## THE CALCULATION OF CUMULANTS VIA CONDITIONING

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The cumulants of random variables are important in deriving, for statistics of interest, exact sampling distributions, approximate sampling distributions (as via Cornish-Fisher expansions) and asymptotic sampling distributions (such as asymptotic normality). This note presents a means of calculating cumulants when two or more stages of sampling may be recognized.

Given the k-variate random variable  $(x_1, \dots, x_k)$ , let A denote an event in the associated probability field. The following properties of first and second order cumulants are well known (see Hansen, Hurwitz and Madow ([4], pp. 61-66) or Feller ([3], p. 164)).

$$(1) Ex_i = E\{E(x_i|A)\}$$

(2) 
$$\operatorname{var} x_i = E_A \{ \operatorname{var} (x_i | A) \} + \operatorname{var}_A \{ E(x_i | A) \}$$

(3) 
$$\operatorname{cov}(x_i, x_j) = E_A \{ \operatorname{cov}(x_i, x_j | A) \} + \operatorname{cov}_A \{ E(x_i | A), E(x_j | A) \}$$

for  $i, j=1, \dots, k$  where |A| indicates that calculations are carried out conditionally on the event A, while the subscript A indicates that calculations are carried out over the various values of A. Let  $\kappa(x_1, \dots, x_k)$  denote the joint kth order cumulant of  $(x_1, \dots, x_k)$  and for integers  $\beta_1, \dots, \beta_k$  let  $\kappa_{\beta_1 \dots \beta_k}(x_1, \dots, x_k) = \kappa(x_1[\beta_1 \text{ times}], \dots, x_k[\beta_k \text{ times}])$ . In this note we generalize (1), (2), (3) to

(4) 
$$\kappa(x_1, \dots, x_k) = \sum_{\alpha} \kappa_{A} \{\kappa(x_{\alpha_1}|A), \dots, \kappa(x_{\alpha_p}|A)\}.$$

The summation in (4) extends over all partitions  $\alpha = (\alpha_1, \dots, \alpha_p)$ ,  $p=1,\dots,k$  of the integers  $(1,\dots,k)$  and  $x_{\alpha_j}=(x_{j_1},\dots,x_{j_j})$  if  $\alpha_j=(j_1,\dots,j_j)$ . We may prove,

THEOREM. Given the k-variate random variable  $(x_1, \dots, x_k)$  with  $E|x_i|^k < \infty$ ,  $i=1, \dots, k$ , the identity (4) is valid.

PROOF.  $\kappa(x_1, \dots, x_k)$  is the coefficient of  $t_1 \dots t_k$  in the Taylor series

expansion of  $\log E(\exp \sum_i x_i t_i)$  about the origin. (This expansion may be carried out because of the assumed finiteness of moments.) However

$$(5) \qquad \log E(\exp \sum x_i t_i) = \log E_A \{ E(\exp \sum x_i t_i | A) \}$$

$$= \log E_A \left\{ \exp \left( \sum \kappa_{\beta_1 \cdots \beta_k} (x_1, \cdots, x_k | A) \frac{t_1^{\beta_1} \cdots t_k^{\beta_k}}{\beta_1! \cdots \beta_k!} + o(\|t\|^k) \right) \right\}$$

where  $||t||^2 = t_1^2 + \cdots + t_k^2$  and the summation extends over integers  $\beta_i$ ,  $0 \le \beta_i \le k$ ,  $i = 1, \dots, k$ , with  $0 < \sum_i \beta_i \le k$ . We note that the expression on the right-hand side of (5) is essentially the cumulant generating function of the random variables  $\kappa_{\beta_1 \cdots \beta_k}(x_1, \cdots, x_k | A)$ . The stated result now follows on identification of the coefficient of  $t_1 \cdots t_k$ .

COROLLARY. The kth order cumulant  $\kappa_k(x)$  of a univariate random variable x, with  $E|x|^k < \infty$ , is given by

where the summation extends over all partitions  $(p_1^{\mu_1}, p_2^{\mu_2}, \cdots)$  of k with  $p_1\mu_1 + p_2\mu_2 + \cdots = k$ .

This corollary follows from the theorem on taking  $x_i=x$ ,  $i=1, \dots, k$ , and counting the identical terms.

We now turn to several examples of the theorem and corollary.

Example 1. Mixtures. Suppose that the probability measure of  $(x_1, \dots, x_k)$  is in fact a mixture, that is its c.d.f.  $F(x_1, \dots, x_k)$  is of the form

(7) 
$$F(x_1, \dots, x_k) = \int G(x_1, \dots, x_k; \theta) dU(\theta)$$

where, for fixed  $\theta$ ,  $G(x_1, \dots, x_k; \theta)$  is a c.d.f. and  $U(\theta)$  is a probability measure in  $\theta$ . The theorem allows us to express the kth order joint cumulant of  $(x_1, \dots, x_k)$  in terms of the cumulants calculated from  $G(x_1, \dots, x_k; \theta)$  for fixed  $\theta$ . The required expression is given by (4) taking A to refer to  $\theta$ .

This result is given for the first and second order cases in Feller ([3], p. 164).

The next example refers to the sum of a random number of random variables.

Example 2. Cumulants of random sums. Let  $x_1, x_2, \cdots$  be a sequence of independent, identically distributed random variables with  $\kappa_j(x) = \kappa_j$ ,  $j = 1, \dots, k$  existing, and n an integer valued random variable

distributed independently of the sequence, whose moments exist up to order k. Let  $S_n = x_1 + \cdots + x_n$ . From (1) and (2) above, letting A refer to n, we see  $ES_n = (Ex)(En)$  and  $var S_n = (En) var x + var n(Ex)^2$ . In general we have from the corollary, taking A to refer to n,

$$\begin{array}{ll} (\ 8\ ) & \qquad \kappa_{\scriptscriptstyle k}(S_{\scriptscriptstyle n}) \! = \! \sum \frac{k!}{\mu_{\scriptscriptstyle 1}! \; \mu_{\scriptscriptstyle 2}! \; \cdots \;} \frac{1}{(p_{\scriptscriptstyle 1}!)^{\mu_{\scriptscriptstyle 1}} (p_{\scriptscriptstyle 2}!)^{\mu_{\scriptscriptstyle 2}} \cdots \;} \kappa_{\scriptscriptstyle \mu_{\scriptscriptstyle 1}\mu_{\scriptscriptstyle 2}} \! \ldots \! \{ n \kappa_{\scriptscriptstyle p_{\scriptscriptstyle 1}}, \; n \kappa_{\scriptscriptstyle p_{\scriptscriptstyle 2}}, \; \cdots \} \\ & = \! \sum \frac{k!}{\mu_{\scriptscriptstyle 1}! \; \mu_{\scriptscriptstyle 2}! \; \cdots \;} \frac{1}{(p_{\scriptscriptstyle 1}!)^{\mu_{\scriptscriptstyle 1}} (p_{\scriptscriptstyle 2}!)^{\mu_{\scriptscriptstyle 2}} \cdots \;} \kappa_{\scriptscriptstyle p_{\scriptscriptstyle 1}}^{\mu_{\scriptscriptstyle 1}} \kappa_{\scriptscriptstyle p_{\scriptscriptstyle 2}}^{\mu_{\scriptscriptstyle 2}} \cdots \; \kappa_{\scriptscriptstyle \mu_{\scriptscriptstyle 1}+\mu_{\scriptscriptstyle 2}+} \ldots (n) \; , \end{array}$$

the summation extending over all partitions  $(p_1^{\mu_1}, p_2^{\mu_2}, \cdots)$  of k with  $p_1\mu_1 + p_2\mu_2 + \cdots = k$ .

The expression (8) may be used to derive a central limit theorem for a random number of random variables. Suppose all moments of x and n exist with  $\kappa_1(x)=0$ . Suppose the distribution of n depends on a parameter N with  $\lim_{N\to\infty}\kappa_1(n)=\infty$ . Consider the standardized variate  $Z_n=S_n/(\operatorname{var} S_n)^{1/2}$ . We see that  $EZ_n=0$ ,  $\operatorname{var} Z_n=1$  and  $\kappa_k(Z_n)=\kappa_k(S_n)/(En\cdot\operatorname{var} x)^{k/2}$ . By inspection we see that if  $\kappa_k(n)/(E(n))^{k/2}\to 0$  as  $N\to\infty$  for  $k=3,4,\cdots$ , then  $\kappa_k(Z_n)\to 0$  as  $N\to\infty$  for  $k=3,4,\cdots$ . We see that  $Z_n$  is asymptotically standardized normal. Central limit theorems for random sums are considered in Robbins [5] and Wittenberg [7].

Robbins also considers an alternate form of standardization of  $S_n$ , namely  $Y_n = (S_n - ES_n)/n^{1/2}$ . Here we see from (1) and (2) that  $EY_n = 0$  and var  $Y_n = \text{var } x$ . From the corollary we have for k > 2,

(9) 
$$\kappa_{k}(Y_{n}) = \sum \frac{k!}{\mu_{1}! \ \mu_{2}! \cdots} \frac{1}{(p_{1}!)^{\mu_{1}} (p_{2}!)^{\mu_{2}} \cdots} \kappa_{p_{1}}^{\mu_{1}} \kappa_{p_{2}}^{\mu_{2}} \cdots \times \kappa_{\mu_{1}\mu_{2}} \dots (n^{-(p_{1}-2)/2}, \ n^{-(p_{2}-2)/2}, \cdots)$$

the summation extending over all partitions  $(p_1^{u_1}, p_2^{u_2}, \cdots)$  of k with  $p_1\mu_1 + p_2\mu_2 + \cdots = k$  and  $p_1, p_2, \cdots > 1$ .

Example 3. Two-stage sampling. Consider a sampling plan involving the selection of n first-stage units with or without replacement and with possibly unequal probabilities, followed by a second stage of sampling, carried out independently within the selected first-stage units, followed by the measurement of the k-variate random variable  $x(j) = (x_1(j), \dots, x_k(j))$  in the jth unit. Define indicator variables as follows;  $a_j = 1$  if the jth unit is in the sample and  $a_j = 0$  otherwise. Consider sample totals. We see that these have the form  $X_i = \sum_j a_j x_i(j)$ , i = 1,  $\dots$ , k where  $(a_1, a_2, \dots)$  is independent of the  $(x_1(j), \dots, x_k(j))$ ,  $j = 1, 2, \dots$ , which are independent of each other.

Letting A refer to the variate  $(a_1, a_2, \cdots)$  and  $X = (X_1, \cdots, X_k)$ , we have from the theorem

(10) 
$$\kappa(X) = \sum_{\alpha} \kappa_{\alpha} \{\kappa(X_{\alpha_1}|A), \dots, \kappa(X_{\alpha_p}|A)\},$$

the summation extending over all partitions  $\alpha = (\alpha_1, \dots, \alpha_p), p = 1, \dots, k$  of the integers  $(1, \dots, k)$ .

Since the  $(x_i(j), \dots, x_k(j))$ ,  $j=1, 2, \dots$  are independent and  $a_j^m=a_j$ ,  $m=1, 2, \dots$ 

(11) 
$$\kappa(X_{\beta}|A) = \sum_{i} a_{j}\kappa(x_{\beta}; j)$$

where  $\kappa(x_{\beta}; j) = \kappa(x_{i_1}(j), \dots, x_{i_I}(j))$  if  $\beta = (i_1, \dots, i_I)$ . We have therefore

(12) 
$$\kappa(X) = \sum_{\alpha} \sum_{j_1} \cdots \sum_{j_p} \kappa(x_{\alpha_1}; j_1) \cdots \kappa(x_{\alpha_p}; j_p) \kappa(a_{j_1}, \cdots, a_{j_p}) .$$

We note that the cumulants of the variate  $(a_1, a_2, \cdots)$  are needed and that these depend solely on the form of sampling employed in the selection of the first-stage units. We see that in order to obtain an unbiased estimate of  $\kappa(X)$ , we require unbiased estimates of the products of the cumulants of the x(j). If the first-stage units are infinite in size and one employs simple random sampling within them, these estimates have been provided in Dressel [1] and Tukey [6].

After this note had been prepared, the author learned that D. S. Robson of Cornell University had previously obtained the result contained in the corollary. Ebner [2] employed it in an investigation of the balanced one-way nested classification and work has continued at Cornell on its use in sampling from finite populations.

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