

## ON EQUAL DISTRIBUTIONS

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It is shown that two distributions both of which have a finite expectation are equal if and only if for every  $n \geq 1$  there exists  $1 \leq k \leq n$  such that the  $k$ th order statistics from samples of size  $n$  of each distribution have equal expectations.

Similarly, it is shown that a distribution with finite expectation is symmetric about zero if and only if for every  $n \geq 0$  there exists  $0 \leq k \leq 2n + 1$  such that the sum of the expectations of the  $k$ th smallest and the  $k$ th largest observations in a sample of size  $2n + 1$  is zero.

The object of this paper is to give a characterization of distributions of random variables whose expectations exist and are finite.

In this paper,  $(X_{(1)}^n, \dots, X_{(n)}^n)$  will denote the order statistic (in increasing order) of  $n$  independent observations of the random variable  $X$ .

**THEOREM 1.** *Let  $X, Y$  be random variables whose expectations exist and are finite, and let  $F, G$  denote their respective distributions. A necessary and sufficient condition for  $F$  to equal  $G$  is that for every positive integer  $n$  there exist a  $k_n, 1 \leq k_n \leq n$ , such that*

$$(1) \quad EX_{(k_n)}^n = EY_{(k_n)}^n .$$

**PROOF.** The necessity of the condition is clear. The sufficiency will be proved by the following two lemmas.

**LEMMA 1.** *Let  $X, Y$  be defined as in Theorem 1. If for every positive integer  $n$  there exists a  $k_n, 1 \leq k_n \leq n$ , such that (1) holds, then for every  $n$  and every  $1 \leq i \leq n$ .*

$$(2) \quad EX_{(i)}^n = EY_{(i)}^n .$$

**PROOF.** By induction on  $n$  and  $i$ . (2) clearly holds for  $n = 1, i = 1$ . Suppose that (2) holds for all  $1 \leq i \leq n, n \leq r$ . It suffices to prove that  $EX_{(i)}^{r+1} = EY_{(i)}^{r+1}$  for all  $1 \leq i \leq r + 1$ . By hypothesis,  $EX_{(k_{r+1})}^{r+1} = EY_{(k_{r+1})}^{r+1}$ .

Let

$A_j^n$  = the set of all subsets of  $\{1, \dots, n\}$  which contain exactly  $j$  different elements

$(S_{(1)}^\sigma, \dots, S_{(j)}^\sigma)$  = the order statistic (in increasing order) of  $\{X_{(i)}^n \mid i \in \sigma, \sigma \in A_j^n\}$

$(T_{(1)}^\sigma, \dots, T_{(j)}^\sigma)$  = the order statistic (in increasing order) of  $\{Y_{(i)}^n \mid i \in \sigma, \sigma \in A_j^n\}$

$$V_i^{r+1} = \sum_{\sigma \in A_{r+1}} S_{(i)}^\sigma$$

$$W_i^{r+1} = \sum_{\sigma \in A_{r+1}} T_{(i)}^\sigma$$

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By the induction hypothesis,  $ES_{(i)}^\sigma = ET_{(i)}^\sigma$  for each  $\sigma \in A_r^{r+1}$ ,  $1 \leq i \leq r$ ; and so for every  $1 \leq i \leq r$ :

$$(3) \quad EV_i^{r+1} = EW_i^{r+1}.$$

Clearly, for  $1 \leq i \leq r + 1$

$$(4) \quad V_i^{r+1} = (r + 1 - i)X_{(i)}^{r+1} + iX_{(i+1)}^{r+1}$$

$$(5) \quad W_i^{r+1} = (r + 1 - i)Y_{(i)}^{r+1} + iY_{(i+1)}^{r+1}.$$

If  $1 \leq k_{r+1} < r + 1$ , then taking  $i = k_{r+1}$  in (3), (4) and (5), one gets that

$$iE(X_{(k_{r+1}+1)}^{r+1} - Y_{(k_{r+1}+1)}^{r+1}) = E(V_{k_{r+1}}^{r+1} - W_{k_{r+1}}^{r+1}) - (r + 1 - i)E(X_{(k_{r+1})}^{r+1} - Y_{(k_{r+1})}^{r+1}) = 0$$

and so  $EX_{(j)}^{r+1} = EY_{(j)}^{r+1}$  for  $j = k_{r+1} + 1$  and inductively for  $j = k_{r+1} + 2, \dots, r + 1$ .

For  $1 < k_{r+1} \leq r + 1$ , taking  $i = k_{r+1} - 1$  in (3), (4) and (5) in the same way gives  $EX_{(j)}^{r+1} = EY_{(j)}^{r+1}$  for  $j = k_{r+1} - 1$ , and inductively for  $j = k_{r+1} - 2, \dots, 1$ .

Therefore (2) holds for  $n \leq r + 1$ ,  $1 \leq i \leq n$ , and by induction for all  $n$ ,  $1 \leq i \leq n$ , proving Lemma 1.

LEMMA 2. Let  $X, Y, F, G$  be defined as in Theorem 1. Let (2) hold for every  $n$  and every  $i$ ,  $1 \leq i \leq n$ . Then  $F = G$ .

PROOF. Let  $0 < \alpha < 1$  be such that there are unique  $x_\alpha$  and  $y_\alpha$  with  $F(x_\alpha -) \leq \alpha \leq F(x_\alpha)$  and  $G(y_\alpha -) \leq \alpha \leq G(y_\alpha)$ . To show  $F = G$  it is enough to show  $x_\alpha = y_\alpha$  for such pairs. Select integers  $\rho_n$  such that  $\rho_n/n \rightarrow \alpha$ . Then if  $F_n$  is the empirical distribution function based on  $n$  observations of  $X$ ,  $F_n(X_{(\rho_n)}^n) = \rho_n/n \rightarrow \alpha$  (if there are ties,  $F_n$  may jump above  $\rho_n/n$  at  $X_{(\rho_n)}^n$  but we give the equation  $F_n(X_{(\rho_n)}^n) = \rho_n/n$  the obvious interpretation) so that  $X_{(\rho_n)}^n \rightarrow_{a.s.} x_\alpha$ . Similarly,  $Y_{(\rho_n)}^n \rightarrow_{a.s.} y_\alpha$ . It is easy to see that  $EX_{(\rho_n)}^n \rightarrow x_\alpha$  and  $EY_{(\rho_n)}^n \rightarrow y_\alpha$ . But since  $EX_{(\rho_n)}^n = EY_{(\rho_n)}^n$  (by hypothesis), this implies that  $x_\alpha = y_\alpha$ .

This completes the proof of Lemma 2 and thus the proof of Theorem 1.

REMARK. In particular,  $F = G$  if  $E \max_{i=1, \dots, K} X_i = E \max_{i=1, \dots, K} Y_i$  for all  $K$  (or for  $K \geq n_0$ , as can be shown by similar methods; this would be enough also if  $EX = -\infty$  but  $E \max_{i=1, \dots, K} X_i$  is finite for  $K \geq n_0$ ,  $n_0$  arbitrary).

THEOREM 2. Let  $X$  be a random variable whose expectation exists. Then the distribution  $F$  of  $X$  is symmetric about zero if and only if for each nonnegative integer  $n$  there exists a  $k_n$ ,  $1 \leq k_n \leq 2n + 1$ , such that  $E(X_{(k_n)}^{2n+1} + X_{(2n+2-k_n)}^{2n+1}) = 0$ .

PROOF. The necessity of the condition is clear. The sufficiency: Denote  $Y = -X$ . As in the proof of Theorem 1, one first proves (by induction on  $n$ ) that

$$(6) \quad EX_{(i)}^n = EY_{(i)}^n = -EX_{(n+1-i)}^n$$

for every  $i, n$ ;  $1 \leq i \leq n$ . (6) clearly holds for  $n = 1, i = 1$ . Suppose (6) holds for all  $i, n$  where  $1 \leq i \leq n, 1 \leq n \leq 2m + 1$ . In particular,  $EX_{(m+1)}^{2m+1} = -EX_{(m+1)}^{2m+1} = 0$ , and so  $EV_{m+1}^{2m+2} = 0$  (where  $V_i^{r+1}$  is as denoted in the proof of

Lemma 1). However  $V_{m+1}^{2m+2} = (m+1)X_{(m+1)}^{2m+2} + (m+1)X_{(m+2)}^{2m+2}$ . Thus (6) holds also for  $n = 2m + 2, i = m + 1$ . By reasoning analogous to the proof of Lemma 1, one gets that (6) holds for  $n = 2m + 2$  and all  $i = 1, \dots, 2m + 2$ , and likewise, also for  $n = 2m + 3$  and all  $i = 1, \dots, 2m + 3$ , and by induction for every  $i, n, n > 0, 1 \leq i \leq n$ . This shows the conditions of Lemma 2 to hold, and so  $X$  and  $-X$  have the same distribution, completing the proof of Theorem 2.

REMARK. In particular,  $F$  is symmetric about zero if and only if

$$E(\max_{i=1, \dots, 2n+1} X_i + \min_{i=1, \dots, 2n+1} X_i) = 0$$

and likewise if and only if  $E \text{ median}_{i=1, \dots, 2n+1} X_i = 0$  for all nonnegative integers  $n$ , where  $X_1, \dots, X_{2n+1}$  are independent observations of  $X$ . Using similar methods, one can prove that if the expectation of  $X$  does not exist, but the expectations of the medians of large enough samples do and are equal to zero, then the distribution of  $X$  is symmetric (about zero).

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