

## A TWO-PARAMETER HOMOGENEOUS MEAN VALUE<sup>1,2</sup>

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**1. Introduction.** The mean of order  $t$  of the positive values  $x = (x_1, x_2, \dots, x_n)$  with positive weights  $w = (w_1, w_2, \dots, w_n)$ ,  $\sum w_i = 1$ , is defined [3], [5] by

$$(1.1) \quad M_t(x, w) = \left[ \sum_{i=1}^n w_i x_i^t \right]^{1/t}, \quad t \neq 0,$$

$$M_0(x, w) = \prod_{i=1}^n x_i^{w_i} = \lim_{t \rightarrow 0} M_t(x, w).$$

Homogeneity in  $x$  distinguishes  $M_t$  from all other means of the form  $\phi^{-1}[\sum_{i=1}^n w_i \phi(x_i)]$ , where  $\phi$  is any function with a unique inverse  $\phi^{-1}$  [5, Theorem 84].

Without losing homogeneity,  $M_t$  has been generalized to the hypergeometric mean value  $M(t, c; x; w)$  [4]. In the present paper we shall make a further generalization while maintaining homogeneity. We construct a two-parameter mean,  $L(s, t; x)$ , by first forming the mean  $M_s(x, u)$ . Using an arbitrary weight function,  $P(u)$ , we then take an integral average over all possible choices of the weights  $u$  satisfying  $\sum_{i=1}^n u_i = 1$ . Because  $u_n = 1 - u_1 - \dots - u_{n-1}$ , the average requires an  $(n-1)$ -fold integration with respect to  $u_1, u_2, \dots, u_{n-1}$ . If the variables  $x$  are all positive, then for any real  $s$  and  $t$  we define

$$(1.2) \quad L(s, t; x) = \left[ \int_E M_s^t(x, u) P(u) du' \right]^{1/t}, \quad t \neq 0,$$

$$L(s, 0; x) = \lim_{t \rightarrow 0} L(s, t; x),$$

where  $u' = (u_1, \dots, u_{n-1})$ ,  $du' = du_1 du_2 \dots du_{n-1}$ ,  $P(u) \geq 0$ ,  $\int_E P(u) du' = 1$  and  $E = \{u' \mid u_i > 0, 1 \leq i \leq n-1 \text{ and } u_n = 1 - u_1 - \dots - u_{n-1} > 0\}$ .

The  $L(s, t; x)$  mean is homogeneous in  $x$ . It can be regarded as a special case of the integral mean [5, Chapter 3]

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$$M_t(f, P) = \left[ \int f^t(u) P(u) du \right]^{1/t},$$

but it is a generalization of previously known ways of constructing a homogeneous mean of the discrete variables  $x_1, x_2, \dots, x_n$ . After defining a set of natural weights  $w$  associated with the function  $P(u)$ , we shall show (Theorem 3) that  $L(s, t; x)$  contains  $M_t(x, w)$  in the special case  $s=t$ . It contains also the hypergeometric mean,  $M(t, c; x; w)$ , in the special case  $s=1$  and  $P(u) = P(cw; u)$ , a particular weight function depending on the parameters  $cw = (cw_1, cw_2, \dots, cw_n)$  [4, Equation (2.2)]. All properties of  $M(t, c; x; w)$  which do not depend explicitly on  $c$  can be generalized to properties of  $L(s, t; x)$ . In this generalization,  $s$  and  $t$  each play roles analogous to that of  $t$  in  $M(t, c; x; w)$ .

**2. Elementary properties of  $L(s, t; x)$ .** If  $P(u)$  is such that  $L^t(s, t; x)$  is an improper integral, it is easily shown to converge uniformly in  $s, t$ , and  $x$  for  $0 < m \leq x_i \leq M, 1 \leq i \leq n$ , all real  $s$ , and  $-T \leq t \leq T$ . As a result of this uniform convergence and the continuity of  $M_s^t(x, u)$  in  $s, t, x$ , and  $u$ ,  $L(s, t; x)$  is continuous in  $s, t$ , and  $x$ .

In considering some limiting values of  $L(s, t; x)$ , we use the notation  $x_{\max} = \max \{x_1, x_2, \dots, x_n\}$  and  $x_{\min} = \min \{x_1, x_2, \dots, x_n\}$ .

**THEOREM 1.**

- (a)  $L(s, 0; x) = \lim_{t \rightarrow 0} L(s, t; x) = \exp \left[ \int_E \ln M_s(x, u) P(u) du' \right],$
- (b)  $\lim_{t \rightarrow \infty} L(s, t; x) = x_{\max},$       (c)  $\lim_{t \rightarrow -\infty} L(s, t; x) = x_{\min},$
- (d)  $\lim_{s \rightarrow \infty} L(s, t; x) = x_{\max},$       (e)  $\lim_{s \rightarrow -\infty} L(s, t; x) = x_{\min}.$

**PROOF.** Part (a) is an application of L'Hôpital's rule, differentiation with respect to  $t$  under the integral sign being permissible because the integral of the derivative converges uniformly for  $-T \leq t \leq T$  and  $0 < m \leq x_i \leq M, 1 \leq i \leq n$ . Parts (b) and (c) follow from properties of the integral mean  $M_t(f, p)$  [5, p. 143] with  $f(u) = M_s(x, u)$ . Parts (d) and (e) follow from properties of  $M_s(x, u)$  [5, Theorem 4].

- THEOREM 2.** (a)  $L(s, t; x)$  is a strictly increasing function of  $x$ ; i.e. if  $x_i \leq y_i$  for all  $i$  and  $x_j < y_j$  for some  $j$ , then  $L(s, t; x) < L(s, t; y)$ .  
 (b) If  $x_{\max} > x_{\min}$ , then  $L(s, t; x)$  is a strictly increasing function of  $t$ .  
 (c) If  $x_{\max} > x_{\min}$ , then  $L(s, t; x)$  is a strictly increasing function of  $s$ .

**PROOF.** (a) From the definition of  $M_s(x, u)$  we see that  $M_s(x, u)$

$< M_s(y, u)$ . The result is evident by inspection of (1.2) and Theorem 1(a). Part (b) is a property of the integral mean  $M_t(f, P)$  [5, p. 144] with  $f(u) = M_s(x, u)$ . For (c) it suffices to observe that  $f = M_s(x, u)$  is a strictly increasing function of  $s$  [5, Theorem 16] and that  $M_t(g, P) > M_t(f, P)$  if  $g(u) > f(u)$  for all  $u$ .

The following theorem shows that the elementary mean  $M_s(x, w)$  is a special case of the  $L(s, t; x)$  mean; the weights  $w = (w_1, w_2, \dots, w_n)$  are the "natural weights" associated with the weight function  $P(u)$ . We define the natural weights by

$$(2.1) \quad w_i = \int_E u_i P(u) du', \quad 1 \leq i \leq n.$$

**THEOREM 3.** *If  $w$  denotes the natural weights, then  $L(s, s; x) = M_s(x, w)$ .*

**PROOF.** If  $s \neq 0$ ,

$$\begin{aligned} L(s, s; x) &= \left[ \int_E M_s^s(x, u) P(u) du' \right]^{1/s} = \left[ \int_E \sum_{i=1}^n u_i x_i^s P(u) du' \right]^{1/s} \\ &= \left[ \sum_{i=1}^n x_i^s \int_E u_i P(u) du' \right]^{1/s} = M_s(x, w). \end{aligned}$$

If  $s = 0$ ,

$$\begin{aligned} L(0, 0; x) &= \exp \left[ \int_E \ln \prod_{i=1}^n x_i^{u_i} P(u) du' \right] \\ &= \exp \left[ \sum_{i=1}^n \ln x_i \int_E u_i P(u) du' \right] \\ &= \exp \left[ \ln \prod_{i=1}^n x_i^{w_i} \right] = M_0(x, w). \end{aligned}$$

A given set of weights  $w$  occurs as the natural weights associated with a large class of functions  $P$ . This class contains the family  $P(cw; u)$  [4, Equation (2.2)] where the natural weights are just the parameters  $w$ . If  $P(u) = P(cw; u)$ ,  $L(s, t; x)$  becomes the "generalized hypergeometric mean"  $L(s, t, c; x; w)$ . In particular,  $L(1, t, c; x; w) = M(t, c; x; w)$ . For  $s \neq 0$ ,  $L(s, t, c; x; w)$  can be expressed in terms of  $M(t, c; x; w)$  by using the identity

$$(2.2) \quad \begin{aligned} L^r(s, t; x) &= \left[ \int_E \left[ \sum_{i=1}^n u_i (x_i)^{s/r} \right]^{(t/r)/(s/r)} P(u) du' \right]^{r/t} \\ &= L(s/r, t/r; x^r), \quad r, s, t \neq 0. \end{aligned}$$

A similar argument gives the same identity if  $s$  or  $t$  is zero. Putting  $r=s$  and  $P(u)=P(cw; u')$ , we have

$$L(s, t, c; x; w) = [M(t/s, c; x^s; w)]^{1/s}.$$

Although Theorems 1 and 2 show at once that  $x_{\min} < L(s, t; x) < x_{\max}$ , the introduction of natural weights allows us to give sharper inequalities:

**COROLLARY 1.** *Let  $w$  denote the natural weights. If  $x_{\max} > x_{\min}$  and  $s < t$ , then  $M_s(x, w) < L(s, t; x) < M_t(x, w)$ . The inequalities are reversed if  $s > t$ .*

**PROOF.** By Theorem 2,  $L(s, t; x)$  is an increasing function of each of the parameters  $s$  and  $t$ . Hence, applying Theorem 3, if  $s < t$ ,

$$M_s(x, w) = L(s, s; x) < L(s, t; x) < L(t, t; x) = M_t(x, w),$$

with reversed inequalities if  $s > t$ .

It is well known [1, p. 9] that if  $\log f(r, u)$  is convex in  $r$ , then  $\log \int f(r, u) du$  is convex in  $r$ . To study the convexity of  $L^s(s, t; x)$  in  $s$ , we need the analogous theorem that if  $r \log f(r, u)$  is convex in  $r$ , then  $r \log \int f(r, u) du$  is convex in  $r$  for  $r > 0$ .

**LEMMA 1.** *Let  $f^r(r, u)$  be continuous in  $r$  and  $u$  and log convex in  $r$ . Then, if  $[\int_E f(r, u) P(u) du']^r$  is continuous, it is log convex in  $r$  for  $r > 0$ .*

**PROOF.** If  $f^r(r, u)$  is continuous and log convex in  $r$ , then  $[f^r(r, u)]^{1/r} = f(r, u)$  is continuous and log convex in the variable  $1/r$  for  $r > 0$  [5, Theorem 119]. Since  $f(r, u)$  is continuous in  $u$  and log convex in  $1/r$ ,  $\int_E f(r, u) P(u) du'$  is log convex in  $1/r$  [1, p. 9]. Hence  $[\int_E f(r, u) P(u) du']^r$  is log convex in  $r$  for  $r > 0$  [5, Theorem 119].

**THEOREM 4.** (a)  $L^s(s, t; x)$  is log convex in  $s$  for  $t/s \geq 0$ . (b)  $L^t(s, t; x)$  is log convex in  $t$ .

**PROOF.** (a) For any fixed  $t \neq 0$ ,  $L^s(s, t; x) = [\int_E M_s^t(x, u) P(u) du']^{s/t} = [\int_E M_r^t(y, u) P(u) du']^r$ , where  $y_i = x_i^t$ ,  $1 \leq i \leq n$ , and  $r = s/t$ . Since  $M_r^t(y, u)$  is log convex in  $r$  [5, Theorem 87], Lemma 1 implies  $[\int_E M_r^t(y, u) P(u) du']^r$  is log convex in  $r$  for  $r = s/t > 0$ . But if a function is log convex in  $r$ , it is log convex in  $tr = s$  for any fixed  $t \neq 0$ , [1, Theorem 1.10].

For  $t=0$ ,  $\ln L^s(s, 0; x) = \int_E \ln M_s^s(x, u) P(u) du'$  is convex in  $s$  since  $M_s^s(x, u)$  is log convex in  $s$ .

(b)  $L^t(s, t; x)$  is log convex in  $t$  since  $M_t^t(f, P)$  is log convex in  $t$  [5, Theorem 197], and  $L^t(s, t; x)$  has this form with  $f(u) = M_s(x, u)$ .

**3. Inequalities for  $L(s, t; x)$ .** The next two theorems are results of

properties of the mean  $M_s(x, u)$  and the integral mean  $M_t(f, P)$ . The first comes from Theorems 24, 186 and 198 of [5]. The second comes from Theorems 12 and 188 of [5], with the special case  $t=0$  as an elementary result of properties of the logarithm. We use the notation  $x+y=(x_1+y_1, x_2+y_2, \dots, x_n+y_n)$  and  $xy=(x_1y_1, x_2y_2, \dots, x_ny_n)$ .

**THEOREM 5 (MINKOWSKI).** *Let  $x$  and  $y$  be vectors with  $x_i > 0$  and  $y_i > 0, 1 \leq i \leq n$ . Then, unless  $s=t=1$  or  $x_i=ky_i, 1 \leq i \leq n$ ,*

$$L(s, t; x + y) < L(s, t; x) + L(s, t; y), \quad (s, t \geq 1),$$

*with reversed inequality if  $s, t \leq 1$ . Equality holds in the exceptional cases.*

**THEOREM 6 (HÖLDER).** *Let  $x$  and  $y$  be vectors with  $x_i > 0$  and  $y_i > 0, 1 \leq i \leq n$ , and let  $p$  and  $q$  be real numbers greater than unity such that  $1/p + 1/q = 1$ . Then, unless  $s=t=0$  or  $x_i^p = ky_i^q, 1 \leq i \leq n$ ,*

$$L(s, t; xy) < L^{1/p}(s, t; x^p) L^{1/q}(s, t; y^q), \quad (s, t \geq 0),$$

*with reversed inequality if  $s, t \leq 0$ . Equality holds in the exceptional cases.*

By defining the mean of a matrix of values  $x_{ij}$ , both of the preceding theorems can be included in an analogue of the Jessen-Ingham inequality [5, Theorems 26 and 203]. The proof [7, p. 20] relies primarily on the Minkowski inequality and uses Hölder's inequality for a special case.

Finally, we shall show that  $L(s, t; x)$  satisfies a Kantorovich inequality [2, p. 208]. The proof proceeds by adapting a method due to Rennie [6, p. 982].

**THEOREM 7 (RENNIE).** *Let  $0 < A \leq x_i \leq B, 1 \leq i \leq n$ . Then if  $t \neq 0$ ,*

$$L^t(s, t; x) + A^t B^t L^{-t}(s, -t; x) = L^t(s, t; x) + A^t B^t L^t(-s, t; 1/x) \leq A^t + B^t,$$

*with equality if and only if  $x_i = A$  or  $x_i = B, 1 \leq i \leq n$ .*

**PROOF.** The equality between the first and second members is seen by (2.2) with  $r = -1$ . To obtain the inequality we notice that  $M_s^t(x, u)$  is bounded between  $A^t$  and  $B^t$  and, following Rennie, we consider

$$[M_s^t(x, u) - A^t][1 - B^t M_s^{-t}(x, u)]P(u) \leq 0, \quad (t \neq 0).$$

Integrating and rearranging, we have

$$\int_B M_s^t(x, u)P(u)du + A^t B^t \int M_s^{-t}(x, u)P(u)du \leq A^t + B^t.$$

Equality holds if and only if  $M'_s(x, u) = A^t$  or  $M'_s(x, u) = B^t$  for all  $u \in E$ .

THEOREM 8 (KANTOROVICH). *If  $t > 0$  and  $0 < A \leq x_i \leq B$ ,  $1 \leq i \leq n$ , then*

$$1 \leq L(s, t; x)/L(s, -t; x) = L(-s, t; x)L(s, t; 1/x) \\ \leq [(A^t + B^t)/2]^{1/t}[(A^{-t} + B^{-t})/2]^{1/t},$$

*with equality on the left if and only if  $x_{\max} = x_{\min}$ , and equality on the right if and only if  $A = B$ .*

PROOF. The left inequality holds because  $L(s, t; x)$  is a strictly increasing function of  $t$  by Theorem 2(b), unless  $x_{\max} = x_{\min}$ . The equality between the second and third members is due to (2.2) with  $r = -1$ . To obtain the right-hand inequality we start with Rennie's inequality (Theorem 7),

$$L^t(s, t; x) + A^t B^t L^{-t}(s, -t; x) \leq A^t + B^t.$$

Dividing by 2, applying the inequality of the arithmetic and geometric means to the left side, and squaring, we find

$$[L(s, t; x)/L(s, -t; x)]^t \leq (A^t + B^t)^2/4A^t B^t \\ = [(A^t + B^t)/2][(A^{-t} + B^{-t})/2].$$

Taking the  $t$ th root gives the desired inequality. If  $A = B$ , we have equality at each step of the proof; if  $A \neq B$ , then the conditions for equality in Theorem 7 imply strict inequality of the arithmetic and geometric means.

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