Hamiltonia

$$H(\lambda, x, u) = \lambda^T Ax + \lambda^T Bu + \frac{1}{2} \|u\|^2$$

the P.M.P. necessary conditions say that there is a *costate* $\lambda(t)$, which together with the optimizing values of "x" and "u" satisfy

$$\dot{\lambda} = -A^T \lambda(t)$$

$$B^T \lambda(t) + u(t) = 0$$

$$\dot{x} = Ax + Bu$$

From the first equation $\lambda(t) = e^{-A^T t} \lambda_0$ for some $\lambda_0 \in \mathbb{R}^n$

$$u(t) = -B^T \lambda(t)$$

$$= -B^T e^{-A^T t} \lambda_0$$

Now λ_0 must be chosen s.t. the boundary conditions $x(0) = x_0$, $x(T) = x_1$ are met, so from the variation of constants formula

$$x(t) = e^{At} x_0 - \int_0^T e^{A(t-s)} B B^T e^{-A^T s} ds \lambda_0$$

having the boundary conditions met requires that

$$x(t) = e^{At}(x_0 - \int_0^T e^{-A(s)}BB^T e^{-A^T s} ds \lambda_0)$$

In other words

$$\int_0^T e^{-A(s)}BB^T e^{-A^T s} ds \lambda_0 = (x_0 - e^{-AT} x_1)$$

Hence,

$$\lambda_0 = W^{-1}(x_0 - e^{-AT} x_1)$$

where

$$W = \int_0^T e^{-A(s)}BB^T e^{-A^T s} ds$$

Note: That W is an invertible n x n matrix if the system is controllable.

Summary: The control law that steers $\dot{x} = Ax + Bu$ from $x(0) = x_0$ to $x(T) = x_1$ so as to minimize the norm of control squared is integrated

$$\int_0^T \|u(t)\|^2 dt$$

Is given by

$$u(t) = -B^T e^{-A^T t} \lambda_0$$

where λ_0 is the solution to $x_0 - e^{-AT} x_1 = W \lambda_0$ with

$$W = \int_0^T e^{-A(s)}BB^T e^{-A^T s} ds$$

Other linear quadratic Optimization problems:

Free Endpoint Problem (not specifying the final endpoint)

$$\dot{x} = Ax + Bu$$

steer this system so as to minimize

$$J = \frac{1}{2} \int_0^T x^T(t)Qx(t) + u^T(t)Ru(t) dt$$

Where R is symmetric and positive definite, and Q is symmetric and positive semi-definite.

For this problem the Hamiltonian is $\lambda^T Ax + \lambda^T Bu + \frac{1}{2}u^T Ru$ the P.M.P necessary conditions are

$$x = A^T \lambda - Qx; \lambda(T) = 0$$

$$\dot{x} = -A^T \lambda - Qx$$

$$Ru + B^T \lambda = 0$$

R is invertable because it is positive definite $u(t) = -R^{-1}B^T \lambda(t)$

$$\dot{x} = Ax - BR^{-1}B^T \lambda$$

$$\dot{\lambda} = -A^T \lambda - Qx$$

$$x(0) = x_0, \lambda(T) = 0 \leftarrow \text{from P.M.P}$$

suppose we can write

$$\lambda(T) = M(t)x(t)$$

$$\dot{\lambda} = M\dot{M}x + M\dot{x}$$

from x equation

$$= \dot{M}x + MA - MBR^{-1}B^T Mx$$

from the λ equation

$$= -A^T Mx - Qx$$

This will be satisfied for any X(t) if M(t) satisfies the matrix Riccati equation

$$\dot{M} + A^T M + MA - MBR^{-1}B^T M + Q = 0(R)$$

To meet the boundary conditions $\lambda(T) = 0$ we require that $M(t)=0$

The question of Existence of M

Consider the Scalar Riccati equation

$$\dot{M} = -2M + M^2 + 1$$

$$M(T) = 0$$

(This would be associated with Optimization)

$$\dot{x} = x + u$$

$$J = \int_0^T u^2(t) - x^2(t) dt$$

note: $\dot{m} = (m - 1)^2$ and get

$$\dot{k} = k^2$$

This is a solution of the form

$$k(t) = \frac{\alpha}{1 - \alpha t}$$

$$\dot{k}(t) = \frac{\alpha^2}{(1 - \alpha t)^2} = k(t)^2$$

One can prove that the solution is unique. Choose α s.t. $M(t) = 0$ i.e. $K(T) = -1$ then

$$\frac{\alpha}{1 - \alpha T} = -1 \rightarrow \alpha = \frac{1}{T - 1}$$

For example when $T=2$ and $\alpha = 1$ and $k(t) = \frac{1}{1-t}$

$$M(t) = \frac{1}{1-t}$$

Which blows up at $t = 1$

So the quadratic is not always the best option since it has a finite blow up times

We state the following without proof

Theorem: If A, B is a controllable pair, $Q = Q^T$ is positive semi-definite, then the solution to the Riccati equation (T) passing through 0 at $t=T$ exists and is positive semi-definite on an interval $[t_0, T]$ for any $t_0 < T$

Preview of the next time

Theorem: Suppose A,B is a controllabe pair. The conrol law that steers

$$\dot{x} = Ax + Bu$$

so as to minimize

$$\int_0^T x^T Qx + u^T R u dt$$

is given by

$$u(t) = -R^{-1}B^T M(t)x(t)$$

where M satisfies

$$\dot{M} + MA + A^T M - MBR^{-1}B^T M + Q = 0$$

$$M(T) = 0$$