# CHARACTERIZATION OF DISTRIBUTIONS BY THE EXPECTED VALUES OF THE ORDER STATISTICS\*

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#### Introduction

Several parametric families of distributions have been characterized by the properties of their order statistics. See, for example, Ferguson [4] and David ([3], p. 19). An interesting non-parametric result is due to Chan [2] and Konheim [6]: Let  $X_1, X_2, \cdots$  be independent random variables with common distribution function F. Let  $X_{1,n} \leq \cdots \leq X_{n,n}$  be the order statistics of  $X_1, \cdots, X_n$ . If  $E | X_1 |$  is finite then F is determined by the sequence  $\{E(X_{1,n}): n=1, 2, \cdots\}$  (and likewise, by  $\{E(X_{n,n}): n=1, 2, \cdots\}$ ). Wang [12] showed that for any  $k \geq 1$ , the sequence  $\{E(X_{k,n}): n=k, k+1, \cdots\}$  (and likewise,  $\{E(X_{n-k+1,n}): n=k, k+1, \cdots\}$ ) determines F, also assuming the finiteness of  $E | X_1 |$ . In this paper we show that under lesser restrictions on the moments of  $X_1$ , the distribution F is characterizable by more general sequences.

## 1. Characterization by a tail sequence

Throughout this paper we shall let  $\{k(n): n=1, 2, \cdots\}$  denote a sequence of integers with  $1 \le k(n) \le n$ .  $\{k(n)\}$  is said to satisfy property (A-m) if furthermore

(1) 
$$k(m) \leq k(n) \leq k(m) + n - m \quad \text{for all } n \geq m.$$

We note that each  $\{k(n)\}$  automatically satisfies property (A-1).

LEMMA 1. If  $\{k(n)\}$  satisfies (A-m) for some m, then

(2) 
$$\int_0^1 f(x)x^{k(n)-1}(1-x)^{n-k(n)}dx=0, \quad n=m, m+1, \cdots$$

implies f(x) = 0 a.e. (0, 1).

PROOF. Letting  $g(x) = f(x)x^{k(m)-1}(1-x)^{m-k(m)}$  and  $P_i(x) = x^{k(m+i)-k(m)}$ .  $(1-x)^{i-k(m+i)+k(m)}$ ,  $i=0,1,\cdots$ , we may write (2) in the form

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(3) 
$$\int_0^1 g(x)P_i(x)dx = 0, \quad i = 0, 1, \dots.$$

It follows from (A-m) that  $P_i(x)$  is a polynomial (of degree i) and therefore (3) is equivalent to  $\int_0^1 g(x)x^idx=0$ ,  $i=0,1,\cdots$ . It is well known (see Sz.-Nagy [11], p. 331) that

(4) 
$$\int_{a}^{1} g(x)x^{i}dx = 0, \quad i = 0, 1, \dots \Rightarrow g(x) = 0 \text{ a.e. } [0, 1].$$

Hence f(x)=0 a.e. [0, 1].

In other words, the lemma asserts the completeness of a certain subfamily of the beta distributions. Since  $E(X_{k,n})=E[F^{-1}(Y)]$ , where Y is the beta random variable with parameters (k, n-k+1),  $F^{-1}$  is the inverse function of F (see Moriguti [7] and Hájek and Šidák [5]) defined by

$$F^{-1}(s) = \inf \{x : F(x) \ge s\}$$
  $0 < s < 1$ ,

it leads to the following application.

THEOREM 1. Let  $X_{1,n} \leq \cdots \leq X_{n,n}$  be the order statistics from the distribution function F. If the sequence  $\{k(n)\}$  satisfies the property (A-m) for some m and  $E \mid X_{k(m),m} \mid < \infty$ , then F is uniquely determined by the sequence of numbers

(5) 
$$\{E(X_{k(n),n}): n=m, m+1,\cdots\}$$
.

PROOF. We first show that each term in (5) is finite. By (A-m) it follows that for  $n \ge m$ ,  $k(n) \ge k(m)$  and  $n-k(n) \ge m-k(m)$ . Thus

$$\begin{split} \mathrm{E} \; |X_{k(n),n}| &= C_n \int_0^1 |F^{-1}(s)| \, s^{k(n)-1} (1-s)^{n-k(n)} ds \\ &\leq C_n \int_0^1 |F^{-1}(s)| \, s^{k(m)-1} (1-s)^{m-k(m)} ds \\ &= C_n C_m^{-1} \, \mathrm{E} \; |X_{k(m),m}| < \infty \; , \end{split}$$

where  $C_i = k(i) \binom{i}{k(i)}$ . Now suppose G is a distribution function whose order statistics  $Y_{k,n}$  satisfy

$$E(X_{k(n),n}) = E(Y_{k(n),n}), \quad n=m, m+1, \cdots.$$

It is equivalent to

$$\int_0^1 [F^{-1}(s) - G^{-1}(s)] s^{k(n)-1} (1-s)^{n-k(n)} ds = 0 , \qquad n = m, \, m+1, \cdots .$$

From Lemma 1 it follows that  $F^{-1}=G^{-1}$  a.e. Finally, it follows from the left continuity of  $F^{-1}$  that F=G. This completes the proof.

We wish to point out that, in order to characterize a distribution F via the sequence (5) it is necessary that each term in (5) is finite. Clearly, there are infinitely many distributions having, say,

$$E |X_{n,n}| = \infty$$
,  $n=1, 2, \cdots$ .

Thus the sequence  $\{E(X_{n,n})\}$  will not characterize, for example, the Cauchy distribution nor the Pareto distribution defined by

(6) 
$$F(x)=1-x^{-1}, x>1.$$

Of course, one notices that  $E(X_{1,2})$  is finite for the case of (6) and thus Theorem 1 is applicable with the choice m=2, k(2)=1.

It is also possible to characterize a distribution for which none of its order statistics has a finite mean. For example, let X be distributed by  $F(x)=1-(\log x)^{-1}$ , x>e. Stoops and Barr [10] showed that not only is  $E\mid X\mid^{\delta}=\infty$  for all  $\delta>0$  but also  $E\mid X_{k,n}\mid^{\delta}=\infty$  for all (k,n) and all  $\delta>0$ . Consider the random variable  $Y=\log(\log X)$ . We see that Y is exponentially distributed and  $E(Y_{k,n})$  is finite for all k,n. Thus one can characterize the distribution of Y via  $\{E(Y_{k,n})\}$ , which in turn characterizes the distribution of X.

In general, if  $X \sim F$  and  $\phi$  is a measurable 1-1 function on the support of F, to characterize X it is equivalent to characterize  $Y = \phi(X)$ . If furthermore  $\phi$  is monotone increasing then  $Y_{k,n} = (\phi(X))_{k,n} = \phi(X_{k,n})$  and thus  $E \mid \phi(X_{k(m),m}) \mid < \infty \Rightarrow E \mid \phi(X_{k(n),n}) \mid < \infty$ ,  $n \ge m$  for those  $\{k(n)\}$  satisfying (A-m). For monotone decreasing  $\phi$  the situation is analogous except for interchanging  $\phi(X_{k,n})$  and  $\phi(X_{n-k+1,n})$ . A simple application of this idea leads to the following result:

COROLLARY 1. Let r be a positive odd integer and let  $E | X_1|^{\delta}$  be finite for some  $\delta > 0$ . If  $\{k(n)\}$  satisfies (A-m) for some  $m \ge 1$ , and if  $r\delta^{-1} \le k(m) \le m+1-r\delta^{-1}$ , then F is determined by

(7) 
$$\{E(X_{k(n),n}^r): n=m, m+1,\cdots\}$$
.

PROOF. Since  $\phi(x)=x^r$  is a monotone increasing function there is no ambiguity about the meaning of (7). Clearly, (7) determines F as long as each term there is finite. It remains to show that this is implied by the finiteness of  $E |X_1|^s$ . That this is true is the consequence of the following lemma which was proved by Sen [9] for absolutely continuous distribution F. His proof continues to hold for arbitrary distribution function F with appropriate modifications.

LEMMA 2 (Sen). If  $E |X_1|^{\delta} < \infty$  for some  $\delta > 0$  then  $E |X_{k,n}|^{r} < \infty$  for all n and k satisfying  $r\delta^{-1} \le k \le n+1-r\delta^{-1}$ .

COROLLARY 2. Let  $E|X_1| < \infty$ . If for some m, either  $\{k(n): n = m, m+1, \cdots\}$  or  $\{n-k(n): n = m, m+1, \cdots\}$  is non-decreasing, then F is determined by (5).

PROOF. It suffices to show that  $\{k(n)\}$  satisfies  $(A-m_0)$  for some  $m_0$ . Suppose k(n) is non-decreasing. There exists  $m_0 \ge m$  such that  $m_0 - k(m_0) \le n - k(n)$  for all  $n \ge m_0$ . Thus  $(A-m_0)$  is satisfied. It thus follows from Theorem 1 that F is determined by the subset of (5):

$$\{E(X_{k(n),n}): n=m_0, m_0+1,\cdots\}$$
.

The case with non-decreasing n-k(n) is analogous.

Wang's result (and, a fortiori, Chan's) is a special case of our next corollary, which is itself a special case of Corollary 2.

COROLLARY 3. If  $E|X_1| < \infty$  then F is determined by  $\{E(X_{k,n}): n=m, m+1, \cdots\}$  for any k and any  $m \ge k$  (and, likewise, by  $\{E(X_{n-k+1,n}): n=m, m+1, \cdots\}$ ).

Chan's result is a special case of our next corollary.

COROLLARY 4. If  $E|X_1| < \infty$  then F is determined by  $\{E(X_{k(n),n}): n=1, 2, \cdots\}$  for any  $\{k(n)\}$ .

PROOF. This is a special case (m=1) of Theorem 1. As was mentioned before, the constraint (A-m) in Theorem 1 is vacuous when m=1 since all  $\{k(n)\}$  are supposed to satisfy  $1 \le k(n) \le n$ .

# 2. Characterization by a subsequence

It seems natural to ask: Is F determined if more than finitely many terms are removed from the sequence  $\{E(X_{k(n),n}): n=1, 2, \cdots\}$ ? Under some conditions the answer is affirmative. For instance, by using (4) alone it can be seen that for any fixed positive integers k, a and b, the equally spaced subsequence  $\{E(X_{k,n}): n=a, a+b, a+2b, \cdots\}$  does determine F. This is a special case of our next theorem, whose proof depends on a generalization of the uniqueness theorem (4). The following lemma is proved in Boas ([1], Theorem 12.4.4, p. 235):

LEMMA 3 (Müntz). Let  $f \in L_1(0, 1)$ . If

(8) 
$$\int_{0}^{1} f(x)x^{n_{i}}dx = 0, \quad i=1, 2, \cdots,$$

where  $n_i$  are distinct positive real numbers with  $\sum_{i=1}^{\infty} n_i^{-1} = \infty$  then f(x) = 0 a.e. (0, 1).

THEOREM 2. Let  $E |X_{k,n}| < \infty$  for some k and n (say,  $= n_1 \ge k$ ). Then F is uniquely determined by

(9) 
$$\{E(X_{k,n}): n=n_1, n_2, \cdots\}$$

for any sequence of integers  $n_2, n_3, \cdots$  such that  $n_1 < n_2 < n_3 < \cdots$  and  $\sum_{i=1}^{\infty} n_i^{-1} = \infty \quad (and, by \quad \{E(X_{n-k+1,n}): n=n_1, n_2, \cdots\} \quad provided \quad E|X_{n_1-k+1,n_1}| < \infty).$ 

PROOF. The proof is essentially the same as in Theorem 1. Instead of (4), we now have the stronger Lemma 3. It suffices to point out that  $\sum_{i=2}^{\infty} (n_i - n_1)^{-1}$  diverges if and only if  $\sum_{i=2}^{\infty} n_i^{-1}$  does. The parenthetical remark is proved by interchanging x and 1-x.

We note that Corollary 3 is also a special case of Theorem 2.

The result of Lemma 3 appears self-strengthening in that if k(n) keeps taking on a fixed value "often" enough, then

(10) 
$$\int_0^1 f(x) x^{k(n)-1} (1-x)^{n-k(n)} dx = 0 , \qquad n = n_1, n_2, \cdots$$

implies f=0 a.e. For instance, let  $\{k(n_i)\}$  be bounded and let  $\sum_{i=1}^{\infty} n_i^{-1} = \infty$ . It follows that there exists a subsequence  $\{m_i\} \subset \{n_i\}$  such that  $k(m_i)$  is constant (say,  $=k^*$ ) and  $\sum m_i^{-1} = \infty$ . By (10) we have  $\int g(x)x^ndx = 0$ ,  $n=m_1, m_2, \cdots$ , where  $g(x)\equiv f(1-x)x^{-k^*}(1-x)^{k^*-1}$ . Thus we see by Lemma 3 that  $g\equiv 0$ , and so is f. Indeed, the boundedness of  $\{k(n)\}$  is not necessary. It suffices to have  $\{k(n)\}$  bounded "sufficiently often" in the sense that the sum of  $n_i^{-1}$  over  $\{i \mid k(n_i) \leq c \text{ or } n_i - k(n_i) \leq c\}$  for some constant c diverges. In terms of application, this means that F is then characterizable by sequences of form

(11) 
$$\{ \mathbb{E} (X_{k(n),n}) : n = n_1, n_2, \cdots \}.$$

Inasmuch as k(n) need not be a constant function (or even bounded), (11) appears to be more general than (9).

This "strengthening", however, is illusory. It amounts to the detection of a subsequence of (11) to which Theorem 2 is applicable (and simply ignore the rest of the terms). In other words, the question is whether or not F is already determined by some subset of (11). Since (10) demands more than (8), it is thus anything but a strengthening.

COROLLARY 5. Let  $\{\min\{k(n), n-k(n)\}\}\$  be bounded on  $\{n_i\}$  with  $\sum n_i^{-1} = \infty$ . Then F is determined by (11).

Remark 1. Without the constraint of min  $\{k(n), n-k(n)\}$  being bounded "sufficiently often", the sequence (11) will not characterize F. Take, for example, two symmetric distributions F and G with the same mean (say, =0). It is clear that the expected values of their medians are all identical, namely,  $E(X_{(n+1)/2,n})=E(Y_{(n+1)/2,n})=0$  for  $n=1,3,5,\cdots$ . Thus  $\{E(X_{(n+1)/2,n}): n=1,3,5,\cdots\}$  does not characterize F.

Remark 2. We wish to conclude this paper with a question: Is Theorem 1 true without the extra constraint (A-m) on the sequence  $\{k(n)\}$ ? Let  $\{k(n)|n\geq m\}$  be such that for each  $m_0\geq m$  the property  $(A-m_0)$  is violated. This means that there exists at least one value of  $n > m_0$  such that either  $k(n) < k(m_0)$  or  $n-k(n) < m_0-k(m_0)$ . If there are "sufficiently many" such n's, F will be determined by virtue of Corollary 5. It is when there are, for each  $m_0$ , some but not "many enough" such n's that we do not know the answer. Our conjecture is that (A-m) is not really needed in Theorem 1.

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### Addendum

After this paper was accepted for publication it came to our attention that Corollary 4 was obtained also by Pollak [8] via a different argument.

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