# SOME INEQUALITIES FOR CONVEXIFIABLE FUNCTION WITH APPLICATIONS

#### Adriana CLIM

Bucharest University, Academiei 14, Bucharest, Romania E-mail: clim.adriana@gmail.com

Some function becomes convex after adding to it a quadratic term. In this paper we extend some properties of convex function to convexifiable case of operators.

*Key words*: Convexify; Convexifiable function; Continuous function; Jensen inequality; Operator on Hilbert space.

# **1. INTRODUCTION**

Convex functions are often used in applied mathematics. They have many uses in optimization and numerical methods. Using convexity, one can also study non-convex problems in two directions: transformation of arbitrary continuous functions to convex-like function and transformation of mathematical programs with such functions to equivalent programs.

For a given function  $f: R \to R$ , defined on a bounded convex set *J* of a real line *R*, we can construct the convex function by adding simple quadratic term  $\alpha \cdot x^T \cdot x$  to *f* where  $\alpha$  is a sufficiently large non-negative number. The numerical value of the "convexifier" ( $\alpha$ ) depends on the function *f* and the interval where *f* is "convexified". For  $\alpha < 0$  the quadratic term is strictly convex, so *f* is called "weakly convex"[7]. Also, if there is  $\alpha$ , convexifier of *f*, then there are a lot of such values  $\alpha^* \leq \alpha$  which are also convexifiers.

Therefore every convexifiable function f can be written as the sum of a convex function  $f(x) - \frac{\alpha}{2} \cdot x^T \cdot x$  and a concave quadratic term  $\frac{\alpha}{2} \cdot x^T \cdot x$ , for every  $\alpha$  which is sufficiently small.

2 Convexifiable functions have been studied also on  $R^n$  and characterized using the fact that for continuous functions a class of convexifiable function is large: beside convex and twice continuously

differentiable functions a class of convexifiable function is large: beside convex and twice continuously differentiable functions with Lipschitz derivative. In [10] Zlobec showed that there exist continuously differentiable functions and also differentiable Lipschitz functions that can not be convexified.

Here we extend some of results from [5] to convexifiable case of convex operator.

### 2. SOME PRELIMINARY RESULTS

**Definition 2.1.** If  $f: \mathbb{R}^n \to \mathbb{R}$  is a continuous function of n variables defined on a convex set  $J, J \subseteq \mathbb{R}^n$  then the function is said to be convex (concave) on J if

$$f(\lambda x + (1 - \lambda)y) \le (\ge) \lambda f(x) + (1 - \lambda)f(y) \ (\forall) x, y \in J, \ (\forall)\lambda \in [0, 1].$$
(1)

**Theorem 2.2** [2]. If f is a continuous, real function on an interval J, the following conditions are equivalent:

(i) f is operator concave.

(ii)  $f(C^*AC+t_0(I-C^*C)) \ge C^*f(A)C + f(t_0)(I-C^*C)$  for an operator C with  $||C|| \le 1$  and a self-adjoint operator A with  $\sigma(A) \subseteq J$  and for fixed real number  $t_0 \in J$ .

(iii) 
$$f\left(\sum_{j=1}^{n} C_{j}^{*} A_{j} C_{j} + t_{0}\left(I - \sum_{j=1}^{n} C_{j}^{*} C_{j}\right)\right) \ge \sum_{j=1}^{n} C_{j}^{*} f(A_{j}) C_{j} + f(t_{0})\left(I - \sum_{j=1}^{n} C_{j}^{*} C_{j}\right)$$
 for operators

 $C_j$  with  $\sum_{j=1}^{n} C_j^* C_j \leq I$  and self-adjoint operator  $A_j$  with  $\sigma(A_j) \subseteq J$ , j=1, 2, ..., n, and for a fixed

*real number*  $t_0 \in J$ .

(iv) 
$$f\left(\sum_{j=1}^{n} C_{j}^{*} A_{j} C_{j}\right) \ge \sum_{j=1}^{n} C_{j}^{*} f(A_{j}) C_{j}$$
 for operators  $C_{j}$  with  $\sum_{j=1}^{n} C_{j}^{*} C_{j} = I$ , self-adjoint operator  $A_{j}$   
with  $\sigma(A_{j}) \subseteq J, j=1, 2, ..., n$ , where  $n \ge 2$ .

(v)  $f(PAP + t_0(I - P)) \ge P \cdot f(A) \cdot P + f(t_0)(I - P)$  for a projection P and a self-adjoint operator A with  $\sigma(A) \subseteq J$  and for a fixed real number  $t_0 \in J$ .

**Definition 2.3.** [8] Given a continuous  $f : \mathbb{R}^n \to \mathbb{R}$  defined on a convex set  $J \subset \mathbb{R}^n$ , consider the parametric function  $\varphi : \mathbb{R}^n \times \mathbb{R} \to \mathbb{R}$  defined by

$$\varphi(x,\alpha) = f(x) - \frac{1}{2}\alpha x^T x , \qquad (2)$$

where  $x^{T}$  is the transposed of x. If  $\varphi(x,\alpha)$  is a convex function on J for some  $\alpha = \alpha^{*}$ , then  $\varphi(x,\alpha)$  is a convexification of f and  $\alpha^{*}$  is its convexifier on J. Function f is convexifiable if it has a convexification.

*Remark 2.4.* If  $\alpha$  is a convexifier of *f*, then so is every  $\alpha^* \leq \alpha$ .

**Theorem 2.5** [9]. If f is a continuous function  $f : \mathbb{R}^n \to \mathbb{R}$  defined on a convex set  $J \subset \mathbb{R}^n$  then f is convex if and only if f is mid-point convex, i.e.,

$$f\left(\frac{x+y}{2}\right) \le \frac{1}{2} \left(f(x) + f(y)\right), \,\forall x, y \in J.$$
(3)

*Remark 2.6.* Every convex function defined on a convex set from Euclidean space is mid-point convex. Over non-Euclidean space (e.g. the scalar field of rational numbers) we can construct a non-convex mid-point convex function.

With every continuous function  $f: \mathbb{R}^n \to \mathbb{R}$  we can associate a particular function  $\psi: \mathbb{R}^n \times \mathbb{R}^n \to \mathbb{R}$ . We denote the norm of  $u \in \mathbb{R}^n$  by  $||u|| = (u^T u)^{1/2}$ .

*Remark 2.7* [10]. Given a continuous function  $f : \mathbb{R}^n \to \mathbb{R}$  and a compact convex set J in  $\mathbb{R}^n$  the midpoint acceleration function of f on J is the function

$$\psi(x,y) = \frac{4}{\|x-y\|^2} \left[ f(x) + f(y) - 2f\left(\frac{x+y}{2}\right) \right], \ (\forall) x, y \in J, x \neq y.$$
(4)

*Remark 2.8.* (Justification of function's name). If we take *x*, *y* in *J* then their mid point is 1/2(x+y) and also is x+1/2(y-x). Using the notation  $\Delta x = 1/2(y-x)$ , the mid point can be written as  $x+\Delta x$ , which is the same as  $y-\Delta x$ . Then the distance from *x* and  $x + \Delta x$ , i.e.  $||\Delta x||$ , so the average displacement of *f* at *x* in the direction of mid-point  $x+\Delta x$ , over distance is  $\Delta f(x) = [f(x+\Delta x) - f(x)] / ||\Delta x||$ .

This is repeated at the mid-point and y, so we obtain  $\Delta f(x+\Delta x) = [f(y) - f(x+\Delta x)] / ||\Delta x||$ . Hence the average "displacement of the displacement", i.e. the "acceleration" is

$$\left[\Delta f(x + \Delta x) - \Delta f(x)\right] / \left\| \Delta x \right\| = \psi(x, y)$$

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**Theorem 2.9** [10]. Given a continuous function  $f : \mathbb{R}^n \to \mathbb{R}$  on a compact convex set J in  $\mathbb{R}^n$ , function f is convexifiable on J if and only if its mid-point acceleration function is bounded on J.

*Proof.* From *f* convexifiable we have  $\varphi(x, \alpha) = f(x) - \frac{1}{2\alpha x^T x}$  convex for some  $\alpha$ . But for  $\varphi$  we have  $\varphi((x+y,\alpha)/2) \le \frac{1}{2} \cdot (\varphi(x,\alpha) + \varphi(y,\alpha)), x, y \in J.$ 

After substitution, this is

 $2 \cdot f((x+y)/2) - [f(x)+f(y)] \le \alpha \cdot \{1/4 \cdot [||x||^2 + 2 \cdot (x, y) + ||y||^2] - 1/2 \cdot [||x||^2 + ||y||^2]\} = -\alpha/4 \cdot ||x - y||^2.$ 

So, finally

 $\alpha \leq \psi(x, y)$ , for every  $x, y \in J, x \neq y$ .

**Theorem 2.10** [11]. [Jensen's inequality for convexifiable functions]. If f is a convexifiable function on a bounded nontrivial convex set  $J \subset \mathbb{R}^n$ , and  $\alpha$  is its convexifier, then

$$f\left(\sum_{i=1}^{p} \lambda_{i} x_{i}\right) \leq \sum_{i=1}^{p} \lambda_{i} f(x_{i}) - \frac{\alpha}{2} \left(\sum_{\substack{i,j=1\\i < j}}^{p} \lambda_{i} \lambda_{j} \| x_{i} - x_{j} \|^{2}\right)$$
(5)

for every set of points  $\{x_i\}_{i=1,...,p}$  from J and all real scalar  $\lambda_I \ge 0$  with i = 1, 2, ..., p and  $\sum_{i=1}^{p} \lambda_i = 1$ .

**Definition 2.11.** Let F be an operator over Hilbert space H. F is convex (concave) operator over  $J \subset H$  if

$$F(\alpha x + \beta y) \le (\ge) \alpha F(x) + \beta F(y) \tag{6}$$

for any real  $\alpha$ ,  $\beta$  with  $\alpha+\beta=1$ ,  $\alpha$ ,  $\beta \ge 0$  and  $x, y \in J$ .

**Definition 2.12.** Let *F* be an operator over  $J \subset H$ , a Hilbert space. We say that *F* is an convexifiable operator if exists some real number  $\alpha$  so that for  $A \in B(H)$  the new operator

$$\varphi(A,\alpha) = F(A) - \frac{1}{2}\alpha A^T A \tag{7}$$

is convex over J.

*Remark 2.13.* (generalization of Jensen inequality). Let *A*, *B* be self-adjoint operators with  $\sigma(A) \subseteq J$  and  $\sigma(B) \subseteq J$ . If *f* is an convexifiable operator on an interval *J* then for *s*,  $t \ge 0$ , s+t=1 we have

$$f(s \cdot A + t \cdot B) \le s \cdot f(A) + t \cdot f(B) - \frac{\alpha}{2} \left[ s \cdot t \cdot \|A - B\|^2 \right].$$
(8)

*Proof.* If f is convexifiable with  $\alpha$  its convexifier then there is a convex operator  $\varphi$  such that  $\varphi(C, \alpha) = f(C) - \alpha/2C^*C$ .

If we apply Jensen's inequality for convex function to  $\varphi$ , for *A*, *B*, *s*, *t* with *s*+*t*=1 we have

$$\varphi(sA + tB) \le s\varphi(A) + t\varphi(B). \tag{9}$$

After substitutions, the inequality is

$$f(s \cdot A + t \cdot B) \leq s \cdot f(A) + t \cdot f(B) - \frac{\alpha}{2} \left[ sA^2 + tB^2 - (sA + tB)^2 \right].$$

Finally, from s + t = 1 the conclusion follows.

# 3. OPERATOR INEQUALITIES FOR CONVEXIFIABLE CASE

**Theorem 2.14.** *The following conditions are equivalent for an operator*  $F: J \rightarrow R, J \subset R$ .

### i1. *F* is a convexifiable operator with $\alpha$ its convexifier.

i2. For an operator *C* with  $||C|| \le 1$  and a self-adjoint operator *A* with  $\sigma(A) \subseteq J$  and for fixed real number  $t_0 \in J$ , the operator *F* with its convexifier  $\alpha$  satisfy

$$F(C^*AC + t_0(I - C^*C)) \le C^*F(A)C + F(t_0)(I - C^*C) + \frac{\alpha}{2}D_1,$$
(10)

where

$$D_{1} = (C^{*}AC + t_{0}(I - C^{*}C))^{2} - C^{*}A^{2}C - t_{0}^{2}(I - C^{*}C).$$
(11)

i3. For operators  $C_j$  with  $\sum_{j=1}^{n} C_j^* C_j \le I$  and self-adjoint operators  $A_j$  with  $\sigma(A_j) \subset J$ , j = 1, 2, ..., n,

and for fixed real number  $t_0 \in J$ , F verify the inequality

$$F\left(\sum_{j=1}^{n} C_{j}^{*} A_{j} C_{j} + t_{0} \left(I - \sum_{j=1}^{n} C_{j}^{*} C_{j}\right)\right) \leq \sum_{j=1}^{n} C_{j}^{*} F(A_{j}) C_{j} + F(t_{0}) \left(I - \sum_{j=1}^{n} C_{j}^{*} C_{j}\right) + \frac{\alpha}{2} D_{2},$$
(12)

where  $\alpha$  is its convexifier and

$$D_{2} = \left(\sum_{j=1}^{n} C_{j}^{*} A_{j} C_{j} + t_{0} \left(I - \sum_{j=1}^{n} C_{j}^{*} C_{j}\right)\right)^{2} - \sum_{j=1}^{n} C_{j}^{*} A_{j}^{2} C_{j} - t_{0}^{2} \left(I - \sum_{j=1}^{n} C_{j}^{*} C_{j}\right).$$
(13)

i4. If we have a particular case when operators  $C_j$  satisfy condition  $\sum_{j=1}^{n} C_j^* C_j = I$  then for self-adjoint operators  $A_j$  with  $\sigma(A_j) \subset J$  for j = 1, 2, ..., n, and for fixed real number  $t_0 \in J$ , F verify the inequality

$$F\left(\sum_{j=1}^{n} C_{j}^{*} A_{j} C_{j}\right) \leq \sum_{j=1}^{n} C_{j}^{*} F(A_{j}) C_{j} + \frac{\alpha}{2} \left[ \left(\sum_{j=1}^{n} C_{j}^{*} A_{j} C_{j}\right)^{2} - \sum_{j=1}^{n} C_{j}^{*} A_{j}^{2} C_{j} \right].$$
(14)

i5. If we consider an operator projection P then for a self-adjoint operator A with  $\sigma(A) \subseteq J$  and for fixed real number  $t_0 \in J$ , the operator F with its convexifier  $\alpha$  satisfy the inequality

$$F(PAP + t_0(I - P)) \le P \cdot F(A) \cdot P + F(t_0)(I - P) + -\frac{\alpha}{2} \Big[ PA^2 P - PAPAP + (t_0^2 - t_0)(I - P) \Big].$$
(15)

*Proof.* The equivalence will be done following  $i1 \Rightarrow i2 \Rightarrow i3 \Rightarrow i4 \Rightarrow i1$  and  $i2 \Rightarrow i5 \Rightarrow i1$ .

i1 $\Rightarrow$  i2 From definition, if F is convexifiable then there is some real number  $\alpha$  so that new operator

$$\varphi(A,\alpha) = F(A) - \frac{\alpha}{2}A^T A$$

is convex. For every convex function the inequality holds (Theorem 2.2)

$$\varphi(C^*AC + t_0(I - C^*C), \alpha) \le C^*\varphi(A, \alpha)C + \varphi(t_0, \alpha)(I - C^*C).$$

So, we have

$$F(C^*AC + t_0(I - C^*C)) - \frac{\alpha}{2} (C^*AC + t_0(I - C^*C))^T (C^*AC + t_0(I - C^*C))$$
  
$$\leq C^* (F(A) - \frac{\alpha}{2} A^T A) C + (F(t_0) - \frac{\alpha}{2} t_0^2) (I - C^*C).$$

Since

$$\left( C^* A C + t_0 (I - C^* C) \right)^T \left( C^* A C + t_0 (I - C^* C) \right) =$$
  
=  $\left( (C^* A C)^T + t_0 (I - C^* C)^T \right) \left( C^* A C + t_0 (I - C^* C) \right) = \left( C^* A C + t_0 (I - C^* C) \right)^2$ 

we obtain

$$F(C^*AC + t_0(I - C^*C)) - \frac{\alpha}{2} (C^*AC + t_0(I - C^*C))^2 \le \le C^*F(A)C - \frac{\alpha}{2}(C^*A^2C) + (F(t_0) - \frac{\alpha}{2}t_0^2)(I - C^*C).$$

So,

$$F(C^*AC + t_0(I - C^*C)) \le C^*F(A)C + F(t_0)(I - C^*C) + \frac{\alpha}{2} \bigg[ (C^*AC + t_0(I - C^*C))^2 - C^*A^2C - t_0^2(I - C^*C) \bigg].$$

i2  $\Rightarrow$  i3. For a convex function  $\varphi$  we can prove the inequality (Theorem 2.2)

$$\begin{split} \varphi \Biggl( \sum_{j=1}^{n} C_{j}^{*} A_{j} C_{j} + t_{0} (I - \sum_{j=1}^{n} C_{j}^{*} C_{j}), \alpha \Biggr) &\leq \sum_{j=1}^{n} C^{*} \varphi(A, \alpha) C + \varphi(t_{0}, \alpha) (I - \sum_{j=1}^{n} C_{j}^{*} C_{j}), \text{ i.e.} \\ F \Biggl( \sum_{j=1}^{n} C_{j}^{*} A_{j} C_{j} + t_{0} \Biggl( I - \sum_{j=1}^{n} C_{j}^{*} C_{j} \Biggr) \Biggr) - \frac{\alpha}{2} \Biggl( \sum_{j=1}^{n} C_{j}^{*} A_{j} C_{j} + t_{0} \Biggl( I - \sum_{j=1}^{n} C_{j}^{*} C_{j} \Biggr) \Biggr)^{T} \cdot \\ \cdot \Biggl( \sum_{j=1}^{n} C_{j}^{*} A_{j} C_{j} + t_{0} \Biggl( I - \sum_{j=1}^{n} C_{j}^{*} C_{j} \Biggr) \Biggr) \leq \\ &\leq \sum_{j=1}^{n} C_{j}^{*} \Biggl( F(A_{j}) - \frac{\alpha}{2} A_{j}^{T} A_{j} \Biggr) C_{j} + \Biggl( F(t_{0}) - \frac{\alpha}{2} t_{0}^{2} \Biggr) \Biggl( I - \sum_{j=1}^{n} C_{j}^{*} C_{j} \Biggr). \end{split}$$

But,

$$\left(\sum_{j=1}^{n} C_{j}^{*} A_{j} C_{j} + \left(I - \sum_{j=1}^{n} C_{j}^{*} C_{j}\right)\right)^{T} \left(\sum_{j=1}^{n} C_{j}^{*} A_{j} C_{j} + \left(I - \sum_{j=1}^{n} C_{j}^{*} C_{j}\right)\right) = \left(\sum_{j=1}^{n} C_{j}^{*} A_{j} C_{j} + \left(I - \sum_{j=1}^{n} C_{j}^{*} C_{j}\right)\right)^{2}.$$

So

$$F\left(\sum_{j=1}^{n} C_{j}^{*} A_{j} C_{j} + t_{0} \left(I - \sum_{j=1}^{n} C_{j}^{*} C_{j}\right)\right) \leq \\ \leq \sum_{j=1}^{n} C^{*} F(A_{j}) C + \frac{\alpha}{2} \left(\sum_{j=1}^{n} C_{j}^{*} A_{j} C_{j} + \left(I - \sum_{j=1}^{n} C_{j}^{*} C_{j}\right)\right)^{2} - \frac{\alpha}{2} \sum_{j=1}^{n} C_{j}^{*} A_{j}^{2} C_{j} + \left(F(t_{0}) - \frac{\alpha}{2} t_{0}^{2}\right) \left(I - \sum_{j=1}^{n} C_{j}^{*} C_{j}\right).$$

If we distribute the term, we obtain

$$F\left(\sum_{j=1}^{n} C_{j}^{*} A_{j} C_{j} + t_{0} \left(I - \sum_{j=1}^{n} C_{j}^{*} C_{j}\right)\right) \leq \sum_{j=1}^{n} C_{j}^{*} F(A_{j}) C_{j} + f(t_{0}) \left(I - \sum_{j=1}^{n} C_{j}^{*} C_{j}\right) + \frac{\alpha}{2} D_{2},$$

where  $D_2$  is defined by (13).

i3 $\Rightarrow$  i4. If we have  $\sum_{j=1}^{n} C_j^* C_j = I$  then for a convexifiable operator, from (12) and (13) we have

$$F\left(\sum_{j=1}^{n} C_{j}^{*} A_{j} C_{j}\right) \leq \sum_{j=1}^{n} C_{j}^{*} F(A_{j}) C_{j} + \frac{\alpha}{2} \left[ \left(\sum_{j=1}^{n} C_{j}^{*} A_{j} C_{j}\right)^{2} - \sum_{j=1}^{n} C_{j}^{*} A_{j}^{2} C_{j} \right]$$

i4 $\Rightarrow$  i1. If we take real numbers  $C_1$ ,  $C_2$  and  $C_i = 0$  for I  $\ge 3$ , we obtain  $C_1^2 + C_2^2 = 1$  and (14) became:

$$F(C_1^2A_1 + C_2^2A_2) \le C_1^2F(A_1) + C_2^2F(A_2) + \alpha/2[(C_1^2A_1 + C_2^2A_2)^2 - C_1^2A_1^2 - C_2^2A_2^2],$$

the convexifiable definition of F.

 $i2 \Rightarrow i5$ . If we consider C=P, a projection operator in (10) and (11), then F satisfy the relation

$$F(P^*AP + t_0(I - P^*P)) \le P^*F(A)P + F(t_0)(I - P^*P) + \frac{\alpha}{2}D,$$

with

$$D = (P^*AP + t_0(I - P^*P))^2 - P^*A^2P - t_0^2(I - P^*P).$$

If *P* is a projection then  $P^2 = P$  and  $P^* = P$  so we obtain

$$D = (PAP + t_0(I - P))^2 - PA^2P - t_0^2(I - P) = PAPAP + t_0(I - P)^2 - PA^2P - t_0^2(I - P) =$$
  
= PAPAP - PA^2P - (t\_0^2 - t\_0)(I - P).

Finally we have (15).

i5  $\Rightarrow$  i1. For the self-adjoint operators *C*, *D* with  $\sigma(C)$ ,  $\sigma(D) \subseteq J$  and  $\lambda \in [0,1]$  we construct some new operators over  $H \oplus H$ 

$$X = \begin{pmatrix} C & 0 \\ 0 & D \end{pmatrix}, \ U = \begin{pmatrix} \lambda^{1/2}I & -(1-\lambda)^{1/2}I \\ -(1-\lambda)^{1/2}I & \lambda^{1/2}I \end{pmatrix}, \ P = \begin{pmatrix} I & 0 \\ 0 & 0 \end{pmatrix}$$

We have that  $\sigma(X) = \sigma(C) \cup \sigma(D) \subseteq J$ , X is an self-adjoint operator, U is unitary and P is projection. Now, relative to F, we proceed as in [1]. Since  $\sigma(\lambda C + (1-\lambda)D) \subseteq J$  we get

 $F(\lambda C+(1-\lambda)D) \le \lambda F(C) + (1-\lambda) F(D) - \alpha/2 \{\lambda C^2 + (1-\lambda)D^2 - [\lambda C+(1-\lambda)D]^2\}.$ 

*Remark.2.15.* On the line of papers [3, 4, 5, 6] we can formulate a problem of such type.

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Received May 12, 2010