

Critical Point and Local Extreme Value

Given a critical point c , the following test helps determine if $f(c)$ is a local extreme value:

Procedure

[Local Extreme Value]: Let c be an isolated critical point of f

- 1 $f(c)$ is a local minimum if $f(x)$ is decreasing in an interval $[c - \epsilon_1, c]$ and increasing in an interval $[c, c + \epsilon_2]$ with $\epsilon_1, \epsilon_2 > 0$.
- 2 $f(c)$ is a local maximum if $f(x)$ is increasing in an interval $[c - \epsilon_1, c]$ and decreasing in an interval $[c, c + \epsilon_2]$ with $\epsilon_1, \epsilon_2 > 0$.

Given a critical point c , *first derivative test* (sufficient condition) helps determine if $f(c)$ is a local extreme value:

Procedure

[First derivative test]: Let c be an isolated critical point of f

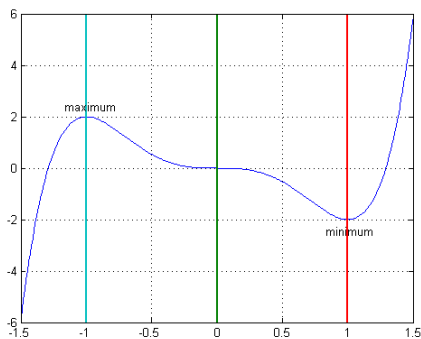
- 1 $f(c)$ is a local minimum if the sign of $f'(x)$ changes from negative in $[c - \epsilon_1, c]$ to positive in $[c, c + \epsilon_2]$ with $\epsilon_1, \epsilon_2 > 0$.
- 2 $f(c)$ is a local maximum if $f'(x)$ the sign of $f'(x)$ changes from positive in $[c - \epsilon_1, c]$ to negative in $[c, c + \epsilon_2]$ with $\epsilon_1, \epsilon_2 > 0$.
- 3 If $f'(x)$ is positive in an interval $[c - \epsilon_1, c]$ and also positive in an interval $[c, c - \epsilon_2]$, or $f'(x)$ is negative in an interval $[c - \epsilon_1, c]$ and also negative in an interval $[c, c - \epsilon_2]$ with $\epsilon_1, \epsilon_2 > 0$, then $f(c)$ is not a local extremum.

First Derivative Test: Critical Point and Local Extreme Value

As an example, the function $f(x) = 3x^5 - 5x^3$ has

First Derivative Test: Critical Point and Local Extreme Value

As an example, the function $f(x) = 3x^5 - 5x^3$ has the derivative $f'(x) = 15x^2(x+1)(x-1)$. The critical points are 0, 1 and -1 . Of the three, the sign of $f'(x)$ changes at 1 and -1 , which are local minimum and maximum respectively. The sign does not change at 0, which is therefore not a local supremum.



First Derivative Test: Critical Point and Local Extreme Value

As another example, consider the function

$$f(x) = \begin{cases} -x & \text{if } x \leq 0 \\ 1 & \text{if } x > 0 \end{cases}$$

Then,

$$f'(x) = \begin{cases} -1 & \text{if } x < 0 \\ 0 & \text{if } x > 0 \end{cases}$$

Note that $f(x)$ is discontinuous at $x = 0$, and therefore $f'(x)$ is not defined at $x = 0$. All numbers $x \geq 0$ are critical numbers. $f(0) = 0$ is a local minimum, whereas $f(x) = 1$ is a local minimum as well as a local maximum $\forall x > 0$.

Strict Convexity and Extremum

- A differentiable function f is said to be *strictly convex* (or *strictly concave up*) on an open interval \mathcal{I} , iff, $f'(x)$ is increasing on \mathcal{I} .
- Recall from theorem 7, the graphical interpretation of the first derivative $f'(x)$; $f'(x) > 0$ implies that $f(x)$ is increasing at x .
- Similarly, $f'(x)$ is increasing when $f''(x) > 0$. This gives us a sufficient condition for the strict convexity of a function:

Claim

If at all points in an open interval \mathcal{I} , $f(x)$ is doubly differentiable and if $f''(x) > 0$, $\forall x \in \mathcal{I}$, then the slope of the function is always increasing with x and the graph is strictly convex. This is illustrated in Figure 8.

- On the other hand, if the function is strictly convex and doubly differentiable in \mathcal{I} , then $f''(x) \geq 0$, $\forall x \in \mathcal{I}$.

Strict Convexity and Extremum (Illustrated)

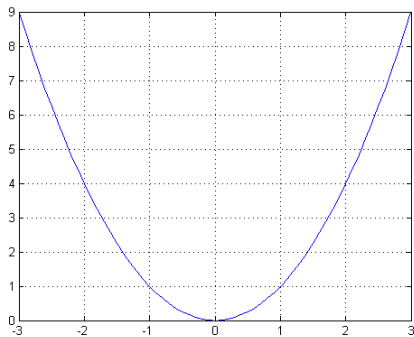


Figure 8:

Strict Convexity and Extremum: Slopeless interpretation (SI)

Claim

A function f is strictly convex on an open interval \mathcal{I} , iff

$$f(ax_1 + (1 - a)x_2) < af(x_1) + (1 - a)f(x_2) \quad (1)$$

whenever $x_1, x_2 \in \mathcal{I}$, $x_1 \neq x_2$ and $0 < a < 1$.

SI: Necessity when f is differentiable

First we will prove the **necessity**.

²For the case $x_2 < x_1$, the proof is very similar.

SI: Necessity when f is differentiable

First we will prove the **necessity**. Suppose f' is increasing on \mathcal{I} . Let $0 < a < 1$, $x_1, x_2 \in \mathcal{I}$ and $x_1 \neq x_2$. Without loss of generality assume that² $x_1 < x_2$. Then, $x_1 < ax_1 + (1 - a)x_2 < x_2$ and therefore $ax_1 + (1 - a)x_2 \in \mathcal{I}$. By the mean value theorem, there exist s and t with $x_1 < s < ax_1 + (1 - a)x_2 < t < x_2$, such that $f(ax_1 + (1 - a)x_2) - f(x_1) = f'(s)(x_2 - x_1)(1 - a)$ and $f(x_2) - f(ax_1 + (1 - a)x_2) = f'(t)(x_2 - x_1)a$. Therefore,

²For the case $x_2 < x_1$, the proof is very similar.

SI: Necessity when f is differentiable

First we will prove the **necessity**. Suppose f is increasing on \mathcal{I} . Let $0 < a < 1$, $x_1, x_2 \in \mathcal{I}$ and $x_1 \neq x_2$. Without loss of generality assume that² $x_1 < x_2$. Then, $x_1 < ax_1 + (1 - a)x_2 < x_2$ and therefore $ax_1 + (1 - a)x_2 \in \mathcal{I}$. By the mean value theorem, there exist s and t with $x_1 < s < ax_1 + (1 - a)x_2 < t < x_2$, such that $f(ax_1 + (1 - a)x_2) - f(x_1) = f'(s)(x_2 - x_1)(1 - a)$ and $f(x_2) - f(ax_1 + (1 - a)x_2) = f'(t)(x_2 - x_1)a$. Therefore,

$$\begin{aligned}(1 - a)f(x_1) - f(ax_1 + (1 - a)x_2) + af(x_2) &= \\ a[f(x_2) - f(ax_1 + (1 - a)x_2)] - (1 - a)[f(ax_1 + (1 - a)x_2) - f(x_1)] &= \\ a(1 - a)(x_2 - x_1)[f'(t) - f'(s)] &= \end{aligned}$$

Since $f(x)$ is strictly convex on \mathcal{I} , $f'(x)$ is increasing on \mathcal{I} and therefore, $f'(t) - f'(s) > 0$. Moreover, $x_2 - x_1 > 0$ and $0 < a < 1$. Thus, $(1 - a)f(x_1) - f(ax_1 + (1 - a)x_2) + af(x_2) > 0$, or equivalently, $f(ax_1 + (1 - a)x_2) < af(x_1) + (1 - a)f(x_2)$, which is what we wanted to prove in inequality (1).

²For the case $x_2 < x_1$, the proof is very similar.

SI: Sufficiency when f is differentiable

Suppose the inequality in inequality (1) holds. Then, taking limits,

SI: Sufficiency when f is differentiable

Suppose the inequality in inequality (1) holds. Then, taking limits,

$$\lim_{a \rightarrow 0} \frac{f(x_2 + a(x_1 - x_2)) - f(x_2)}{a} \leq f(x_1) - f(x_2). \text{ That is,}$$

SI: Sufficiency when f is differentiable

Suppose the inequality in inequality (1) holds. Then, taking limits,

$$\lim_{a \rightarrow 0} \frac{f(x_2 + a(x_1 - x_2)) - f(x_2)}{a} \leq f(x_1) - f(x_2). \text{ That is,}$$

$$f'(x_2)(x_1 - x_2) \leq f(x_1) - f(x_2) \quad (2)$$

Similarly, we can show that

$$f'(x_1)(x_2 - x_1) \leq f(x_2) - f(x_1) \quad (3)$$

Adding the left and right hand sides of inequalities in (2) and (3), and multiplying the resultant inequality by -1 gives us

SI: Sufficiency when f is differentiable

Suppose the inequality in inequality (1) holds. Then, taking limits,

$$\lim_{a \rightarrow 0} \frac{f(x_2 + a(x_1 - x_2)) - f(x_2)}{a} \leq f(x_1) - f(x_2). \text{ That is,}$$

$$f'(x_2)(x_1 - x_2) \leq f(x_1) - f(x_2) \quad (2)$$

Similarly, we can show that

$$f'(x_1)(x_2 - x_1) \leq f(x_2) - f(x_1) \quad (3)$$

Adding the left and right hand sides of inequalities in (2) and (3), and multiplying the resultant inequality by -1 gives us

$$(f'(x_2) - f'(x_1))(x_2 - x_1) \geq 0 \quad (4)$$

We now need to prove that the inequality in (4) is strict.

SI: Sufficiency when f is differentiable (contd)

Using the mean value theorem, $\exists z = x_1 + t(x_2 - x_1)$ for $t \in (0, 1)$ such that

SI: Sufficiency when f is differentiable (contd)

Using the mean value theorem, $\exists z = x_1 + t(x_2 - x_1)$ for $t \in (0, 1)$ such that

$$f(x_2) - f(x_1) = f'(z)(x_2 - x_1) \quad (5)$$

Since (4) holds for any $x_1, x_2 \in \mathcal{I}$, it also hold for $x_2 = z$. Therefore,

SI: Sufficiency when f is differentiable (contd)

Using the mean value theorem, $\exists z = x_1 + t(x_2 - x_1)$ for $t \in (0, 1)$ such that

$$f(x_2) - f(x_1) = f'(z)(x_2 - x_1) \quad (5)$$

Since (4) holds for any $x_1, x_2 \in \mathcal{I}$, it also holds for $x_2 = z$. Therefore,

$$(f'(z) - f'(x_1))(x_2 - x_1) = \frac{1}{t}(f'(z) - f'(x_1))(z - x_1) \geq 0$$

Additionally using (5), we get

SI: Sufficiency when f is differentiable (contd)

Using the mean value theorem, $\exists z = x_1 + t(x_2 - x_1)$ for $t \in (0, 1)$ such that

$$f(x_2) - f(x_1) = f'(z)(x_2 - x_1) \quad (5)$$

Since (4) holds for any $x_1, x_2 \in \mathcal{I}$, it also holds for $x_2 = z$. Therefore,

$$(f'(z) - f'(x_1))(x_2 - x_1) = \frac{1}{t}(f'(z) - f'(x_1))(z - x_1) \geq 0$$

Additionally using (5), we get

$$f(x_2) - f(x_1) = (f'(z) - f'(x_1))(x_2 - x_1) + f'(x_1)(x_2 - x_1) \geq f'(x_1)(x_2 - x_1) \quad (6)$$

SI: Sufficiency when f is differentiable (contd)

Suppose equality holds in (4) for some $x_1 \neq x_2$. Then it holds in (6) for the same x_1 and x_2 . That is,

SI: Sufficiency when f is differentiable (contd)

Suppose equality holds in (4) for some $x_1 \neq x_2$. Then it holds in (6) for the same x_1 and x_2 . That is,

$$f(x_2) - f(x_1) = f'(x_1)(x_2 - x_1) \quad (7)$$

Substituting x_2 with $x_1 + a(x_2 - x_1)$ and applying (6), we get

SI: Sufficiency when f is differentiable (contd)

Suppose equality holds in (4) for some $x_1 \neq x_2$. Then it holds in (6) for the same x_1 and x_2 . That is,

$$f(x_2) - f(x_1) = f'(x_1)(x_2 - x_1) \quad (7)$$

Substituting x_2 with $x_1 + a(x_2 - x_1)$ and applying (6), we get

$$f(x_1) + af'(x_1)(x_2 - x_1) \leq f(x_1 + a(x_2 - x_1)) \quad (8)$$

Further using (1) and 7, we can derive that

SI: Sufficiency when f is differentiable (contd)

Suppose equality holds in (4) for some $x_1 \neq x_2$. Then it holds in (6) for the same x_1 and x_2 . That is,

$$f(x_2) - f(x_1) = f'(x_1)(x_2 - x_1) \quad (7)$$

Substituting x_2 with $x_1 + a(x_2 - x_1)$ and applying (6), we get

$$f(x_1) + af'(x_1)(x_2 - x_1) \leq f(x_1 + a(x_2 - x_1)) \quad (8)$$

Further using (1) and 7, we can derive that

$$f(x_1 + a(x_2 - x_1)) < (1 - a)f(x_1) + af(x_2) = f(x_1) + af'(x_1)(x_2 - x_1) \quad (9)$$

Contradiction between (8) and (9)!

SI: Sufficiency when f is differentiable (contd)

Thus, equations 8 and 9 contradict each other.

$$f(x_1) + af'(x_1)(x_2 - x_1) \leq f(x_1 + a(x_2 - x_1))$$

$$f(x_1 + a(x_2 - x_1)) < (1 - a)f(x_1) + af(x_2) = f(x_1) + af'(x_1)(x_2 - x_1)$$

Therefore, equality in 4 cannot hold for any $x_1 \neq x_2$, implying that

$$(f'(x_2) - f'(x_1))(x_2 - x_1) > 0$$

that is, $f'(x)$ is increasing and therefore f is convex on \mathcal{I} . □

Strict Concavity

- A differentiable function f is said to be *strictly concave* on an open interval \mathcal{I} , iff, $f'(x)$ is decreasing on \mathcal{I} .
- Recall from theorem 7, the graphical interpretation of the first derivative $f'(x)$; $f'(x) < 0$ implies that $f(x)$ is decreasing at x .
- Similarly, $f'(x)$ is (strictly) monotonically decreasing when

Strict Concavity

- A differentiable function f is said to be *strictly concave* on an open interval \mathcal{I} , iff, $f'(x)$ is decreasing on \mathcal{I} .
- Recall from theorem 7, the graphical interpretation of the first derivative $f'(x)$; $f'(x) < 0$ implies that $f(x)$ is decreasing at x .
- Similarly, $f'(x)$ is (strictly) monotonically decreasing when $f''(x) > 0$. This gives us a sufficient condition for the concavity of a function:

Claim

If at all points in an open interval \mathcal{I} , $f(x)$ is doubly differentiable and if $f''(x) < 0$, $\forall x \in \mathcal{I}$, then the slope of the function is always decreasing with x and the graph is strictly concave.

Strict Concavity

On the other hand, if the function is strictly concave and doubly differentiable in \mathcal{I} , then $f''(x) \leq 0$, $\forall x \in \mathcal{I}$. This is illustrated in Figure 9.

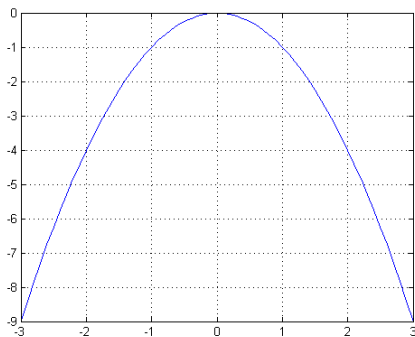


Figure 9:

Strict Concavity (slopeless interpretation)

There is also a slopeless interpretation of concavity as stated in the following theorem:

Claim

A differentiable function f is strictly concave on an open interval \mathcal{I} , iff

$$f(ax_1 + (1 - a)x_2) > af(x_1) + (1 - a)f(x_2) \quad (10)$$

whenever $x_1, x_2 \in \mathcal{I}$, $x_1 \neq x_2$ and $0 < a < 1$.

The proof is similar to that for theorem 12.

Convex & Concave Regions and Inflection Point

Study the function $f(x) = x^3 - x + 2$.

Convex & Concave Regions and Inflection Point

Study the function $f(x) = x^3 - x + 2$. It's slope decreases as x increases to 0 ($f'(x) < 0$) and then the slope increases beyond $x = 0$ ($f'(x) > 0$). The point 0, where the $f'(x)$ changes sign is called the *inflection point*; the graph is strictly concave for $x < 0$ and strictly convex for $x > 0$. See Figure 10.

