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# A Generalization of Hyperbolic Special Mean and its Schur Power Convexity

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*Abstract.* By combining hyperbolic special mean with power average, a more general mean is studied. The necessary and sufficient conditions for the determination of Schur are also given.

Keywords: Hyperbolic function; Schur convexity; Schur power convexity

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#### 1. Introduction

Let x, y be two positive numbers, then

$$M_{p}(x, y) = \begin{cases} (\frac{x^{p}}{2} + \frac{y^{p}}{2})^{1/p}, p \neq 0\\ \sqrt{xy}, p = 0 \end{cases}$$
  
$$G(x, y) = M_{0}(x, y) = \sqrt{xy},$$
  
$$A(x, y) = M_{1}(x, y) = \frac{x + y}{2}$$

are called Power mean, geometric mean and arithmetic mean of x, y, respectively.

In 2003, American mathematical monthly, problem 11031, proposed a strong mean and Inequality conjecture as below.

Problem 11031: Let x, y > 0, define  $M(x, y) = \ln N(x, y)$ , where

$$N = N(x, y) = \frac{1 + \ln(\sqrt{1 + f} + \sqrt{f})}{1 - \ln(\sqrt{1 + f} - \sqrt{f})}, f = f(x, y) = \frac{1}{4} \left(e^{\frac{2(e^x - 1)}{e^x + 1}} - 1\right) \left(e^{\frac{2(e^y - 1)}{e^y + 1}} - 1\right) e^{-\left(\frac{e^x - 1}{e^x + 1} + \frac{e^y - 1}{e^y + 1}\right)},$$

to prove or disprove  $M(x, y) \leq G(x, y)$ .

Zhang [14] studied the Schur- geometric convexity of

 $m(x, y) = \sqrt{1 + f(x, y)} + \sqrt{f(x, y)}$ , and give the positive proof of the above problem.

Li and Shi [6] adapted 
$$M(x, y)$$
 as  $M(x, y) = 2 \operatorname{th}^{-1} \operatorname{sh}^{-1} \sqrt{\operatorname{sh}(\operatorname{th} \frac{x}{2}) \operatorname{sh}(\operatorname{th} \frac{y}{2})}$ , then by

geometric convexity of sh(thx), they also solved the above Inequality conjecture.

Shi [7] discussed the Schur- convexity and Schur- geometric convexity of M(x, y)

He [5] further defined  $H(x, y) = 2 \tan^{-1} \sin^{-1} \sqrt{\sin(\tan \frac{x}{2}) \sin(\tan \frac{y}{2})}$  ( $x, y \in (0, 2 \tan^{-1} \pi/2)$ ), by polynomial discriminant system [9-10], discussed the Schur power convexity of

M(x, y) and H(x, y) [2,11-13,17].

Similarly, Chen et. al. [1] defined

$$M^{*}(x, y) = 2 \operatorname{sh}^{-1} \operatorname{th}^{-1} \sqrt{\operatorname{th}(\operatorname{sh} \frac{x}{2}) \operatorname{th}(\operatorname{sh} \frac{y}{2})} (x, y \in (0, +\infty)),$$
  
$$H^{*}(x, y) = 2 \operatorname{sin}^{-1} \operatorname{tan}^{-1} \sqrt{\operatorname{tan}(\operatorname{sin} \frac{x}{2}) \operatorname{tan}(\operatorname{sin} \frac{y}{2})} (x, y \in (0, \pi)),$$

and then discussed their Schur power convexity.

In this paper, we generalize the above means and define

$$M_{p}^{*}(x, y) = 2 \operatorname{sh}^{-1} \operatorname{th}^{-1} \left[ M_{p} \left( \operatorname{th}(\operatorname{sh} \frac{x}{2}), \operatorname{th}(\operatorname{sh} \frac{y}{2}) \right) \right]$$
$$= \begin{cases} 2 \operatorname{sh}^{-1} \operatorname{th}^{-1} \left\{ \left[ \frac{1}{2} \operatorname{th}^{p}(\operatorname{sh} \frac{x}{2}) + \frac{1}{2} \operatorname{th}^{p}(\operatorname{sh} \frac{y}{2}) \right]^{1/p} \right\}, p > 0 \\ 2 \operatorname{sh}^{-1} \operatorname{th}^{-1} \sqrt{\operatorname{th}(\operatorname{sh} \frac{x}{2}) \operatorname{th}(\operatorname{sh} \frac{y}{2})}, p = 0 \end{cases} (x, y \in (0, +\infty)),$$

then discussed its Schur power convexity.

#### 2. Definition and lemma

For  $x = (x_1, x_2, \dots, x_n) \in \mathbb{R}^n$ , We rearrange its components in descending order, and denote  $x_{[1]} \ge x_{[2]} \ge \dots \ge x_{[n]}$ . When  $x_i \le y_i$   $(i = 1, \dots, n)$ , we write  $x \le y$  for short.

**Definition 1.** [8] Suppose  $x, y \in R^n$  satisfy:

(i) 
$$\sum_{i=1}^{k} x_{i} \le \sum_{i=1}^{k} y_{i} (k = 1, 2, \dots, n-1)$$
  
(E)  $\sum_{i=1}^{n} x_{i} = \sum_{i=1}^{n} y_{i}$ ,

then we say x is controlled by y, denoted by  $x \prec y$ .

**Definition 2.** [8] Suppose  $\Omega \subset R^n$ ,  $\varphi : \Omega \to R$ ,

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(i) If for any  $x, y \in \Omega$ ,  $x \le y \Rightarrow \varphi(x) \le \varphi(y)$ , then  $\varphi$  is called increasing function on  $\Omega$ ; if  $-\varphi$  is a increasing function on  $\Omega$ , then  $\varphi$  is called reduction function on  $\Omega$ .  $\varphi$  is called Schur convex function on  $\Omega$ 

(ii) If for any  $x, y \in \Omega$ ,  $x \prec y \Rightarrow \varphi(x) \le \varphi(y)$ , then  $\varphi$  is called Schur-convex function on  $\Omega$ ; if  $-\varphi$  is called Schur-convex function on  $\Omega$ , then  $\varphi$  is called Schur-concave function on  $\Omega$ .

**Lemma 1.** [8] Let  $E(\subseteq \mathbb{R}^n)$  be a symmetric convex set with certain interior points,  $f: E \to \mathbb{R}$  is continuous and differentiable in int *E*, then *f* is called a Schur- convex (concave) function on *E* if and only iff *f* is symmetrical on *E* and for all  $x \in \text{int } E$ ,

$$\left(x_{1}-x_{2}\right)\left(\frac{\partial f}{\partial x_{1}}-\frac{\partial f}{\partial x_{2}}\right)\geq0\left(\leq0\right).$$
(1)

**Definition 3.** [14] Let  $E \subseteq R_{++}^n$ , For any two-vector  $x, y \in E$ , when  $(\ln x_1, \ln x_2, \dots, \ln x_n) \prec (\ln y_1, \ln y_2, \dots, \ln y_n)$ , there are  $f(x) \le f(y)$ . Then f are the Schurgeometric convex function on E; f is the Schurgeometry concave function on E, if and only if f is schurgeometric convex function.

**Lemma 2.** [16] Let  $E(\subseteq R^n)$  is a symmetric set with interior points,

 $\{(\ln x_1, \ln x_2, \dots, \ln x_n) | x \in E\}$  is convex set,  $f : E \to R$  continuation, and differentiable in the **int** E. Then the necessary and sufficient condition for f to be a convex (concave) function of Schur- geometry is f is symmetric on E, and for all  $x \in \text{int } E$ , both of

$$\left(x_1 - x_2\right) \left(x_1 \frac{\partial f}{\partial x_1} - x_2 \frac{\partial f}{\partial x_2}\right) \ge (\le) 0.$$
(2)

**Definition 4.** [3,4] Let  $E \subset \mathbb{R}^n_{++}$ ,  $f: E \to \mathbb{R}$ , If you take it at will  $x, y \in E$ , when

$$\left(\frac{1}{x_1}, \frac{1}{x_2}, \dots, \frac{1}{x_n}\right) \prec \left(\frac{1}{y_1}, \frac{1}{y_2}, \dots, \frac{1}{y_n}\right)$$

there are  $f(x) \le f(y)$ , Then f are the Schur- harmonic convex function on E; If -f is harmonic convex function on E, then f are the Schur- harmonic concave function on E.

**Lemma 3.** [4] Let  $E(\subset \mathbb{R}^n_{++})$  is a symmetric set with interior points,

 $\{(1/x_1, 1/x_2, \dots, 1/x_n) | x \in E\}$  is convex set,  $f : E \to R$  continuation, and differentiable in the **int** E. Then the necessary and sufficient condition for f to be a convex (concave) function of Schur- harmonic is f is symmetric on E, and for all  $x \in int E$ , both of

$$\left(x_{1}-x_{2}\right)\left(x_{1}^{2}\frac{\partial\varphi}{\partial x_{1}}-x_{2}^{2}\frac{\partial\varphi}{\partial x_{2}}\right)\geq0(\leq0).$$
(3)

**Definition 5.** [11-13] (i) Let  $f: R_{++} \to R$  is strictly monotone function,  $\Omega \subset R^n$ . If for any  $x, y \in \Omega$ , always  $f^{-1}(\alpha f(x) + \beta f(y)) \in \Omega$ , call  $\Omega$  is f-convex set, among  $\alpha, \beta \in [0,1]$  and  $\alpha + \beta = 1$ .

(ii) Let  $\Omega \subset \mathbb{R}^n$ ,  $\Omega$  internal is not empty.  $\varphi : \Omega \to \mathbb{R}$ , for any  $x, y \in \Omega$ , when  $f(x) \prec f(y)$  there are  $\varphi(x) \le \varphi(y)$ , Then  $-\varphi$  are  $\Omega$  the Schur- **f** convex function on E. If  $-\varphi$  is Schur- *f* convex function on  $\Omega$ , then  $\varphi$  are the Schur- *f* concave function on  $\Omega$ .

According to the definition of Schur- f convex function. If g is monotonously increasing (decrease),  $g(\varphi(x))$  make sense, then  $\varphi$  is Schur- f convex function, if and only if  $g \circ \varphi$  is Schur- f convex (concave) function.

Definition 6. [11-13] In definition 5, we take

$$f: x \in (0, +\infty) \to \begin{cases} \frac{x^m - 1}{m}, & m \neq 0; \\ \ln x, & m = 0. \end{cases}$$

then  $\varphi$  are the Schur-*m* order power convex function on  $\Omega$ ; If  $-\varphi$  is Schur-*m* order power convex function on  $\Omega$ , then  $\varphi$  are the Schur-*m* order power concave function on  $\Omega$ .

**Lemma 4.** [11-13] Let  $f: R \to R$  is strictly monotone differentiable functions,  $\Omega(\subset R_{++}^n)$  is symmetry with interior points f -convex set,  $\varphi: \Omega \to R$  on  $\Omega$  is continuation, Inside of  $\Omega$ ,  $\Omega^0$  is differentiable, then the necessary and sufficient condition for  $\varphi$  to be a Schur*f* bulge (Schur- *f* concave) is  $\varphi$  is symmetric on  $\Omega$ , and for  $\forall x \in \Omega^0$ , we have

$$\Delta := \left( f(x_1) - f(x_2) \right) \left( \frac{1}{f'(x_1)} \frac{\partial \varphi}{\partial x_1} - \frac{1}{f'(x_2)} \frac{\partial \varphi}{\partial x_2} \right) \ge 0 \left( \le 0 \right).$$
(4)

For Schur-*m* order power convex function, if  $m \neq 0$ , the corresponding Schur condition is

$$\frac{x_1^m - x_2^m}{m} \left( x_1^{1-m} \frac{\partial \varphi}{\partial x_1} - x_2^{1-m} \frac{\partial \varphi}{\partial x_2} \right) \ge 0 \left( \le 0 \right).$$
(5)

It's not hard to find, formula (4) Synthetic formulae (1-5).

**Remarks:** Owing to  $sgn(\frac{x_1^m - x_2^m}{m}) = sgn(x_1 - x_2)$ , so the above Schur condition is equivalent to

$$(x_1 - x_2) \left( x_1^{1-m} \frac{\partial \varphi}{\partial x_1} - x_2^{1-m} \frac{\partial \varphi}{\partial x_2} \right) \ge 0 (\le 0) (m \in R).$$

Lemma 5.  $g_1(x) = x^{-1} \operatorname{th}(\operatorname{sh} \frac{x}{2})$  on  $(0, +\infty)$  is monotone decreasing and  $0 < g_1(x) < \frac{1}{2}$ .

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**Proof:** 
$$g'_1(x) = -\frac{1}{x^2} \left\{ \operatorname{th}(\operatorname{sh}\frac{x}{2}) - \frac{1}{2}x[1 - \operatorname{th}^2(\operatorname{sh}\frac{x}{2})]\operatorname{ch}\frac{x}{2} \right\} = -\frac{2\operatorname{sh}(\operatorname{sh}\frac{x}{2})\operatorname{ch}(\operatorname{sh}\frac{x}{2}) - x\operatorname{ch}\frac{x}{2}}{2x^2\operatorname{ch}^2(\operatorname{sh}\frac{x}{2})},$$

owing to  $p_1(t) = \operatorname{sh} t$ ,  $p_2(t) = \operatorname{ch} t$  all about t on  $(0, +\infty)$  is monotone increment, and  $\operatorname{sh} t > t(t > 0) t$ , then

$$2\operatorname{sh}(\operatorname{sh}\frac{x}{2})\operatorname{ch}(\operatorname{sh}\frac{x}{2}) - \operatorname{xch}\frac{x}{2} > 2\operatorname{sh}\frac{x}{2}\operatorname{ch}\frac{x}{2} - \operatorname{xch}\frac{x}{2} > 2 \times \frac{x}{2}\operatorname{ch}\frac{x}{2} - \operatorname{xch}\frac{x}{2} = 0.$$

Thus  $g'_1(x) < 0$ ,  $g_1(x)$  about x on  $(0, +\infty)$  is monotone decrease. Obvious  $g_1(x) > 0$ , again

$$\lim_{x \to 0^+} g_1(x) = \lim_{x \to 0^+} \frac{\operatorname{th}(\operatorname{sh} \frac{x}{2})}{x} = \lim_{x \to 0^+} \frac{\operatorname{sh} \frac{x}{2}}{x} = \lim_{x \to 0^+} \frac{\frac{x}{2}}{x} = \frac{1}{2}.$$

Then for any  $x \in (0, +\infty)$ , we have  $g_1(x) < g_1(0) = \frac{1}{2}$ .

**Lemma 6.** [1]  $g_2(x) = 2x \operatorname{ch} \frac{x}{2} [\operatorname{sh}(2\operatorname{sh} \frac{x}{2})]^{-1}$  on  $(0, +\infty)$  is monotone decrease, and  $0 < g_2(x) < 2$ .

#### 3. Main results and proof

**Theorem 1.**  $M_p^*(x, y)$  about (x, y) on  $(0, +\infty)^2$  Schur-m order concave, if and only if  $m \ge p$ .

**Proof:** When p = 0,  $M_p^*(x, y) = M^*(x, y)$ , on document[12] certified  $M^*(x, y)$  about (x, y) on  $(0, +\infty)^2$  Schur-m order concave, if and only if  $m \ge p$ .

When p > 0, calculated

$$\frac{\partial M_{p}^{*}(x,y)}{\partial x} = \frac{f_{1}(x,y)}{f_{2}(x,y)} \operatorname{th}^{p-1}(\operatorname{sh}\frac{x}{2})[1-\operatorname{th}^{2}(\operatorname{sh}\frac{x}{2})]\operatorname{ch}\frac{x}{2},$$
$$\frac{\partial M_{p}^{*}(x,y)}{\partial y} = \frac{f_{1}(x,y)}{f_{2}(x,y)} \operatorname{th}^{p-1}(\operatorname{sh}\frac{y}{2})[1-\operatorname{th}^{2}(\operatorname{sh}\frac{y}{2})]\operatorname{ch}\frac{y}{2}.$$

Among

$$f_{1}(x, y) = \frac{\left[\frac{1}{2} \operatorname{th}^{p}(\operatorname{sh}\frac{x}{2}) + \frac{1}{2} \operatorname{th}^{p}(\operatorname{sh}\frac{y}{2})\right]^{1/p-1}}{2\sqrt{1 + \left\{\operatorname{th}^{-1}\left[\left(\frac{1}{2} \operatorname{th}^{p}(\operatorname{sh}\frac{x}{2}) + \frac{1}{2} \operatorname{th}^{p}(\operatorname{sh}\frac{y}{2})\right)^{1/p}\right]\right\}^{2}}},$$
$$f_{2}(x, y) = 1 - \left[\left(\frac{1}{2} \operatorname{th}^{p}(\operatorname{sh}\frac{x}{2}) + \frac{1}{2} \operatorname{th}^{p}(\operatorname{sh}\frac{y}{2})\right)^{1/p}\right]^{2}.$$

Obvious  $f_1(x, y) > 0$ , owing to  $th(sh\frac{x}{2})$ ,  $th(sh\frac{y}{2}) \in (0,1)$ , then

 $th^{p}(\operatorname{sh} \frac{x}{2}), th^{p}(\operatorname{sh} \frac{y}{2}) \in (0,1), \left(\frac{1}{2}th^{p}(\operatorname{sh} \frac{x}{2}) + \frac{1}{2}th^{p}(\operatorname{sh} \frac{y}{2})\right)^{1/p} \in (0,1),$ Thus  $f_{2}(x, y) > 0$ .

$$\begin{split} &\Delta_{M_{p}^{*}}(x,y) = (x-y) \left( x^{1-m} \frac{\partial M_{p}^{*}(x,y)}{\partial x} - y^{1-m} \frac{\partial M_{p}^{*}(x,y)}{\partial y} \right), \text{ then } \\ &\Delta_{M_{p}^{*}}(x,y) \\ &= \frac{(x-y)f_{1}(x,y)}{f_{2}(x,y)} \left\{ x^{1-m} \text{th}^{p-1}(\text{sh}\frac{x}{2})[1-\text{th}^{2}(\text{sh}\frac{x}{2})] \text{ch}\frac{x}{2} - y^{1-m} \text{th}^{p-1}(\text{sh}\frac{y}{2})[1-\text{th}^{2}(\text{sh}\frac{y}{2})] \text{ch}\frac{y}{2} \right\} \\ &- (x-y)^{2} f_{1}(x,y) - x^{1-m} \text{th}^{p-1}(\text{sh}\frac{x}{2})[1-\text{th}^{2}(\text{sh}\frac{x}{2})] \text{ch}\frac{x}{2} - y^{1-m} \text{th}^{p-1}(\text{sh}\frac{y}{2})[1-\text{th}^{2}(\text{sh}\frac{y}{2})] \text{ch}\frac{y}{2} \end{split}$$

x - y

owing to

 $f_2(x, y)$ 

$$x^{1-m} \operatorname{th}^{p-1}(\operatorname{sh}\frac{x}{2})[1-\operatorname{th}^{2}(\operatorname{sh}\frac{x}{2})]\operatorname{ch}\frac{x}{2} = x^{p-m}[x^{-1}\operatorname{th}(\operatorname{sh}\frac{x}{2})]^{p} \frac{x[1-\operatorname{th}^{2}(\operatorname{sh}\frac{x}{2})]\operatorname{ch}\frac{x}{2}}{\operatorname{th}(\operatorname{sh}\frac{x}{2})}$$
$$= x^{p-m}[x^{-1}\operatorname{th}(\operatorname{sh}\frac{x}{2})]^{p} \frac{2x\operatorname{ch}\frac{x}{2}}{\operatorname{sh}(2\operatorname{sh}\frac{x}{2})}$$
$$= x^{p-m}[g_{1}(x)]^{p}g_{2}(x).$$

When  $m \ge p$ , apparently  $x^{p-m} > 0$ , and  $y_1 = x^{p-m}$  about x on  $(0, +\infty)$  is monotone decrease. By Lemma 5 and p > 0, we have  $[g_1(x)]^p > 0$  and  $y_2 = [g_1(x)]^p$  about x on  $(0, +\infty)$  is monotone decrease. Then according to lemma 6, obvious  $g_2(x) > 0$ , and  $g_2(x)$  about x on  $(0, +\infty)$  is monotone decrease. Comprehensive, function  $h(x) = x^{p-m}[g_1(x)]^p g_2(x)$  about x on  $(0, +\infty)$  is monotone decrease, then

$$\Delta_{M_p^*}(x,y) = \frac{(x-y)^2 f_1(x,y)}{f_2(x,y)} \cdot \frac{h(x) - h(y)}{x-y} \le 0.$$

According to lemma 6, we can get  $M_p^*(x, y)$  about (x, y) on  $(0, +\infty)^2$  is Schur-m order concave.

When p > 0 and m < p, owing to

$$\lim_{x \to +\infty} [g_1(x)]^p = [\lim_{x \to +\infty} g_1(x)]^p = (\lim_{x \to +\infty} \frac{1}{x})^p = 0,$$
  
$$0 < \frac{2x^{1+p-m} \operatorname{ch} \frac{x}{2}}{\operatorname{sh}(2\operatorname{sh} \frac{x}{2})} < \frac{2x^{1+p-m} \operatorname{ch} \frac{x}{2}}{\operatorname{sh} x} = \frac{x^{1+p-m}}{\operatorname{sh} \frac{x}{2}} < \frac{2x^{1+p-m}}{e^{\frac{x}{2}} - 1} \to 0 (x \to +\infty).$$

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Then  $\lim_{x \to +\infty} h(x) = \lim_{x \to +\infty} [g_1(x)]^p \cdot \lim_{x \to +\infty} \frac{2x^{1+p-m} \operatorname{ch} \frac{x}{2}}{\operatorname{sh}(2\operatorname{sh} \frac{x}{2})} = 0 \times 0 = 0.$ 

So 
$$\lim_{x \to +\infty} [h(x) - h(1)] = \lim_{x \to +\infty} \left[ h(x) - \operatorname{th}^{p}(\operatorname{sh}\frac{1}{2}) \frac{2\operatorname{ch}\frac{1}{2}}{\operatorname{sh}(2\operatorname{sh}\frac{1}{2})} \right] = -\operatorname{th}^{p}(\operatorname{sh}\frac{1}{2}) \frac{2\operatorname{ch}\frac{1}{2}}{\operatorname{sh}(2\operatorname{sh}\frac{1}{2})} < 0.$$

Then  $\exists x_0 \in (1, +\infty)$  make  $\frac{h(x_0) - h(1)}{x_0 - 1} < 0, \Delta_{M_p^*}(x_0, 1) < 0.$ 

Associative Lemma 5 and Lemma 6, we can get

$$0 < h(y) = y^{p-m} [g_1(y)]^p g_2(y) < y^{p-m} (\frac{1}{2})^p \times 2 \to 0(y \to 0^+) .$$

Then  $\lim_{y\to 0^+} h(y) = 0$ , so

$$\lim_{y \to 0^+} [h(1) - h(y)] = \lim_{y \to 0^+} [th^p(sh\frac{1}{2}) \frac{2ch\frac{1}{2}}{sh(2sh\frac{1}{2})} - h(y)] = th^p(sh\frac{1}{2}) \frac{2ch\frac{1}{2}}{sh(2sh\frac{1}{2})} > 0$$

Then  $\exists y_0 \in (0,1)$  make  $\frac{h(1) - h(y_0)}{1 - y_0} > 0, \Delta_{M_p^*}(1, y_0) > 0.$ 

Because at this point,  $\Delta_{M_p^*}(x, y)$  on  $(0, +\infty)^2$  is symbol uncertainty, thus  $M_p^*(x, y)$  is

not  $(0, +\infty)^2$  Schur-m power concave (convex) function.

In summary, the theorem can be proved.

## 4. Two unresolved issues

**Question 1.** To average  $M_p^*(x, y)$ , Where the range of values of p can be extended to R try to give  $M_p^*(x, y)$  ( $p \in R$ ) about (x, y) on  $(0, +\infty)^2$  necessary and sufficient conditions for power convexity of Schur-m order.

**Question 2.** Similarly the form of  $H^*(x, y)$ . More general averages involving trigonometric functions can be defined as follows

$$H_{p}^{*}(x, y) = 2\sin^{-1}\tan^{-1}\left[M_{p}\left(\tan(\sin\frac{x}{2}), \tan(\sin\frac{y}{2})\right)\right]$$
$$= \begin{cases} 2\sin^{-1}\tan^{-1}\left\{\left[\frac{1}{2}\tan^{p}(\sin\frac{x}{2}) + \frac{1}{2}\tan^{p}(\sin\frac{y}{2})\right]^{1/p}\right\}, p \neq 0\\ 2\sin^{-1}\tan^{-1}\sqrt{\tan(\sin\frac{x}{2})}\tan(\sin\frac{y}{2}), p = 0 \end{cases} (x, y \in (0, \pi)),$$

try to give  $H_p^*(x, y)$  about (x, y) on  $(0, \pi)^2$  necessary and sufficient conditions for power convexity of Schur-m order.

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