
Classical Optimization Theory: Unconstrained Optimization

1. Preliminaries

- **Framework:**

- Consider a set $A \subset \mathbb{R}^n$, and a function $f : A \rightarrow \mathbb{R}$.
- We are interested in identifying points in A at which the function f attains a (local or global) maximum and/or minimum.

- **Local Maximum:**

Let $A \subset \mathbb{R}^n$, and let f be a function from A to \mathbb{R} . A point $x^* \in A$ is said to be a *point of local maximum* of f if there exists an open ball around x^* , $B_\epsilon(x^*)$, such that $f(x^*) \geq f(x)$ for all $x \in B_\epsilon(x^*) \cap A$.

- That is, a point x^* is a local maximum if there are no nearby points at which f takes on a larger value.

- **Global Maximum:**

Let $A \subset \mathbb{R}^n$, and let f be a function from A to \mathbb{R} . A point $x^* \in A$ is said to be a *point of global maximum* of f if $f(x^*) \geq f(x)$ for all $x \in A$.

- **Local and global minimum** can be defined similarly by just reverting the inequalities.

- Two issues:

- (a) Existence of a solution: Weierstrass Theorem;

- (b) Uniqueness of the solution: Strict Quasi-concavity.

- **Theorem 1 (Weierstrass Theorem: Existence of a Solution):**

Let A be a compact subset of \mathbb{R}^n and $f : A \rightarrow \mathbb{R}$ be continuous on A . Then there exists x_m and x_M in A such that $f(x_m) \leq f(x) \leq f(x_M)$ for all $x \in A$.

- That is, x_M is a point of global maximum and x_m is a point of global minimum of f .

- **Theorem 2 (Uniqueness of the Solution):**

Suppose A is a non-empty, compact and convex subset of \mathbb{R}^n . Suppose $f : A \rightarrow \mathbb{R}$ is a continuous and strictly quasi-concave (respectively, quasi-convex) function on A . Then there exists $x^ \in A$ such that $f(x^*) > f(x)$ (respectively, $f(x^*) < f(x)$) for all $x \in A$ where $x \neq x^*$.*

- That is, x^* is the unique global maximum (respectively, minimum) of f .

- Proof: To be discussed in class.

2. Necessary Conditions for Local Maximum & Minimum

- We will present two necessary conditions for local maximum and minimum:
 - one is a condition on the first-order partial derivatives of the relevant function (called the “first-order conditions”),
 - the other is a condition on the second-order partial derivatives of the relevant function (called the “second-order necessary conditions”).

- **Theorem 3:**

Let A be an open set in \mathbb{R}^n , and let $f : A \rightarrow \mathbb{R}$ be a continuously differentiable function on A . If $x^ \in A$ is a point of local maximum or minimum of f , then*

$$\nabla f(x^*) = 0. \tag{1}$$

- Proof: To be discussed in class.
- Remark: The n equations given by (1) are called the *first-order conditions* for a local maximum or minimum.

- **Theorem 4:**

Let A be an open set in \mathbb{R}^n , and let $f : A \rightarrow \mathbb{R}$ be a twice continuously differentiable function on A . If $x^* \in A$ is a point of local maximum (respectively, minimum) of f , then

$$H_f(x^*) \text{ is negative (respectively, positive) semi-definite.} \quad (2)$$

– Proof: To be discussed in class. [Theorem 30.11, pages 840-841, of the textbook.]

– Remark: Condition (2) is called the *second-order necessary condition* for a local maximum (respectively, minimum).

- *Necessary* conditions like (1) and (2) help us to rule out points where a local maximum *cannot* occur, thereby narrowing our search for points where a local maximum does occur. The following two examples illustrate this point.

#1. Let $f : \mathbb{R} \rightarrow \mathbb{R}$ be given by $f(x) = 1 - x^2$ for all $x \in \mathbb{R}$. Does f have a local maximum at the point $x = 1$?

#2. Let $f : \mathbb{R} \rightarrow \mathbb{R}$ be given by $f(x) = 1 - 2x + x^2$ for all $x \in \mathbb{R}$. Does f have a local maximum at the point $x = 1$?

3. Sufficient Conditions for Local Maximum and Minimum

- We present below a set of sufficient conditions for local maximum and minimum.

- **Theorem 5:**

Let A be an open set in \mathfrak{R}^n , and let $f : A \rightarrow \mathfrak{R}$ be a twice continuously differentiable function on A .

- (a) If $x^* \in A$ and

$$\nabla f(x^*) = 0, \text{ and} \tag{3a}$$

$$H_f(x^*) \text{ is negative definite,} \tag{4a}$$

then x^ is a point of local maximum of f .*

- (b) If $x^* \in A$ and

$$\nabla f(x^*) = 0, \text{ and} \tag{3b}$$

$$H_f(x^*) \text{ is positive definite,} \tag{4b}$$

then x^ is a point of local minimum of f .*

(c) If $x^* \in A$ and

$$\nabla f(x^*) = 0, \text{ and} \quad (3c)$$

$$H_f(x^*) \text{ is indefinite,} \quad (4c)$$

then x^* is neither a point of local maximum nor a point of local minimum of f .

- Proof: To be discussed in class. [Theorem 30.10, pages 836-840, of the textbook.]
- Remark: Condition (4a) (respectively, (4b)) is called the *second-order sufficient condition* for a local maximum (respectively, minimum).
- It should be noted that in the statement of Theorem 5 condition (4) *cannot* be weakened to condition (2).

#3. Provide an example to establish this point.

- It should also be observed that in the statement of Theorem 4 condition (2) *cannot* be strengthened to condition (4).

#4. Provide an example to establish this point.

- The outcome of this discussion is the following:
 - The second-order *necessary* conditions for a local maximum and minimum are different from (*weaker* than) the second-order *sufficient* conditions.
 - This simply reflects the fact that, in general, the first and second derivatives of a function at a point do not capture all aspects relevant to the occurrence of a local maximum or minimum of the function at that point.
 - The sufficient conditions of Theorem 5 enable us to find points of local maximum or minimum of a function as the following example shows.
- #5. Let $f : \mathfrak{R}^2 \rightarrow \mathfrak{R}$ be given by $f(x_1, x_2) = 2x_1x_2 - 2x_1^2 - x_2^2$ for all $(x_1, x_2) \in \mathfrak{R}^2$. Find out all the points of local maximum of f .

4. Sufficient Conditions for Global Maximum & Minimum

- **Theorem 6:**

Let A be an open convex set in \mathfrak{R}^n , and let $f : A \rightarrow \mathfrak{R}$ be a continuously differentiable function on A . If $x^ \in A$ satisfies $\nabla f(x^*) = 0$, and f is a concave (respectively, convex) function on A , then x^* is a point of global maximum (respectively, minimum) of f .*

– Proof: To be discussed in class.

- **Theorem 7:**

Let A be an open convex set in \mathfrak{R}^n , and let $f : A \rightarrow \mathfrak{R}$ be a twice continuously differentiable function on A .

(a) If $x^* \in A$ and

$$\nabla f(x^*) = 0, \text{ and}$$

$H_f(x)$ is negative semi-definite for all $x \in A$,

then x^* is a point of global maximum of f .

(b) If $x^* \in A$ and

$$\nabla f(x^*) = 0, \text{ and}$$

$H_f(x)$ is positive semi-definite for all $x \in A$,

then x^* is a point of global minimum of f .

– Proof: To be discussed in class.

- It is worth noting that Theorem 6 or 7 might be applicable in cases where Theorem 5 is not applicable.

– Example: Consider $f : \mathfrak{R} \rightarrow \mathfrak{R}$ be given by $f(x) = -x^4$.

- Note that $f'(0) = 0$ and $f''(x) = -12x^2 \leq 0$ for all $x \in \mathfrak{R}$.

\Rightarrow Applying Theorem 6 or 7 we can conclude that $x = 0$ is a point of global maximum, and hence of local maximum.

- But the conclusion that $x = 0$ is a point of local maximum cannot be derived from Theorem 5, since $f''(0) = 0$.

References

- Must read the following chapter and section from the textbook:
 - Chapter 17 (pages 396 – 410): Unconstrained Optimization,
 - Section 30.4 (pages 836 – 841): Second Order Optimization Conditions.
- This material is based on
 1. Bartle, R., *The Elements of Real Analysis*, (chapter 7),
 2. Apostol, T., *Mathematical Analysis: A Modern Approach to Advanced Calculus*, (chapter 6).
 3. Takayama, A., *Mathematical Economics*, (chapter 1).