

THE DISTRIBUTION OF LINEAR COMBINATIONS OF ORDER STATISTICS FROM THE UNIFORM DISTRIBUTION

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1. Introduction. In this paper we derive an algorithm for computing the distribution function of an arbitrary linear combination of order statistics from a uniform distribution. Suppose $U_{(i)}$ is the i th smallest observation from a sample of size n from the uniform distribution on $[0, 1]$, with the convention $U_0 \equiv 0, U_{n+1} \equiv 1$. Consider a set of S integers $\{k_i\}$ such that

$$(1.1) \quad k_0 = 0 < k_1 < k_2 < \cdots < k_S \leq n.$$

For any set of constants $d_i > 0$ and any x , we seek

$$P\{\sum_{s=1}^S d_s U_{(k_s)} \leq x\}.$$

Our approach is to generalize a formula derived by Dempster and Kleyle (1968).

2. Derivation of the algorithm. Let $X_i = U_{(i)} - U_{(i-1)}, i = 1, 2, \dots, n$. Let $c_{n+1} = 0$.

Define c_1, c_2, \dots, c_n by

$$\begin{aligned} c_{k_i} &= c_{k_{i+1}} + d_i && \text{for } i = 1, \dots, S \\ c_j &= c_{j+1} && \text{for } j \notin \{k_1, \dots, k_S\}. \end{aligned}$$

Then we have

$$(2.1) \quad \sum_{s=1}^S d_s U_{(k_s)} = \sum_{i=1}^n c_i X_i.$$

For the special case $S = n$, Dempster and Kleyle (1968) have shown that

$$(2.2) \quad P\left\{\sum_{i=1}^n c_i X_i \leq x\right\} = 1 - \sum_{j=1}^r \frac{(c_j - x)^n}{c_j \prod_{i \neq j} (c_j - c_i)}$$

for $0 \leq x \leq c_1$, where r is the largest positive integer such that $x \leq c_r$. In the general case $S \leq n$, we wish to allow

$$c_{k_{s-1}+1} = c_{k_{s-1}+2} = \cdots = c_{k_s} = c_{(s)},$$

for $s = 1, 2, \dots, S$.

Let $k_s - k_{s-1} = r_s, s = 1, \dots, S$; and $n - k_S = r_{S+1}$. Then we wish to let the first r_1 c_1 's take the value $c_{(1)}$, the next r_2 take the value $c_{(2)}$, etc. Let $c_{(s+1)} = 0$. In this situation (2.2) is not applicable unless $r_s = 1$ for all s .

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Suppose, however, that we define

$$(2.3) \quad \begin{aligned} b_{(k_{s-1}+i)}(h) &= c_{(s)} + (r_s - i)h, & \text{for } h > 0, i = 1, \dots, r_s; s = 1 \dots S. \\ b_{(k_s+i)}(h) &= (r_{S+1} + 1 - i)h & i = 1, 2, \dots, r_{S+1}. \end{aligned}$$

Then we have

LEMMA 1. $\lim_{h \rightarrow 0} P\{\sum_{i=1}^n b_i(h)X_i \leq x\} = P\{\sum_{s=1}^S d_s U_{(k_s)} \leq x\}.$

PROOF. Let $A_r = \{\sum_{i=1}^n b_i(1/r)X_i \leq x\}$

$$A = \{\sum_{s=1}^n d_s U_{(k_s)} \leq x\} = \{\sum_{i=1}^{k_1} c_{(1)}X_i + \sum_{i=k_1+1}^{k_2} c_{(2)}X_i + \dots \leq x\}.$$

Now $\left\{ \sum_{i=1}^n b_i\left(\frac{1}{r}\right)X_i \leq x \right\} \Rightarrow \left\{ \sum_{i=1}^n b_i\left(\frac{1}{r+1}\right)X_i \leq x \right\}$, so that

$A_r \subset A_{r+1}$ and $A = \bigcup_{r=1}^{\infty} A_r$. Therefore

$$P(A) = P\left(\bigcap_{r=1}^{\infty} A_r\right) = \lim_{r \rightarrow \infty} P\{A_r\}.$$

Suppressing for convenience the dependence of b_i on h we have from (2.2) that

$$\begin{aligned} P\left\{ \sum_{i=1}^n b_i X_i \leq x \right\} &= 1 - \sum_{j=1}^{k_1} \frac{(b_j - x)^n}{b_j \prod_{i \neq j} (b_j - b_i)} - \sum_{j=k_1+1}^{k_2} \frac{(b_j - x)^n}{b_j \prod_{i \neq j} (b_j - b_i)} \\ &\quad - \dots - \sum_{j=k_{m-1}+1}^{k_m} \frac{(b_j - x)^n}{b_j \prod_{i \neq j} (b_j - b_i)} \end{aligned}$$

where m is the largest integer such that $x \leq c_{(m)}$. Let

$$T_s = \sum_{j=k_{s-1}+1}^{k_s} \frac{(b_j - x)^n}{b_j \prod_{i \neq j} (b_j - b_i)}$$

so that

$$(2.4) \quad P\{\sum_{i=1}^n b_i X_i \leq x\} = 1 - \sum_{s=1}^m T_s.$$

LEMMA 2.

$$T_s = \frac{\Delta^{r_s-1} f_s(c_{(s)})}{h^{r_s-1} (r_s - 1)!}$$

where

$$f_s(c) = \frac{(c - x)^n}{c \prod_{i \leq k_{s-1}, i > k_s} (c - b_i)},$$

and Δ is the forward difference operator defined by

$$\Delta^k f(x) = \Delta^{k-1} f(x+h) - \Delta^{k-1} f(x), \quad k = 1, 2, \dots.$$

PROOF.

$$\begin{aligned} T_s &= \sum_{j=k_{s-1}+1}^{k_s} \frac{(b_j-x)^n}{b_j \prod_{i \neq j} (b_j - b_i)} \\ &= \sum_{j=k_{s-1}+1}^{k_s} \frac{(b_j-x)^n}{b_j \prod_{i \leq k_{s-1}, i > k_s} (b_j - b_i) \prod_{k_{s-1} < i < j \leq k_s, k_{s-1} < j < i \leq k_s} (b_j - b_i)} \\ &= \sum_{j=k_{s-1}+1}^{k_s} \frac{f_s(b_j)}{\prod_{k_{s-1} < i < j \leq k_s, k_{s-1} < j < i \leq k_s} (b_j - b_i)} \\ &= \sum_{\rho=1}^{r_s} \frac{f_s(c_{(s)} + (r_s - \rho)h)}{h^{r_s-1} (r_s - \rho)! (\rho - 1)!} (-1)^{\rho-1}. \end{aligned}$$

Making the transformation $j' = r_s - \rho$, this can be written

$$\sum_{j'=0}^{r_s-1} \frac{f_s(c_{(s)} + j'h)}{h^{r_s-1} (r_s - 1)!} (r_s - j') (-1)^{r_s-1-j'},$$

which is equivalent to (see, for example [3] page 46)

$$\frac{\Delta^{r_s-1} f_s(c_{(s)})}{h^{r_s-1} (r_s - 1)!}.$$

We are now ready to prove the main result.

THEOREM.

$$P \left\{ \sum_{s=1}^S d_s U_{(k_s)} \leq x \right\} = 1 - \sum_{s=1}^m \frac{g_s^{(r_s-1)}(c_{(s)})}{(r_s - 1)!}$$

where m is the largest integer such that $x \leq c_{(m)}$ and

$$g_s(c) = \frac{(c-x)^n}{c \prod_{\mu \neq s} (c - c_{(\mu)})^r}.$$

PROOF. It is clear that for any function f whose k th derivative exists at x ,

$$\lim_{h \rightarrow 0} \frac{\Delta^k f(x)}{h^k} = f^{(k)}(x).$$

Note also that

$$\lim_{h \rightarrow 0} b_{k_\mu+i} = c_{(\mu)} \quad \text{for } i = 1, 2, \dots, r_\mu.$$

It follows from Lemma 2 that

$$\lim_{h \rightarrow 0} T_s = \frac{g_s^{(r_s-1)}(c)}{(r_s - 1)!}$$

evaluated at $c_{(s)}$. The theorem then follows from Lemma 1 and (2.4).

To make use of this formula in practice we must be able to evaluate the high order derivatives of the functions g_s . We can write

$$\log g_s(c) = n \log(c-x) - \log c - \sum_{\mu \neq s} r_\mu \log(c - c_{(\mu)}).$$

Differentiating both sides we obtain

$$(2.5) \quad g'_s(c) = g_s(c)h(c)$$

where

$$(2.6) \quad h(c) = \frac{n}{c-x} - \frac{1}{c} - \sum_{\mu \neq s} \frac{r_\mu}{c - c_{(\mu)}}.$$

Using Leibniz's rule for the k th derivative of a product, we obtain the recurrence relation:

$$(2.7) \quad \begin{aligned} g_s^{(k)}(c) &= \frac{d^{k-1}}{dc^{k-1}} g'_s(c) = \frac{d^{k-1}}{dc^{k-1}} (g_s(c)h(c)) \\ &= \sum_{i=0}^{k-1} \binom{k-1}{i} g_s^{(i)}(c)h^{(k-1-i)}(c). \end{aligned}$$

We also have from (2.6)

$$h^{(i)}(c) = (-1)^i i! \left[\frac{n}{(c-x)^{i+1}} - \frac{1}{c^{i+1}} - \sum_{\mu \neq s} \frac{r_\mu}{(c - c_{(\mu)})^{i+1}} \right].$$

Thus (2.7) can be used recursively to obtain $g_s^{(k)}(c)$ for any k .

Note that although we have been assuming $d_i > 0$ for all i , the general problem can be handled by reordering and shifting variables, making use of the symmetry in the situation. For example $2U_{(3)} - U_{(1)} = X_1 + 2X_2 + 2X_3$ has the same distribution as $2X_1 + 2X_2 + X_3 = U_{(2)} + U_{(3)}$.

3. Application. Following the notation of Wilks (1962) we define the $(k-1)$ -variate Dirichlet distribution $D(v_1, v_2, \dots, v_{k-1}; v_k)$ by the density

$$\begin{aligned} f(x_1, \dots, x_{k-1}) &= \frac{\Gamma(v_1 + v_2 + \dots + v_k)^{k-1}}{\prod_{i=1}^k \Gamma(v_i)} \prod_{i=1}^{k-1} x_i^{v_i-1} \left(1 - \sum_1^{k-1} x_i\right)^{v_k-1} \\ &\quad \text{for } x_i \geq 0, i = 1, \dots, k \quad \sum_1^{k-1} x_i \leq 1 \\ &= 0 \quad \text{otherwise.} \end{aligned}$$

It is easily shown that the joint distribution of $U_{(k_1)}, U_{(k_1+k_2)} - U_{(k_1)}, \dots, U_{(k_1+\dots+k_s)} - U_{(k_1+\dots+k_{s-1})}$, for k_i 's as in (1.1), is $D(k_1, k_2, \dots, k_s; n - \sum_1^s k_i + 1)$.

Let $p_1, p_2, \dots, p_{k-1}, p_k$ represent the cell probabilities for a multinomial population with k categories. For a Bayesian analysis it is common to assume a conjugate prior of the form $D(\eta_1, \eta_2, \dots; \eta_k)$ for p_1, \dots, p_{k-1} . Suppose $\eta_1, \eta_2, \dots, \eta_k$ are integers. Let $n_i, i = 1, \dots, k$, be the observed frequency for the i th category and $n = \sum_{i=1}^k n_i$. Then the posterior distribution of p_1, \dots, p_{k-1} is $D(n_1 + \eta_1, \dots; n_k + \eta_k)$.

Suppose we wish to make posterior probability statements about events of the form $\{\sum_1^k a_i p_i \leq x\}$ for real numbers a_1, \dots, a_k and x . Let $v_j = \sum_{i=1}^j (n_i + \eta_i)$, $j = 1, \dots, k$. Then we have

$$(3.1) \quad P\{\sum_1^k a_i p_i \leq x\} = P\{\sum_1^k a_i [U_{(v_i)} - U_{(v_{i-1})}] \leq x\} \\ = P\{a_k + \sum_1^{k-1} (a_i - a_{i+1}) U_{(v_i)} \leq x\},$$

where $U_{(j)}$ is the j th smallest observation from a sample of size $(v_k - 1)$ from the uniform distribution on $[0, 1]$. Thus the algorithm of Section 2 can be applied.

For example, suppose we have $k = 5$, and we assume the improper prior $D(0, 0, 0, 0; 0)$ suggested by Lindley (1964) for (p_1, p_2, p_3, p_4) . Suppose also that

$$\begin{array}{ll} a_1 = -5 & n_1 = 10 \\ a_2 = -2 & n_2 = 15 \\ a_3 = 0 & n_3 = 10 \\ a_4 = +2 & n_4 = 10 \\ a_5 = +5 & n_5 = 6. \end{array}$$

From (3.1)

$$P\{\sum_1^5 a_i p_i \leq x\} = P\{5 - 3U_{(10)} - 2U_{(25)} - 2U_{(35)} - 3U_{(45)} \leq x\} \\ = P\{3U_{(10)} + 2U_{(25)} + 2U_{(35)} + 3U_{(45)} \geq 5 - x\},$$

where the order statistics are from a sample of size 50.

A computer program to implement the algorithm of Section 2 has been successfully run and used to obtain the following results for this example.

x	$P\{\sum_1^5 a_i p_i \leq x\}$
1.0	.9998
.8	.9992
.6	.9967
.4	.9885
.2	.9660
0	.9150
-.2	.8196
-.4	.6738
-.6	.4929
-.8	.3119
-1.0	.1669
-1.2	.0741
-1.4	.0269
-1.6	.0079
-1.8	.0018
-2.0	.0003.

REFERENCES

- [1] DEMPSTER, A. P. and KLEYLE, R. M. (1968). Distributions determined by cutting a simplex with hyperplanes. *Ann. Math. Statist.* **39** 1473-78.
- [2] LINDLEY, D. V. (1964). Bayesian analysis of contingency tables. *Ann. Math. Statist.* **35** 1622-43
- [3] RALSTON, A. R. (1965). *A First Course in Numerical Analysis*. McGraw Hill, New York.
- [4] WILKS, S. S. (1962). *Mathematical Statistics*. Wiley, New York.