# Concave and convex functions

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**Abstract**: This research is a review of the topic of convex and concave functions. We presented some important theories and definitions related to this topic and presented some of the relationships that link these concepts.

**Keywords:** convex set, concave function, GA-convex function

#### 1. Introduction

A function that translates a convex subset of a vector space to the set of real numbers is called a convex function. There are numerous uses for functions' concavity and convexity in demonstrating inequality. In 2004, Cha studied formulas pertaining to convex function theorems and derived a number of significant inequalities that were then used to solve conditional extremum problems and prove other inequalities [2]. Xia used the concavity, convexity, and continuity of functions to derive Jensen's inequality in 2005 [7]. Song and Wan's investigation of GA-convex functions in 2010 yielded a more succinct Hadamard-type inequality for GA-convex functions [5]. Convexity, monotonicity, and non-negativity are some crucial characteristics of convex functions that aid in the derivation and demonstration of inequalities.

#### 2. Basic definitions and theorems

**Definition 2.1:** [4] A set Y is convex where  $Y \subset \mathbb{R}^n$  if there is  $y', y'' \in Y$  the straight line segment between y' and y'' belongs entirely to Y, in another way  $\forall h \in [0,1]$  then

$$y^h = (1 - h)y' + hy'' \in Y \ \forall \ h \in [0,1].$$

**Definition 2. 2:** [4] A function g defined on a convex set N is concave where  $g: N \to R$ ,  $N \subset R^n$  if

$$(1-h)g(y') + hg(y'') \le g((1-h)y' + hy''),$$

where y'and y''  $\in N \ \forall \ h \in [0,1]$ .

g is completely concave if

$$(1-h)g(y') + hg(y'') < g((1-h)y' + hy'').$$

**Definition 2. 3:** [4] A function  $g: \mathbb{N} \to \mathbb{R}$ ,  $\mathbb{N} \subset \mathbb{R}^n$  is convex if

$$(1-h)g(y') + hg(y'') \ge g((1-h)y' + hy''),$$

where y'and y''  $\in N \ \forall \ h \in [0,1]$ .

g is completely convex if

$$(1-h)g(y') + hg(y'') > g((1-h)y' + hy'')$$

**Theorem 2.4:** [4] Let functions  $g_1, ..., g_n$  are convex (concave) and b1 > 0, ..., bn > 0, then  $H = b1 g_1 + ... + bn g_n$  is convex (concave).

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**Theorem 2.5:** [4] A  $\mathbb{C}^1$  function  $g: \mathbb{N} \to \mathbb{R}$ ,  $\mathbb{N} \subset \mathbb{R}^n$  is concave iff

$$g(z) - g(y) \le Dg(y)(z - y),$$

 $\forall$  y, z  $\in$  N, in other words

$$G(z) - g(y) \le \frac{\partial g}{\partial y_1}(y)(z_1 - y_1) + \dots + \frac{\partial g}{\partial y_n}(y)(z_n - y_n),$$

and g is convex iff  $g(z) - g(y) \ge Dg(y)(z - y)$ .

**Theorem 2. 6:** [4] A function g defined on a convex set N is concave where  $g: N \to R$ ,  $N \subset R^n$  if

$$Dg(y^*)(z-y^*) \le 0$$

 $\forall z \in \mathbb{N}$ , then  $y^*$  is a global maximizer of g on N.

Also,  $\forall z \in N$  if

$$Dg(y^*)(z - y^*) \ge 0$$

and g is convex then  $y^*$  is a global minimizer of g on N.

**Definition 2.7:** [4] A function g(y) be quasi concave if  $\mathcal{L}_{\eta} = \{y : g(y) \leq \eta\}$  be a convex set  $\forall$  constant  $\eta$ , where g(y) defined on  $\mathbb{N} \subset \mathbb{R}^n$ .

Also, g be quasi convex if N  $\eta = \{y : g(y) \ge \eta\}$  be a convex set  $\forall$  constant  $\eta$ .

**Definition 2. 8:** [4] A function g(y) be quasi concave if

$$g(hy + (1-h)z) \ge min(g(y), g(z))$$

 $\forall y, z \in N \text{ and } h \in [0, 1].$ 

And g is quasi convex if

$$g(hy + (1-h)z) \le max(g(y), g(z)).$$

**Theorem 2.9:** [1] Let  $Y, W \subseteq \mathbb{R}^n$  be convex sets and  $s \in \mathbb{R}$ . Then the following sets are convex

- 1.  $sY = \{q \in \mathbb{R}^n : \text{there is } y \in Y \ni q = sy\}$ 2.  $Y + W = \{q \in \mathbb{R}^n : \text{there is } \mathbf{y} \in Y \text{ and } z \in W \ni q = y + z\}$ 3.  $Y \cap W$ .
- **Definition 2.10**: [1]  $y \in \mathbb{R}^n$  be a convex combination of  $y_1, ..., y_r \in \mathbb{R}^n$  if

$$y = \sum_{i=1}^{r} \delta^{i} y^{i} = \begin{bmatrix} \sum_{i=1}^{r} \delta^{i} y_{1}^{i} \\ \vdots \\ \sum_{i=1}^{r} \delta^{i} y_{n}^{i} \end{bmatrix}$$

where  $\delta^i \in [0,1]$  and  $\sum_{i=1}^r \delta^i = 1$ .

**Definition 2.11:** [1]  $\forall y, y' \in D$  and  $\sigma, \delta > 0$ ,  $\sigma y + \delta y' \in D$  then  $D \subseteq \mathbb{R}^n$  be a convex cone.

**Definition 2.12:** [1] let  $v \in \mathbb{R}^n$ ,  $v \neq 0$ ,  $k \in \mathbb{R}$ , the hyperplane span by v and k is the (n-1) dimensional plane

$$H_{v,k} = \{ y \in \mathcal{R}^n : v \cdot y = k \}.$$

The sets  $\{y \in \mathbb{R}^n : v \cdot y \geq k\}$  and  $\{y \in \mathbb{R}^n : v \cdot y \leq k\}$  be half-space above and the half-space below the hyperplane  $H_{v,k}$ .

**Definition 2.13 :** [1] Let  $Y, W \subseteq \mathbb{R}^n$  are two nonempty sets

- There is  $v \neq 0 \in \mathbb{R}^n$  and  $k \in \mathbb{R}$  where  $v \cdot y \geq k \geq v \cdot z$  iff Y and W are separated via a hyperplane  $H_{v,k}$ , for all  $y \in Y$ ,  $z \in W$ .
- There is  $v \neq 0 \in \mathbb{R}^n$  and  $k \in \mathbb{R}$  where  $v \cdot y > k > v \cdot z$  iff Y and W are strictly separated via a hyperplane  $H_{v,k}$ , for all  $y \in Y$ ,  $z \in W$ .

**Definition 2.14:** [1] Let  $Y \subseteq \mathbb{R}^n$  such that  $Y \neq 0$ .

- There is  $v \neq 0 \in \mathbb{R}^n$  where  $v \cdot y \geq v \cdot y^*$  iff Y is supported at  $y^*$  for all  $y \in Y$
- There is  $v \neq 0 \in \mathbb{R}^n$  where  $v \cdot y > v \cdot y^*$  iff Y is strictly supported at  $y^*$  for all  $y \in Y$ ,  $y \neq y^*$

**Theorem 2.15:** [1]  $\hbar yp g$  is convex iff g is concave, such that

$$\hbar yp g = \{(y, z) \in \mathcal{R}^{n+1} : y \in \mathcal{R}^n \text{ and } z \leq g(y)\}.$$

**Theorem 2.16:** [1] the following are satisfies

- 1. *g* be convex.
- 2. epi g be a convex set , where  $epi g = \{(y, z) \in \mathbb{R}^{n+1} : y \in \mathbb{R}^n \text{ and } z \geq g(y)\}$  .
- 3.  $\forall y', y'' \in K, y^{\sigma} = \sigma y' + (1 \sigma)y'',$

$$g(y^{\sigma}) \leq \sigma g(y') + (1 - \sigma)g(y'').$$

**Theorem 2.17:** [8] If g(y) is concave downwards or upwards in  $[\partial, f]$ , then

$$g(\sigma y_1 + \delta y_2) \le (\sigma r \ge) \sigma g(y_1) + \delta g(y_2).$$

Where (x) is concave downwards or upwards in  $[\partial, f]$  if

$$g(y) \le (or \ge) \frac{y_2 - y}{y_2 - y_1} g(y_1) + \frac{y - y_1}{y_2 - y_1} g(y_2).$$

**Definition 2.18:** [8] g(y) is concave down or concave up on  $[\partial, f]$  if

$$g(\sigma y_1 + \delta y_2) \le (\sigma r \ge) \sigma g(y_1) + \delta g(y_2)$$

where  $y_1$ ,  $y_2 \in [\partial, f]$ ,  $\sigma > 0$ ,  $\delta > 0$   $\sigma + \delta = 1$ .

**Lemma 2.19:** [8] For any  $y_1$ ,  $y_2$ , ...,  $y_n \in [\partial, f]$  there is

$$g\left(\frac{y_1 + y_2 + \dots + y_n}{n}\right) \le or \ge \frac{g(y_1) + g(y_2) + \dots + g(y_n)}{n},$$

where g(y) is convex downwards or upwards on  $[\partial, f]$ .

**Definition 2.20:** [6]  $\forall y_1, y_2 \in J \in (0, +\infty)$  and  $h \subseteq (0,1)$ , there is

$$g(y_1^h y_2^{1-h}) \le hg(y_1) + (1-h)g(y_2).$$

Then g(y) be a GA-sub convex function on J.

**Theorem 2.21:** [8] If g(y) be a GA-convex function on  $(\partial, f) \in (0, +\infty)$ , then  $g(e^y)$  be a GA-sub convex function on  $(\ln \partial, \ln f)$ ,  $\forall y_1, y_2 \in (\partial, f)$  and  $h \in (0,1)$ .

**Theorem 2.22:** [8] If g(y) be a GA-concave on  $J \in (0, +\infty)$ ,  $y_i \in J$ ,  $\lambda_i \in \mathcal{R}$  (j = 1, 2, ..., r),  $\lambda_1 + \lambda_2 + \cdots + \lambda_r = 1$  then

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$$g(y_1^{\lambda_1}y_2^{\lambda_2} \dots y_k^{\lambda_k}) \leq \lambda_1 g(y_1) + \lambda_2 g(y_2) + \dots + \lambda_k g(y_k) (\sum_{j=1}^k \lambda_j = 1, \lambda_j > 0).$$

**Theorem 2.23:** [3] If  $g: [\partial, f] \to (0, +\infty)$  be a GA-Concave, then  $\left(\frac{1}{e} \left(\frac{f^f}{\partial^{\partial}}\right) \frac{1}{f - \partial}\right) \le \frac{1}{f - \partial} \int_{\partial}^{f} g(y) \ dy \le \left(\frac{1}{\ln f - \ln \partial} - \frac{\partial}{\partial - f}\right) g(\partial) + \left(\frac{f}{\partial - f} - \frac{1}{\ln f - \ln \partial}\right) g(f).$ 

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