# ROUGHLY d-CONVEX FUNCTIONS ON UNDIRECTED TREE NETWORKS

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Abstract. In this paper we establish some properties of roughly d-convex functions on undirected tree networks. It is pointed out that these roughly d-convex functions have the following properties concerning the property of minimum: each local minimum of a midpoint  $\delta$ -d-convex or lightly  $\gamma$ -d-convex function is a global minimum, where a local minimizer has to yield the minimal function value in its neighborhood with radius equal to the roughness degree. Since every  $\rho$ -d-convex or  $\delta$ -d-convex function is midpoint  $\delta$ -d-convex and every  $\gamma$ -d-convex function is lightly  $\gamma$ -d-convex, this conclusion holds for them, too. We also state weaker but sufficient conditions for roughly d-convex functions. We adopt the definition of network as metric space introduced by Dearing P.M. and Francis R.L. in 1974.

### 1. Introduction

We recall first the definitions of undirected networks as metric space introduced in [1] by Dearing and Francis.

We consider an undirected, connected graph G = (W, A), without loops or multiple edges. To each vertex  $w_i \in W = \{w_1, ..., w_m\}$  we associate a point  $v_i$  from an euclidian space X. This yields a finite subset  $V = \{v_1, ..., v_m\}$  of X, called the **vertex set** of the network. We also associate to each edge  $(w_i, w_j) \in A$  a rectifiable arc  $[v_i, v_j] \subset X$  called **edge** of the network. We assume that any two edges have no interior common points. Consider that  $[v_i, v_j]$  has the positive length  $l_{ij}$  and denote by U the set of all edges. We define the **network** N = (V, U) by

$$N = \{x \in X \mid \exists (w_i, w_j) \in A \text{ such that } x \in [v_i, v_j]\}.$$

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It is obvious that N is a geometric image of G, which follows naturally from an embedding of G in X. Suppose that for each  $[v_i, v_j] \in U$  there is a continuous one-to-one mapping  $\theta_{ij} : [v_i, v_j] \to [0, 1]$  with  $\theta_{ij} (v_i) = 0, \theta_{ij} (v_j) = 1$ , and  $\theta_{ij} ([v_i, v_j]) = [0, 1]$ . We denote by  $T_{ij}$  the inverse function of  $\theta_{ij}$ .

Any connected and closed subset of an edge bounded by two points x and y of  $[v_i, v_j]$  is called a **closed subedge** and is denoted by [x, y]. If one or both of x, y are missing we say than the subedge is open in x, or in y or is open and we denote this by (x, y], [x, y) or (x, y), respectively. Using  $\theta_{ij}$ , it is possible to compute the length of [x, y] as

$$l([x,y]) = |\theta_{ij}(x) - \theta_{ij}(y)| \cdot l_{ij}.$$

Particularly we have

$$l([v_i, v_j]) = l_{ij}, \qquad l([v_i, x]) = \theta_{ij}(x) l_{ij}$$

and

$$l([x, v_j]) = (1 - \theta_{ij}(x)) l_{ij}.$$

A path L(x, y) linking two points x and y in N is a sequence of edges and at most two subedges at extremities, starting at x and ending at y. If x = y then the path is called **cycle**. The **length of a path (cycle)** is the sum of the lengths of all its component edges and subedges and will be denoted by l(L(x, y)).

A network is connected if for any points  $x, y \in N$  there is a path  $L(x, y) \subset N$ .

A connected network without cycles is called tree.

Let  $L^*(x, y)$  be a shortest path between the points  $x, y \in N$ . This path is also called **geodesic**.

**Definition 1.** [1] For any  $x, y \in N$ , the distance from x to y, d(x, y) in the network N is the length of a shortest path from x to y:

$$d(x,y) = l(L^*(x,y)).$$

It is obvious that (N, d) is a metric space.

For  $x, y \in N$ , we denote

$$\langle x,y\rangle = \left\{z\in N\mid d\left(x,z\right)+d\left(z,y\right)=d\left(x,y\right)\right\},$$

and  $\langle x, y \rangle$  is called the **metric segment** between x and y.

**Definition 2.** [1] A set  $D \subset N$  is called d-convex if  $\langle x, y \rangle \subset D$  for all  $x, y \in D$ .

Roughly d-convex functions are a generalization of roughly convex functions and respective of d-convex functions proposed by V. P. Soltan and P. S. Soltan in [14]. We recall that there is several kinds of roughly convex functions:  $\rho$ -convex functions, proposed by Klötzler and investigated by Hartwig and Söllner in [2], [13],  $\delta$ -convex and midpoint  $\delta$ -convex functions established by Hu, Klee, Larman in [3] and  $\gamma$ -convex, strictly  $\gamma$ -convex, lightly  $\gamma$ -convex, midpoint  $\gamma$ -convex, strictly  $\tau$ -convexlike functions, proposed and investigated by Phu in [6], [7], [8], [9], [10], [11] etc.

In the following lines we consider a network N=(V,U) endowed with the metric defined in Definition 1. We denote  $\overline{\mathbf{R}} = \mathbf{R} \cup \{+\infty\}$ .

**Definition 3.** [14] The function  $f: N \to \overline{\mathbb{R}}$  is called d-convex on N if for any pair of points  $x, y \in N, x \neq y$ , and for every  $z \in \langle x, y \rangle$  is satisfied the inequality

(2) 
$$f(z) \le \frac{d(z,y)}{d(x,y)} f(x) + \frac{d(x,z)}{d(x,y)} f(y).$$

Extending Phu's observation at this function, we remark in [5] that the inequality (2) can be satisfied just for the points  $x, y \in N$  with  $d(x, y) \geq r$ , r being a fixed positive real number convenient selected.

We consider the positive real numbers  $r_{\rho}, r_{\delta}, r_{\gamma}, r$  and a d-convex set  $D \subset N$ .

**Definition 4.** [5] The function  $f: D \to R$  is called:

- 1.  $\rho$ -d-convex on D with the roughness degree  $r_{\rho}$  if for any pair of points  $x, y \in D$  with  $d(x, y) \geq r_{\rho}$ , is satisfied the inequality (2) for all  $z \in \langle x, y \rangle$ ;
- 2.  $\delta$ -d-convex on D with the roughness degree  $r_{\delta}$  if for any pair of points  $x, y \in D$  with  $d(x, y) \geq r_{\delta}$ , is satisfied the inequality (2) for all  $z \in \langle x, y \rangle$  with  $d(x, z) \geq r_{\delta}/2$ , and  $d(z, y) \geq r_{\delta}/2$ ;
- 3. **midpoint**  $\delta$ -d-convex on D with the roughness degree  $r_{\delta}$  if for any pair of points  $x, y \in D$  with  $d(x, y) \geq r_{\delta}$ , is satisfied the inequality (2) for all  $z \in \langle x, y \rangle$  with d(x, z) = d(x, y) = d(x, y)/2;
- 4.  $\gamma$ -d-convex on D with the roughness degree  $r_{\gamma}$  if for any pair of points  $x, y \in D$  with  $d(x, y) \geq r_{\gamma}$ , is satisfied the inequality

(3) 
$$f\left(x'\right) + f\left(y'\right) \le f\left(x\right) + f\left(y\right)$$

for all pair of points  $x^{'},y^{'}\in\langle x,y\rangle$  with  $d\left(x,x^{'}\right)=d\left(y,y^{'}\right)=r_{\gamma};$ 

5. **lightly**  $\gamma$ -d-convex on D with the roughness degree  $r_{\gamma}$  if for any pair of points  $x, y \in D$  with  $d(x, y) \geq r_{\gamma}$ , is satisfied the inequality (2) for all  $z \in \langle x, y \rangle$  with  $d(x, z) = r_{\gamma}$  or for all  $z \in \langle x, y \rangle$  with  $d(z, y) = r_{\gamma}$ ;

- 6. **midpoint**  $\gamma$ -d-**convex** on D with the roughness degree  $r_{\gamma}$  if for any pair of points  $x, y \in D$  with  $d(x, y) = 2r_{\gamma}$ , is satisfied the inequality (2) for all  $z \in \langle x, y \rangle$  with  $d(x, z) = d(z, y) = r_{\gamma}$ ;
- 7. **strictly**  $\gamma$ -d-**convex** on D with the roughness degree  $r_{\gamma}$  if for any pair of points  $x, y \in D$  with  $d(x, y) > r_{\gamma}$ , is satisfied the inequality

$$(4) f\left(x'\right) + f\left(y'\right) < f\left(x\right) + f\left(y\right),$$

for all pair of points  $x^{'},y^{'}\in\langle x,y\rangle$  with  $d\left(x,x^{'}\right)=d\left(y,y^{'}\right)=r_{\gamma}$  ;

8. strictly r-d-convexlike (or strictly roughly d-convexlike) on D with the roughness degree r if for any pair of points  $x, y \in D$  with d(x,y) > r there is  $z \in \langle x,y \rangle$ ,  $z \neq x, z \neq y$  such that is satisfied the inequality:

(5) 
$$f(z) < \frac{d(z,y)}{d(x,y)} f(x) + \frac{d(x,z)}{d(x,y)} f(y).$$

The functions who satisfy one of the conditions (1)-(8) are called roughly d-convex.

We compared this kinds of roughly convex functions and we got the following scheme for the relation between them:

**Theorem 1.** [5]Between some different kinds of roughly d-convex functions there is the following relations:

$$\begin{array}{cccc} f \text{ $d$-convex} \overset{\forall r_{\rho} > 0}{\Longrightarrow} f \text{ $\rho$-$d$-convex} \overset{r_{\rho} \leq r_{\delta}}{\Longrightarrow} f \text{ $\delta$-$d$-convex} \Longrightarrow & f \text{ $midpoint} \\ & & & & & & & & & \\ & & & & & & & \\ & & & & & & & \\ & & & & & & \\ & & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & \\ & & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & \\ & &$$

## 2. Some properties of roughly *d*-convex functions on tree networks

We consider now a tree network N=(V,U) and a d-convex set  $D\subset N$ . We recall that in a tree network the metric segment  $\langle x,y\rangle$  contain an unique path between x and y, for every  $x,y\in N$ .

**Definition 5.** We say that the function  $f: D \to R$  attains a r-local minimum at a point  $x^* \in D$  if

$$f(x) \ge f(x^*)$$
 for all  $x \in D$  satisfying  $d(x, y) < r$ .

**Theorem 2.** [5] If  $f: D \to R$  is a midpoint  $\delta$ -d-convex function with the roughness degree  $r_{\delta} > 0$ ,  $x^* \in D$  and

$$(6) f(x) \ge f(x^*)$$

for all  $x \in U_{r_{\delta}}(x^*) := \{z \in D \mid d(x^*, z) < r_{\delta}\}, \text{ then } f(x) \geq f(x^*) \text{ for all } x \in D \text{ (f attains its global minimum in } D \text{ at } x^*).$ 

**Remark.** Since  $\rho$ -d-convexity and  $\delta$ -d-convexity imply midpoint  $\delta$ -d-convexity,  $\rho$ -d-convex functions and  $\delta$ -d-convex functions have this property, too.

**Theorem 3.** If  $f: D \to R$  is a lightly  $\gamma$ -d-convex function with the roughness degree  $r_{\gamma} > 0$ ,  $x^* \in D$  and

$$f\left(x\right) \geq f\left(x^{*}\right)$$

for all  $x \in \overline{U_{r_{\gamma}}(x^*)} := \{z \in D \mid d(x^*, z) \leq r_{\delta}\}, \text{ then } f(x) \geq f(x^*) \text{ for all } x \in D \text{ ($f$ attains its global minimum in $D$ at $x^*$).}$ 

**Proof.** Assume the contrary that f does not attain its global minimum at  $x^*$ , then there is  $x_0 \in D \setminus \overline{U_{r_{\gamma}}(x^*)}$  such that  $f(x^*) > f(x_0)$ . We consider now the points  $s, x_1 \in \langle x_0, x^* \rangle$  such that

$$d(x^*,s)=r_{\gamma}$$
 and  $d(x_1,x_0)=r_{\gamma}$ .

Since  $f(x_0) < f(x^*) \le f(s)$ , the definition of lightly  $\gamma$ -d-convexity imply

$$f(x_1) \le \frac{d(x_0, x_1)}{d(x_0, x^*)} f(x^*) + \frac{d(x_1, x^*)}{d(x_0, x^*)} f(x_0) < f(x^*).$$

We repeat this construction, and we get  $x_i, i \in I \subset N$ , with  $f(x^*) > f(x_i)$  for all  $i \in I$ . Since  $d(x_i, x^*) = d(x_{i-1}, x^*) - r_{\gamma}$ , there is  $i^* \in I$  such that  $d(x_{i^*}, x^*) < r_{\delta}$  and hence for  $x_{i^*}$  we have  $f(x_{i^*}) \ge f(x^*)$ , which contradicts the relation  $f(x^*) > f(x_i)$  for all  $i \in I$ . This contradiction completes our proof.

**Remark.** Since every  $\gamma$ -d-convex function is lightly  $\gamma$ -d-convex, this conclusion holds for  $\gamma$ -d-convex functions, too.

In the following line we will establish weaker but sufficient conditions for roughly d-convex functions  $f: D \to R$ , where D is a d-convex subset of a tree network N = (V, U).

We consider a tree network N = (V, U) and a d-convex set  $D \subset N$ .

**Theorem 4.** [5] The function  $f: D \to R$  is  $\gamma$ -d-convex on D with the roughness degree  $r_{\gamma} > 0$  if and only if there is a  $\sigma > 0$  such that

(7) 
$$f\left(x'\right) + f\left(y'\right) \le f\left(x\right) + f\left(y\right)$$

is satisfied for any pair of points  $x, y \in D$  with

$$r_{\gamma} \le d(x, y) < r_{\gamma} + \sigma$$

and for 
$$x^{'},y^{'}\in\langle x,y\rangle$$
 with  $d\left(x,x^{'}\right)=d\left(y,y^{'}\right)=r_{\gamma}.$ 

**Theorem 5.** The function  $f: D \to R$  is midpoint  $\delta$ -d-convex on D with the roughness degree  $r_{\gamma} > 0$  if and only if the inequality (2) is satisfied for  $z \in \langle x, y \rangle$  with d(x, z) = d(z, y) = d(x, y)/2, for any pair of points  $x, y \in D$  satisfying

$$r_{\delta} \leq d(x,y) < 2r_{\delta}$$
.

**Proof.** It is clear that we only need to prove the sufficiency. This is done by induction. We are going to show that (2) holds for any pair of points  $x, y \in D$  satisfying

$$r_{\delta} \leq d(x,y) < 2^{i} r_{\delta}, i = 1, 2, \dots$$

and for  $z \in \langle x, y \rangle$  with d(x, z) = d(z, y) = d(x, y)/2.

By assumption, it holds for i = 1. We assume now that the assertion is true for some  $n \in \mathbb{N}$ . Let x, y be a pair of points in D with

$$2^n r_{\delta} \le d(x,y) < 2^{n+1} r_{\delta}.$$

We denote by  $z_1, z_2, z_3$  the points from  $\langle x, y \rangle$  such that

$$d(x, z_1) = d(z_1, z_2) = d(z_2, z_3) = d(z_3, y) = d(x, y)/4.$$

Then

$$r_{\delta} \le d(z_2, x) = d(z_3, z_1) = d(y, z_2) < 2^n r_{\delta}$$

implies

$$f(z_1) \leq (1/2)f(x) + (1/2)f(z_2)$$
  

$$2f(z_2) \leq f(z_1) + f(z_3)$$
  

$$f(z_2) \leq (1/2)f(z_2) + (1/2)f(y).$$

By addition of these three inequalities we get

$$f(z_2) \le (1/2)f(x) + (1/2)f(y).$$

Hence the assertion also holds for this pair of points  $x, y \in D$ .

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