Theorem 6.6.3

Assume the function *f* is either increasing or decreasing and differentiable at *y*. Let $y = f^{-1}(x)$ and $f'(y) \neq 0$. Then f^{-1} is differentiable at *x* and we have

$$(f^{-1})'(x) = \frac{1}{f'(f^{-1}(x))}$$

Proof. By definition,

$$(f^{-1})'(x) = \lim_{\Delta x \to 0} \frac{f^{-1}(x + \Delta x) - f^{-1}(x)}{\Delta x}.$$

Let $y + \Delta y = f^{-1}(x + \Delta x)$, i.e $x + \Delta x = f(y + \Delta y)$. Thus, $\Delta x = f(y + \Delta y) - x = f(y + \Delta y) - f(y)$. Therefore,

$$\frac{f^{-1}(x+\Delta x)-f^{-1}(x)}{\Delta x}=\frac{(y+\Delta y)-y}{\Delta x}=\frac{\Delta y}{\Delta x}.$$

Notice that

$$\lim_{\Delta y \to 0} \frac{\Delta x}{\Delta y} = \lim_{\Delta y \to 0} \frac{f(y + \Delta y) - f(y)}{\Delta y} = f'(y) \neq 0$$

On the other hand, since f is differentiable, it is continuous thus $\lim_{\Delta y\to 0} \Delta x = \lim_{\Delta y\to 0} (f(y + \Delta y) - f(y)) = f(y) - f(y) = 0$, i.e when $\Delta y \to 0$ we have $\Delta x \to 0$. Therefore,

$$(f^{-1})'(x) = \lim_{\Delta x \to 0} \frac{f^{-1}(x + \Delta x) - f^{-1}(x)}{\Delta x} = \lim_{\Delta x \to 0} \frac{\Delta y}{\Delta x}$$
$$= \lim_{\Delta x \to 0} \frac{1}{\frac{\Delta x}{\Delta y}}$$
$$= \frac{1}{\lim_{\Delta x \to 0} \frac{\Delta x}{\Delta y}}$$
$$= \frac{1}{\lim_{\Delta x \to 0} \frac{\Delta x}{\Delta y}}$$
$$= \frac{1}{f'(y)} = \frac{1}{f'(f^{-1}(x))}.$$

■ Example 6.5

Find $\frac{d}{dx} \arcsin x$.

<u>Solution</u>: we have $\arcsin x = \sin^{-1}(x)$ thus

$$\frac{d}{dx} \arcsin x = \frac{d}{dx} \sin^{-1}(x) = \frac{1}{\frac{d}{dy}(\sin y)} \bigg|_{y = \arcsin x} = \frac{1}{\cos(\arcsin x)}$$

Since $\cos x = \sqrt{1 - \sin^2 x}$, we get

$$\frac{d}{dx} \arcsin x = \frac{1}{\cos(\arcsin x)} = \frac{1}{\sqrt{1 - \sin^2(\arcsin x)}} = \frac{1}{\sqrt{1 - x^2}}$$

6.7 Mean Value Theorem

Theorem 6.7.1 — Fermat's theorem

If *f* has a local maximum or minimum at x_0 and *f* is differentiable at x_0 then we have $f'(x_0) = 0$.

Proof. Let us prove it when f has a local maximum. The minimum case can be proven in a similar way.

Since f is differentiable at x_0 , we have

$$f'(x_0) = \lim_{h \to 0} \frac{f(x_0 + h) - f(x_0)}{h} = \lim_{h \to 0+} \frac{f(x_0 + h) - f(x_0)}{h} = \lim_{h \to 0-} \frac{f(x_0 + h) - f(x_0)}{h}$$

Moreover, since f has a maximum at x_0 , we have $f(x_0 + h) \le f(x_0)$, i.e $f(x_0 + h) - f(x_0) \le 0$, if |h| is sufficiently small. Therefore,

• if h > 0, we have $\frac{f(x_0+h)-f(x_0)}{h} \le 0 \Rightarrow \lim_{h \to 0+} \frac{f(x_0+h)-f(x_0)}{h} \le 0$. • if h < 0, we have $\frac{f(x_0+h)-f(x_0)}{h} > 0 \Rightarrow \lim_{h \to 0-} \frac{f(x_0+h)-f(x_0)}{h} > 0$.

• If
$$h < 0$$
, we have $\frac{f(x_0 + h)}{h} \ge 0 \Rightarrow \lim_{h \to 0^-} \frac{f(x_0 + h)}{h}$
Thus

$$\lim_{h \to 0} \frac{f(x_0 + h) - f(x_0)}{h} = \lim_{h \to 0+} \frac{f(x_0 + h) - f(x_0)}{h} \le 0$$

and

$$\lim_{h \to 0} \frac{f(x_0 + h) - f(x_0)}{h} = \lim_{h \to 0^-} \frac{f(x_0 + h) - f(x_0)}{h} \ge 0$$

i.e $(f'(x_0) \le 0) \land (f'(x_0) \ge 0) \Leftrightarrow f'(x_0) = 0.$

Theorem 6.7.2 — Rolle's theorem

Let f a function continuous on [a,b] and differentiable in (a,b) such that f(a) = f(b). Then $\exists c \in (a,b), f'(c) = 0$.

Proof. The case when f is constant on [a,b] is obvious since $\forall x \in [a,b], f'(x) = 0$. Assume now that f is not constant, continuous on [a,b], differentiable on (a,b) such that f(a) = f(b). By the extreme value theorem, we know that f attains a minimum and maximum on [a,b]. These values cannot be equal to f(a) = f(b) at the same time otherwise it would mean that f is constant which contradicts the assumption. Therefore, if the maximum of $f(x) \neq f(a)$ (and f(b)), $\exists c \in (a,b)$ such that f(c) is this maximum of f and from Fermat's theorem we have f'(c) = 0. The same reasoning holds if we consider the minimum of f. Therefore $\exists c \in (a,b), f'(c) = 0$.

Theorem 6.7.3 — Mean Value Theorem

Let *f* a function continuous on [a,b] and differentiable on (a,b). Then $\exists c \in (a,b)$ such that

$$f(b) - f(a) = f'(c)(b - a).$$

Note that the previous equality can be rewritten as

$$f'(c) = \frac{f(b) - f(a)}{b - a},$$

the left-hand side corresponds to the slope of the line passing through points (a, f(a)) and (b, f(b)) while f'(c) is the slope of the tangent line to f at the point (c, f(c)). The Mean value theorem means that there exists a tangent line to f which is parallel to the line passing through the endpoints of f.

Proof. Let denote g the secant line passing through (a, f(a)) and (b, f(b)). We know that the slope of this line is $\frac{f(b)-f(a)}{b-a}$ and the equation of g is given by

$$g(x) = f(a) + \frac{f(b) - f(a)}{b - a}(x - a).$$

Now, we set

$$h(x) = f(x) - g(x) = f(x) - f(a) - \frac{f(b) - f(a)}{b - a}(x - a).$$

Clearly, *h* is continuous on [a,b] and differentiable on (a,b) and we have h(a) = h(b) = 0. Therefore, by Rolle's theorem $\exists c \in (a,b), h'(c) = 0$. But

$$h'(x) = f'(x) - \frac{f(b) - f(a)}{b - a},$$

and the condition h'(c) = 0 is equivalent to

$$f'(c) - \frac{f(b) - f(a)}{b - a} = 0 \Leftrightarrow f'(c) = \frac{f(b) - f(a)}{b - a}.$$

Theorem 6.7.4 — Generalized (Cauchy's) Mean Value Theorem Let two functions f and g continuous on [a,b] and differentiable on (a,b). Then $\exists c \in (a,b)$ such that

$$f'(c)[g(b) - g(a)] = g'(c)[f(b) - f(a)].$$

Proof. Set

$$h(x) = [f(x) - f(a)][g(b) - g(a)] - [f(b) - f(a)][g(x) - g(a)].$$

Clearly, h is continuous on [a,b] and differentiable on (a,b) and h(a) = h(b) = 0. Therefore, Rolle's theorem holds and we get $\exists c \in (a,b), h'(c) = 0$. But,

$$h'(x) = f'(x)[g(b) - g(a)] - [f(b) - f(a)]g'(x),$$

thus the condition h'(c) = 0 is equivalent to

$$f'(c)[g(b) - g(a)] - [f(b) - f(a)]g'(c) = 0 \Leftrightarrow f'(c)[g(b) - g(a)] = [f(b) - f(a)]g'(c).$$

The Mean Value Theorem and its generalized version have several important consequences we will investigate now.

Theorem 6.7.5

Let a function f continuous on [a,b] and differentiable on (a,b).

- If $\forall x \in (a,b), f'(x) > 0$ then *f* is increasing on [a,b],
- If $\forall x \in (a,b), f'(x) < 0$ then *f* is decreasing on [a,b]

Proof. We have $\forall x_1, x_2 \in [a, b], x_1 < x_2$. We have $[x_1, x_2] \subset [a, b]$ and $(x_1, x_2) \subset [a, b]$. Thus, f is continuous on $[x_1, x_2]$ and differentiable on (x_1, x_2) and the Mean Value Theorem can be applied to f on $[x_1, x_2]$:

$$\exists c \in (x_1, x_2), f(x_2) - f(x_1) = f'(c)(x_2 - x_1).$$

Therefore

- if $\forall x \in [a,b], f'(x) > 0$ then f'(c) > 0 and we get that $f(x_2) f(x_1) = f'(c)(x_2 x_1) > 0 \Leftrightarrow f(x_2) > f(x_1)$, i.e *f* is increasing,
- if $\forall x \in [a,b], f'(x) < 0$ then f'(c) < 0 and we get that $f(x_2) f(x_1) = f'(c)(x_2 x_1) < 0 \Leftrightarrow f(x_2) < f(x_1)$, i.e *f* is decreasing.

Proposition 6.7.6

Let *f* a function continuous on [a,b] and differentiable on (a,b). Assume that $\forall x \in (a,b), f'(x) = 0$ then *f* is constant on [a,b].

Proof. We have $\forall x_1, x_2 \in (a, b)$, f is continuous on $[x_1, x_2]$ and differentiable on (x_1, x_2) and the Mean Value Theorem can be applied to f on $[x_1, x_2]$:

$$\exists c \in (x_1, x_2), f(x_2) - f(x_1) = f'(c)(x_2 - x_1).$$

Since $\forall x \in [a,b], f'(x) = 0$, we have f'(c) = 0 thus $f(x_2) - f(x_1) = 0 \Leftrightarrow f(x_1) = f(x_2)$, i.e *f* is a constant function.

Corollary 6.7.7

Assume that $\forall x \in [a,b], f'(x) = g'(x)$ then $\exists C$ (constant) such that $\forall x \in [a,b], f(x) = g(x) + C$.

Proof. Denote $\forall x \in [a,b], h(x) = f(x) - g(x)$ then h'(x) = f'(x) - g'(x) = 0. From Proposition 6.7.6, h is a constant function, i.e $\exists C, h(x) = C$. Therefore, $f(x) - g(x) = C \Leftrightarrow f(x) = g(x) + C$.

6.8 Convexity

Definition 6.8.1

A function f is convex over [a,b] if

$$\forall x, y \in [a, b], x < y, \forall \lambda \in [0, 1], f(\lambda x + (1 - \lambda)y) \le \lambda f(x) + (1 - \lambda)f(y)$$

A function f is concave over [a,b] if

$$\forall x, y \in [a, b], x < y, \forall \lambda \in [0, 1], f(\lambda x + (1 - \lambda)y) \ge \lambda f(x) + (1 - \lambda)f(y).$$

These definitions are illustrated in Figure 6.2.

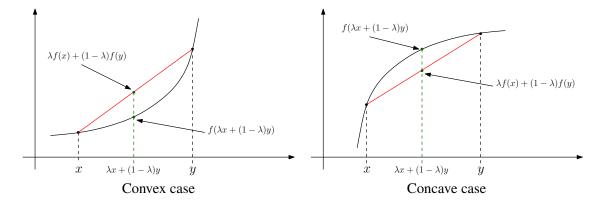


Figure 6.2: Example of convex and concave functions.

Proposition 6.8.1

A function f is a convex function on [a,b] if and only if $\forall x, y \in [a,b], x < y$ and $\forall z \in [x,y]$, we have

$$\frac{f(z) - f(x)}{z - x} \le \frac{f(y) - f(z)}{y - z}$$

5

Proof. \Rightarrow : Since $\forall x, y \in [a, b]$ we choose $z \in [x, y]$, we can write $z = \lambda x + (1 - \lambda)y$ where $\lambda \in [0, 1]$. If we solve for λ , we find

$$\lambda = \frac{y-z}{y-x}$$
 and $1-\lambda = \frac{z-x}{y-x}$.

Since f is convex, we have

$$f(z) = f(\lambda x + (1 - \lambda)y) \le \lambda f(x) + (1 - \lambda)f(y).$$

Therefore, we have

$$\begin{aligned} f(z) - f(x) &\leq \lambda f(x) + (1 - \lambda)f(y) - f(x) = (1 - \lambda)(f(y) - f(x)) = \frac{z - x}{y - x}(f(y) - f(x)) \\ &\Leftrightarrow \frac{f(z) - f(x)}{z - x} \leq \frac{f(y) - f(x)}{y - x}. \end{aligned}$$

On the other hand, we have

$$-f(z) \ge -\lambda f(x) - (1 - \lambda)f(y)$$

$$\Leftrightarrow f(y) - f(z) \ge f(y) - \lambda f(x) - (1 - \lambda)f(y) = \lambda (f(y) - f(x)) = \frac{y - z}{y - x} (f(y) - f(x))$$

$$\Leftrightarrow \frac{f(y) - f(x)}{y - x} \le \frac{f(y) - f(z)}{y - z}.$$

Therefore, combining the two obtained inequalities, we get the expected result:

$$\frac{f(z) - f(x)}{z - x} \le \frac{f(y) - f(z)}{y - z}$$

⇐: Now assume that $\forall x, y \in [a, b]$ and $\forall z \in [x, y]$ and we have

$$\frac{f(z) - f(x)}{z - x} \le \frac{f(y) - f(z)}{y - z}.$$

As in the first part of the proof, we can write $z = \lambda x + (1 - \lambda)y, \lambda \in [0, 1]$ and then

c()

$$\lambda = \frac{y-z}{y-x}$$
 and $1-\lambda = \frac{z-x}{y-x}$.

c()

Thus, we get

$$\begin{split} \frac{f(z) - f(x)}{(1 - \lambda)(y - x)} &\leq \frac{f(y) - f(z)}{\lambda(y - x)} \\ \Leftrightarrow \lambda(f(z) - f(x)) &\leq (1 - \lambda)(f(y) - f(z)) \\ \Leftrightarrow f(z)(\lambda + (1 - \lambda)) &\leq \lambda f(x) + (1 - \lambda)f(y) \\ \Leftrightarrow f(z) &\leq \lambda f(x) + (1 - \lambda)f(y) \\ \Leftrightarrow f(\lambda x + (1 - \lambda)y) &\leq \lambda f(x) + (1 - \lambda)f(y), \end{split}$$

i.e f is convex.