

## Research Article

# New Means of Cauchy's Type

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We will introduce new means of Cauchy's type  $M_{r,l}^s(f, \mu)$  defined, for example, as  $M_{r,l}^s(f, \mu) = ((l(l-s)/r(r-s))(M_r^r(f, \mu) - M_s^r(f, \mu)/M_l^l(f, \mu) - M_s^l(f, \mu)))^{1/(r-l)}$ , in the case when  $l \neq r \neq s$ ,  $l, r \neq 0$ . We will show that this new Cauchy's mean is monotonic, that is, the following result. *Theorem.* Let  $t, r, u, v \in \mathbb{R}$ , such that  $t \leq v$ ,  $r \leq u$ . Then for  $M_{r,l}^s(f, \mu)$ , one has  $M_{t,r}^s \leq M_{v,u}^s$ . We will also give some related comparison results.

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### 1. Introduction

Let  $\Omega$  be a convex set equipped with a probability measure  $\mu$ . Then for a strictly monotonic continuous function  $f$ , the integral power mean of order  $r \in \mathbb{R}$  is defined as follows:

$$M_r(f, \mu) = \begin{cases} \left( \int_{\Omega} (f(u))^r d\mu(u) \right)^{1/r}, & r \neq 0, \\ \exp \left( \int_{\Omega} \log(f(u)) d\mu(u) \right), & r = 0. \end{cases} \quad (1.1)$$

Throughout our present investigation, we tacitly assume, without further comment, that all the integrals involved in our results exist.

The following inequality for differences of power means was obtained (see [1, Remark 8]):

$$\left| \frac{r(r-s)}{l(l-s)} \right| m \leq \left| \frac{M_r^r(f, \mu) - M_s^r(f, \mu)}{M_l^l(f, \mu) - M_s^l(f, \mu)} \right| \leq \left| \frac{r(r-s)}{l(l-s)} \right| M, \quad (1.2)$$

where  $r, l, s \in \mathbb{R}$ ,  $l \neq r \neq s$ ,  $r, l \neq 0$  and where  $m$  and  $M$  are, respectively, the minimum and the maximum values of the function  $x^{r-l}$  on the image of  $f(u)$  ( $u \in \Omega$ ).

Let us note that (1.2) was obtained as consequence of the following result (see, e.g., [1, Corollary 1]).

**Theorem 1.1.** *Let  $r, s, l \in \mathbb{R}$ , and let  $\Omega$  be a convex set equipped with a probability measure  $\mu$ . Then,*

$$\frac{M_r^r(f, \mu) - M_s^r(f, \mu)}{M_l^l(f, \mu) - M_s^l(f, \mu)} = \frac{r(r-s)}{l(l-s)} \eta^{r-l} \quad (1.3)$$

for some  $\eta$  in the image of  $f(u)$  ( $u \in \Omega$ ), provided that the denominator on the left-hand side of (1.3) is non-zero.

We can also note that from (1.3) we can get the following form of (1.2):

$$\inf_{u \in \Omega} f(u) \leq \left( \frac{l(l-s)}{r(r-s)} \frac{M_r^r(f, \mu) - M_s^r(f, \mu)}{M_l^l(f, \mu) - M_s^l(f, \mu)} \right)^{1/(r-l)} \leq \sup_{u \in \Omega} f(u), \quad (1.4)$$

where  $r, l, s \in \mathbb{R}$ ,  $r \neq l \neq s$ ,  $r, l \neq 0$ . Moreover, (1.4) suggests introducing a new mean of Cauchy type. We will prove in Section 3 a comparison theorem for these means. Finally we will, in Section 4, give some applications.

## 2. New Cauchy's mean

From (1.4), we can define a new mean  $M_{r,l}^s$  as follows:

$$M_{r,l}^s(f, \mu) = \left( \frac{l(l-s)}{r(r-s)} \frac{M_r^r(f, \mu) - M_s^r(f, \mu)}{M_l^l(f, \mu) - M_s^l(f, \mu)} \right)^{1/(r-l)}, \quad l \neq r \neq s, \quad l, r \neq 0. \quad (2.1)$$

Now by taking  $\lim_{l \rightarrow 0} M_{r,l}^s(f, \mu)$ , we will get

$$\begin{aligned} M_{r,0}^s(f, \mu) &= M_{0,r}^s(f, \mu) = \lim_{l \rightarrow 0} M_{r,l}^s(f, \mu) \\ &= \left( \frac{s[M_r^r(f, \mu) - M_s^r(f, \mu)]}{r(r-s)[\log M_s(f, \mu) - \log M_0(f, \mu)]} \right)^{1/r}, \quad r \neq s, \quad r, s \neq 0. \end{aligned} \quad (2.2)$$

Now by taking  $\lim_{r \rightarrow s} M_{r,l}^s(f, \mu)$ , we will get

$$\begin{aligned} \lim_{r \rightarrow s} M_{r,l}^s(f, \mu) &= M_{s,l}^s(f, \mu) = M_{l,s}^s(f, \mu) \\ &= \left( \frac{l(l-s)}{s} \frac{[\int f(u)^s \log f(u) d\mu(u) - M_s^s(f, \mu) \log M_s(f, \mu)]}{M_l^l(f, \mu) - M_s^l(f, \mu)} \right)^{1/(s-l)}, \quad l \neq s, \quad l, s \neq 0. \end{aligned} \quad (2.3)$$

By similar way, we can calculate all the cases for  $r, s, l \in \mathbb{R}$ . Finally, we get the following definition of  $M_{r,l}^s(f, \mu)$ :

$$\begin{aligned}
M_{r,l}^s(f, \mu) &= \left( \frac{l(l-s)}{r(r-s)} \frac{M_r^r(f, \mu) - M_s^r(f, \mu)}{M_l^l(f, \mu) - M_s^l(f, \mu)} \right)^{1/(r-l)}, \quad l \neq r \neq s, l, r \neq 0; \\
M_{r,0}^s(f, \mu) &= M_{0,r}^s(f, \mu) = \left( \frac{s[M_r^r(f, \mu) - M_s^r(f, \mu)]}{r(r-s)[\log M_s(f, \mu) - \log M_0(f, \mu)]} \right)^{1/r}, \quad r \neq s, r, s \neq 0; \\
M_{s,l}^s(f, \mu) &= M_{l,s}^s(f, \mu) = \left( \frac{l(l-s) \int f(u)^s \log f(u) d\mu(u) - M_s^s(f, \mu) \log M_s(f, \mu)}{s(M_l^l(f, \mu) - M_s^l(f, \mu))} \right)^{1/(s-l)}, \\
& \hspace{25em} l \neq s, l, s \neq 0; \\
M_{s,0}^s(f, \mu) &= M_{0,s}^s(f, \mu) = \left( \frac{\int f(u)^s \log f(u) d\mu(u) - M_s^s(f, \mu) \log M_s(f, \mu)}{\log M_s(f, \mu) - \log M_0(f, \mu)} \right)^{1/s}, \quad s \neq 0; \\
M_{r,l}^0(f, \mu) &= \left( \frac{l^2(M_r^r(f, \mu) - M_0^r(f, \mu))}{r^2(M_l^l(f, \mu) - M_0^l(f, \mu))} \right)^{1/(r-l)}, \quad l, r \neq 0; \\
M_{r,0}^0(f, \mu) &= M_{0,r}^0(f, \mu) = \left( \frac{2[M_r^r(f, \mu) - M_0^r(f, \mu)]}{r^2[M_2^2(\log f, \mu) - M_1^2(\log f, \mu)]} \right)^{1/r}, \quad r \neq 0; \\
M_{t,t}^s &= \exp \left( -\frac{2t-s}{t(t-s)} + \frac{\int f^t \log f d\mu(u) - M_s^t(f, \mu) \log M_s(f, \mu)}{M_t^t(f, \mu) - M_s^t(f, \mu)} \right), \quad t \neq s; \\
M_{t,t}^0 &= \exp \left( -\frac{2}{t} + \frac{\int f^t \log f d\mu(u) - M_0^t(f, \mu) \log M_0(f, \mu)}{M_t^t(f, \mu) - M_0^t(f, \mu)} \right), \quad t \neq 0; \\
M_{0,0}^0 &= \exp \left( \frac{1 \int (\log f)^3 d\mu(u) - (\log M_0(f, \mu))^3}{3 \int (\log f)^2 d\mu(u) - (\log M_0(f, \mu))^2} \right), \\
M_{s,s}^s &= \exp \left( -\frac{1}{s} + \frac{\int f^s (\log f)^2 d\mu(u) - M_s^s(f, \mu) (\log M_s(f, \mu))^2}{2(\int f^s \log f d\mu(u) - (M_s^s(f, \mu) \log M_s(f, \mu)))} \right), \quad s \neq 0; \\
M_{0,0}^s &= \exp \left( \frac{1}{s} + \frac{\int (\log f)^2 d\mu(u) - (\log M_s(f, \mu))^2}{2(\int \log f d\mu(u) - \log M_s(f, \mu))} \right), \quad s \neq 0.
\end{aligned} \tag{2.4}$$

### 3. Monotonicity of new means

In this section, we will prove the monotonicity of (2.4). We need the following lemmas for log-convex function.

**Lemma 3.1.** Let  $f$  be log-convex function and if  $x_1 \leq y_1$ ,  $x_2 \leq y_2$ ,  $x_1 \neq x_2$ ,  $y_1 \neq y_2$ , then the following inequality is valid:

$$\left(\frac{f(x_2)}{f(x_1)}\right)^{1/(x_2-x_1)} \leq \left(\frac{f(y_2)}{f(y_1)}\right)^{1/(y_2-y_1)}. \quad (3.1)$$

*Proof.* In [2, page 3] we have the following result for convex function  $f$ , with  $x_1 \leq y_1$ ,  $x_2 \leq y_2$ ,  $x_1 \neq x_2$ ,  $y_1 \neq y_2$ :

$$\frac{f(x_2) - f(x_1)}{x_2 - x_1} \leq \frac{f(y_2) - f(y_1)}{y_2 - y_1}. \quad (3.2)$$

Putting  $f = \log f$ , we get

$$\log \left(\frac{f(x_2)}{f(x_1)}\right)^{1/(x_2-x_1)} \leq \log \left(\frac{f(y_2)}{f(y_1)}\right)^{1/(y_2-y_1)}, \quad (3.3)$$

after applying exponential function we get (3.1).  $\square$

The following two lemmas are proved (for functionals) in [3] (Theorem 4 and Lemma 2, for Lemma 3.2 see also [4, Theorem 1]).

**Lemma 3.2.** Let us consider  $\Lambda_t$  defined as

$$\Lambda_t(g, \mu) = \begin{cases} \frac{M_t^t(g, \mu) - M_1^t(g, \mu)}{t(t-1)}, & t \neq 0, 1; \\ \log M_1(g, \mu) - \log M_0^t(g, \mu), & t = 0; \\ \int g \log g \mu - M_0(g, \mu) \log M_0(g, \mu), & t = 1. \end{cases} \quad (3.4)$$

Then,  $\Lambda_t$  is a log-convex function.

**Lemma 3.3.** Let us consider  $\Lambda_t$  defined as

$$\Lambda_t = \begin{cases} \frac{1}{t^2} (M_t^t(f, \mu) - M_0^t(f, \mu)), & t \neq 0; \\ \frac{1}{2} (M_2^2(\log f, \mu) - M_1^2(\log f, \mu)), & t = 0. \end{cases} \quad (3.5)$$

Then,  $\Lambda_t$  is a log-convex function.

**Theorem 3.4.** Let  $t, r, u, v \in \mathbb{R}$ , such that,  $t \leq v$ ,  $r \leq u$ . Then for (2.4), we have

$$M_{t,r}^s \leq M_{v,u}^s. \quad (3.6)$$

*Proof*

*Case 1* ( $s \neq 0$ ). Let us consider  $\Lambda_t$  defined as in Lemma 3.2.  $\Lambda_t$  is a continuous and log-convex. So, Lemma 3.1 implies that for  $t, r, u, v \in \mathbb{R}$ , such that,  $t \leq v$ ,  $r \leq u$ ,  $t \neq r$ ,  $v \neq u$ , we have

$$\left(\frac{\Lambda_t}{\Lambda_r}\right)^{1/(t-r)} \leq \left(\frac{\Lambda_v}{\Lambda_u}\right)^{1/(v-u)}. \quad (3.7)$$

For  $s > 0$  by substituting  $g = f^s$ ,  $t = t/s$ ,  $r = r/s$ ,  $u = u/s$ ,  $v = v/s \in \mathbb{R}$ , such that,  $t/s \leq v/s$ ,  $r/s \leq u/s$ ,  $t \neq r$ ,  $v \neq u$ , in (3.4), we get

$$\Lambda_{t,s}(f, \mu) = \begin{cases} \frac{s^2}{t(1-s)} [M_t^t(f, \mu) - M_s^t(f, \mu)], & t \neq 0, s; \\ s(\log M_s(f, \mu) - \log M_0(f, \mu)), & t = 0; \\ s \left( \int f^s \log f - M_0^s(f, \mu) \log M_0(f, \mu) \right), & t = s. \end{cases} \quad (3.8)$$

And (3.7) becomes

$$\left(\frac{\Lambda_{t,s}}{\Lambda_{r,s}}\right)^{1/(t-r)} \leq \left(\frac{\Lambda_{v,s}}{\Lambda_{u,s}}\right)^{1/(v-u)}. \quad (3.9)$$

From (3.9), we get our required result.

Now when  $s < 0$  by substituting  $g = f^s$ ,  $t = t/s$ ,  $r = r/s$ ,  $u = u/s$ ,  $v = v/s \in \mathbb{R}$ , such that,  $v/s \leq t/s$ ,  $u/s \leq r/s$ ,  $t \neq r$ ,  $v \neq u$ , in (3.4) we get (3.8).

And (3.7) becomes

$$\left(\frac{\Lambda_{v,s}}{\Lambda_{u,s}}\right)^{s/(v-u)} \leq \left(\frac{\Lambda_{t,s}}{\Lambda_{r,s}}\right)^{s/(t-r)}. \quad (3.10)$$

Now  $s < 0$ , from (3.10), by raising power  $-s$ , we get

$$\left(\frac{\Lambda_{t,s}}{\Lambda_{r,s}}\right)^{1/(t-r)} \leq \left(\frac{\Lambda_{v,s}}{\Lambda_{u,s}}\right)^{1/(v-u)}. \quad (3.11)$$

From (3.11), we get our required result.

*Case 2* ( $s = 0$ ). In this case, we can get our result by taking limit  $s \rightarrow 0$  in (3.8) and also in this case we can consider  $\Lambda_t$  defined as in Lemma 3.3.

$\Lambda_t$  is log-convex function. So, Lemma 3.1 implies that for  $t, r, u, v \in \mathbb{R}$ , such that,  $t \leq v$ ,  $r \leq u$ ,  $t \neq r$ ,  $v \neq u$ , we have

$$\left(\frac{\Lambda_t}{\Lambda_r}\right)^{1/(t-r)} \leq \left(\frac{\Lambda_v}{\Lambda_u}\right)^{1/(v-u)}. \quad (3.12)$$

Therefore, we have for  $t, r, u, v \in \mathbb{R}$ , such that,  $t \leq v$ ,  $r \leq u$ ,  $t \neq r$ ,  $v \neq u$ :

$$M_{t,r}^0 \leq M_{v,u}^0 \quad (3.13)$$

which completes the proof.  $\square$

#### 4. Further consequences and applications

In this section, we will represent the various applications of our previous definition of a new Cauchy mean and monotonicity of this above defined a new Cauchy mean.

##### 4.1. Tobey and Stolarsky-Tobey means

Let  $E_{n-1}$  represent the  $(n-1)$ -dimensional Euclidean simplex given by

$$E_{n-1} = \left\{ (u_1, u_2, \dots, u_{n-1}) : u_i \geq 0, 1 \leq i \leq n-1, \sum_{i=1}^{n-1} u_i \leq 1 \right\}, \quad (4.1)$$

and set  $u_n = 1 - \sum_{i=1}^{n-1} u_i$ . Moreover, with  $u = (u_1, \dots, u_n)$ , let  $\mu(u)$  be a probability measure on  $E_{n-1}$ . The power mean of order  $p$  ( $p \in \mathbb{R}$ ) of the positive  $n$ -tuple  $x = (x_1, \dots, x_n) \in \mathbb{R}_+^n$ , with the weights  $u = (u_1, \dots, u_n)$ , is defined by

$$\overline{M}_p(x, \mu) = \begin{cases} \left( \sum_{i=1}^n u_i x_i^p \right)^{1/p}, & p \neq 0; \\ \prod_{i=1}^n x_i^{u_i}, & p = 0. \end{cases} \quad (4.2)$$

Then, the Tobey mean  $L_{p,r}(x; \mu)$  is defined as follows:

$$L_{p,r}(x; \mu) = M_r(\overline{M}_p(x, \mu); \mu), \quad (4.3)$$

where  $M_r(g, \mu)$  denotes the integral power mean, in which  $\Omega$  is now the  $(n-1)$ -dimensional Euclidean simplex  $E_{n-1}$ . We note that, since  $\overline{M}_p(x, \mu)$  is a mean we have  $\min\{x_i\} \leq \overline{M}_p(x, \mu) \leq \max\{x_i\}$ . Now setting  $f(x, \mu) = \overline{M}_p(x, \mu)$  in (2.4) we get

$$\Gamma_{p,r,l}^s(x, \mu) = \left( \frac{l(l-s) L_{p,r}^r(x, \mu) - L_{p,s}^r(x, \mu)}{r(r-s) L_{p,l}^l(x, \mu) - L_{p,s}^l(x, \mu)} \right)^{1/(r-l)}, \quad l \neq r \neq s, l, r \neq 0;$$

$$\Gamma_{p,r,0}^s(x, \mu) = \Gamma_{p,0,r}^s(x, \mu) = \left( \frac{s[L_{p,r}^r(x, \mu) - L_{p,s}^r(x, \mu)]}{r(r-s)[\log L_{p,s}(x, \mu) - \log L_{p,0}(x, \mu)]} \right)^{1/r}, \quad r \neq s, r, s \neq 0;$$

$$\Gamma_{p,s,l}^s(x, \mu) = \Gamma_{p,l,s}^s(x, \mu) = \left( \frac{l(l-s) \int \overline{M}_p(x, \mu)^s \log d\mu(u) - L_{p,s}^s(x, \mu) \log L_{p,s}(x, \mu)}{s [L_{p,l}^l(x, \mu) - L_{p,s}^l(x, \mu)]} \right)^{1/(s-l)},$$

$l \neq s, l, s \neq 0;$

$$\Gamma_{p,s,0}^s(x, \mu) = \Gamma_{p,0,s}^s(x, \mu) = \left( \frac{\int \overline{M}_p(x, \mu)^s \log \overline{M}_p(x, \mu) d\mu(u) - L_{p,s}^s(x, \mu) \log L_{p,s}(x, \mu)}{\log L_{p,s}(x, \mu) - \log L_{p,0}(x, \mu)} \right)^{1/s}, \quad s \neq 0;$$

$$\begin{aligned}
\Gamma_{p,r,l}^0(x,\mu) &= \left( \frac{l^2(L_{p,r}^r(x,\mu) - L_{p,0}^r(x,\mu))}{r^2(L_{p,l}^l(x,\mu) - L_{p,0}^l(x,\mu))} \right)^{1/(r-l)}, \quad l, r \neq 0; \\
\Gamma_{p,r,0}^0(x,\mu) &= \Gamma_{p,0,r}^0(x,\mu) = \left( \frac{2[L_{p,r}^r(x,\mu) - L_{p,0}^r(x,\mu)]}{r^2[M_2^2(\log \bar{M}_p(x,\mu), \mu) - M_1^2(\log \bar{M}_p(x,\mu), \mu)]} \right)^{1/r}, \quad r \neq 0; \\
\Gamma_{p,t,t}^s(x,\mu) &= \exp \left( -\frac{2t-s}{t(t-s)} + \frac{\int \bar{M}_p(x,\mu)^t \log \bar{M}_p(x,\mu) d\mu(u) - L_{p,s}^t(x,\mu) \log L_{p,s}(x,\mu)}{L_{p,t}^t(x,\mu) - L_{p,s}^t(x,\mu)} \right), \quad t \neq s; \\
\Gamma_{p,t,t}^0(x,\mu) &= \exp \left( -\frac{2}{t} + \frac{\int \bar{M}_p(x,\mu)^t \log \bar{M}_p(x,\mu) d\mu(u) - L_{p,0}^t(x,\mu) \log L_{p,0}(x,\mu)}{L_{p,t}^t(x,\mu) - L_{p,0}^t(x,\mu)} \right), \quad t \neq 0; \\
\Gamma_{p,0,0}^0(x,\mu) &= \exp \left( \frac{1 \int (\log \bar{M}_p(x,\mu))^3 d\mu(u) - (\log L_{p,0}(x,\mu))^3}{3 \int (\log \bar{M}_p(x,\mu))^2 d\mu(u) - (\log L_{p,0}(x,\mu))^2} \right), \\
\Gamma_{p,s,s}^s(x,\mu) &= \exp \left( -\frac{1}{s} + \frac{\int \bar{M}_p(x,\mu)^s (\log \bar{M}_p(x,\mu))^2 d\mu(u) - L_{p,s}^s(x,\mu) (\log L_{p,s}(x,\mu))^2}{2(\int \bar{M}_p(x,\mu)^s \log \bar{M}_p(x,\mu) d\mu(u) - (L_{p,s}^s(x,\mu) \log L_{p,s}(x,\mu)))} \right), \quad s \neq 0; \\
\Gamma_{p,0,0}^s(x,\mu) &= \exp \left( \frac{1}{s} + \frac{\int (\log \bar{M}_p(x,\mu))^2 d\mu(u) - (\log L_{p,s}(x,\mu))^2}{2(\int \log \bar{M}_p(x,\mu) d\mu(u) - \log M_s(x,\mu))} \right), \quad s \neq 0.
\end{aligned} \tag{4.4}$$

**Theorem 4.1.** Let  $t, r, u, v \in \mathbb{R}$ , such that,  $t < v$ ,  $r < u$ . Then for (4.4), we have

$$\Gamma_{p,t,r}^s \leq \Gamma_{p,v,u}^s. \tag{4.5}$$

*Proof.* It is a simple consequence of Theorem 3.4.  $\square$

Pečarić and Šimić (see [5, Definition 1]) introduced the Stolarsky-Tobey mean  $\varepsilon_{p,q}(x,\mu)$  defined by

$$\varepsilon_{p,q}(x,\mu) = L_{p,q-p}(x,\nu) = M_{q-p}(\bar{M}_p(x,\mu); \mu), \tag{4.6}$$

where  $L_{p,r}(x,\nu)$  is the Tobey mean already introduced above.

For the Stolarsky-Tobey mean and (2.4), we get the following:

$$\begin{aligned}
\Upsilon_{p,r,l}^s(x,\mu) &= \left( \frac{l(l-s) \varepsilon_{p,p+r}^r(x,\mu) - \varepsilon_{p,p+s}^r(x,\mu)}{r(r-s) \varepsilon_{p,p+l}^l(x,\mu) - \varepsilon_{p,p+s}^l(x,\mu)} \right)^{1/(r-l)}, \quad l \neq r \neq s, \quad l, r \neq 0; \\
\Upsilon_{p,r,0}^s(x,\mu) &= \Upsilon_{p,0,r}^s(x,\mu) = \left( \frac{s[\varepsilon_{p,p+r}^r(x,\mu) - \varepsilon_{p,p+s}^r(x,\mu)]}{r(r-s)[\log \varepsilon_{p,p+s}(x,\mu) - \log \varepsilon_{p,p}(x,\mu)]} \right)^{1/r}, \quad r \neq s, \quad r, s \neq 0;
\end{aligned}$$

$$\Upsilon_{p,s,l}^s(x, \mu) = \Upsilon_{p,l,s}^s(x, \mu) = \left( \frac{l(l-s) \int \overline{M}_p(x, \mu)^s \log d\mu(u) - \varepsilon_{p,p+s}^s(x, \mu) \log \varepsilon_{p,p+s}(x, \mu)}{s (\varepsilon_{p,p+l}^l(x, \mu) - \varepsilon_{p,p+s}^l(x, \mu))} \right)^{1/(s-l)},$$

$l \neq s, l, s \neq 0;$

$$\Upsilon_{p,s,0}^s(x, \mu) = \Upsilon_{p,0,s}^s(x, \mu) = \left( \frac{\int \overline{M}_p(x, \mu)^s \log \overline{M}_p(x, \mu) d\mu(u) - \varepsilon_{p,p+s}^s(x, \mu) \log \varepsilon_{p,p+s}(x, \mu)}{\log \varepsilon_{p,p+s}(x, \mu) - \log \varepsilon_{p,p}(x, \mu)} \right)^{1/s},$$

$s \neq 0;$

$$\Upsilon_{p,r,l}^0(x, \mu) = \left( \frac{l^2 (\varepsilon_{p,p+r}^r(x, \mu) - \varepsilon_{p,p}^r(x, \mu))}{r^2 (\varepsilon_{p,p+l}^l(x, \mu) - \varepsilon_{p,p}^l(x, \mu))} \right)^{1/(r-l)}, \quad l, r \neq 0;$$

$$\Upsilon_{p,r,0}^0(x, \mu) = \Upsilon_{p,0,r}^0(x, \mu) = \left( \frac{2[\varepsilon_{p,p+r}^r(x, \mu) - \varepsilon_{p,p}^r(x, \mu)]}{r^2 [M_2^2(\log \overline{M}_p(x, \mu), \mu) - M_1^2(\log \overline{M}_p(x, \mu), \mu)]} \right)^{1/r}, \quad r \neq 0;$$

$$\Upsilon_{p,t}^s(x, \mu) = \exp \left( -\frac{2t-s}{t(t-s)} + \frac{\int \overline{M}_p(x, \mu)^t \log \overline{M}_p(x, \mu) d\mu(u) - M_s^t \log \varepsilon_{p,p+s}(x, \mu)}{\varepsilon_{p,p+t}^t(x, \mu) - \varepsilon_{p,p+s}^t(x, \mu)} \right), \quad t \neq s;$$

$$\Upsilon_{p,t}^0(x, \mu) = \exp \left( -\frac{2}{t} + \frac{\int \overline{M}_p(x, \mu)^t \log \overline{M}_p(x, \mu) d\mu(u) - \varepsilon_{p,p}^t(x, \mu) \log \varepsilon_{p,p}(x, \mu)}{\varepsilon_{p,p+t}^t(x, \mu) - \varepsilon_{p,p}^t(x, \mu)} \right), \quad t \neq 0;$$

$$\Upsilon_{p,0,0}^0(x, \mu) = \exp \left( \frac{1 \int (\log \overline{M}_p(x, \mu))^3 d\mu(u) - (\log \varepsilon_{p,p}(x, \mu))^3}{3 \int (\log \overline{M}_p(x, \mu))^2 d\mu(u) - (\log \varepsilon_{p,p}(x, \mu))^2} \right),$$

$$\Upsilon_{p,s,s}^s(x, \mu) = \exp \left( -\frac{1}{s} + \frac{\int \overline{M}_p(x, \mu)^s (\log \overline{M}_p(x, \mu))^2 d\mu(u) - \varepsilon_{p,p+s}^s(x, \mu) (\log \varepsilon_{p,p+s}(x, \mu))^2}{2(\int \overline{M}_p(x, \mu)^s \log \overline{M}_p(x, \mu) d\mu(u) - (\varepsilon_{p,p+s}^s(x, \mu) \log \varepsilon_{p,p+s}(x, \mu)))} \right),$$

$s \neq 0;$

$$\Upsilon_{p,0,0}^s(x, \mu) = \exp \left( \frac{1}{s} + \frac{\int (\log \overline{M}_p(x, \mu))^2 d\mu(u) - (\log \varepsilon_{p,p+s}(x, \mu))^2}{2(\int \log \overline{M}_p(x, \mu) d\mu(u) - \log \varepsilon_{p,p+s}(x, \mu))} \right), \quad s \neq 0.$$

(4.7)

**Theorem 4.2.** Let  $t, r, u, v \in \mathbb{R}$ , such that,  $t < v$ ,  $r < u$ . Then for (4.7), we have

$$\Upsilon_{p,t,r}^s \leq \Upsilon_{p,v,u}^s. \quad (4.8)$$

*Proof.* It is a simple consequence of Theorem 3.4. □



#### 4.2. The complete symmetric mean

The  $r$ th complete symmetric polynomial mean (the complete symmetric mean) of the positive real  $n$ -tuple  $x$  is defined by (see [6, pages 332,341])

$$Q_n^{[r]}(x) = \left( q_n^{[r]}(x) \right)^{1/r} = \left( \frac{c_n^{[r]}(x)}{\binom{n+r-1}{r}} \right)^{1/r}, \quad (4.9)$$

where  $c_n^{[0]}(x) = 1$  and  $c_n^{[r]}(x) = \sum_{j=1}^n \left( \prod_{i=1}^n x_i^{i_j} \right)$  and the sum is taken over all  $\binom{n+r-1}{r}$  nonnegative integer  $n$ -tuples  $(i_1, \dots, i_n)$  with  $\sum_{j=1}^n i_j = r$  ( $r \neq 0$ ). The complete symmetric polynomial mean can also be written in an integral form as follows:

$$Q_n^{[r]} = \left( \int_{E_{n-1}} \left( \sum_{i=1}^n x_i u_i \right)^r d\mu(u) \right)^{1/r}, \quad (4.10)$$

where  $\mu$  represents a probability measure such that  $d\mu(u) = (n-1)! du_1 \cdots du_{n-1}$ . We can see this as a special case of the integral power mean  $M_r(f, \mu)$ , where  $f(u) = \sum_{i=1}^n x_i u_i$ ,  $\mu$  is a probability measure as above, and  $\Omega$  is the above defined  $(n-1)$ -dimensional simplex  $E_{n-1}$ . Thus from (2.4), we have the following result:

$$\Theta_{n,r,l}^s(x, \mu) = \left( \frac{l(l-s)}{r(r-s)} \frac{(Q_n^{[r]})^r(x, \mu) - (Q_n^{[s]})^r(x, \mu)}{(Q_n^{[l]})^l(x, \mu) - (Q_n^{[s]})^l(x, \mu)} \right)^{1/(r-l)}, \quad l \neq r \neq s, l, r \in \mathbb{N}. \quad (4.11)$$

A simple consequence of Theorem 3.4 is the following result.

**Theorem 4.3.** Let  $t, r, u, v \in \mathbb{N}$ , such that,  $t < v$ ,  $r < u$ . Then for (4.11), we have

$$\Theta_{n,t,r}^s \leq \Theta_{n,v,u}^s. \quad (4.12)$$

#### 4.3. Whiteley means

Let  $x$  be a positive real  $n$ -tuple,  $s \in \mathbb{R}$  ( $s \neq 0$ ) and  $r \in \mathbb{N}$ . Then, the  $s$ th function of degree  $r$  is defined by the following generating function (see [6, pages 341–344]):

$$\sum_{r=0}^{\infty} t_n^{[r,s]}(x) t^r = \begin{cases} \prod_{i=1}^n (1 + x_i t)^s, & s > 0, \\ \prod_{i=1}^n (1 - x_i t)^s, & s < 0. \end{cases} \quad (4.13)$$

The Whiteley mean is now defined by

$$\mathcal{W}_n^{[r,s]}(x) = \left( w_n^{[r,s]}(x) \right)^{1/r} = \begin{cases} \left( \frac{t_n^{[r,s]}(x)}{\binom{nr}{s}} \right)^{1/r}, & s > 0, \\ \left( \frac{t_n^{[r,s]}(x)}{(-1)^r \binom{nr}{s}} \right)^{1/r}, & s < 0. \end{cases} \quad (4.14)$$

For  $s < 0$ , the Whiteley mean can be further generalized if we slightly change the definition of  $t_n^{[r,s]}(x)$  and define  $h_n^{[r,\sigma]}(x)$  as follows:

$$\sum_{r=0}^{\infty} h_n^{[r,\sigma]}(x)t^r = \prod_{i=1}^n \frac{1}{(1-x_it)^{\sigma_i}}, \quad (4.15)$$

where  $\sigma = (\sigma_1, \dots, \sigma_n)$ ;  $\sigma \in \mathbb{R}_+$ ;  $i = 1, \dots, n$ . The following generalization of the Whiteley mean for  $s < 0$  is defined by (see [7, Lemma 2.3])

$$\mathcal{H}_n^{[r,\sigma]}(x) = \left( \frac{h_n^{[r,\sigma]}(x)}{(\sum_{i=1}^n \sigma_i + r - 1)} \right)^{1/r}. \quad (4.16)$$

If we denote by  $\mu$  a measure on the simplex  $\Delta_{n-1} = \{(u_1, \dots, u_n) : u_i \geq 0, i = 1, \dots, n - 1, \sum_{i=1}^n u_i \leq 1\}$  such that

$$d\mu(u) = \frac{\Gamma(\sum_{i=1}^n \sigma_i)}{\prod_{i=1}^n \Gamma(\sigma_i)} \prod_{i=1}^n u_i^{\sigma_i-1} du_1 \cdots du_{n-1}, \quad (4.17)$$

where  $u_n = 1 - \sum_{i=1}^{n-1} u_i$ , then we have  $\mu$  as a probability measure and we can also write the mean  $\mathcal{H}_n^{[r,\sigma]}(x)$  in integral form as follows:

$$\mathcal{H}_n^{[r,\sigma]}(x) = \left( \int_{\Delta_{n-1}} \left( \sum_{i=1}^n x_i u_i \right)^r d\mu(u) \right)^{1/r}. \quad (4.18)$$

Finally, just as we did above in this investigation, we can develop the following analogous definition:

$$\mathfrak{H}_{n,r,l}^s(x, \mu) = \left( \frac{l(l-s) \left( \mathcal{H}_n^{[r,\sigma]} \right)^r(x, \mu) - \left( \mathcal{H}_n^{[s,\sigma]} \right)^r(x, \mu)}{r(r-s) \left( \mathcal{H}_n^{[l,\sigma]} \right)^l(x, \mu) - \left( \mathcal{H}_n^{[s,\sigma]} \right)^l(x, \mu)} \right)^{1/(r-l)}, \quad l \neq r \neq s, l, r \in \mathbb{N}. \quad (4.19)$$

A simple consequence of Theorem 3.4 is the following result.

**Theorem 4.4.** Let  $t, r, u, v \in \mathbb{N}$ , such that,  $t < v$ ,  $r < u$ . Then for (4.19), we have

$$\mathfrak{H}_{n,t,r}^s \leq \mathfrak{H}_{n,v,u}^s. \quad (4.20)$$

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