

**THE INTEGRAL OF AN INVARIANT UNIMODAL  
FUNCTION OVER AN INVARIANT CONVEX SET  
—AN INEQUALITY AND APPLICATIONS<sup>1</sup>**

GOVIND S. MUDHOLKAR

**1. Introduction and summary.** It is well known that the integral  $\int_{-a+\theta}^{a+\theta} f(x) dx$ ,  $a > 0$ , of a nonnegative function  $f(x)$  on the real line, which is unimodal, i.e.,  $f(kx) \geq f(x)$ ,  $0 \leq k \leq 1$ , and symmetrical about the origin, is a monotonically decreasing function of  $|\theta|$ . An immediate probabilistic consequence of this is the fact that, if a random variable  $X$  has a unimodal probability density function symmetric about the origin, and  $Y$  is any independently distributed random variable, then  $\Pr\{|X| \geq a\} \leq \Pr\{|X+Y| \geq a\}$  for any real  $a$ . T. W. Anderson [1] has extended the aforementioned monotonicity property to integrals of functions on a Euclidean  $n$ -space  $\mathcal{L}_n$  by replacing the symmetric interval of the real line by a convex set of  $\mathcal{L}_n$  symmetric about the origin, and formulating the following definition of unimodality of functions on  $\mathcal{L}_n$ .

**DEFINITION 1.** A function  $f(x)$  on  $\mathcal{L}_n$  is said to be unimodal if the set  $K_u = \{x | f(x) \geq u\}$  is convex for each  $u \geq 0$ .

More specifically, he has proved the following:

**THEOREM 1.** Let  $E$  be a convex set in  $\mathcal{L}_n$ , symmetric about the origin. Let  $f(x) > 0$  be a function such that (i)  $f(x) = f(-x)$ , (ii)  $\{x | f(x) > u\} = K_u$  is convex for every  $u$ , ( $0 < u < \infty$ ), and (iii)  $\int_E f(x) dx < \infty$  (in the Lebesgue sense). Then

$$(1) \quad \int_E f(x + ky) dx \geq \int_E f(x + y) dx$$

for  $0 \leq k \leq 1$ .

Anderson has also discussed some analogues of the probability inequality mentioned above and many other probabilistic and statistical applications.

In §2 we have obtained a generalization of the Theorem 1 by relaxing the condition of symmetry about the origin, on the function  $f$  and the set  $E$ , to a restriction of invariance with respect to finite groups of linear transformations of  $\mathcal{L}_n$ , and we have indicated the analogues of some of the probability inequalities in [1].

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In §3 we have discussed some particular cases by considering invariance with respect to the group of reflections in the origin, the permutation group, and the cyclic permutation group in  $n$ -space. It has been shown that the integral of a symmetric unimodal function over a symmetric convex region, is an  $S$ -concave (in Ostrowski's sense) function of the translation parameter.

In the final §4, we have stated and have outlined a somewhat different proof of a slightly different version of the inequality without assuming the group of transformations to be finite.

**2. The inequality with invariance w.r.t. a finite group  $G$ .** Let  $G = \{g_i, i = 1, 2, \dots, N\}$  be a finite group of Lebesgue measure-preserving linear transformations of  $\mathcal{L}_n$  onto  $\mathcal{L}_n$ . Let  $E$  be a convex set of  $n$ -space, invariant under  $G$ , or  $G$ -invariant, i.e.  $x \in E$  implies  $g_i x \in E$ ,  $i = 1, 2, \dots, N$ . Let  $f(x) \geq 0$  be a function on  $n$ -space satisfying

- (i) the unimodality condition:  $\{x | f(x) \geq u\} = K_u$  is convex for every  $u$ ,  $0 < u < \infty$ ,
- (2) (ii)  $G$ -invariance condition:  $f(g_i x) = f(x)$ ,  $i = 1, 2, \dots, N$ , for each  $x$  in  $\mathcal{L}_n$ , and
- (iii)  $\int_E f(x) dx < \infty$  in the Lebesgue sense.

For a set  $\alpha = \{\alpha_1, \alpha_2, \dots, \alpha_N\}$ ,  $\alpha_i \geq 0$ ,  $i = 1, 2, \dots, N$ ,  $\sum_{i=1}^N \alpha_i = 1$ , and a vector  $y$  of  $n$ -space let us define

$$(3) \quad \alpha(y) = \sum_{i=1}^N \alpha_i g_i y.$$

Then we have, as a generalization of the Theorem 1, the following:

**THEOREM 2.** For each set  $\alpha = \{\alpha_1, \dots, \alpha_N\}$ ,  $\alpha_i \geq 0$ ,  $\sum \alpha_i = 1$ , and vector  $y$  of  $\mathcal{L}_n$  we have

$$(4) \quad \int_E f(x + \alpha(y)) dx \geq \int_E f(x + y) dx,$$

where  $f \geq 0$  and  $E$  are  $G$ -invariant,  $f$  is unimodal,  $E$  is convex and  $\alpha(y)$  is defined by (3), provided  $\int_E f(x) dx < \infty$ .

**PROOF.** We have to show, equivalently, that

$$(5) \quad \int_{E+\alpha(y)} f(x) dx \geq \int_{E+y} f(x) dx,$$

where  $E+y$  is the set  $E$  translated by the vector  $y$ . Now it is easy to verify that, because of the convexity of  $E$  and  $K_u$

$$(6) \quad \{E + \alpha(y)\} \cap K_u \supset \sum_{i=1}^N \alpha_i [\{E + g_i y\} \cap K_u],$$

where the summation symbol  $\sum$  on the right hand side of the inclusion relation corresponds to the sum, called the Minkowski sum [4], defined by

$$(7) \quad A + B = \{a + b \mid a \in A, b \in B\}$$

for any two sets  $A, B$  of  $\mathfrak{L}_n$ , and where multiple  $cA$  of a set  $A$  of  $\mathfrak{L}_n$  by a real  $c$  is defined by  $cA = \{ca \mid a \in A\}$ . Let  $\mu(\cdot)$  be the Lebesgue measure of sets in  $\mathfrak{L}_n$ . Then we have by the Brunn-Minkowski theorem [2], [3], [4],

$$(8) \quad \mu^{1/n} \left( \sum_{i=1}^N \alpha_i [\{E + g_i y\} \cap K_u] \right) \geq \sum_{i=1}^N \alpha_i \mu^{1/n} (\{E + g_i y\} \cap K_u).$$

But because of the invariance of  $f$  and  $E$  and linearity and measure preserving properties of the transformations  $g_i, i=1, 2, \dots, N$ , we have

$$(9) \quad \mu(\{E + g_i y\} \cap K_u) = \mu(\{E + y\} \cap K_u).$$

Combining (6), (8) and (9) we get

$$(10) \quad H(u) = \mu(\{E + \alpha(y)\} \cap K_u) \geq \mu(\{E + y\} \cap K_u) = H^*(u).$$

Because of the definition of Lebesgue-Stieltjes integrals we can write

$$(11) \quad \int_{E+\alpha(y)} f(x) dx - \int_{E+y} f(x) dx = \int_0^\infty u d[H^*(u) - H(u)].$$

The right hand side of (11) is nonnegative, which may be verified by using integration by parts as in the proof of the Theorem 1 of [1]. This completes the proof of the Theorem 2.

As in [1], it may be noted that we obtain strict inequality in (4) if, and only if, for at least one  $u$ ,  $H(u) > H^*(u)$ , since  $H(u)$  is continuous on the left. For  $H(u) = H^*(u)$  we need equality in (8), which is a consequence of the Brunn-Minkowski theorem. The condition for equality may, therefore, be stated as:

**COROLLARY 1.** *In the Theorem 2, the equality in (4) holds if, and only if,  $(E + g_i y) \cap K_u$  are similarly oriented for each  $u$ .*

**COROLLARY 2.** *If the probability density function  $f(x)$  of a random  $n$ -vector  $X$  satisfies the conditions (2) and  $E$  is a convex set of  $n$ -space invariant under  $G$ , then for any  $n$ -vector  $y$  and set  $\alpha$ ,  $\Pr\{X + \alpha(y) \in E\}$*

$\geq \Pr\{X+y \in E\}$ . Furthermore, if  $h(x)$  is a  $G$ -invariant function such that  $\{x \mid h(x) \leq v\}$  is convex, then  $\Pr\{h(X+\alpha(y)) \leq v\} \geq \Pr\{h(X+y) \leq v\}$ .

The proof of the following corollary is analogous to the proof of the Theorem 2 of [1].

**COROLLARY 3.** *Let the probability density function  $f(x)$  of a random  $n$ -vector  $X$  satisfy the conditions (2) and let  $Y$  be any independently distributed random  $n$ -vector. Then for any set  $\alpha = \{\alpha_1, \dots, \alpha_N\}$ ,  $\alpha_i \geq 0$ ,  $\sum \alpha_i = 1$ , and any convex  $G$ -invariant set  $E$  of  $n$ -space*

$$(12) \quad \Pr\{X + \alpha(Y) \in E\} \geq \Pr\{X + Y \in E\}.$$

Furthermore, if  $h(x)$  is a  $G$ -invariant function such that  $\{x \mid h(x) \leq v\}$  is convex, then

$$(13) \quad \Pr\{h(X + \alpha(Y)) \leq v\} \geq \Pr\{h(X + Y) \leq v\}.$$

**3. Some particular cases.** (i) If the group  $G$  in §2 is the group of reflections in the origin the Theorem 2 reduces to the above stated Theorem 1 of Anderson [1].

(ii) An important particular case of the Theorem 2 is obtained if the group  $G$  is the permutation group in  $\mathcal{L}_n$ .

**DEFINITION 2 (OSTROWSKI [6]).** *A function  $G(y)$  on  $\mathcal{L}_n$  is said to be  $S$ -concave if, for each doubly stochastic matrix  $S$ , of order  $n$ , and each  $y$  in  $\mathcal{L}_n$*

$$(14) \quad G(Sy) \geq G(y).$$

Now Birkhoff's theorem [5] states that the set of doubly stochastic matrices of order  $n$  is a convex polyhedron with  $N = n!$  permutation matrices  $P_i, i = 1, 2, \dots, N$ , as the vertices. Thus every doubly stochastic matrix  $S = \sum_{i=1}^N \alpha_i P_i$  for some set  $\alpha = \{\alpha_1, \dots, \alpha_N\}$ ,  $\alpha_i \geq 0$ ,  $\sum_{i=1}^n \alpha_i = 1$ . Hence we have the following Theorem 3 as a particular case of the Theorem 2.

**THEOREM 3.** *Let a function  $f(x) \geq 0$  on  $\mathcal{L}_n$  be symmetric (w.r.t. permutations) and satisfy the conditions (i) and (iii) of (2). Let  $E$  be a convex, symmetric (w.r.t. permutations) set of  $\mathcal{L}_n$ . Then  $\int_E f(x+y) dx$  is an  $S$ -concave function of  $y$ , i.e.,*

$$(15) \quad \int_E f(x + Sy) dx \geq \int_E f(x + y) dx,$$

for any doubly stochastic matrix  $S$ .

It is also well known [6] that, given two  $n$ -vectors  $y$  and  $z$ , there exists a doubly stochastic matrix  $S$ , such that  $z = Sy$  if, and only if,

$$(16) \quad \begin{aligned} z_{(1)} + \cdots + z_{(k)} &\leq y_{(1)} + \cdots + y_{(k)}, & k = 1, 2, \dots, n - 1, \\ z_{(1)} + \cdots + z_{(n)} &= y_{(1)} + \cdots + y_{(n)}, \end{aligned}$$

where  $y_{(i)}$  and  $z_{(i)}$ ,  $i = 1, 2, \dots, n$  are the coordinates of  $y$  and  $z$ , ordered in nonincreasing order of magnitude. The conclusion (15) of the Theorem 3 may thus be expressed as: For any two vectors  $y$  and  $z$  of  $n$ -space, satisfying (16) we have

$$(17) \quad \int_E f(x + z) \, dx \geq \int_E f(x + y) \, dx.$$

**COROLLARY 4.** *If the probability density function  $f(x)$  of a random  $n$ -vector  $X$  is unimodal and symmetric w.r.t. permutations of the coordinates of  $x$ , and  $Y$  is an independently distributed random vector then for any convex symmetric set  $E$  of  $\mathcal{L}_n$ , and any doubly-stochastic matrix  $S$ , we have*

$$(18) \quad \Pr\{X + SY \in E\} \geq \Pr\{X + Y \in E\}.$$

Furthermore, if  $h$  is a symmetric function such that  $\{x | h(x) \leq v\}$  is convex then

$$(19) \quad \Pr\{h(X + SY) \leq v\} \geq \Pr\{h(X + Y) \leq v\}.$$

(iii) Now let the transformation group of §2 be the group of the cyclic permutations of  $n$  coordinates. One has the matrix representation for this group as  $g_i = p^{i-1}$ ,  $i = 1, 2, \dots, n$ , where  $P$  is a permutation matrix given by

$$P = \begin{bmatrix} 0 & 0 & \cdots & 0 & 1 \\ 1 & 0 & \cdots & 0 & 0 \\ 0 & 1 & \cdots & 0 & 0 \\ \cdot & \cdot & \cdot & \cdot & \cdot \\ 0 & 0 & \cdots & 1 & 0 \end{bmatrix}.$$

Thus for any vector  $y$  of  $\mathcal{L}_n$  and a set  $\alpha = \{\alpha_1, \alpha_2, \dots, \alpha_n\}$ ,  $\alpha_i \geq 0$ ,  $i = 1, 2, \dots, n$ ,  $\sum \alpha_i = 1$  we have

$$\alpha(y) = \sum_{i=1}^n \alpha_i g_i y = \sum_{i=1}^n \alpha_i P^{i+1} y = C(\alpha)y,$$

where  $C(\alpha)$  is the doubly stochastic circulant matrix given by  $C(\alpha) = (c_{ij})$ ,  $c_{ij} = \alpha_k$ ,  $k = k(i, j) = i + j - 1 \pmod{n}$  or more explicitly,

$$C(\alpha) = \begin{pmatrix} \alpha_1 & \alpha_2 & \cdots & \alpha_{n-1} & \alpha_n \\ \alpha_2 & \alpha_3 & \cdots & \alpha_n & \alpha_1 \\ \alpha_3 & \alpha_4 & \cdots & \alpha_1 & \alpha_2 \\ \cdot & \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot & \cdot \\ \alpha_n & \alpha_1 & \cdots & \alpha_{n-2} & \alpha_{n-1} \end{pmatrix} .$$

Also, it is easy to verify that the set of all doubly stochastic circulant matrices form a convex polyhedron with the permutation matrices  $p^{i-1}$ ,  $i = 1, 2, \dots, n$ , as the vertices. We have, therefore, the following Theorem 4 as another particular case of the Theorem 2.

**THEOREM 4.** *Let a function  $f(x) \geq 0$  on  $n$ -space be unimodal and invariant under cyclic permutations. Let  $E$  be a convex set of  $n$ -space symmetric with respect to cyclic permutations. Let  $\int_E f(x) dx < \infty$ . Then for any doubly stochastic circulant matrix  $C$  and any vector  $y$  of  $n$ -space we have*

$$(18) \quad \int_E f(x + Cy) dx \geq \int_E f(x + y) dx.$$

It is easy to write down analogues of the Corollary 4 for this case.

One may similarly write down, with ease, the particular case of the Theorem 2 when the function  $f(x)$  and the set  $E$  are invariant under the group of  $2^n$  reflections in the coordinate planes.

**4. The inequality.** We shall now outline a somewhat different proof of a somewhat different and generalized version of the inequality of §2 without the finiteness condition on the group  $G$ .

**THEOREM 5.** *Let  $G = \{g\}$  be a group of linear Lebesgue measure-preserving transformations of  $\mathfrak{L}_n$  onto  $\mathfrak{L}_n$ . Let  $E$  be a convex,  $G$ -invariant region of  $\mathfrak{L}_n$ . Let  $f$  be a nonnegative real-valued,  $G$ -invariant and unimodal function on  $\mathfrak{L}_n$ . Then for arbitrary  $y$  in  $\mathfrak{L}_n$  we have*

$$(19) \quad \int_E f(x + z) dx \geq \int_E f(x + y) dx,$$

where  $z$  is any point in the convex-hull of the  $G$ -orbit of  $y$ .

**PROOF.** The theorem can be proved along the lines of the proof of the Theorem 2, by using the generalized version of the Brunn-Minkowski theorem due to Dinghas [2], [3], [4]. However, the argument may be simplified as follows, by using a twist suggested by

Kemperman in a personal communication.

The crucial step in the proof of the Theorem 2 is the statement (10), which holds without requiring  $G$  to be finite. To see this let us fix  $y$  and write,

$$(20) \quad Z = \{z \mid \phi(z) \geq \phi(y)\},$$

where

$$(21) \quad \phi(z) = \mu^{1/n}((E + z) \cap K_u).$$

Then by the Brunn-Minkowski theorem it follows that,

$$(22) \quad \phi(\lambda z_1 + (1 - \lambda) z_2) \geq \lambda \phi(z_1) + (1 - \lambda) \phi(z_2), \quad \text{for } 0 \leq \lambda \leq 1.$$

Hence for any  $y$  in  $\mathcal{L}_n$  the set  $Z$  of (20) is convex. Furthermore as in the proof of the Theorem 2 it can be verified that  $gy \in Z$  for each  $g \in G$ . Therefore, for any point  $z$  in the convex-hull of  $\{gy \mid g \in G\}$  we have

$$(23) \quad \mu((E + z) \cap K_u) \geq \mu((E + y) \cap K_u),$$

which is analogous to the statement (10). The proof from here on is the same as the proof of the Theorem 2.

The analogue of the Corollary 2 for the Theorem 5 is easy to formulate. A group  $G$  of special interest in probability and statistics is the group of orthogonal transformations. The  $G$ -orbit of any  $y$  for this group is the sphere  $\sum_{i=1}^n x_i^2 = \|y\|^2$  in  $\mathcal{L}_n$ . This special case may be studied as in §3 without any difficulty.

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