# OPTIMAL CONTROL

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## 1 Optimal Control Introduction

## 1.1 Deterministic Finite-Dimensional Continuous-Time Problem

$$\inf_{u \in \mathcal{U}} J(u; t_0, t_f, X_0) \equiv K(t_f, X_f) + \int_{t_0}^{t_f} L(s, X_s, u_s) \, ds \tag{1}$$

$$dX_t \equiv f(t, X_t, u_t) dt \quad X_0 \in \mathbb{R}^m$$
 (2)

$$\psi(t, X_t, u_t) = 0 \in \mathbb{R}^l, \quad \forall t \in [t_0, t_f]$$
(3)

$$\phi(t_f, X_f, u_t) \le 0 \in \mathbb{R}^k, \quad \forall t \in [t_0, t_f]$$
(4)

#### 1.2 Definitions

Summary. Fundamental definitions

1. Metric Space: (M, d)

2. Inner Product Induced Metric:  $(M, \langle \cdot, \cdot \rangle)$ 

3. Topology:  $\mathcal{T} \equiv (A_i)$ 

4. Open & Closed Sets:  $A, A^c$ 

5. Open & Closed Balls:  $B(x,\epsilon;d), \bar{B}(x,\epsilon;d)$ 

6. Metric Topology:  $\mathcal{T}(M)$ 

7. Set Closure:  $\bar{A}$ 

8. Set Interior:  $A^{\circ}$ 

9. Open Neighborhood:  $A \subseteq M$ 

#### **Definition 1.1** (Metric Space):

Defining

• (M,d): a metric space

• M: a set with topology induced by d

•  $d: M \times M - > [0, \infty)$ 

Then

1.  $d(x,y) = d(y,x) \ \forall x,y \in M$ 

Symmetric

2.  $d(x,x) = 0 \ \forall x \in M$ 

3. d(x,y) > 0,  $\forall x,y \in M$ ,  $x \neq y$ 

Non-Negative

4.  $d(x,y) \le d(x,z) + d(z,y) \forall x,y,z \in M$ 

Triangle Inequality

## **Definition 1.2** (Inner Product Induced Metric):

Defining

- $(M, \langle \cdot, \cdot \rangle)$ : an inner product space
- M: a vector space
- $\langle \cdot, \cdot \rangle$ : an inner product

This induces the metric

$$d(x,y) = |x-y| \equiv \langle x-y, x-y \rangle^{1/2}, \quad \forall x, y \in M$$
 (5)

## **Definition 1.3** (Topology):

Defining

•  $\mathcal{T} \equiv (A_i)$ : a collection of subsets of M

 $\mathcal{T}$  forms a topology for M if the following hold

- 1.  $M, \emptyset \in \mathcal{T}$
- 2. If  $(E_i) \subseteq \mathcal{T}$  is a countable collection
- 3. If  $(E_i) \subseteq \mathcal{T}$  is a finite collection

### **Definition 1.4** (Open & Closed Sets):

- Open Set: elements of a topology (i.e.,  $A \in \mathcal{T}$ )
- Closed Set: complement of an open set (i.e.,  $A^c$ )

## **Definition 1.5** (Open & Closed Balls):

Defining

- (M, d): a metric space
- $\epsilon > 0$ : Radius
- $x \in M$ : Center

Open and closed balls are defined as

- 1. Open Ball:  $B(x, \epsilon; d) \equiv \{y \in M | d(x, y) < \epsilon\}$
- 2. Closed Ball:  $\bar{B}(x, \epsilon; d) \equiv \{y \in M | d(x, y) \le \epsilon\}$

## **Definition 1.6** (Metric Topology):

Given

• (M, d): a metric space

The metric can induce a topology by considering a collection of open balls.

Example 1.0.1 (Borel Topology):

- standard topology on  $\mathbb{R}^m$
- all open balls centered at rational numbers  $\mathbb Q$
- radius is positive rational

## **Definition 1.7** (Set Closure):

Given

- $A \subseteq M$ : a subset of a metric space
- $D_i$ : collection of all closed sets that contain A
- $\bar{A} \supseteq A$ : the closure of A

Closure is defined as

$$\bar{A} \equiv \bigcap_{i} D_{i} \tag{6}$$

Remark. In Borel topology, isolated points are closed

#### **Definition 1.8** (Set Interior):

Given

- $A \subseteq M$ : a subset of a metric space
- $E_i$ : collection of all open sets that contain A
- $A^{\circ} \subseteq A$ : the interior of A

The interior is defined as

$$A^{\circ} \equiv \bigcup_{i} E_{i} \tag{7}$$

# **Definition 1.9** (Open Neighborhood): Given

•  $x \in M$ : a point in a metric space

•  $\mathcal{T}(M)$ : the topology of M

 $A \subseteq M$  is an open neighborhood of x if

- $x \in A$
- $A \in \mathcal{T}(M)$

Remark. The neighborhood is implied to be small, with motivation from metric topology implying that A looks like a small open ball centered at x.

## 2 Parameter Optimization Conditions

## 2.1 Defining Optimality

Summary. Defining local and global minimum on a metric space for a cost function

- 1. Local minimum
- 2. Global minimum
- 3. Local minimum;  $(\mathbb{R}, d)$ : Local minimum in standard 1D metric space
- 4. Global minimum;  $(\mathbb{R}, d)$ : Global minimum in standard 1D metric space
- 5. Extremum

## **Definition 2.1** (Local Minimum):

Given

- (M,d): a metric space
- $\mathcal{T}$ : a topology on M induced by the metric d
- $f: M \to \mathbb{R}$ : a cost function

 $x^* \in M$  is a local minimum of f if  $\exists$  neighborhood A of  $x^*$  such that

$$f(x^*) \le f(x) \quad \forall x \in A \tag{8}$$

Remark.  $x^*$  is a strict local minimum if

$$f(x^*) < f(x) \quad \forall x \in A \backslash x^*$$
 (9)

## **Definition 2.2** (Global Minimum):

Given

- (M, d): a metric space
- $\mathcal{T}$ : a topology on M induced by the metric d
- $f: M \to \mathbb{R}$ : a cost function

 $x^* \in M$  is a global minimum of f if

$$f(x^*) \le f(x) \quad \forall x \in M \tag{10}$$

**Definition 2.3** (Local Minimum;  $(\mathbb{R}, d)$ ):

 $x^* \in \mathbb{R}$  is a local minimum of  $f : \mathbb{R} \to \mathbb{R}$  if  $\exists \epsilon > 0$  such that

$$f(x^*) \le f(x) \quad \forall x \in (x^* - \epsilon, x^* + \epsilon) = B(x^*, \epsilon) \tag{11}$$

 $x^*$  is a strict local minimum if

$$f(x^*) < f(x) \quad \forall x \in (x^* - \epsilon, x^* + \epsilon) \setminus \{x^*\}$$
(12)

**Definition 2.4** (Global Minimum;  $(\mathbb{R}, d)$ ):

 $x^* \in \mathbb{R}$  is a global minimum of  $f: \mathbb{R} \to \mathbb{R}$  if

$$f(x^*) \le f(x) \quad \forall x \in \mathbb{R} \tag{13}$$

 $x^*$  is a strict global minimum if

$$f(x^*) < f(x) \quad \forall x \in \mathbb{R} \setminus \{x^*\}$$
 (14)

**Definition 2.5** (Extremum):

If  $x^*$  is a local minimum or maximum, then  $x^*$  is an extremum.

## 2.2 Unconstrained Smooth Parameter Optimization

**Summary.** Deriving first and second order necessary and sufficient optimiality conditions in Euclidean space  $\mathbb{R}^n$ .

#### Definitions:

- 1. Continuously Bounded Differentiable Function
- 2. Compact Set
- 3. Stationary Point
- 4. Unit Sphere
- 5. Hessian Matrix

#### Theorems:

- 1. Heine-Borel Property
- 2. Weierstrass Extreme Value
- 3. First Order Necessary Condition;  $(\mathbb{R}, d)$
- 4. Taylor's Formula; Using Lagrange Form of the Mean-Value of the Remainder
- 5. Second Order Necessary Condition;  $(\mathbb{R}, d)$
- 6. Second Order Sufficient Condition;  $(\mathbb{R}, d)$
- 7. First Order Necessary Condition;  $(\mathbb{R}^n, d)$

## $\begin{tabular}{ll} \textbf{Definition 2.6} & (Continuously Bounded Differentiable Function): \\ \end{tabular}$

#### Defining

- $C^k(\Omega; \mathbb{R})$ : real-valued k-times continuously differentiable functions on the set  $\Omega$ . So  $f \in C^k(\Omega; \mathbb{R})$  is k-times differentiable, and each derivative is continuous. e.g.,  $\Omega \subseteq \mathbb{R}^n$
- $C_b^k(\Omega;\mathbb{R})$ : real-valued k-times continuously differentiable functions on the set  $\Omega$  that are bounded

## **Definition 2.7** (Compact Set):

Given

- $(M, \mathcal{T})$ : a topological space
- $\Omega \subseteq M$ : a set

 $\Omega$  is compact if

- $(E_i)_{i \in I}$ : open cover of  $\Omega$
- $(E_j)_{j\in J}$ : finite subcover where  $J\subseteq I$
- For any  $(E_i)_{i\in I}$ , there exists  $(E_j)_{j\in J}$ .

## Theorem 2.1 (Heine-Borel Property):

A metric space has the Heine-Borel property iff every compact set is closed and bounded. In particular,  $\mathbb{R}^n$  with the sandard metric has this property.

### Theorem 2.2 (Weierstrass Extreme Value):

Let

- $\Omega$ : a compact set
- $f: \omega \to \mathbb{R}$  a continous function.

Then f has extremums on  $\Omega$ .

*Remark.* If  $\Omega$  is compact, then the class of functions is automatically  $C_b^k(\Omega;\mathbb{R})$ 

## **Theorem 2.3** (First Order Necessary Condition; $(\mathbb{R}, D)$ ):

Given

- $f \in C^1(\Omega; \mathbb{R})$
- $x^* \in \Omega^{\circ} \subseteq \mathbb{R}$

If  $x^*$  is a local minimum for f, then

$$\frac{\partial f}{\partial x_{\,x^*}} = 0$$

#### **Definition 2.8** (Stationary Point):

 $x^*$  is a stationary point for f if

$$\partial_x f|_{x^*} = 0 \tag{15}$$

**Theorem 2.4** (Taylor's Formula; Using Language Form of the Mean-Value of the Remainder):

Assume

- $f \in C_b^2(\mathbb{R}; \mathbb{R})$
- $x, x^*$
- $\delta x \equiv x x^*$

there exists a point y between x and  $x^*$  such that

$$f(x) = f(x^*) + \delta_x f|_{x^*} \delta x + \frac{1}{2} \delta_x^2 f|_y \delta x^2$$
 (16)

**Theorem 2.5** (Second Order Necessary Condition;  $(\mathbb{R}, d)$ ):

If  $x^* \in \Omega^{\circ} \subseteq \mathbb{R}$  is a local minimum for  $f \in C_b^2(\Omega; \mathbb{R})$ , then

$$\frac{\partial^2 f}{\partial x^2}\big|_{x^*} \ge 0 \tag{17}$$

**Theorem 2.6** (Second Order Sufficient Condition;  $(\mathbb{R}, d)$ ):

Prerequisites

- Let  $x^* \in \Omega^{\circ} \subseteq \mathbb{R}$
- Assume  $f \in C_h^2(\Omega; \mathbb{R})$
- If  $\frac{\partial f}{\partial x}|_{x^*} = 0$  and  $\frac{\partial^2 f}{\partial x^2}|_{x^*} > 0$  hold

Then  $x^*$  is a strict local minimum for f.

## **Definition 2.9** (Unit Sphere):

Defining

- $S^n(x,d) \subset \mathbb{R}^{n+1}$  for  $n \in \mathbb{N}$ : unit sphere
- $x \in \mathbb{R}^{n+1}$ : a point
- d: a metric on  $\mathbb{R}^{n+1}$

Then the complete definition is  $S^n(x,d) \equiv \{y \in \mathbb{R}^{n+1}: d(x,y) = 1\}$ 

Remark. If  $S^n$  is denoted without x and d, then it can be assumed that x=0 and d is the standard  $l_2$  metric.

**Theorem 2.7** (First Order Necessary Condition;  $(\mathbb{R}^n, d)$ ): Given

- $f \in C^1(\Omega; \mathbb{R})$
- $x^* \in \Omega^{\circ} \subseteq \mathbb{R}^n$

If  $x^*$  is a local minimum for f, then

$$\nabla f\big|_{x^*} \equiv \left(\frac{\partial f}{\partial x_1}\bigg|, \cdots, \frac{\partial f}{\partial x_n}\bigg|_{x^*}\right) = 0 \in \mathbb{R}^n$$

## **Definition 2.10** (Hessian Matrix):

Given

- $f \in C^2(\mathbb{R}^n; \mathbb{R})$
- $\mathbb{S}(n;\mathbb{R}) = \mathbb{S}(n)$ : an  $n \times n$  symmetric matrix

The Hessian matrix of f evaluated at  $x \in \mathbb{R}$  is

$$\nabla_x^{\otimes 2} f = \nabla_x^2 = \begin{bmatrix} \frac{\partial^2 f}{\partial x_1^2} & \cdots & \frac{\partial f}{\partial x_1 x_n} \\ \vdots & \ddots & \vdots \\ \frac{\partial f}{\partial x_n x_1} & \cdots & \frac{\partial^2 f}{\partial x_n^2} \end{bmatrix}$$
(18)

Remark. Every  $A \in \mathbb{S}(n)$  can be expressed as  $A = U\Lambda U^T$  where the columns of U are the eigenvectors of A and the diagonal of  $\Lambda$  is the eigenvalues of A.

## 3 Constrained Parameter Optimization

## 3.1 Second Order Conditions in Arbitrary Dimensions

Summary. Theorems: Figure out what constrained vs unconstrained is

- 1. Second Order Necessary Condition;  $(\mathbb{R}^n, d)$
- 2. Second Order Sufficent Condition;  $(\mathbb{R}^n, d)$

#### Summary. Definitions

1. Big O Notation

## 3.2 Equality Constrained Smooth Parameter Optimization

- 1. Regular Point of M
- 2. Tangent Space (Geometric)
- 3. Tangent Space (Curves)
- 4. First Order Necessary Condition (Geometric)
- 5. Lagrange Multipliers
- 6. First Order Necessary Condition (Analytic)