

# A Brief Introduction to Optimal Control

Ismail Hossain

Department of Mathematics  
University of Manitoba

January 25, 2017

- Basic Concepts

# Outlines

- Basic Concepts
- Controllability

- Basic Concepts
- Controllability
- The Pontryagin Maximum Principle

# Notations

We use the following notations throughout the discussion,

$$\mathbf{f}(x) = \begin{pmatrix} f_1(x) \\ \vdots \\ f_n(x) \end{pmatrix} \text{ and } \mathbf{x}(t) = \begin{pmatrix} x_1(t) \\ \vdots \\ x_n(t) \end{pmatrix}$$

## Definition 1.1.

**Dynamics:** Let us consider a differential equation of the form

$$\begin{cases} \mathbf{x}'(t) = \mathbf{f}(\mathbf{x}(t)) \\ \mathbf{x}(0) = \mathbf{x}_0 \end{cases} \quad (1)$$

Where,  $t > 0$ ,  $\mathbf{x}_0 \in \mathbb{R}^n$  is the initial point and the function  $\mathbf{f} : \mathbb{R}^n \rightarrow \mathbb{R}^n$ . The unknown is the curve  $\mathbf{x} : [0, \infty) \rightarrow \mathbb{R}^n$ , which is regarded as the dynamical evolution of the state of a system.

## Definition 1.2.

**Controlled Dynamics:** Suppose the function  $\mathbf{f}$  depends on some *control* parameters belonging to a set  $A \subset \mathbb{R}^m$ ; so that  $\mathbf{f} : \mathbb{R}^n \times A \rightarrow \mathbb{R}^n$ . Then if we choose  $a \in A$  and consider the corresponding dynamics,

$$\begin{cases} \mathbf{x}'(t) = \mathbf{f}(\mathbf{x}(t), a) \\ \mathbf{x}(0) = \mathbf{x}_0 \end{cases} \quad (2)$$

we obtain the evolution of our system when the parameter value ' $a$ ' is fixed. We can change the parameter values as the system evolves.

## Definition 1.3.

**Control:** We call a function  $\alpha : [0, \infty) \rightarrow A$  a control. Corresponding to each control, we consider the differential equation

$$\begin{cases} \mathbf{x}'(t) = \mathbf{f}(\mathbf{x}(t), \alpha(t)) \\ \mathbf{x}(0) = x_0 \end{cases} \quad (3)$$

and regard the trajectory  $\mathbf{x}(\cdot)$  as the corresponding *response* of the system.



## Definition 1.4.

**Admissible Control:** We introduce

$$\mathcal{A} = \{\alpha : [0, \infty) \rightarrow A \mid \alpha(\cdot) \text{ measurable}\}$$

to denote the collection of all *admissible controls*, where

$$\alpha(t) = \begin{pmatrix} \alpha_1(t) \\ \vdots \\ \alpha_m(t) \end{pmatrix}$$

## Definition 1.5.

**Payoffs:** We always look forward to determine the best control for our system. For this we need to specify a specific payoff(or reward)criterion. Let us define the *payoff functional*

$$P[\alpha(\cdot)] := \int_0^T r(\mathbf{x}(t), \alpha(t)) dt + g(\mathbf{x}(T)) \quad (P)$$

where  $\mathbf{x}(\cdot)$  solves the system for the control  $\alpha(t)$ . Here  $r : \mathbb{R}^n \times A \rightarrow \mathbb{R}$  and  $g : \mathbb{R}^n \rightarrow \mathbb{R}$  are given, and we call  $r$  the running payoff and  $g$  the terminal payoff. The terminal time  $T > 0$  is also given.

# Basic Control Problem

**Basic Control Problem:** Our goal is to find a control  $\alpha^*(\cdot)$ , which *maximizes* the payoff. That is, we want

$$P[\alpha^*(\cdot)] \geq P[\alpha(\cdot)]$$

for all controls  $\alpha(\cdot) \in \mathcal{A}$ . Such a control  $\alpha^*(\cdot)$  is called *optimal*.

**Controllability Problem:** The controllability question is, “Given the initial point  $x_0$  and a *target set*  $S \subset \mathbb{R}^n$ , does there exist a control steering the system to  $S$  in finite time?

In this section we consider the problem of driving the system to the origin  $S = \{0\}$ .

**Controllability Problem:** The controllability question is, “Given the initial point  $x_0$  and a *target set*  $S \subset \mathbb{R}^n$ , does there exist a control steering the system to  $S$  in finite time?

In this section we consider the problem of driving the system to the origin  $S = \{0\}$ .

## Definition 2.1.

**Reachable Set:** The reachable set for time  $t$  is defined as

$\mathcal{C}(t)$  = set of initial points  $x_0$  for which there exists a control such that  $\mathbf{x}(t) = 0$ .

and the overall reachable set is the union of all  $\mathcal{C}(t)$  for all  $t \geq 0$ . That is,

$$\mathcal{C} = \bigcup_{t \geq 0} \mathcal{C}(t)$$

# Controllability of Linear System

**Controllability of Linear System:** Let us consider that our system is linear in both the state  $\mathbf{x}(\cdot)$  and the control  $\alpha(\cdot)$  and consequently has the form

$$\begin{cases} \mathbf{x}'(t) = M\mathbf{x}(t) + N\alpha(t) \\ \mathbf{x}(0) = \mathbf{x}_0 \end{cases} \quad (4)$$

where  $t > 0$ ,  $M \in \mathbb{M}^{n \times n}$  and  $N \in \mathbb{M}^{n \times m}$ . We also assume that the set of control parameters  $A$  is a cube in  $\mathbb{R}^m$ :

$$A = [-1, 1]^m = \{a \in \mathbb{R}^m \mid |a_i| \leq 1, i = 1, \dots, m\}$$

## Definition 2.2.

**Controllability Matrix:** The controllability matrix is defined as,

$$G = G(M, N) := [N, MN, M^2N, \dots, M^{n-1}N]$$

and  $G$  is a  $n \times (mn)$  matrix.

## Theorem 2.3.

*The rank of the controllability matrix  $G$  is,*

$$\text{rank } G = n$$

*if and only if,  $0 \in \mathcal{C}^0$ , where  $\mathcal{C}^0$  represents the interior of the reachable set  $\mathcal{C}$ .*



## Theorem 2.4.

**Controllability Criterion:** *Let  $A$  be the cube  $[-1, 1]^n$  in  $\mathbb{R}^n$ . Also suppose that  $\text{rank } G = n$ , and  $\text{Re } \lambda < 0$  for each eigenvalue  $\lambda$  of the matrix  $M$ . Then the system (4) is controllable.*

# Controllability of Linear System: Example

**Example:** Let us consider the following linear system

$$\mathbf{x}' = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} \mathbf{x} + \begin{pmatrix} 0 \\ 1 \end{pmatrix} \alpha$$

where,

$$M = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} \text{ and } N = \begin{pmatrix} 0 \\ 1 \end{pmatrix}.$$

Then

$$G = [N, MN] = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}.$$

Therefore,

$$\text{rank } G = 2 = n.$$

## Example Continued

Also, the characteristic equation of the matrix  $M$  is

$$p(\lambda) = 0 = \det(\lambda I - M) = \det \begin{pmatrix} \lambda & -1 \\ 0 & \lambda \end{pmatrix} \\ \implies \lambda^2 = 0$$

we can see that both of the eigenvalues are 0, so it fails to satisfy the hypotheses of theorem (2.4).

## Theorem 2.5.

**Improved Criterion for Controllability:** Assume that  $\text{rank } G = n$ , and  $\text{Re } \lambda \leq 0$  for each eigenvalue  $\lambda$  of the matrix  $M$ . Then the system (4) is controllable.

# Motivation for the Pontryagin Maximum Principle

## Definition 3.1.

**Lagrangian:** Consider a smooth function  $L : \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R}$ ,  $L = L(x, v)$ ;  $L$  is called the *Lagrangian*.

# Motivation for the Pontryagin Maximum Principle

## Definition 3.1.

**Lagrangian:** Consider a smooth function  $L : \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R}$ ,  $L = L(\mathbf{x}, \mathbf{v})$ ;  $L$  is called the *Lagrangian*.

**Basic Problem of Calculus of Variations:** Find a curve  $\mathbf{x}^*(\cdot) : [0, T] \rightarrow \mathbb{R}^n$  that minimizes the functional

$$I[\mathbf{x}(\cdot)] := \int_0^T L(\mathbf{x}(t), \mathbf{x}'(t)) dt \quad (5)$$

among all functions  $\mathbf{x}(\cdot)$  satisfying  $\mathbf{x}(0) = \mathbf{x}_0$  and  $\mathbf{x}(T) = \mathbf{x}_1$ .

# Motivation for the Pontryagin Maximum Principle

## Theorem 3.2.

**Euler-Lagrange Equation:** *Let  $\mathbf{x}^*(\cdot)$  solves the calculus of variations problem. Then  $\mathbf{x}^*(\cdot)$  solves the Euler-Lagrange differential equations*

$$\frac{d}{dt}[\nabla_v L(\mathbf{x}^*(t), \mathbf{x}'^*(t))] = \nabla_x L(\mathbf{x}^*(t), \mathbf{x}'^*(t)) \quad (E-L)$$

# Motivation for the Pontryagin Maximum Principle

## Theorem 3.2.

**Euler-Lagrange Equation:** Let  $\mathbf{x}^*(\cdot)$  solves the calculus of variations problem. Then  $\mathbf{x}^*(\cdot)$  solves the Euler-Lagrange differential equations

$$\frac{d}{dt}[\nabla_v L(\mathbf{x}^*(t), \mathbf{x}'^*(t))] = \nabla_x L(\mathbf{x}^*(t), \mathbf{x}'^*(t)) \quad (E-L)$$

## Definition 3.3.

**Generalized Momentum:** For a given curve  $\mathbf{x}(\cdot)$ , we define the generalized momentum  $\mathbf{p}(\cdot)$  as follows

$$\mathbf{p}(t) := \nabla_v L(\mathbf{x}(t), \mathbf{x}'(t)) \quad (0 \leq t \leq T)$$



# Motivation for the Pontryagin Maximum Principle

## Definition 3.4.

**Hamiltonian:** We define the *Hamiltonian*  $H : \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R}$  of a dynamical systems by the formula

$$H(x, p) = p \cdot \mathbf{v}(x, p) - L(x, \mathbf{v}(x, p))$$

# Motivation for the Pontryagin Maximum Principle

## Definition 3.4.

**Hamiltonian:** We define the *Hamiltonian*  $H : \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R}$  of a dynamical systems by the formula

$$H(x, p) = p \cdot v(x, p) - L(x, v(x, p))$$

## Theorem 3.5.

**Hamiltonian Dynamics:** Suppose  $x(\cdot)$  solve the Euler-Lagrange equations (E-L) and define  $p(\cdot)$  as above. Then the pair  $(x(\cdot), p(\cdot))$  solves the Hamilton's equations

$$\begin{cases} x'(t) = \nabla_p H(x(t), p(t)) \\ p'(t) = -\nabla_x H(x(t), p(t)) \end{cases}$$

Furthermore the mapping  $t \rightarrow H(x(t), p(t))$  is constant.

# Fixed Time, Free Endpoint Problem

**Fixed Time, Free Endpoint Problem:** Consider the system (3),  $(P)$ , admissible controls defined earlier. The basic problem is to find a control  $\alpha^*(\cdot)$  such that

$$P[\alpha^*(\cdot)] = \max_{\alpha(\cdot) \in \mathcal{A}} P[\alpha(\cdot)]$$

# Fixed Time, Free Endpoint Problem

**Fixed Time, Free Endpoint Problem:** Consider the system (3),  $(P)$ , admissible controls defined earlier. The basic problem is to find a control  $\alpha^*(\cdot)$  such that

$$P[\alpha^*(\cdot)] = \max_{\alpha(\cdot) \in \mathcal{A}} P[\alpha(\cdot)]$$

## Definition 3.6.

The control theory Hamiltonian is the function

$$H(x, p, a) := f(x, a) \cdot p + r(x, a) \quad (x, p \in \mathbb{R}^n, a \in A).$$

# Statement of Pontryagin Maximum Principle

## Theorem 3.7.

**Pontryagin Maximum Principle:** Let  $\alpha^*(\cdot)$  be optimal for system (3), (P) and  $\mathbf{x}^*(\cdot)$  be the corresponding trajectory. Then there exists a function  $\mathbf{p}^* : [0, T] \rightarrow \mathbb{R}^n$  such that

$$\mathbf{x}^{*'}(t) = \nabla_p H(\mathbf{x}^*(t), \mathbf{p}^*(t), \alpha^*(t)) \quad (ODE)$$

$$\mathbf{p}^{*'}(t) = -\nabla_x H(\mathbf{x}^*(t), \mathbf{p}^*(t), \alpha^*(t)) \quad (ADJ)$$

and

$$H(\mathbf{x}^*(t), \mathbf{p}^*(t), \alpha^*(t)) = \max_{a \in A} H(\mathbf{x}^*(t), \mathbf{p}^*(t), a) \quad (0 \leq t \leq T) \quad (M)$$

Additionally,  $H(\mathbf{x}^*(t), \mathbf{p}^*(t), \alpha^*(t)) \equiv \text{constant}$

Finally, we have the terminal condition

$$\mathbf{p}^*(T) = \nabla g(\mathbf{x}^*(T)) \quad (T)$$

**Free Time, Fixed Endpoint Problem:** Consider the system (3) defined earlier. Assuming that a target point  $x_1 \in \mathbb{R}^n$  is given, we introduce the following payoff functional

$$P[\alpha(\cdot)] = \int_0^\tau r(\mathbf{x}(t), \alpha(t)) dt$$

Here,  $\tau = \tau[\alpha(\cdot)] \leq \infty$  denotes the first time the solution of the system hits the target point  $x_1$ .

# Statement of Pontryagin Maximum Principle

## Theorem 3.8.

**Pontryagin Maximum Principle:** Let  $\alpha^*(\cdot)$  be optimal for system (3), (P) and  $\mathbf{x}^*(\cdot)$  be the corresponding trajectory. Then there exists a function  $\mathbf{p}^* : [0, \tau^*] \rightarrow \mathbb{R}^n$  such that

$$\mathbf{x}'^*(t) = \nabla_{\mathbf{p}} H(\mathbf{x}^*(t), \mathbf{p}^*(t), \alpha^*(t))$$

$$\mathbf{p}'^*(t) = -\nabla_{\mathbf{x}} H(\mathbf{x}^*(t), \mathbf{p}^*(t), \alpha^*(t))$$

and

$$H(\mathbf{x}^*(t), \mathbf{p}^*(t), \alpha^*(t)) = \max_{a \in A} H(\mathbf{x}^*(t), \mathbf{p}^*(t), a) \quad (0 \leq t \leq \tau^*) \quad (M)$$

Also,

$$H(\mathbf{x}^*(t), \mathbf{p}^*(t), \alpha^*(t)) \equiv 0 \quad (0 \leq t \leq \tau^*)$$

- Lawrence C. Evans, *An introduction to mathematical optimal control theory*, Version 0.2. Lectures at the University of Maryland, 1983.  
<https://math.berkeley.edu/~evans/control.course.pdf>



**Thank You for Your Attention**