

# Notes in Dynamics and Manuevering

## ECEN 5008

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November 10, 2015

### 1 explaining some of the diff eqs from paper

$$\min \int_0^T l(x(\tau), u(\tau), \tau) d\tau + m(x(T))$$

such that  $\dot{x} = f(x, u, t), x(0) = x_0$ .

$\min(f(x))$  Thm:  $x^*$  is a local min  $\implies \nabla f(x^*) = 0$ . First order necessary condition.

assume these functions are differentiable.

Pontryagin's maximum (or minimum) principle.

Baby PMP THM:

Suppose that  $(\bar{x}(t), \bar{u}(t)), t \in [0, T]$  satisfies the dynamics ( $\dot{x} = f(x, u, t)$ ) and locally minimizes the cost ( $\int_0^T l d\tau + m$ ) (over trajectory).

THM: there is an absolutely continuous curve  $\bar{p}(t), t \in [0, T]$  such that:

$$\begin{aligned}\dot{\bar{x}}(t) &= H_p^T(\bar{x}(t), \bar{u}(t), \bar{p}(t), t), \bar{x}(0) = x_0 \\ \dot{\bar{p}}(t) &= -H_x^T(\bar{x}(t), \bar{u}(t), \bar{p}(t), t), \bar{p}(T) = m_x^T(x(T)) \\ 0 &= H_u^T(\bar{x}(t), \bar{u}(t), \bar{p}(t), t)\end{aligned}$$

where the control Hamiltonian is:

$$H(x, u, p, t) = l(x, u, t) + p^T f(x, u, t)$$

Written more directly,

$$\begin{aligned}\dot{\bar{x}}(t) &= f(\bar{x}(t), \bar{u}(t), t), \bar{x}(0) = x_0 \\ \dot{\bar{p}}(t) &= -A^T(\bar{x}(t), \bar{u}(t), t)\bar{p}(t) - l_x^T(\bar{x}(t), \bar{u}(t), t), \bar{p}(T) = m_x^T(x(T)) \\ 0 &= B^T(\bar{x}(t), \bar{u}(t), t)\bar{p}(t) + l_u^T(\bar{x}(t), \bar{u}(t), t)\end{aligned}$$

two point boundary value problem (TPBVP)

BVP 4c and BVP 5c solvers will do this, but need a good initial guess to use properl.

need x to be linear and cost to be quadratic.

### 2 Linear Dynamics, Quadratic Cost

(LQ opt ctrl LQR)

$$\begin{aligned}\min(\int_0^T \frac{1}{2}x^T Q x + \frac{1}{2}u^T R u d\tau + \frac{1}{2}x(T)^T P_1 x(T)) \\ Q = Q^T \geq 0\end{aligned}$$

$$R = R^T > 0$$

$$P_1 = P_1^T \geq 0$$

take to be constant, but they don't have to be.  
All of this such that:

$$\begin{aligned}\dot{x} &= A(t)x + B(t)u, x(0) = x_0 \\ H(x, u, p, t) &= \frac{1}{2}x^T Q x + \frac{1}{2}u^T R u + p^T (Ax + Bu) \\ \dot{x} &= H_p^T = Ax + Bu \\ -\dot{p} &= H_x^T = A^T p + Qx \\ 0 &= H_u^T = B^T p + Ru \implies u = -R^{-1}B^T p \\ \begin{bmatrix} \dot{x} \\ \dot{p} \end{bmatrix} &= \begin{bmatrix} A & -BR^{-1}B^T \\ -Q & -A^T \end{bmatrix} \begin{bmatrix} x \\ p \end{bmatrix} \\ x(0) &= x_0 \\ p(T) &= P_1 x(T) \\ \begin{bmatrix} x(T) \\ p(T) \end{bmatrix} &= \begin{bmatrix} I \\ P_1 \end{bmatrix} x(T) \\ p(T) &= P_1 x(T)\end{aligned}$$

maybe  $p(t) = P(t)x(t)$ (if we are lucky). Because of the specifications on definiteness of Q,R,P this is true here.

He likes to call this the Ricatti Transformation.

$$\begin{aligned}p &= Px \\ \dot{p} &= \dot{P}x + P\dot{x} = \dot{P}x + P(Ax - BR^{-1}B^T Px) \\ &= -Qx - A^T Px \\ 0 &= (\dot{P} + A^T P + PA - PBR^{-1}B^T P + Q)x\end{aligned}$$

this is a ricotti equation.

If  $P(t), t \in [0, T]$  satisfies:

$$\dot{P} + A^T P + PA - PBR^{-1}B^T P + Q = 0 \quad P(T) = P_1$$

Then,  $p(t) = P(t)x(t) \forall t \in [0, T]$ .

and  $u = -R^{-1}B^T Px = -K(t)x$ .

so that the optimal trajectories one obtained using a linear feedback.

$$\dot{x} = (A - Bk)x, x(0) = x_0$$

### 3 Minimize this optimization

$$\min \int_0^T a^T(\tau)z(\tau) + b^T(\tau)v(\tau) + \frac{1}{2} \begin{bmatrix} z(\tau) \\ v(\tau) \end{bmatrix}^T \begin{bmatrix} Q(\tau) & S(\tau) \\ S^T(\tau) & R(\tau) \end{bmatrix} \begin{bmatrix} z(\tau) \\ v(\tau) \end{bmatrix} d\tau + \frac{1}{2} z(T)^T P_1 z(T) + r_1^T z(T)$$

$$\dot{z} = A(t)z + B(t)v, z(0) = 0$$

$$a(t) = l_x^T(t)$$

$$b(t) = L_u^T(t)$$

$$-\dot{r} == (A - BK_0)^T r + a - K_0^T b = A^T r + a - K_0^T (B^T r + b)$$

For the cheat sheet, we are at a trajectory, compute some minimums to figure out what is going downhill.

### 4 Newton Method (Pronto):

start with initial trajectory:  $\epsilon \in \mathfrak{T}$

for  $k = 0, 1, 2, \dots$

redesign  $k(t)$ .

descent direction:

$$\zeta_k = \arg \min_{\zeta \in T_{\epsilon_k} \mathfrak{T}} Dh(\epsilon_k)\zeta + \frac{1}{2} D^2 g(\epsilon_k) * (\zeta_1 \zeta)$$

or  $\frac{1}{2} \|\zeta\|_{L^2(\tilde{Q}, \tilde{R})}^2$

or  $\frac{1}{2} D^2 h(\epsilon_{12})(\zeta_1, \zeta)$ .

line search (Aruijo backtracking)

$$\gamma_{12} = \arg \min_{\gamma \in [0,1]} h(\mathfrak{P}(\epsilon_k \gamma \zeta_k))$$

update:

$$\epsilon_{k+1} = \mathfrak{P}(\epsilon_k + \gamma_k \zeta_k)$$

$$\int_x (\tau) z(\tau) + l_u(\tau) v(t) d\tau + m_x(x(T)) * z(T)$$