

Numerical Methods / Indirect Methods

Note Title

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Neighboring Extremal Algorithms

Case: Some state variables fixed at fixed t_f

$$\min J = \Phi[x(t_f)] + \int_{t_0}^{t_f} L(x, u, t) dt$$

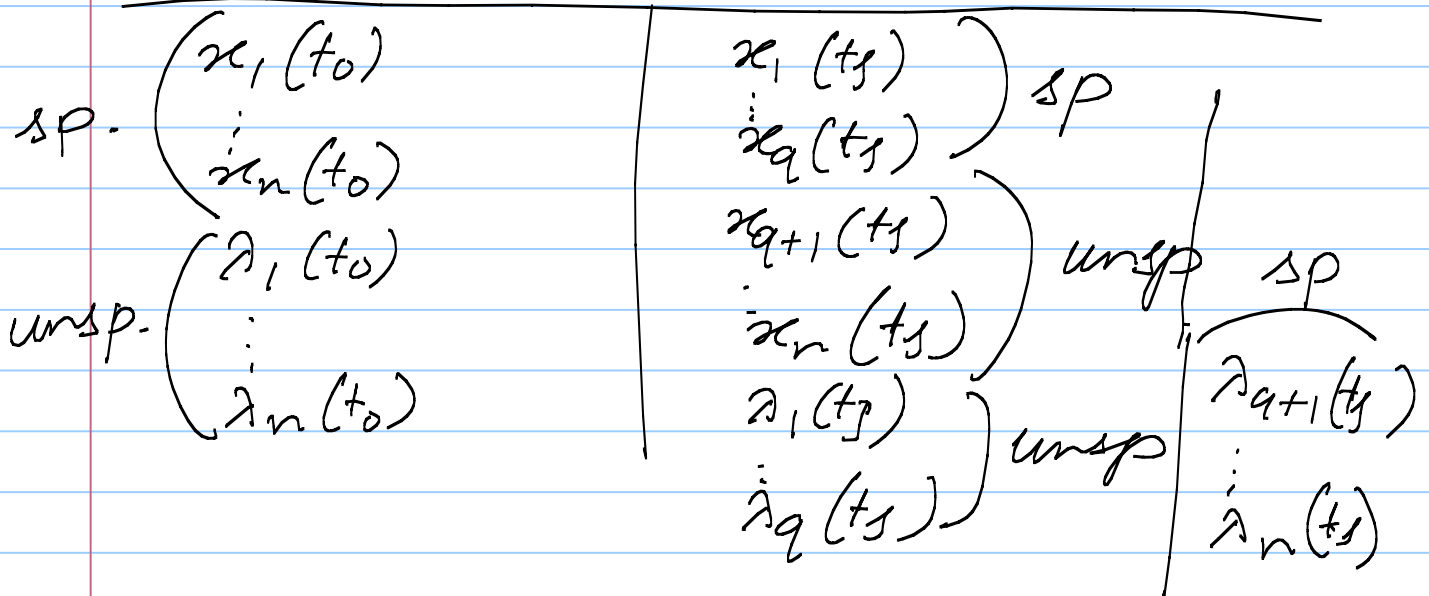
where $\dot{x} = f(x, u, t)$
 $x(t_0) \leftarrow$ fixed $\left| \begin{array}{l} t_0, t_f \\ \rightarrow \text{fixed} \end{array} \right.$

$$\Psi[x(t_f)] = \begin{bmatrix} x_1(t_f) - x_1^f \\ \vdots \\ x_q(t_f) - x_q^f \end{bmatrix} = 0$$

E-L eqns: $\dot{\lambda}^T = - \frac{\partial L}{\partial x} - \lambda^T \frac{\partial f}{\partial x}$

$$\lambda_j^0(t_f) = \left(\frac{\partial \Phi}{\partial x_j} \right)_{t=t_f} \quad j = q+1, \dots, n$$

$$0 = \frac{\partial L}{\partial u} + \lambda^T \frac{\partial f}{\partial u}$$



Step 1: Guess $\lambda_1(t_0), \dots, \lambda_n(t_0)$

Step 2: Integrate
$$\dot{x}_i = f(x, u, t)$$
$$\dot{\lambda}^T = -\frac{\partial H}{\partial x}$$

forward from t_0 to t_f using
 $\frac{\partial H}{\partial u} = 0$ to determine $u(t)$ at
each step.

Step 3: Calculate $[x_1(t_f) \dots x_q(t_f)]$
and $[\lambda_{q+1}(t_f) \dots \lambda_m(t_f)]$

Step 4: Determine the $n \times n$
transition matrix $\frac{\partial \mu(t_f)}{\partial \lambda(t_0)}$, where

def
$$\delta \mu(t_f) = \begin{bmatrix} \delta x_1(t_f) \\ \vdots \\ \delta x_q(t_f) \\ \delta \lambda_{q+1}(t_f) \\ \vdots \\ \delta \lambda_m(t_f) \end{bmatrix} = \frac{\partial \mu(t_f)}{\partial \lambda(t_0)} \delta \lambda(t_0)$$

↓
② How to compute this? → later

Step 5: Choose $\delta \mu(t_f) = -\varepsilon [\mu(t_f) - \mu^*]$
 $0 \leq \varepsilon \leq 1$

Step 6: For chosen $\delta \mu(t_f)$, find
$$\delta \lambda(t_0) = \left[\frac{\partial \mu(t_f)}{\partial \lambda(t_0)} \right]^{-1} \delta \mu(t_f)$$

Step 7: Using $\lambda(t_0)_{\text{new}} = \lambda(t_0)_{\text{old}} + \delta\lambda(t_0)$

repeat steps 1 to 7 until $\mu(t_f)$ has the specified values.

Q) How to calculate $\frac{\partial \mu(t_f)}{\partial \lambda(t_0)}$?

Many methods are possible. The simplest one is:

→ Direct numerical integration:
Integrate n -times $\left. \begin{array}{l} \dot{x}_i = f(x, u, t) \\ \dot{\lambda}^T = -\frac{\partial H}{\partial x} \end{array} \right\} \text{ using } \frac{\partial H}{\partial u} = 0$

→ For each int. change one component of $\lambda_i(t_0)$ by a small amount from the original guess in step 1.
→ Compute $\delta\mu(t_f)$ and divide by $\delta\lambda_i(t_0)$.

Functions of state variables specified at free t_f

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

$$\left. \begin{array}{l} \dot{x}_i = f(x, u, t) \\ x(t_0) = x^0, t_0 \in \text{fixed} \\ \Psi[x(t_f), t_f] = 0 \end{array} \right\} t_f \in \underline{\underline{\text{free}}}$$

$$\dot{\lambda}^T = - \left[\frac{\partial H}{\partial x} \right] \quad 0 = \frac{\partial H}{\partial u}$$

$$\lambda^T(t_f) = \frac{\partial \Phi}{\partial x} \Big|_{t=t_f} \quad \left. \begin{array}{l} \\ \\ \end{array} \right\} \underline{\Phi} = \Phi + \lambda^T \psi$$

$$\Omega = \left(\frac{\partial \Phi}{\partial t} + \frac{\partial \Phi}{\partial x} f + L \right)_{t=t_f} = 0$$

Step 1: Guess $x(t_f)$ (n values), v (q values) and t_f

Step 2: Calculate 1) $\psi(x(t_f), t_f)$ — directly
2) $\lambda(t_f)$ from $\lambda^T(t_f) = \frac{\partial \Phi}{\partial x} \Big|_{t=t_f}$

3) $\Omega[x(t_f), u(t_f), v, t_f]$

This $u(t_f)$ is calculated from $\frac{\partial H}{\partial u} = 0$ (using $\lambda(t_f)$)

Step 3: $\left. \begin{array}{l} \dot{x} = f(x, u, t) \\ \dot{\lambda}^T = - \frac{\partial H}{\partial x} \end{array} \right\} \text{Integrate backwards from } t_f \text{ to } t_0 \text{ using}$
1) $u(t)$ from $\frac{\partial H}{\partial u} = 0$
2) $x(t_f) + \lambda(t_f) \leftarrow$ from steps 1 & 2

Step 4: Calculate $x(t_0)$

Step 5: Calculate $[(n+q+1) \times (n+q+1)]$ transition matrix $\frac{\partial [x(t_0), \psi, \Omega]}{\partial [x(t_f), v, t_f]}$ where

$$\begin{bmatrix} \delta x(t_0) \\ d\psi \\ d\Omega \end{bmatrix} = \frac{\partial [x(t_0), \psi, \Omega]}{\partial [x(t_f), v, t_f]} \begin{bmatrix} \delta x(t_f) \\ dv \\ dt_f \end{bmatrix}$$

Step 6: Choose $\delta x(t_0)$, $d\psi$, $d\Omega$ so as to bring the next state closer to $x(t_0)$, $\psi = 0$, $\Omega = 0$.

$$\begin{bmatrix} \delta x(t_0) \\ d\psi \\ d\Omega \end{bmatrix} = -\epsilon \begin{bmatrix} x(t_0) - x^0 \\ \psi [x(t_f), t_f] \\ \Omega [x(t_f), t_f] \end{bmatrix} \quad 0 < \epsilon \leq 1$$

Step 7: With chosen values of $\delta x(t_0)$, $d\psi$ & $d\Omega$, insert the transition matrix to calculate $\delta x(t_f)$, dv , dt_f

Step 8: Using $\begin{bmatrix} x(t_f) \\ v \\ t_f \end{bmatrix}_{\text{new}} = \begin{bmatrix} x(t_f) \\ v \\ t_f \end{bmatrix}_{\text{old}} + \begin{bmatrix} dx(t_f) \\ dv \\ dt_f \end{bmatrix}$

$$dx(t_f) = \delta x(t_f) + \dot{x}(t_f) dt_f$$

repeat steps 1-8 until $x(t_f) = x^0$
 $\psi [x(t_f), t_f] = 0$ & $\Omega(\dots) = 0$

First Order Gradient Algorithms:

Some state variable specified at fixed terminal time:

Step 1: Guess $u(t)$

Step 2: Integrate $\dot{x} = f(x, u, t)$ forward
 with $x(t_0)$ & $u(t)$ from step 1.
 Record $x(t)$, $u(t)$, $\psi(x(t_f))$

Step 3: Backward integrate the following eqns:

$$\dot{p}^T = -\frac{\partial L}{\partial x} - p^T \frac{\partial f}{\partial x} \quad \Bigg| \quad p_i(t_f) = \begin{cases} 0 & i=1, \dots, q \\ \frac{\partial \psi}{\partial x_i} & i=q+1, \dots, n \end{cases} \Bigg|_{t=t_f}$$

$$\dot{R}^T = -R^T \left[\frac{\partial f}{\partial u} \right] \quad R_{ij}^o(t_f) = \begin{cases} 1 & i=j \\ 0 & i \neq j \end{cases}$$

$i=1, \dots, n$
 $j=1, \dots, q$

Step 4: Simultaneously with step 3, compute the following integrals:

$$I_{\psi\psi} = \int_{t_0}^{t_f} R^T \frac{\partial f}{\partial u} W^{-1} \left(\frac{\partial f}{\partial u} \right)^T R dt \quad (q \times q)$$

$$I_{J\psi} = I_{\psi J}^T = \int_{t_0}^{t_f} \left(p^T \frac{\partial f}{\partial u} + \frac{\partial L}{\partial u} \right) W^{-1} \left(\frac{\partial f}{\partial u} \right)^T R dt$$

(q-row vector)

$$I_{JJ} = \int_{t_0}^{t_f} \left(p^T \frac{\partial f}{\partial u} + \frac{\partial L}{\partial u} \right) W^{-1} \left[\left(\frac{\partial f}{\partial u} \right)^T p + \left(\frac{\partial L}{\partial u} \right)^T \right] dt$$

↪ scalar

where $W(t)$ is a $(m \times m)$ positive definite weighting matrix.

Step 5: Choose $\delta\psi$ to cause next
 solⁿ to be closer to $\psi[x(t_f)] = 0$
 $\rightarrow \delta\psi = -\epsilon \psi[x(t_f)]$, $0 < \epsilon \leq 1$
 \rightarrow Then calculate v from

$$v = - [I_{\psi\psi}]^{-1} (\delta\psi + I_{\psi J})$$

Step 6: Repeat steps 1-6, using
 improved $(u(t))_{\text{new}} = (u(t))_{\text{old}} + \delta u(t)$
 where

$$\delta u(t) = - [W(t)]^{-1} \left[\frac{\partial L}{\partial u} + [P(t) + R(t)v]^T \frac{\partial f}{\partial u} \right]^T$$

Stop when $\psi[x(t_f)] = 0$ and

$$I_{JJ} - I_{J\psi} I_{\psi\psi}^{-1} I_{\psi J} = 0$$