

# Static and Dynamic Optimization (42111)

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Lecture 9: End Point constraints

## Outline of lecture

- Recap F8
- Solution to Free C problem
- Simple EPC
- Simple partial EPC
- Linear EPC
- General EPC
- Continuous time DO with EPC
- Reading guidance (DO chapter 3).

## Dynamic Optimization (D, free)

Find a sequence  $u_i, i = 0, \dots, N - 1$  which takes the system

$$x_{i+1} = f_i(x_i, u_i) \quad x_0 = \underline{x}_0$$

from its initial state  $\underline{x}_0$  along a trajectory such that the performance index

$$J = \phi_N[x_N] + \sum_{i=0}^{N-1} L_i(x_i, u_i)$$

is optimized. Define the Hamiltonian function as:

$$H_i = L_i(x_i, u_i) + \lambda_{i+1}^T f_i(x_i, u_i)$$

Then the Euler-Lagrange equations are:

$$x_{i+1} = f_i(x_i, u_i) \quad \lambda_i^T = \frac{\partial}{\partial x_i} H_i$$

$$0 = \frac{\partial}{\partial u_i} H_i$$

with boundary conditions:

$$x_0 = \underline{x}_0 \quad \lambda_N^T = \frac{\partial}{\partial x_N} \phi_N(x_N)$$

## Dynamic Optimization (C, free)

Find a function  $u_t, t \in [0; T]$  which takes the system

$$\dot{x}_t = f_t(x_t, u_t) \quad x_0 = \underline{x}_0$$

from its initial state  $\underline{x}_0$  along a trajectory such that the performance index

$$J = \phi_T[x_T] + \int_0^T L_t(x_t, u_t) dt$$

is optimized. Define the Hamilton function as:

$$H_t(x_t, u_t, \lambda_t) = L_t(x_t, u_t) + \lambda_t^T f_t(x_t, u_t)$$

Then the Euler-Lagrange equations are:

$$\dot{x}_t = f_t(x_t, u_t) \quad -\dot{\lambda}_t^T = \frac{\partial}{\partial x_t} H_t$$

$$0 = \frac{\partial}{\partial u_t} H_t$$

with boundary conditions:

$$x_0 = \underline{x}_0 \quad \lambda_T^T = \frac{\partial}{\partial x_T} \phi_T(x_T)$$

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# Solutions for the C problem

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Type of solutions:

- Analytical solutions (for very simple problems)
- Semi analytical solutions (eg. the LQ problem)
- numerical solutions

$$H_t = L_t(x, u) + \lambda_t f_t(x, u)$$

## Euler-Lagrange Equations I

$$\dot{x} = f_t(x_t, u_t) \quad x_0 = \underline{x}_0$$

$$-\dot{\lambda}^T = \frac{\partial}{\partial x} H_t \quad \lambda_T^T = \frac{\partial}{\partial x} \phi_T(x_T)$$

$$0 = \frac{\partial}{\partial u} H_t$$

## Euler-Lagrange Equations II

$$\dot{x} = f_t(x_t, u_t)$$

$$-\dot{\lambda}^T = \frac{\partial}{\partial x_t} L_t(x, u) + \lambda^T \frac{\partial}{\partial x} f_t(x, u)$$

$$0 = \frac{\partial}{\partial u} L_t(x, u) + \lambda^T \frac{\partial}{\partial u} f_t(x, u)$$

## Costate equation

$$\dot{\lambda} = -\left[\frac{\partial}{\partial x} H_t\right]^T = g_t(x, \lambda, u)$$

## Stationarity equation

$$u_t = h_t(x, \lambda)$$

Guess  $\lambda_0$  and use the knowledge  $x_0$  and integrate (use e.g. ode45)

$$\frac{d}{dt} \begin{bmatrix} x \\ \lambda \end{bmatrix} = \begin{bmatrix} f_t(x, u) \\ g_t(x, \lambda, u) \end{bmatrix} \quad u_t = h_t(x, \lambda)$$

i.e.

$$\frac{d}{dt} \begin{bmatrix} x \\ \lambda \end{bmatrix} = \begin{bmatrix} \underline{f}_t(x, \lambda) \\ \underline{g}_t(x, \lambda) \end{bmatrix}$$

At the end check the condition:

$$\lambda_T^T = \frac{\partial}{\partial x_T} \phi_T(x_T)$$

Use e.g. fsolve to adjust  $\lambda_0$  such that the condition is satisfied.

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# End point constraints (EPC)

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Find a sequence  $u_i$ ,  $i = 0, \dots, N - 1$  which takes the system

$$x_{i+1} = f_i(x_i, u_i) \quad x_0 = \underline{x}_0$$

from its initial state,  $\underline{x}_0$ , along a trajectory to

$$x_N = \underline{x}_N \quad (\text{Simple EPC})$$

such that the performance index

$$J = \phi_N[x_N] + \sum_{i=0}^{N-1} L_i(x_i, u_i)$$

is optimized.

In general:

$$\psi_N(x_N) = 0 \qquad \psi : \mathbb{R}^{n+1} \rightarrow \mathbb{R}^p \qquad p \leq n + 1$$

Linear EPC

$$Cx_N = \underline{r} \qquad \text{e.g.} \quad C = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \end{bmatrix} \qquad r = \begin{bmatrix} 1.4 \\ 2.3 \end{bmatrix}$$

Simple partial EPC

$$x_N = \begin{bmatrix} \tilde{x}_N \\ \bar{x}_N \end{bmatrix} \qquad \tilde{x}_N = \underline{\tilde{x}}_N \in \mathbb{R}^p \qquad p \leq n$$

# Investment planning

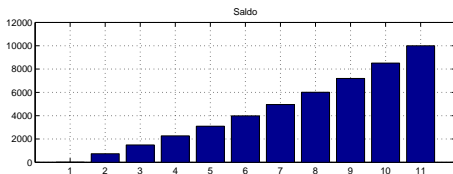
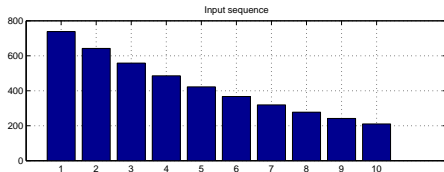
**Plan:** During a period of time ( $N$  intervals) to invest a amount of money  $u_i$  to obtain a specified sum ( $\underline{x}_N$ ) at the end of the period.

Dynamics:

$$x_{i+1} = (1 + \alpha)x_i + u_i \quad x_0 = 0 \quad x_N = 10.000 \text{ Dkr}$$

Objective:

$$\text{Min } J \quad J = \sum_{i=0}^{N-1} \frac{1}{2} u_i^2$$



Consider the discrete time system (for  $i = 0, 1, \dots, N - 1$ )

$$x_{i+1} = f_i(x_i, u_i) \quad x_0 = \underline{x}_0 \quad (1)$$

the performance index

$$J = \phi_N(x_N) + \sum_{i=0}^{N-1} L_i(x_i, u_i) \quad (2)$$

and the simple terminal constraint

$$x_N = \underline{x}_N \quad (3)$$

where  $\underline{x}_N$  (and  $\underline{x}_0$ ) is given. Introduce the **multiplier** (vector with same length as  $x$ )  $\nu$  and form the **Lagrange** function:

$$J_L = \phi_N(x_N) + \lambda_0^T (\underline{x}_0 - x_0) + \nu^T (x_N - \underline{x}_N) + \sum_{i=0}^{N-1} \left[ L_i(x_i, u_i) + \lambda_{i+1}^T (f_i(x_i, u_i) - x_{i+1}) \right]$$

**New conditions:** Stationarity w.r.t.  $x_N$  (for  $i = N - 1$ ) gives:

$$0 = \frac{\partial}{\partial x_N} \phi_N + \nu^T - \lambda_N^T \quad \lambda_N^T = \nu^T + \frac{\partial}{\partial x_N} \phi_N$$

Stationarity w.r.t.  $\nu$  gives

$$x_N = \underline{x}_N$$

The rest is as usual (as for the free case).

# Simple end point constraints

Defining the Hamiltonian function

$$H_i(x_i, u_i, \lambda_{i+1}) = L_i(x_i, u_i) + \lambda_{i+1}^T f_i(x_i, u_i)$$

The Euler-Lagrange equations:

$$x_{i+1} = f_i(x_i, u_i) \quad \lambda_i^T = \frac{\partial}{\partial x} H_i \quad 0 = \frac{\partial}{\partial u} H_i$$

with boundary conditions:

$$x_0 = \underline{x}_0 \quad x_N = \underline{x}_N \quad \lambda_N^T = \nu^T + \frac{\partial}{\partial x_N} \phi_N$$

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Conditions:  $3 \times n$  (of which  $2 \times n$  are trivial and  $n$  are very simple)

Unknowns:  $x_0$ ,  $x_N$  and  $\nu$  (results:  $3 \times n$ )

Conditions on states rather than on costates (for simple EPC). Trade conditions on states for costates.

Let us return to the system

$$x_{i+1} = x_i + u_i$$

The task is to bring the system from the initial position,  $\underline{x}_0$  to a given final position,  $\underline{x}_N$ , in a fixed number,  $N$  of steps, such that the performance index

$$J = \sum_{i=0}^{N-1} \frac{1}{2} u_i^2$$

is minimized. The Hamiltonian function is in this case

$$H_i = \frac{1}{2} u_i^2 + \lambda_{i+1} (x_i + u_i)$$

and the Euler-Lagrange equations are simply (as in the unconstrained situation) :

$$x_{i+1} = x_i + u_i \tag{4}$$

$$\lambda_i = \lambda_{i+1} \tag{5}$$

$$0 = u_i + \lambda_{i+1} \tag{6}$$

with the (slightly changed) boundary conditions:

$$x_0 = \underline{x}_0 \quad x_N = \underline{x}_N \quad \lambda_N = \nu$$

Firstly, we notice that the costates are constant, i.e.

$$\lambda_i = c = \lambda_N = \nu$$

Secondly, from the stationarity condition we have:

$$u_i = -c$$

and inserted in the state equation (4)

$$x_i = x_0 - ic \quad \text{and finally} \quad x_N = x_0 - Nc$$

From the latter equation and boundary condition we can determine the constant to be

$$c = \frac{\underline{x}_0 - \underline{x}_N}{N}$$

Notice, the solution to the Free problem tends to this for  $p \rightarrow \infty$  (and  $\underline{x}_N = 0$ ).

Also notice, the Lagrange multiplier to the terminal conditions is equal

$$\nu = \lambda_N = c = \frac{\underline{x}_0 - \underline{x}_N}{N}$$

and has an interpretation as a shadow price.

## Partial simple end point constraints

Consider the system ( $i = 0, \dots, N - 1$ )

$$x_{i+1} = f_i(x_i, u_i) \quad x_0 = \underline{x}_0 \quad (7)$$

the performance index

$$J = \phi_N(x_N) + \sum_{i=0}^{N-1} L_i(x_i, u_i) \quad (8)$$

and the simple but partial simple terminal constraints

$$x_N = \begin{bmatrix} \tilde{x}_N \\ \bar{x}_N \end{bmatrix} \quad \tilde{x}_N = \underline{\tilde{x}}_N \in \mathbb{R}^p \quad p < n \quad \lambda_N = \begin{bmatrix} \tilde{\lambda}_N \\ \bar{\lambda}_N \end{bmatrix}$$

where  $\underline{\tilde{x}}_N$  (and  $\underline{x}_0$ ) are given. Introduce the **multiplier** (vector)  $\nu$  and form the **Lagrange** function:

$$J_L = \phi_N(x_N) + \lambda_0^T(\underline{x}_0 - x_0) + \nu^T(\tilde{x}_N - \underline{\tilde{x}}_N) + \sum_{i=0}^{N-1} \left[ L_i(x_i, u_i) + \lambda_{i+1}^T(f_i(x_i, u_i) - x_{i+1}) \right]$$

**New conditions:** Stationarity w.r.t.  $x_N$  (i.e.  $\tilde{x}$  and  $\bar{x}$ ) gives:

$$\tilde{\lambda}_N^T = \nu^T + \frac{\partial}{\partial \tilde{x}} \phi \quad \bar{\lambda}_N^T = \frac{\partial}{\partial \bar{x}} \phi$$

Stationarity w.r.t.  $\nu$  gives

$$\tilde{x}_N = \underline{\tilde{x}}_N$$

The rest is as usual (free dyn. opt.).

Defining the **Hamiltonian** function

$$H_i = L_i(x_i, u_i) + \lambda_{i+1}^T f_i(x_i, u_i)$$

The Euler-Lagrange equations:

$$x_{i+1} = f_i(x_i, u_i) \quad \lambda_i^T = \frac{\partial}{\partial x} H_i \quad 0 = \frac{\partial}{\partial u} H_i$$

with boundary conditions:

$$x_0 = \underline{x}_0 \quad \tilde{x}_N = \underline{\tilde{x}}_N \quad \tilde{\lambda}_N^T = \nu^T + \frac{\partial}{\partial \tilde{x}_N} \phi(x_N) \quad \bar{\lambda}_N^T = \frac{\partial}{\partial \bar{x}_N} \phi(x_N)$$

Conditions:  $n + p + p + (n - p) = 2 \times n + p$ .

Unknowns:  $x_0, \tilde{x}_N, \nu$  and  $\bar{\lambda}_N$  (results:  $n + p + p + (n - p)$ )

$$Cx_N = \underline{r}$$

E.g.

$$C = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \end{bmatrix} \quad \underline{r} = \begin{bmatrix} r_1 \\ r_2 \end{bmatrix}$$

## Linear end point constraints

Consider the system ( $i = 0, \dots, N - 1$ )

$$x_{i+1} = f_i(x_i, u_i) \quad x_0 = \underline{x}_0 \quad (9)$$

the performance index

$$J = \phi(x_N) + \sum_{i=0}^{N-1} L_i(x_i, u_i) \quad (10)$$

and the linear terminal constraints

$$Cx_N = \underline{r}_N \quad C : p \times n \quad (11)$$

where  $C$  and  $\underline{r}_N$  (and  $\underline{x}_0$ ) are given. The **Lagrange** function and the **multiplier** (vector with length  $p$ )  $\nu$ :

$$J_L = \phi(x_N) + \sum_{i=0}^{N-1} \left[ L_i(x_i, u_i) + \lambda_{i+1}^T (f_i(x_i, u_i) - x_{i+1}) \right] + \lambda_0^T (\underline{x}_0 - x_0) + \nu^T (Cx_N - \underline{r}_N)$$

**New conditions:** Stationarity w.r.t.  $x_N$  gives:

$$\lambda_N^T = \nu^T C + \frac{\partial}{\partial x_N} \phi(x_N)$$

Stationarity w.r.t.  $\nu$  gives

$$Cx_N = \underline{r}$$

The rest is as usual (free dyn. opt.).

Defining the **Hamiltonian** function

$$H_i = L_i(x_i, u_i) + \lambda_{i+1}^T f_i(x_i, u_i)$$

The Euler-Lagrange equations:

$$x_{i+1} = f_i(x_i, u_i) \quad \lambda_i^T = \frac{\partial}{\partial x} H_i \quad 0 = \frac{\partial}{\partial u} H_i$$

with boundary conditions:

$$x_0 = \underline{x}_0 \quad Cx_N = \underline{r}_N \quad \lambda_N^T = \nu^T C + \frac{\partial}{\partial x_N} \phi$$

Conditions:  $2 \times n + p$ .

Unknowns:  $x_0$ ,  $x_N$  and  $\nu$  (results:  $2 \times n + p$ )

## General end point constraints

Consider the system ( $i = 0, \dots, N - 1$ )

$$x_{i+1} = f_i(x_i, u_i) \quad x_0 = \underline{x}_0 \quad (12)$$

the performance index

$$J = \phi_N(x_N) + \sum_{i=0}^{N-1} L_i(x_i, u_i) \quad (13)$$

and the general terminal constraints

$$\psi_N(x_N) = 0 \quad \psi : \mathbb{R}^{n+1} \rightarrow \mathbb{R}^p \quad (14)$$

where  $\psi$  (and  $\underline{x}_0$ ) are given. Introduce the **multiplier** (vector of length  $p$ )  $\nu$  and form the **Lagrange** function:

$$J_L = \phi_N(x_N) + \lambda_0^T (\underline{x}_0 - x_0) + \nu^T \psi_N(x_N) + \sum_{i=0}^{N-1} \left[ L_i(x_i, u_i) + \lambda_{i+1}^T (f_i(x_i, u_i) - x_{i+1}) \right]$$

**New conditions:** Stationarity w.r.t.  $x_N$  gives:

$$\lambda_N^T = \nu^T \frac{\partial}{\partial x_N} \psi + \frac{\partial}{\partial x_N} \phi$$

Stationarity w.r.t.  $\nu$  gives

$$\psi_N(x_N) = 0$$

The rest is as usual (free dyn. opt.).

Defining the **Hamiltonian** function

$$H_i = L_i(x_i, u_i) + \lambda_{i+1}^T f_i(x_i, u_i)$$

The Euler-Lagrange equations:

$$x_{i+1} = f_i(x_i, u_i) \quad \lambda_i^T = \frac{\partial}{\partial x} H_i \quad 0 = \frac{\partial}{\partial u} H_i$$

with boundary conditions:

$$x_0 = \underline{x}_0 \quad \psi(x_N) = 0 \quad \lambda_N^T = \nu^T \frac{\partial}{\partial x_T} \psi + \frac{\partial}{\partial x_T} \phi$$

Conditions:  $n + p + n$ .

Unknowns:  $x_0$ ,  $x_N$  and  $\nu$  (results:  $2 \times n + p$ )

## End point constraints (C)

In this section we consider the continuous case in which  $t \in [0; T] \in \mathbb{R}$ . The problem is to find the input function  $u_t$  to the system

$$\dot{x} = f_t(x_t, u_t) \quad x_0 = \underline{x}_0$$

such that the performance index

$$J = \phi_T(x_T) + \int_0^T L_t(x_t, u_t) dt$$

is optimized and the end point constraints in

$$\psi_T(x_T) = 0$$

are met.

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$$J_L = \phi_T(x_T) + \lambda_0^T x_0 - \lambda_T^T x_T + \nu^T \psi_T(x_T) + \int_0^T \left( L_t(x_t, u_t) + \lambda_t^T f_t(x_t, u_t) + \dot{\lambda}_t^T x_t \right) dt$$

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Stationarity w.r.t.  $x_T$  gives:

$$\lambda_T^T = \nu^T \frac{\partial}{\partial x_T} \psi_T + \frac{\partial}{\partial x_T} \phi_T$$

stationarity w.r.t.  $\nu$  gives

$$\psi_T(x_T) = 0$$

If we introduce the Hamiltonian function as

$$H_t(x_t, u_t) = L_t(x_t, u_t) + \lambda_t^T f_t(x_t, u_t) \quad (15)$$

we can express the necessary conditions as

$$\dot{x} = f_t(x_t, u_t) \quad -\dot{\lambda}^T = \frac{\partial}{\partial x} H \quad 0^T = \frac{\partial}{\partial u} H$$

with the (split) boundary conditions

$$x_0 = \underline{x}_0 \quad \psi_T(x_T) = 0 \quad \lambda_T^T = \nu^T \frac{\partial}{\partial x} \psi_T + \frac{\partial}{\partial x} \phi_T$$

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Simple EPC:

$$\psi_T(x_T) = (x_T - \underline{x}_T) = 0$$

$$x_0 = \underline{x}_0 \quad x_T = \underline{x}_T \quad \lambda_T^T = \nu^T + \frac{\partial}{\partial x} \phi_T(x_T)$$

Partial simple EPC:

$$x_T = \begin{bmatrix} \tilde{x}_T \\ \bar{x}_T \end{bmatrix} \quad \tilde{x}_T = \underline{\tilde{x}}_T$$

$$x_0 = \underline{x}_0 \quad \tilde{x}_T = \underline{\tilde{x}}_T \quad \tilde{\lambda}_T = \nu^T + \frac{\partial}{\partial \tilde{x}_T} \phi \quad \bar{\lambda}_T = \frac{\partial}{\partial \bar{x}_T} \phi$$

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Linear EPC

$$Cx_T = \underline{r}$$

$$x_0 = \underline{x}_0 \quad Cx_T = \underline{r} \quad \lambda_T = \nu^T C + \frac{\partial}{\partial x} \phi_T(x_T)$$

The problem is to bring the system

$$\dot{x} = u_t \quad x_0 = \underline{x}_0$$

from  $\underline{x}_0$  to  $\underline{x}_T$  such that the performance index

$$J = \int_0^T \frac{1}{2} u^2 dt$$

is minimized. The Hamiltonian function is

$$H = \frac{1}{2} u^2 + \lambda u$$

and the Euler-Lagrange equations are simply

$$\begin{aligned} \dot{x} &= u_t & x_0 &= \underline{x}_0 \\ -\dot{\lambda} &= 0 \\ 0 &= u + \lambda \end{aligned}$$

with boundary conditions:

$$x_0 = \underline{x}_0 \quad x_T = \underline{x}_T \quad \lambda_T = \nu$$

Solution:

$$\lambda = c \quad u = -c \quad x_t = \underline{x}_0 - ct$$

and

$$\underline{x}_T = \underline{x}_0 - cT \quad \text{or} \quad c = \frac{\underline{x}_0 - \underline{x}_T}{T}$$

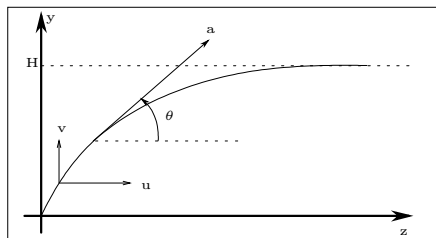
Finally, we have

$$x_t = \underline{x}_0 + \frac{\underline{x}_T - \underline{x}_0}{T} t \quad u_t = \frac{\underline{x}_T - \underline{x}_0}{T} \quad \lambda = \frac{\underline{x}_0 - \underline{x}_T}{T}$$

► Finish

## Orbit injection problem - Simplified

A body is initially at rest in the origin. A constant specific thrust force,  $a$ , is applied to the body in a direction that makes an angle  $\theta_t$  with the  $z$ -axis. Let  $u$  and  $v$  be the velocity in the  $z$  and  $y$  direction, respectively.



The task is to find an input function of angles of direction,  $\theta_t$  such that the body in a finite period,  $T$ ,

- 1 is injected into orbit i.e. reach a specific height  $H$

$$y_T = H$$

- 2 has zero vertical speed ( $y$ -direction)

$$v_T = 0$$

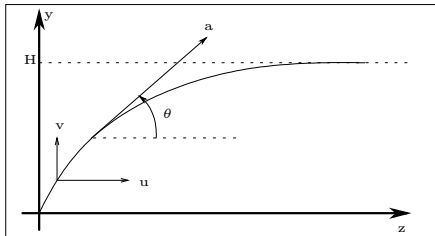
- 3 has maximum horizontal speed ( $z$ -direction)

$$\text{Max } u_T$$

This is also denoted as a Thrust Direction Programming (TDP) problem.

## Orbit injection - The dynamic

The problem is to find the input function,  $\theta_t$ , such that the terminal horizontal velocity,  $u_T$ , (at a specific altitude  $H$ ) is maximized.



The dynamic is:

$$\frac{d}{dt} \begin{bmatrix} u_t \\ v_t \\ z \\ y \end{bmatrix} = \begin{bmatrix} a \cos(\theta_t) \\ a \sin(\theta_t) \\ u_t \\ v_t \end{bmatrix} \quad \begin{bmatrix} u_0 \\ v_0 \\ z_0 \\ y_0 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

The terminal constraints are

$$v_T = 0 \quad y_T = H$$

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The objective is to maximize:

$$J = \phi(x_T) = u_T$$

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More condensed:

$$J = \phi(x_T) = u_T \quad \begin{bmatrix} v \\ y \end{bmatrix}_T = \begin{bmatrix} 0 \\ H \end{bmatrix}$$

$$x = \begin{bmatrix} u \\ v \\ z \\ y \end{bmatrix}$$

# Orbit injection - Euler-Lagrange equations

The Hamilton functions is (since  $L = 0$ )

$$H_t = \lambda_t^T f_t = \begin{bmatrix} \lambda_t^u & \lambda_t^v & \lambda_t^z & \lambda_t^y \end{bmatrix} \begin{bmatrix} a \cos(\theta_t) \\ a \sin(\theta_t) \\ u_t \\ v_t \end{bmatrix}$$

$$H_t = \lambda_t^u a \cos(\theta_t) + \lambda_t^v a \sin(\theta_t) + \lambda_t^z u_t + \lambda_t^y v_t$$

The Euler-Lagrange equations consists of the **state** equation,

$$\frac{d}{dt} \begin{bmatrix} u_t \\ v_t \\ z \\ y \end{bmatrix} = \begin{bmatrix} a \cos(\theta_t) \\ a \sin(\theta_t) \\ u_t \\ v_t \end{bmatrix} \quad \begin{bmatrix} u_0 \\ v_0 \\ z_0 \\ y_0 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

(just cut and paste)

the **costate** equation

$$-\frac{d}{dt} \begin{bmatrix} \lambda_t^u & \lambda_t^v & \lambda_t^z & \lambda_t^y \end{bmatrix} = \begin{bmatrix} \lambda_t^z & \lambda_t^y & 0 & 0 \end{bmatrix}$$

and the **stationarity** condition

$$0 = -\lambda_t^u a \sin(\theta_t) + \lambda_t^v a \cos(\theta_t)$$

Since

$$\phi_T(x_t) = u_t \quad \begin{bmatrix} v \\ y \end{bmatrix}_T = \begin{bmatrix} 0 \\ H \end{bmatrix}$$

we have the boundary conditions

$$\lambda_T^v = \nu_v \quad \lambda_T^y = \nu_y$$

$$\lambda_T^u = 1 \quad \lambda_T^z = 0$$

The **stationarity** condition

$$0 = -\lambda_t^u a \sin(\theta_t) + \lambda_t^v a \cos(\theta_t)$$

gives the tangent law:

$$\tan(\theta_t) = \frac{\lambda_t^v}{\lambda_t^u}$$

It turns out (later on) to be a linear tangent law.

The Costate equations

$$-\frac{d}{dt} \begin{bmatrix} \lambda_t^u & \lambda_t^v & \lambda_t^z & \lambda_t^y \end{bmatrix} = \begin{bmatrix} \lambda_t^z & \lambda_t^y & 0 & 0 \end{bmatrix}$$

and the boundary conditions

$$\lambda_T^v = \nu_v \quad \lambda_T^y = \nu_y \quad (\text{just a copy})$$

$$\lambda_T^u = 1 \quad \lambda_T^z = 0$$

gives us:

$$\lambda_t^z = 0 \quad \lambda_t^y = \nu_y \quad \text{constant in time}$$

$$\lambda_t^u = 1 \quad \text{constant in time}$$

$$\lambda_t^v = \nu_v + \nu_y(T - t)$$

$$\tan(\theta_t) = \nu_v + \nu_y(T - t)$$

Find  $\nu_v$  and  $\nu_y$  such that

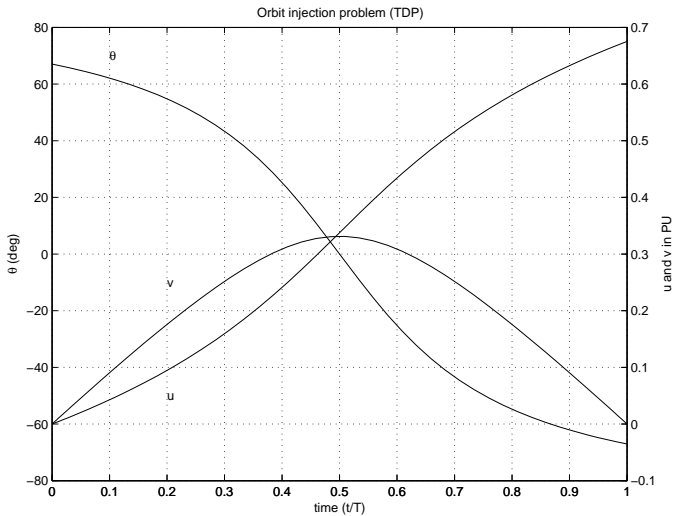
$$\tan(\theta_t) = \nu_v + \nu_y(T - t)$$

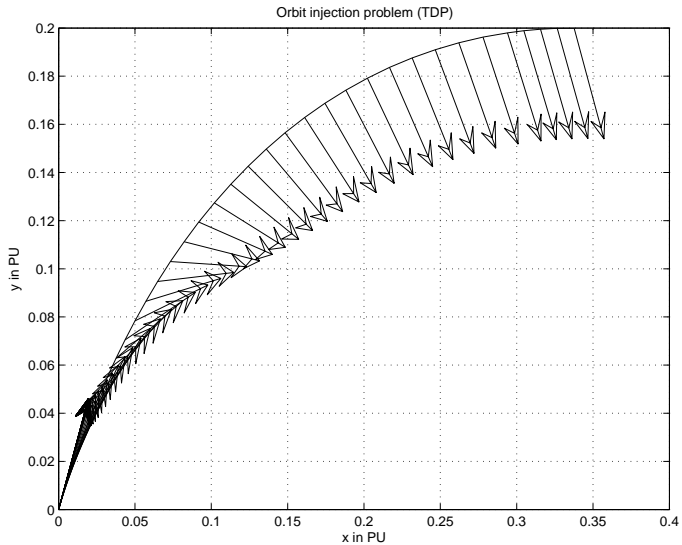
in the dynamics

$$\frac{d}{dt} \begin{bmatrix} u_t \\ v_t \\ z \\ y \end{bmatrix} = \begin{bmatrix} a \cos(\theta_t) \\ a \sin(\theta_t) \\ u_t \\ v_t \end{bmatrix} \qquad \begin{bmatrix} u_0 \\ v_0 \\ z_0 \\ y_0 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

results in

$$\begin{bmatrix} v \\ y \end{bmatrix}_T = \begin{bmatrix} 0 \\ H \end{bmatrix}$$





```

% -----
function main1
% -----
T=1; % parametes
a=1;
H=0.2;
parm=[-2.4 4.7]'; % Initial guess on parametes
x0=zeros(4,1); % Initial state variable

opt=optimset; % Options for fsolve
opt=optimset(opt,'Display','iter');
parm=fsolve(@erf,parm,opt,T,a,x0,H); % Call fsolve for finding parameters

[err,time,xt]=erf(parm,T,a,x0,H); % Call erf ones more for getting the
tht=atan(parm(1)+parm(2)*(T-time)); % optimal input solution

% Here goes the plotting commands. Omitted here.
% file on databar: ~nkpo/02711/dist3/main1.m

```

```

% -----
function [err,time,xt]=erf(parm,T,a,x0,H)
% -----
% Determine the end point error (err) given the EPC Lagrange multipliers
% in parm (and the constants that specifies the problem).
Tspan=0:T;
[time,xt]=ode45(@tdp,Tspan,x0,[],parm,T,a);
xT=xt(end,:)' ;
err=[xT(2);
     xT(4)-H];

```

```

% -----
function dx=tdp(t,x,parm,T,a)
% -----
% System model. Determine the (time) derivative of the state vector
% given the time, state (x) and the EPC Lagrange multipliers.
u=x(1); v=x(2); z=x(3); y=x(4);
nuu=parm(1); nuy=parm(2);
th=atan(nuu+nuy*(T-t));
dx=[a*cos(th);
    a*sin(th);
    u;
    v];

```

DO Chapter 3