

Convex sets and (quasi)-concave/convex functions

1.A. Convex Sets

Definition 1.A.1 (Convex Set). A set \mathcal{S} of points in n -dimensional space is called *convex* if, given any two points $x^a = (x_1^a, x_2^a, \dots, x_n^a)$ and $x^b = (x_1^b, x_2^b, \dots, x_n^b)$ in \mathcal{S} and any real number $\alpha \in [0, 1]$, the point $\alpha x^a + (1 - \alpha)x^b = (\alpha x_1^a + (1 - \alpha)x_1^b, \dots, \alpha x_n^a + (1 - \alpha)x_n^b)$ is also in \mathcal{S} .

A geometric test of convexity is that given any two points of the set, the whole line segment joining them should lie in the set.

Figure 1 and 2 are examples of *convex* sets. Please be aware that to apply the geometric test of convexity, we need to ensure that for *any* two points of the set, the whole line segment lie in the set.

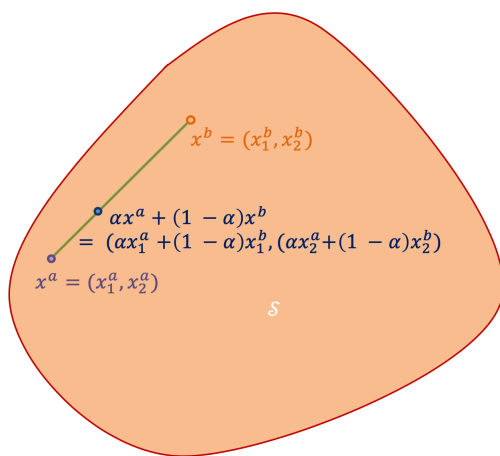


Figure 1: Convex Set (a)

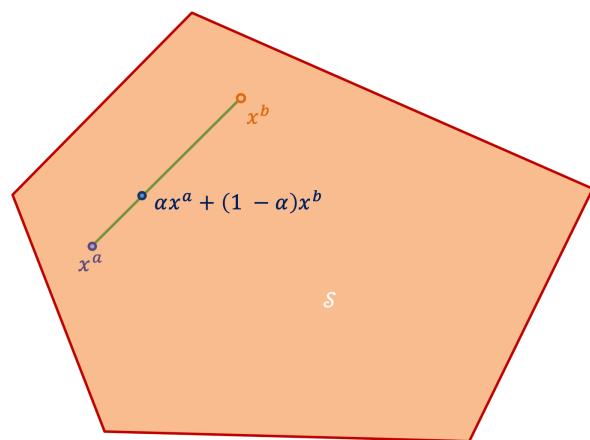


Figure 2: Convex Set (b)

Figure 3 and 4 are examples of *non-convex* sets. The sets are *non-convex*, since there exist points x^a and x^b and a real number α , such that the point $\alpha x^a + (1 - \alpha)x^b$ is not inside the set.

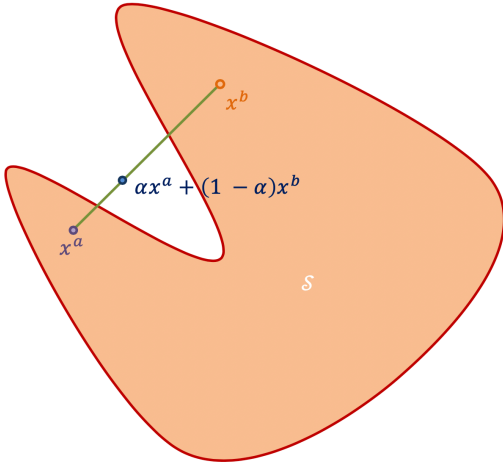


Figure 3: Non-Convex Set (a)

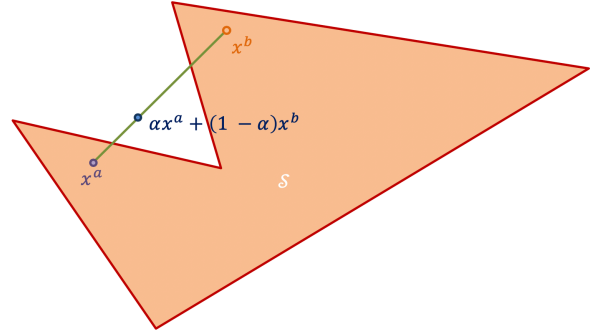


Figure 4: Non-Convex Set (b)

1.B. Quasi-convex/concave functions

Definition 1.B.1 (Quasi-convex Function). A function $f : \mathcal{S} \rightarrow \mathbb{R}$, defined on a convex set $\mathcal{S} \subset \mathbb{R}^N$, is *quasi-convex* if the set $\{x | f(x) \leq c\}$ is convex for all $c \in \mathbb{R}$, or equivalently, if

$$f(\alpha x^a + (1 - \alpha)x^b) \leq \max\{f(x^a), f(x^b)\}, \quad (1)$$

for all x^a, x^b and for all $\alpha \in [0, 1]$.

We show the equivalence of

- (a) The set $\{x | f(x) \leq c\}$ is convex for all $c \in \mathbb{R}$;
- (b) $f(\alpha x^a + (1 - \alpha)x^b) \leq \max\{f(x^a), f(x^b)\}$, for all x^a, x^b and for all $\alpha \in [0, 1]$.

Proof. (a) \implies (b): Since (a) holds for all $c \in \mathbb{R}$, for any x^a and x^b , we could set $c = \max\{f(x^a), f(x^b)\}$. Then since $f(x^a) \leq \max\{f(x^a), f(x^b)\} = c$, $f(x^b) \leq \max\{f(x^a), f(x^b)\} = c$, by (a), we have $f(\alpha x^a + (1 - \alpha)x^b) \leq c = \max\{f(x^a), f(x^b)\}$ for any $\alpha \in [0, 1]$. Thus, (b) holds.

(b) \implies (a): Equivalently, we show “not (a) \implies not (b)”.

If (a) fails, then there exists x^a, x^b, c and $\alpha \in [0, 1]$ such that $f(x^a) \leq c$ and $f(x^b) \leq c$ but $f(\alpha x^a + (1 - \alpha)x^b) > c$. Then $f(\alpha x^a + (1 - \alpha)x^b) > c \geq \max\{f(x^a), f(x^b)\}$. Thus, (b) fails for these values of x^a, x^b and α . \square

The definition of *quasi-concave* is given in Definition 1.B.2 below.

Definition 1.B.2 (Quasi-concave Function). A function $f : \mathcal{S} \rightarrow \mathbb{R}$, defined on a convex set $\mathcal{S} \subset \mathbb{R}^N$, *quasi-concave* if the set $\{x | f(x) \geq c\}$ is convex for all $c \in \mathbb{R}$, or equivalently, if $f(\alpha x^a + (1 - \alpha)x^b) \geq \min\{f(x^a), f(x^b)\}$, for all x^a, x^b and for all $\alpha \in [0, 1]$.

1.C. Quasi-convexity (quasi-concavity) and convexity (concavity)

The *quasi* in Definition 1.B.1 and 1.B.2 serves to distinguish them from stronger properties of *convexity* and *concavity*. Formally, we define convexity as follows.

Definition 1.C.1 (Convex Function). A function $f : \mathcal{S} \rightarrow \mathbb{R}$, defined on a convex set $\mathcal{S} \subset \mathbb{R}^N$, is *convex* if

$$f(\alpha x^a + (1 - \alpha)x^b) \leq \alpha f(x^a) + (1 - \alpha)f(x^b), \quad (2)$$

for all x^a, x^b and for all $\alpha \in [0, 1]$.

(2) *convexity* implies (1) *quasi-convexity* since

$$\begin{aligned} f(\alpha x^a + (1 - \alpha)x^b) &\stackrel{(2)}{\leq} \alpha f(x^a) + (1 - \alpha)f(x^b) \\ &\leq \alpha \max\{f(x^a), f(x^b)\} + (1 - \alpha) \max\{f(x^a), f(x^b)\} \\ &= \max\{f(x^a), f(x^b)\}. \end{aligned}$$

In other words, a convex function must be quasi-convex.

Similarly, we could define concavity and compare it with *quasi-concavity*.

Definition 1.C.2 (Concave Function). A function $f : \mathcal{S} \rightarrow \mathbb{R}$, defined on a convex set $\mathcal{S} \subset \mathbb{R}^N$, is *concave* if

$$f(\alpha x^a + (1 - \alpha)x^b) \geq \alpha f(x^a) + (1 - \alpha)f(x^b), \quad (3)$$

for all x^a, x^b and for all $\alpha \in [0, 1]$.

Following the same logic, we could show that a concave function must be quasi-concave.

Figure 5 provides a graphical illustration of a *concave* function for the one-variable case.

The red dot (LHS of (3)) is always higher than the green dot (RHS of (3)).

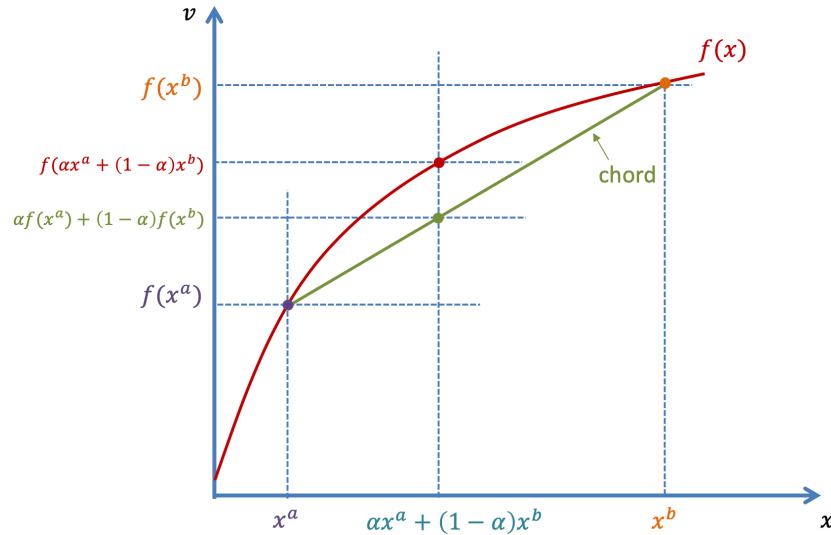


Figure 5: Concave Function

The graph of the function lies on or above the chord joining any two points of it.

1.D. More on concave functions

An alternative interpretation of a concave function is sometimes useful. Consider the $(n + 1)$ -dimensional space consisting of points like (x, v) where x is an n -dimensional vector and v is a scalar. Define the set $\mathcal{F} = \{(x, v) | v \leq f(x)\}$.

Then, we make the following claim:

Claim. f is a concave function if and only if \mathcal{F} is a convex set.

Proof. “ \implies ”: To prove that \mathcal{F} is a convex set, we need to show that for all (x^a, v^a) and (x^b, v^b) that satisfy $v^a \leq f(x^a)$ and $v^b \leq f(x^b)$ and any real number $\alpha \in [0, 1]$, we have $\alpha v^a + (1 - \alpha)v^b \leq f(\alpha x^a + (1 - \alpha)x^b)$.

By concavity of f , we know that for all x^a and x^b and for all $\alpha \in [0, 1]$, (3) holds.

Therefore, for all (x^a, v^a) and (x^b, v^b) that satisfy $v^a \leq f(x^a)$ and $v^b \leq f(x^b)$ and any real number $\alpha \in [0, 1]$,

$$\begin{aligned} & \alpha v^a + (1 - \alpha)v^b - [f(\alpha x^a + (1 - \alpha)x^b)] \\ & \stackrel{(3)}{\leq} \underbrace{\alpha v^a + (1 - \alpha)v^b}_{(3)} - [\alpha f(x^a) + (1 - \alpha)f(x^b)] \\ & \stackrel{v^a \leq f(x^a) \text{ and } v^b \leq f(x^b)}{\leq} \underbrace{\alpha v^a + (1 - \alpha)v^b}_{v^a \leq f(x^a) \text{ and } v^b \leq f(x^b)} - [\alpha v^a + (1 - \alpha)v^b] = 0 \end{aligned}$$

Therefore, $\alpha v^a + (1 - \alpha)v^b \leq f(\alpha x^a + (1 - \alpha)x^b)$ and convexity of set \mathcal{F} follows.

“ \Leftarrow ”: To prove that F is concave, we need to show that for all x^a, x^b and all $\alpha \in [0, 1]$, (3) holds.

For any x^a and x^b , set $v^a = f(x^a)$ and $v^b = f(x^b)$, so that $v^a \leq f(x^a)$ and $v^b \leq f(x^b)$ are satisfied, i.e., $(x^a, v^a) \in \mathcal{F}$ and $(x^b, v^b) \in \mathcal{F}$. Then by convexity of set \mathcal{F} , for any real number $\alpha \in [0, 1]$, we have $\alpha v^a + (1 - \alpha)v^b \leq f(\alpha x^a + (1 - \alpha)x^b) \underbrace{\implies}_{v^a=f(x^a), v^b=f(x^b)} \alpha f(x^a) + (1 - \alpha)f(x^b) \leq f(\alpha x^a + (1 - \alpha)x^b)$, and concavity of the function f follows. \square

The claim could be more easily understood graphically. Figure 6 illustrates the case with a scalar variable x . The function f is the red curve. The set \mathcal{F} is the area shaded in orange. The claim means that the concave function f traps a convex set \mathcal{F} underneath its graph. And it is clear from Figure 6.

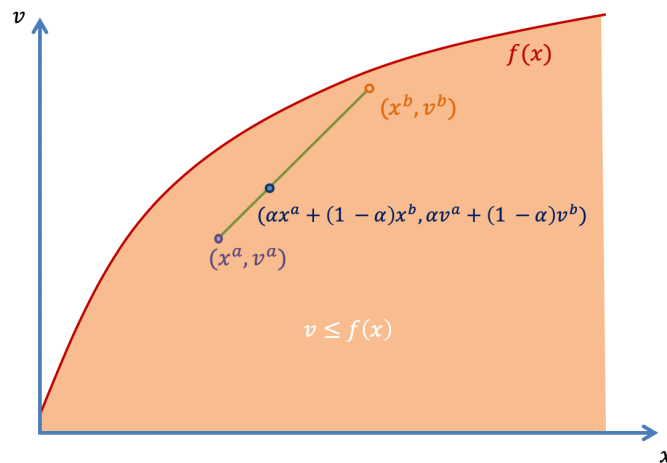


Figure 6: Concave Function

For differentiable functions, the concavity property could be interpreted in terms of first-order derivatives. We have also shown a similar graph to Figure 7 below, and interpreted *concavity* graphically: the graph of the function lies on or above the chord joining any two points of it. To express the concavity of $f(x)$ in terms of its derivative, we now draw the tangent to $f(x)$ at x^a . The requirement of concavity says that the graph of the function should lie on or below the tangent. Or expressed differently,

$$f_x(x^a)(x^b - x^a) \geq f(x^b) - f(x^a),$$

where $f_x(x^a)$ is the slope of the tangent to $f(x)$ at x^a .

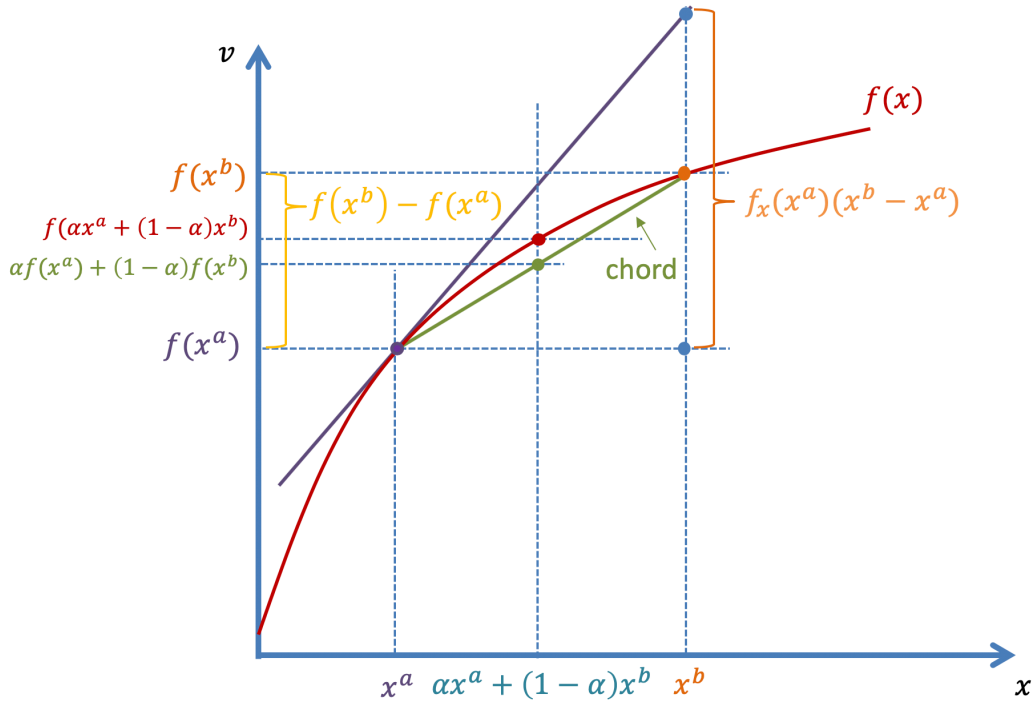


Figure 7: Concave Function

Such an expression holds for higher dimensions. The result is summarized in Proposition 1.D.1 below.

Proposition 1.D.1 (Concave Function). *A differentiable function $f : \mathcal{S} \rightarrow \mathbb{R}$, defined on a convex set $\mathcal{S} \subset \mathbb{R}^N$, is concave if and only if*

$$f_x(x^a)(x^b - x^a) \geq f(x^b) - f(x^a), \quad (7.1)$$

for all $x^a, x^b \in \mathcal{S}$.

For twice continuously differentiable functions, this concavity property could be interpreted in terms of second-order derivatives.

Proposition 1.D.2. *The (twice continuously differentiable) function $f : \mathcal{S} \rightarrow \mathbb{R}$ is concave if and only if f_{xx} is negative semi-definite for every $x \in \mathcal{S}$. If f_{xx} is negative definite for every $x \in \mathcal{S}$, then the function is strictly concave.*