## RECURRENCE RELATIONS BETWEEN MOMENTS OF ORDER STATISTICS FOR EXCHANGEABLE VARIATES

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Let  $X_{i:n}$   $(i=1, 2, \dots, n)$  be the order statistics obtained by re-arranging in non-decreasing order of magnitude the variates  $X_i$  having common marginal cdf P(x). Denote by  $F_{i:n}(x)$  and  $\mu_{i:n}$  the cdf and expected value of  $X_{i:n}$ . Recurrence relations for moments and other functions of the  $X_{i:n}$  have been derived by many authors, usually on the assumption that the  $X_i$  are independent continuous variates. The most basic of these relations states that for  $r=1, 2, \dots, n-1$ ,

$$n\mu_{r:n-1} = r\mu_{r+1:n} + (n-r)\mu_{r:n}.$$

In a paper which has appeared since a longer version of this note was submitted for publication, Young [6] shows (in effect) that (1) and hence results deducible from (1) continue to hold if the  $X_i$  are exchangeable, continuous or discrete variates, i.e., if  $\Pr\{X_1 \leq x_1, X_2 \leq x_2, \dots, X_n \leq x_n\}$  is symmetric in  $x_1, x_2, \dots, x_n$ . In this note we point out a simple argument which establishes (1) and multivariate generalizations thereof for exchangeable variates. Also, we give an application of the result

(2) 
$$F_{n-1:n}(x) = nF_{n-1:n-1}(x) - (n-1)F_{n:n}(x),$$

which is just the special case r = n - 1 of the counterpart of (1) for cdf's.

We illustrate our argument on the bivariate case. Let  $F_{r,s:n}(x, y)$  denote the joint cdf of  $X_{r:n}$  and  $X_{s:n}$  ( $1 \le r < s \le n$ ;  $x \le y$ ) and let  $\mu_{r,s:n} = \mathcal{E}(X_{r:n}X_{s:n})$ . Now of the n variates  $X_i$  drop one at random and let  $Y_{i:n-1}$  ( $i=1,2,\cdots,n-1$ ) denote the ith order statistic in the reduced set of n-1 exchangeable variates. Then according as the variate dropped is one of the (a) first r, (b) next s-r, (c) last n-s, of the  $X_{i:n}$  ( $1 \le r < s \le n-1$ ), we see that  $Y_{r:n-1}$ ,  $Y_{s:n-1}$  are distributed jointly as (a)  $X_{r+1:n}$ ,  $X_{s+1:n}$  or (b)  $X_{r:n}$ ,  $X_{s+1:n}$  or (c)  $X_{r:n}$ ,  $X_{s:n}$ . Since the events (a), (b), (c), have respective probabilities r/n, (s-r)/n, (n-s)/n, it follows that for any x, y ( $x \le y$ )

(3) 
$$nF_{r,s:n-1}(x,y) = rF_{r+1,s+1:n}(x,y)$$

$$+ (s-r)F_{r,s+1:n}(x,y) + (n-s)F_{r,s:n}(x,y).$$

Differentiating or differencing, multiplying by  $e^{itx+iuy}$  and integrating or summing, we obtain the same relation between pdf's, characteristic functions, and

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TABLE 1									
Upper 5 and 1% points of $X_{n-1:n}$ , the	ie secon	d largest an	iong n equi-	-correlated	standard normal				
variates	with	correlation	coefficient						

n	2	3	4	5	6	7
5%	1.100	1.400	1.569	1.685	1.773	1.843
1%	1.713	1.981	2.134	2.242	2.324	2.390
n	8	9	10	11	12	
5%	1.901	1.950	1.993	2.031	2.065	
1%	2.443	2.490	2.532	2.569	2.597	

hence raw moments of any order (provided these moments exist). In particular, this gives the result

(4) 
$$n\mu_{r,s:n-1} = r\mu_{r+1,s+1:n} + (s-r)\mu_{r,s+1:n} + (n-s)\mu_{r,s:n},$$

established by Govindarajulu [3] for independent identically distributed continuous variates. For the equi-correlated multivariate normal case (with common marginal cdf). (4) may also be proved with the help of expressions for the moments of order statistics given by Owen and Steck [5].

As an application of (2) consider the problem of testing n "treatment" means against a control "mean" (Dunnett [1]). Let  $Z_{ij}$  and  $Z_{0h}$  ( $i=1, 2, \dots, n$ ;  $j=1, 2, \dots, k$ ;  $h=1, 2, \dots, l$ ) be mutually independent normal variates,  $Z_{ij}$  and  $Z_{0h}$  being respectively  $N(\mu_i, \sigma^2)$  and  $N(\mu_0, \sigma^2)$ , with  $\sigma^2$  assumed known. In order to test simultaneously whether any of the treatment means  $\bar{Z}_i$  differ from the control mean  $\bar{Z}_0$  we may use the statistic

(5) 
$$X_{n:n} = \max X_i = \max_{i=1,2,\dots,n} (\bar{Z}_i - \bar{Z}_0) / \sigma (1/k + 1/l)^{\frac{1}{2}}.$$

Here the  $X_i$  are equi-correlated standard normal variates with  $\rho = k/(k+l)$ . The cdf of  $X_{n:n}$  for various  $\rho$  has been tabulated by Gupta [4] whose tables may therefore be used to obtain  $F_{n-1:n}(x)$  and hence percentage points of  $X_{n-1:n}$ . For the case k=l, i.e.,  $\rho=\frac{1}{2}$ , upper 5 and 1% points are given in Table 1. Young [6] also tabulates upper percentage points of  $X_{n-1:n}$  for  $n \leq 8$  but we disagree with some of his values. The more general statistic, in which  $\sigma$  is replaced by an estimator S such that  $\nu S^2/\sigma^2$  is distributed as  $\chi^2$  with  $\nu$  d.f., independently of the numerator in (5), can now be handled by studentization. As pointed out by Fisher [2] in connection with harmonic analysis a test of the second largest variate becomes of special interest when the test on the largest is inconclusive, that is, close to the chosen level of significance.

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