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SOME MULTIVARIATE DISTRIBUTIONS ARISING IN FAULTY
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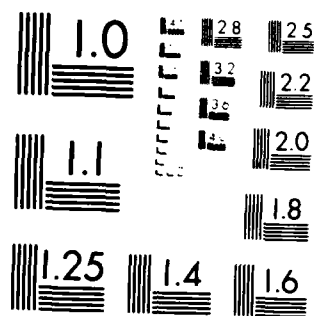
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SOME MULTIVARIATE DISTRIBUTIONS ARISING
IN FAULTY SAMPLING INSPECTION

Norman L. Johnson

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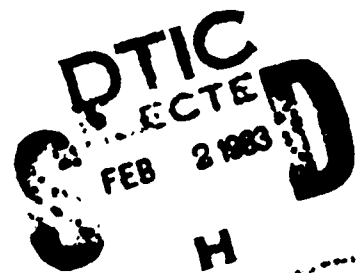
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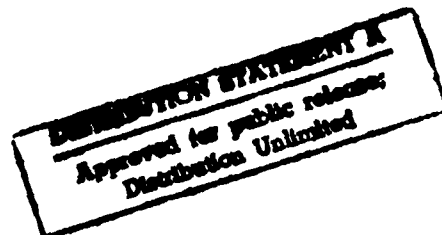
University of Maryland
at College Park

An Invited Paper for the Sixth International
Symposium on Multivariate Analysis

University of Pittsburgh, PA
July 1983



January 1983



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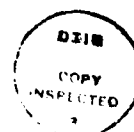
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ABSTRACT

In the present paper we extend in two ways some results presented in Kotz and Johnson (Comm. in Statistics (1982), A11) relating to the study of distributional aspects of effects of errors in inspection sampling:

(1) Multistage sampling with k successive samples involving the possibility of two types of errors in inspection (classifying a defective individual as non-defective, or a non-defective as defective); (2) Single-stage sampling considering several types of defects of which only one is tested on inspection. Both (1) and (2) lead to novel multivariate distributions. Their structural properties are analysed in some detail and some applications, in particular those in quality control are discussed.

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1. INTRODUCTION

We have recently studied effects of inaccuracies in inspection on the properties of acceptance sampling procedures (Johnson & Kotz (1981), Kotz & Johnson (1982a,b)). In particular we have considered two-stage sampling, wherein up to two successive samples of sizes n_1, n_2 may be taken from a lot of size N , containing D defective items. (The second sample is taken only if the number (Z_1) of items found to be defective is within certain limits defined by the sampling schemes.) It was supposed that inspection is not perfect, so that some defective items may not be noticed as such, while some nondefective items may be classified as 'defective'.

Similar investigations have been described in quality control literature by Hoag et al. (1975) Dorris and Foote (1978) and Rahali and Foote (1982) (see also Armstrong (1982)). In the present paper we shall extend this work along two directions, each of which introduces apparently novel multivariate distributions.

First we shall generalize our result to multistage sampling with k successive samples of sizes $n_1, n_2, \dots, n_k$ and suppose that for the i -th sample

p_i = probability that a defective item is correctly classified

p_i' = probability that a nondefective item is classified as 'defective'.

Secondly, returning to single-stage sampling we shall consider the case of two types of defects one of which is relatively easy to detect, and investigate the distribution of the second kind of defect when selection is on

the basis of the first kind. As before we will suppose that p_1 denotes the probability of correct classification of a defective item and p_1' the probability of classification of a non-defective item as 'defective'.

2. MULTISTAGE SAMPLING ANALYSIS

We will be interested in the joint distribution of $Z_1, Z_2, \dots, Z_k$, the numbers of items classified as defective as a result of inspection of the sequence of k samples. This distribution seems to be new, and presents some features of theoretical interest, as well as having the possible practical applications we have indicated. Chief among the latter is the possible calculation of acceptance probabilities for multi-stage (or even sequential) sampling schemes under imperfect inspection. Some practical comments, in the two-stage case are given in Kotz & Johnson (1982b).

Let $Y_1, Y_2, \dots, Y_k$ denote the actual numbers of defectives in the 1st, 2nd, ..., k -th samples respectively. The joint distribution of $\underline{Y} = (Y_1, \dots, Y_k)$ is a multivariate hypergeometric with

$$\Pr\left[\prod_{i=1}^k (Y_i = y_i)\right] = \binom{N}{D}^{-1} \left\{ \prod_{i=1}^k \binom{n_i}{y_i} \right\} \frac{\binom{N-Dn_i}{D-\sum y_i}}{\binom{N-Dn_i}{D-\sum y_i}} \quad (1)$$

$$(0 \leq y_i \leq n_i (i=1, \dots, k); D-N + \sum_{i=1}^k n_i \leq \sum_{i=1}^k y_i).$$

Symbolically, we write

$$\underline{Y} \sim \text{Mult.Hyp}_{\underline{n}}(\underline{n}; D, N).$$

We note that

$$E\left[\prod_{i=1}^k \binom{r_i}{Y_i} \binom{s_i}{n_i - Y_i}\right] = D \binom{r_i}{N-D} \binom{s_i}{N-D} \left\{ \prod_{i=1}^k \binom{r_i + s_i}{n_i} \right\} / N \quad (2)$$

where $a^{(b)} = a(a-1)\dots(a-b+1)$ is the b -th descending factorial of a .

Conditionally on $\underline{Y}$, the Z 's are mutually independent, with Z_i distributed as

the sum of two independent binomial variables with parameters (Y_i, p_i) and $(n_i - Y_i, p'_i)$ respectively.

$$\begin{aligned} \Pr\left[\bigcap_{i=1}^k (Z_i = z_i) \mid \underline{Y} = \underline{y}\right] &= \prod_{i=1}^k \left[\sum_{h=0}^{z_i} \binom{y_i}{h} \binom{n_i - y_i}{z_i - h} p_i^h p_i'^{z_i - h} (1 - p_i)^{y_i - h} (1 - p_i')^{n_i - y_i - z_i + h} \right] \\ &= \prod_{i=1}^k b(z_i; y_i, p_i; n_i - y_i, p_i') \end{aligned} \quad (3)$$

where $b(z_i; y_i, p_i; n_i - y_i, p_i')$ is the probability function for the convolution of the two binomial distributions with parameters (y_i, p_i) and $(n_i - y_i, p_i')$ respectively.

The unconditional distribution of $\underline{Z}$ is

$$\Pr[\underline{Z} = \underline{z}] = \binom{N}{D}^{-1} \sum_{y_1} \dots \sum_{y_k} \left\{ \binom{N - n_i}{D - y_i} \prod_{i=1}^k \binom{n_i}{y_i} \prod_{i=1}^k b(z_i; y_i, p_i; n_i - y_i, p_i') \right\} \quad (4)$$

The limits for $\underline{y}$ are as in (1).

Symbolically

$$\underline{Z} \sim \left[\begin{array}{c} \text{Bin}(Y_1, p_1) * \text{Bin}(n_1 - Y_1, p_1') \\ \vdots \\ \text{Bin}(Y_k, p_k) * \text{Bin}(n_k - Y_k, p_k') \end{array} \right] \underset{\underline{Y}}{\sim} \text{Mult. Hypg}_k (n; D, \underline{p}) \quad (4a)$$

where $*$ stands for convolution and $\underset{\underline{Y}}{\sim}$ is the compounding operator (as defined, for example, in Johnson & Kotz (1969, p. 184)).

This distribution might be called "Multivariate Hypergeometric-Convolved Binomial(s)".

The conditional (on $\underline{Y}$) r_i -th factorial moment of Z_i is

$$E[Z_i^{(r_i)} \mid \underline{Y}] = \sum_{h=0}^{r_i} \binom{r_i}{h} p_i^h p_i'^{r_i - h} y_i^{(h)} (n_i - y_i)^{(r_i - h)}.$$

The unconditional $r = (r_1, \dots, r_k)$ -th factorial moment of Z is

$$\begin{aligned} \mu_r(Z) &= E\left[\prod_{i=1}^k Z_i^{(r_i)}\right] = E_Y E\left[\prod_{i=1}^k Z_i^{(r_i)} | Y\right] = E_Y \left[\prod_{i=1}^k E[Z_i^{(r_i)} | Y]\right] \\ &= E\left[\prod_{i=1}^k \sum_{h=0}^{r_i} \binom{r_i}{h} p_i^h p_i'^{r_i-h} Y_i^{(h)} (n_i - Y_i)^{(r_i-h)}\right] \quad (6) \\ &= \sum_{r_1=0}^{r_1} \dots \sum_{r_k=0}^{r_k} \prod_{i=1}^k \binom{r_i}{h_i} p_i^{h_i} p_i'^{r_i-h_i} E[Y_i^{(h_i)} (n_i - Y_i)^{(r_i-h_i)}] \\ &= \frac{\prod_{i=1}^k n_i^{(r_i)}}{N^{(\sum_{i=1}^k r_i)}} \sum_{r_1=0}^{r_1} \dots \sum_{r_k=0}^{r_k} \prod_{i=1}^k \binom{r_i}{h_i} p_i^{h_i} p_i'^{r_i-h_i} D^{(h_i)} (N-D)^{(r_i-h_i)} g_i \\ &= \frac{\prod_{i=1}^k n_i^{(r_i)}}{N^{(\sum_{i=1}^k r_i)}} \sum_{r_1=0}^{r_1} \dots \sum_{r_k=0}^{r_k} \prod_{i=1}^k D^{(h_i)} (N-D)^{(r_i-h_i)} g_i \quad (6') \end{aligned}$$

where g_i is the coefficient of x^i in $\prod_{i=1}^k (n_i + p_i x)^{r_i}$. (Note that

(i) if sampling were with replacement, (6) would be

$$\mu_r(Z) = \frac{\prod_{i=1}^k n_i^{(r_i)}}{N^{(\sum_{i=1}^k r_i)}} \left[\prod_{i=1}^k D p_i + (N-D) p_i' \right]^{r_i} / N^{r_i};$$

(ii) if $p_i = p$ and $p_i' = p'$ for all i -corresponding to constant quality of inspection throughout - then

$$g_i = \binom{r_i}{h_i} p^h p'^{r_i-h}$$

In particular

$$E[Z_i] = n_i [DN^{-1} p_i + (1-DN^{-1}) p_i'] = n_i \bar{p}_i \quad (7.1)$$

where $\bar{p}_i = DN^{-1} p_i + (1-DN^{-1}) p_i'$ is the probability that an individual chosen at random in the i -th sample will be classified as defective (whether it really is so, or not). Also

$$\text{var}(Z_i) = n_i \bar{p}_i (1 - \bar{p}_i) - n_i (n_i - 1) (N - 1)^{-1} D N^{-1} (1 - D N^{-1}) (p_i - p_i')^2 \quad (7.2)$$

and

$$\text{cov}(Z_i, Z_j) = -n_i n_j (N - 1)^{-1} D N^{-1} (1 - D N^{-1}) (p_i - p_i') (p_j - p_j') \quad (8)$$

If $p_i > p_i'$ and $p_j > p_j'$ as one would hope, if inspection is to be of any use at all, the covariance is negative, as might be expected, since the covariance of Y_i and Y_j is negative. If $p_i = p_i'$ or $p_j = p_j'$, so that it is irrelevant whether an item in the corresponding sample(s) is defective or not, then the covariance is zero - in fact the corresponding Z (or Z 's) is independent of all other Z 's.

It is easy to write down the conditional distribution of Z_i , given Z_j , but it is rather complicated in form. We can, however, derive the regression function in the following way.

We have
$$E[Z_i | Y_i] = Y_i p_i + (n_i - Y_i) p_i'$$

so
$$E[Z_i | Z_j] = p_i E[Y_i | Z_j] + p_i' (n_i - E[Y_i | Z_j]) \quad (9)$$

Now
$$E[Y_i | Y_j] = n_i (D - Y_j) / (N - n_j)$$

so
$$E[Y_i | Z_j] = n_i (N - n_j)^{-1} (D - E[Y_j | Z_j]) \quad (10)$$

In order to evaluate $E[Y_j | Z_j]$ note that, for the j -th sample,

$$\text{Pr}[\text{item is defective} | \text{item classified as defective}] = D N^{-1} p_j / \bar{p}_j$$

and

$$\text{Pr}[\text{item is defective} | \text{item classified as nondefective}] = D N^{-1} (1 - p_j) / (1 - \bar{p}_j)$$

so
$$E[Y_j | Z_j] = D N^{-1} [Z_j (p_j / \bar{p}_j) + (n_j - Z_j) \{(1 - p_j) / (1 - \bar{p}_j)\}] \quad (11)$$

From (9), (10) and (11) we obtain

$$\begin{aligned}
 E[Z_i|Z_j] &= n_i p_i' + (p_i - p_i') E[Y_i|Z_j] \\
 &= n_i p_i' + n_i (p_i - p_i') (N - n_j)^{-1} (D - E[Y_j|Z_j]) \\
 &= n_i p_i' + n_i (p_i - p_i') D N^{-1} \left(1 - \frac{(1 - D N^{-1})(n_j - p_j')}{(N - n_j) \bar{p}_j (1 - \bar{p}_j)} \cdot (Z_j - n_j \bar{p}_j) \right) \quad (12)
 \end{aligned}$$

The regression is linear, and the sign of the regression coefficient is opposite to that of $(p_i - p_i')(p_j - p_j')$ agreeing with the sign of $\text{cov}(Z_i, Z_j)$ in (8).

3. MULTISTAGE SAMPLING: A SPECIAL CASE

If we take $n_1 = n_2 = \dots = n_k = 1$ we have the first k stages of a fully sequential sampling procedure. In this situation the only possible values of each of the Y 's and Z 's are 0 and 1. Formula (3) becomes

$$\Pr[\underline{Z} = \underline{z}] = \binom{N}{D}^{-1} \sum_{y_1} \dots \sum_{y_k} \binom{N-k}{D-y} \prod_{i=1}^k y_i p_i^{z_i} (1-p_i)^{1-z_i} + (1-y_i) p_i^{z_i} (1-p_i)^{1-z_i} \quad (13)$$

Collecting together terms with the same value of $y_i = y$ (corresponding to the total number of defective items selected) we get

$$\Pr[\underline{Z} = \underline{z}] = \binom{N}{D}^{-1} \sum_{y=0}^k \binom{N-k}{D-y} \sum_{\underline{z}} \prod_{i=1}^k p_i^{z_i} (1-p_i)^{1-z_i} \quad (13)'$$

where summation with respect to $\underline{z} = (z_1, \dots, z_k)$ and $\underline{z}' = (z_1', \dots, z_k')$ is constrained by

$$\sum_{i=1}^k (z_i + z_i') = y, \text{ and } \sum_{i=1}^k (z_i + z_i'') = \sum_{i=1}^k z_i = z$$

(corresponding to the total number of items classified as defectives) and for each i , one of (z_i, z_i', z_i'') is 1, the other three are each zero.

Yet another way of writing (13) is

$$\Pr[Z=z] = \binom{N}{D}^{-1} \sum_y \binom{N-k}{D-y} \cdot [\text{coefficient of } x^z u^y \text{ in} \\ \prod_{j=1}^k (p_j x u + p'_j x + (1-p_j) u + 1 - p'_j)] \quad (13)''$$

It may be noted that this shows that $\sum_{i=1}^k Z_i$ is a sufficient statistic (for D, supposing N, p's and p's are known). If $p_i = p$ and $p'_i = p'$ for all i, we have the coefficient of $x^z u^y$ in $(pxu + p'x + (1-p)u + (1-p'))^k$ on the right hand side of (13)''.

4. ASSOCIATED DEFECTS: ANALYSIS

We now suppose that there are two types of defects - (1) and (2) - and that in a population of size N there are D_{gh} individuals with g type (1), and h type (2) defects ($g, h = 0, 1$). (Of course $D_{00} + D_{01} + D_{10} + D_{11} = N$.) For example (1) might represent surface irregularity, with (2) corresponding to internal flaws in material, such as metal bars or plates. In many situations the different types of defect correspond to different modes of failure.

A random sample of size n is taken (without replacement) from the population; and each of the chosen individuals is examined for presence of type (1) defect. We are interested in the number (Z_1) of individuals classified as possessing defect (1); and among these Z_1 individuals, the number (Z_1^*) actually possessing defect (1) and the number (Z_2^*) possessing defect (2).

The distribution of Z_1 is given in Section 2, and also in Kotz & Johnson (1982b). The distribution of Z_1^* is the hypergeometric-binomial

$$\text{Binomial}(Y, p_1) \wedge \text{Hypg.}(n; D_{10} + D_{11}; N) \quad (14)$$

The probability that the sample will contain y_{gh} individuals with g type (1) defects and h type (2) defects is

$$\Pr\left[\sum_{g=0}^1 \sum_{h=0}^1 (Y_{gh} = y_{gh})\right] = \left[\sum_{g=0}^1 \sum_{h=0}^1 \binom{D_{gh}}{y_{gh}} \right] / \binom{N}{n} \quad (y_{00} + y_{01} + y_{10} + y_{11} = n) \quad (15)$$

(A multivariate hypergeometric $(n; D, N)$ distribution -cf (4a) where $\underline{n}$ is a vector but D is not.) For the $\underline{y} = (y_{00}, y_{01}, y_{10}, y_{11})$ -th joint factorial moment of $\underline{Y}$ we have the expression

$$\mu_{(\underline{y})}(\underline{Y}) = n^{(\underline{y})} \left[\sum_{g=0}^1 \sum_{h=0}^1 \binom{D_{gh}}{y_{gh}} \right] / N^{(\underline{y})} \quad \text{where } \underline{y} = (y_{00}, y_{01}, y_{10}, y_{11}) \quad (16)$$

Given $\underline{y} = (y_{00}, y_{01}, y_{10}, y_{11})$, Z_1 is distributed as the sum of two independent binomial variables, with parameters $(y_{10} + y_{11}, p_1)$ and $(y_{00} + y_{01}, p'_1)$ respectively. The first of these variables is, in fact Z_1^* . Similarly, Z_2^* is distributed as the sum of two independent binomial variables with parameters (y_{11}, p_1) and (y_{01}, p'_1) respectively. Introducing four independent binomial variables W_{gh} with parameters $(y_{gh}, s_g p_1 + (1-s_g) p'_1)$ for $g, h=0,1$ with $s_g = \frac{1}{2}[1 - (-1)^g]$, then conditionally on $\underline{Y} = \underline{y}$

$$Z_1 = W_{00} + W_{01} + W_{10} + W_{11} \quad (17.1)$$

$$Z_1^* = W_{10} + W_{11} \quad (17.2)$$

$$Z_2^* = W_{01} + W_{11} \quad (17.3)$$

We have

$$E[Z_1 | \underline{y}] = (y_{10} + y_{11}) p_1 + (y_{00} + y_{01}) p'_1; \text{var}(Z_1 | \underline{y}) = (y_{10} + y_{11}) p_1 (1-p_1) + (y_{00} + y_{01}) p'_1 (1-p'_1) \quad (18.1)$$

$$E[Z_2^* | \underline{y}] = y_{11} p_1 + y_{01} p'_1; \text{var}(Z_2^* | \underline{y}) = y_{11} p_1 (1-p_1) + y_{01} p'_1 (1-p'_1) \quad (18.2)$$

$$\text{cov}(Z_1, Z_2^* | \underline{y}) = \text{var}(W_{11} | \underline{y}) = \text{var}(Z_2^* | \underline{y}) \quad (18.3)$$

$$\text{corr}(Z_1, Z_2^* | \underline{y}) = \{\text{var}(Z_2^* | \underline{y}) / \text{var}(Z_1 | \underline{y})\}^{1/2} \quad (18.4)$$

$$\text{cov}(Z_1^*, Z_2^* | \underline{y}) = \text{var}(W_{11} | \underline{y}) = y_{11} p_1 (1-p_1) \quad (18.5)$$

$$\text{corr}(Z_1^*, Z_2^* | y) = \left[\frac{y_{11}^2 p_1 (1-p_1)}{(y_{10}+y_{11}) \cdot y_{11} p_1 (1-p_1) + y_{01} p_1' (1-p_1')} \right]^{-1/2} \quad (18.6)$$

The distribution of Z_2^* is the multivariate hypergeometric-convoluted binomial

$$\{ \text{Binomial}(Y_{11}, p_1) * \text{Binomial}(Y_{10}, p_1') \} \bigwedge_{Y_{11}, Y_{10}} \text{Mult. Hypg}(n; D_{11}, D_{10}; N) \quad (19)$$

Moments can be obtained similarly as in Section 2.

For an individual chosen at random from the population

$$\begin{aligned} \text{Pr}[\text{classified as having defect (1)}] &= N^{-1} \cdot (D_{01} + D_{00}) p_1' + (D_{10} + D_{11}) p_1 \\ &= N^{-1} (D_0 p_1' + D_1 p_1) = \bar{p}_1, \text{ say} \end{aligned} \quad (19.1)$$

($D_1 = D_{10} + D_{11}$ = total number of individuals in the population with defect (1);

$D_0 = N - D_1$)

and

$$\begin{aligned} \text{Pr}[\text{having defect (2) classified as having defect (1)}] &= \frac{N^{-1} (D_{01} p_1' + D_{11} p_1)}{\bar{p}_1} \\ &= \frac{\bar{p}_{2|1}}{\bar{p}_1}, \text{ say} \end{aligned} \quad (19.2)$$

$$\text{Hence } E[Z_2^* | Z_1] = Z_1 \bar{p}_{2|1} / \bar{p}_1 \quad (20.1)$$

and (from (7.1))

$$E[Z_2^*] = n \bar{p}_{2|1} \quad (20.2)$$

Formula (20.1) is also valid conditionally on $\underline{Y} = \underline{y}$, so we have

$$\text{cov}(Z_1, Z_2^*) = \frac{\bar{p}_{2|1}}{\bar{p}_1} E[Z_1^2] - n^2 \bar{p} \bar{p}_{2|1} \quad (21.1)$$

and from (7.2) and (7.1)

$$\text{cov}(Z_1, Z_2^*) = \frac{n\bar{p}_2|1}{\bar{p}_1} (\bar{p}_1(1-\bar{p}_1) - \frac{n-1}{N-1} \frac{D_1}{N} (1 - \frac{D_1}{N})(p_1-p_1')^2) \quad (21.2)$$

We must have $\text{cov}(Z_1, Z_2^*) > 0$ because from (20.1), $E[Z_2^* | Z_1]$ is an increasing function of Z_1 .

Extension to situations in which there are m (≥ 1) types of defects - (2), (3), ..., $(m+1)$ - in addition to the type (1) which is inspected directly, is straightforward. The only essentially new problems are the joint distribution of $Z_2^*, \dots, Z_{m+1}^*$ - the numbers of individuals with type (2), ..., $(m+1)$ defects among those classified as having defect (1); and also the distributions of variables like Z_{ij}^* , the number among these individuals, with both (i) & (j) type defects ($i \neq j \geq 2$). Using an obvious notation (with subscripts 0(1) indicating absence (presence) of the corresponding type of defect) we have, for example, corresponding to (19.2)

$$\Pr[\text{having defects (2) and (3), but not (4), ..., (m+1) classified as having defect (1)}] = \frac{D_{0110...0}p_1' + D_{1110...0}p_1}{D_{0...}p_1' + D_{1...}p_1} \quad (22)$$

where $D_{g...} = \sum_{a_2=0}^1 \dots \sum_{a_{m+1}=0}^1 D_g a_2 \dots a_{m+1}$ ($g = 0, 1$) ($D_{g...}$ is the quantity

previously represented by D_g .)

Considering, for simplicity, the case $m = 2$, we now obtain an expression for $\text{cov}(Z_2^*, Z_3^*)$. We have, analogously to (17.3)

$$\begin{aligned} Z_2^* &= W_{111} + W_{011} + W_{110} + W_{010} \\ Z_3^* &= W_{111} + W_{011} + W_{101} + W_{001} \end{aligned} \quad (23)$$

where the W 's are independent binomial variables and the parameters of W_{ghi} are $(y_{ghi}, \delta_g p_1 + (1-\delta_g)p_1')$.

Symbolically

$$\begin{matrix} Z_2^* \\ Z_3^* \end{matrix} \sim \begin{matrix} \text{Bin.}(Y_{111}, p_1) * \text{Bin.}(Y_{011}, p_1') \\ \text{Bin.}(Y_{101}, p_1) * \text{Bin.}(Y_{001}, p_1') \end{matrix} \quad \begin{matrix} \text{Bin.}(Y_{110}, p_1) * \text{Bin.}(Y_{010}, p_1') \\ \text{Bin.}(Y_{100}, p_1) * \text{Bin.}(Y_{000}, p_1') \end{matrix} \quad \text{Mult. Hypo}(n; D, \underline{y}) \quad (24)$$

(Here $\underline{y} = (Y_{111}, Y_{011}, Y_{110}, Y_{010}, Y_{101}, Y_{001})$.)

A general expression for the joint factorial moment $E[Z_2^{*(s_2)} Z_3^{*(s_3)}]$ can be obtained in the following way.

$$\begin{aligned} Z_2^{*(s_2)} Z_3^{*(s_3)} &= \sum_{i_1=0}^{(4)} \sum_{i_2=0}^{(4)} \sum_{i_3=0}^{(4)} \sum_{i_4=0}^{(4)} w_{111}^{(i_1)} w_{011}^{(i_2)} w_{110}^{(i_3)} w_{010}^{(i_4)} \sum_{j_1=0}^{(4)} \sum_{j_2=0}^{(4)} \sum_{j_3=0}^{(4)} \sum_{j_4=0}^{(4)} w_{111}^{(j_1)} w_{011}^{(j_2)} w_{101}^{(j_3)} w_{001}^{(j_4)} \\ &= \sum_{i_1=0}^{(4)} \sum_{i_2=0}^{(4)} \sum_{j_1=0}^{(4)} \sum_{j_2=0}^{(4)} \sum_{g=0}^{j_1} \sum_{h=0}^{j_2} \binom{j_1}{g} \binom{j_2}{h} \binom{j_1-g}{i_1} \binom{j_2-h}{i_2} w_{111}^{(i_1+g)} w_{011}^{(i_2+h)} \\ &\quad + w_{110}^{(i_3)} w_{010}^{(i_4)} w_{101}^{(j_3)} w_{001}^{(j_4)} \end{aligned} \quad (25)$$

where $\sum_{i_1=0}^{(4)} \sum_{i_2=0}^{(4)} \sum_{i_3=0}^{(4)} \sum_{i_4=0}^{(4)} = \sum_{i_1=0}^{(4)} \sum_{i_2=0}^{(4)} \sum_{i_3=0}^{(4)} \sum_{i_4=0}^{(4)}$ and utilizing the

$$\text{identity } a^{(u_1)} a^{(u_2)} = a^{(u_1)} a^{(u_2-u_1)} a^{(u_1+u_2-u_1)} = \sum_{u=0}^{u_2} \binom{u_2}{u} a^{(u_2-u)} a^{(u_1+u)}$$

An expression for $E[Z_2^{*(s_2)} Z_3^{*(s_3)} | \underline{y} = \underline{y}]$ is obtained by replacing $w_{ghi}^{(\cdot)}$ in (25)

by $y_{ghi}^{(\cdot)} \tilde{p}_g$ where $\tilde{p}_g = y_g p_1 + (1-y_g) p_1'$. Then taking expectations with respect to $\underline{y}$, we obtain

$$\begin{aligned} E[Z_2^{*(s_2)} Z_3^{*(s_3)}] &= \sum_{i_1=0}^{(4)} \sum_{i_2=0}^{(4)} \sum_{j_1=0}^{(4)} \sum_{j_2=0}^{(4)} \sum_{g=0}^{j_1} \sum_{h=0}^{j_2} \binom{j_1}{g} \binom{j_2}{h} \binom{j_1-g}{i_1} \binom{j_2-h}{i_2} p_1^{g+i_1+i_3+j_3} p_1'^{h+i_2+j_2+j_4} \\ &\quad \times \binom{j_1-g}{i_1} \binom{j_2-h}{i_2} \cdot \frac{n}{N} \frac{(s_2+s_3+g+h)}{(s_2+s_3+g+h)} D_{111}^{(i_1+g)} D_{011}^{(i_2+h)} D_{110}^{(i_3)} D_{010}^{(i_4)} D_{101}^{(j_3)} D_{001}^{(j_4)} \end{aligned} \quad (26)$$

If $m > 2$, D_{ghi} is replaced by $D_{ghi} \dots$; joint factorial moments of three or more Z_i^* 's ($i > 1$) can be obtained by similar techniques, though the formulas rapidly become more cumbersome.

For lower order moments, direct calculation is often simpler than using general formulas.

We now outline the calculation of $\text{cov}(Z_2^*, Z_3^*)$. We have from (23) and (24)

$$\text{cov}(Z_2^*, Z_3^* | y) = y_{111} p_1 (1-p_1) + y_{011} p_1' (1-p_1') \quad (27)$$

whence

$$E[Z_2^* Z_3^* | y] = y_{111} p_1 (1-p_1) + y_{011} p_1' (1-p_1') + (y_{111} + y_{110}) p_1 + (y_{011} + y_{010}) p_1'$$

$$+ (y_{111} + y_{101}) p_1 + (y_{011} + y_{001}) p_1'$$

so (using (15))

$$\text{cov}(Z_2^*, Z_3^*) = E[Z_2^* Z_3^*] - E[Z_2^*] E[Z_3^*]$$

$$\begin{aligned} &= \frac{n}{N} \{ D_{111} p_1 (1-p_1) + D_{011} p_1' (1-p_1') \} + \left\{ \frac{n}{N} - \frac{n(n-1)}{N(N-1)} \right\} (D_{111} p_1^2 + D_{011} p_1'^2) \\ &\quad + \left\{ \frac{n}{N} - \frac{n(n-1)}{N(N-1)} \right\} - \frac{n^2}{N^2} \{ (D_{111} + D_{110}) p_1 + (D_{011} + D_{010}) p_1' \} \{ (D_{111} + D_{101}) p_1 + (D_{011} + D_{001}) p_1' \} \\ &= \frac{n}{N} \{ D_{111} p_1 (1 - \frac{n-1}{N-1} p_1) + D_{011} p_1' (1 - \frac{n-1}{N-1} p_1') \} \quad (28) \\ &\quad - \frac{n(N-n)}{N^2(N-1)} \{ (D_{111} + D_{110}) p_1 + (D_{011} + D_{010}) p_1' \} \{ (D_{111} + D_{101}) p_1 + (D_{011} + D_{001}) p_1' \} \end{aligned}$$

This covariance can be positive or negative. Both Z_2^* and Z_3^* are positively correlated with Z_1 (the total number of individuals, classified as 'defective', of which Z_2^* and Z_3^* are subsets), but they may be negatively correlated in the population. The latter situation corresponds to values of D_{g10} and D_{g01} which are large relative to D_{g11} . ($g = 0, 1$).

When N is large compared with n ,

$$\text{cov}(Z_2^*, Z_3^*) \doteq n[\bar{p}_{1|11} - (\bar{p}_{1|11} + \bar{p}_{1|10})(\bar{p}_{1|11} + \bar{p}_{1|01})] \quad (29)$$

where $\bar{p}_{1|hi} = N^{-1}(D_{1hi} p_1 + D_{0hi} p_1')$.

Taking as an example $p_1 = 0.90$, $p'_1 = 0.10$

$$D_{111}/N = 0.01; D_{110}/N = 0.1; D_{101}/N = 0.15$$

$$D_{011}/N = 0.04; D_{010}/N = 0.15; D_{001}/N = 0.2$$

we have

$$\bar{p}_{1|11} = 0.010; \bar{p}_{1|10} = 0.105; \bar{p}_{1|01} = 0.155$$

so

$$\bar{p}_{1|11} - (\bar{p}_{1|11} + \bar{p}_{1|10})(\bar{p}_{1|11} + \bar{p}_{1|01}) = 0$$

corresponding to negative $\text{cov}(Z_2^*, Z_3^*)$.

On the other hand, if D_{111}/N and D_{011}/N are each increased to 0.05, the other parameters remaining the same, we have $\bar{p}_{1|11} = 0.050$, while $\bar{p}_{1|10}$ and $\bar{p}_{1|01}$ remain unchanged, so

$$\bar{p}_{1|11} - (\bar{p}_{1|11} + \bar{p}_{1|10})(\bar{p}_{1|11} + \bar{p}_{1|01}) = 0.050 - (0.155 \times 0.205) > 0$$

corresponding to positive $\text{cov}(Z_2^*, Z_3^*)$.

The same formulas apply when there are $m(2)$ types of defect other than (1), replacing D_{ghi} by $D_{ghi\dots}$.

At the cost of some elaboration in the formulas, we can allow for the possibility that presence or absence of a defect of type (2) may affect the probability of correct classification in regard to defects of type (1). Introducing the notation

${}_2p_1$ (${}_2\bar{p}_1$) for probabilities of detection of (1) in the presence (absence) of (2)

and

${}_2p'_1$ (${}_2\bar{p}'_1$) for probabilities of incorrect assignment of (1) when no (1) is present, in the presence (absence) of (2) we would still have a model of form (17) but the parameters of the binomial distributions of the

w_{gh} 's would now be $(y_{g0} \cdot {}_2p_1 + (1 - y_{g0}) \cdot {}_2\bar{p}'_1)$ for $h = 0$; $g = 1, 2$

$$(y_{g1}, \dots, y_{g2} \cdot 2^{p_1} + (1 - y_{g2}) \cdot 2^{p_1}) \quad \text{for } h = 1; g = 1, 2.$$

The probability of having defect (2), if classified as having (1) would be

$$\frac{D_{01} \cdot 2^{p_1} + D_{11} \cdot 2^{p_1}}{D_{00} \cdot 2^{p_1} + D_{01} \cdot 2^{p_1} + D_{10} \cdot 2^{p_1} + D_{11} \cdot 2^{p_1}} = \frac{\bar{p}_{2|1}^*}{\bar{p}_1^*} \quad (30)$$

where $\bar{p}_1^* = \frac{1}{N}(D_{00} \cdot 2^{p_1} + D_{01} \cdot 2^{p_1} + D_{10} \cdot 2^{p_1} + D_{11} \cdot 2^{p_1}) = \text{Pr}[\text{classified as having defect (1)}]$

$$\text{and } \bar{p}_{2|1}^* = \frac{1}{N}(D_{01} \cdot 2^{p_1} + D_{11} \cdot 2^{p_1})$$

Formulas (20.1) and (20.2) would still be valid, with $\bar{p}_1, \bar{p}_{2|1}$ replaced by $\bar{p}_1^*, \bar{p}_{2|1}^*$ respectively.

Extension to situations with $m(>1)$ types of defect, other than the one (type (1)) which is inspected directly is, again, straightforward.

In view of the model (17) which applies, with appropriate adjustments, to all the cases mentioned above, the joint distribution of Z's and Z*'s is asymptotically multinomial as the population size N increases indefinitely, with the ratios $n:D's:N$ remaining constant, or tending to fixed values.

5. SOME APPLICATIONS

Although this paper is concerned primarily with some novel compound multivariate discrete distribution which can arise in connection with faulty inspection rather than in specific applications, we shall indicate in this section a few circumstances in which knowledge of these distributions may be useful and directly applicable to specific investigations and inquiries.

The results in Section 2 are relevant to studies of robustness of multistage sample procedures to errors incurred in inspection and consequently to the

actual construction of such procedures. They would also be directly relevant to construction of tests for inequalities among the p_i 's and/or p_i' 's which could be one aspect of attempting to detect the existence of faulty items. (Evidently if p_i and/or p_i' vary with i , they cannot be identically equal to 1 or 0 respectively for all i). The distributions derived in this paper are also indirectly relevant to construction of tests of hypothesis of no faults ($p = 1, p' = 0$) assuming p_i, p_i' do not depend on i . Some attempts in this direction have been made in Johnson & Kotz (1982) while analyses of ways in which cost consideration can be allowed for in faulty inspection problems are sketched in Kotz & Johnson (1982b).

The results in Section 4 are relevant to assessment of performance of procedures for identifying individuals with defects of type (2), say (especially in those cases when these defects are not easily detectable) by observing the existence or non-existence of defects of type (1), and the robustness of this assessment to actual numbers of faults among inspected items. In these circumstances it may sometimes be appropriate to carry out a 100% inspection - that is, $n = N$ - though the more general formulas we have derived are of course of greater flexibility and are useful in various situations when total inspection is either not feasible or too costly. Indeed, studies in this direction will involve introduction of cost functions allowing for costs of sampling and losses due to erroneous retention of defective individuals of type (2) and the erroneous rejection of non-defective type (2) items. See Kotz & Johnson (1982b) for an appropriate model and some preliminary results for the case of two-stage sampling with defects of a single type. Finally questions of choice of which type(s) of defects to inspect for can also arise in this context.

ACKNOWLEDGEMENTS

Norman L. Johnson's work was supported by the National Science Foundation under Grant MCS-8021704. Samuel Kotz's work was supported by the U.S. Office of Naval Research under Contract N00014-81-K-0301.

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SECURITY CLASSIFICATION OF THIS PAGE (When Data Entered)

REPORT DOCUMENTATION PAGE		READ INSTRUCTIONS BEFORE COMPLETING FORM	
1. REPORT NUMBER 83-1	2. GOVT ACCESSION NO. A124018	3. REPORT CATEGORY NUMBER	
4. TITLE (and Subtitle) Some Multivariate Distributions Arising in Faulty Sampling Inspection		5. TYPE OF REPORT & PERIOD COVERED TECHNICAL	
7. AUTHOR(s) Norman L. Johnson and Samuel Kotz		6. PERFORMING ORG. REPORT NUMBER	
9. PERFORMING ORGANIZATION NAME AND ADDRESS Department of Management Sciences and Statistics University of Maryland College Park, MD 20742		8. CONTRACT OR GRANT NUMBER(s) ONR Contract N00014-81-K-0301 NSF Grant MCS-8-21704	
11. CONTROLLING OFFICE NAME AND ADDRESS U.S. Office of Naval Research Statistics and Probability Program (Code 436)		10. PROGRAM ELEMENT, PROJECT, TASK AND WORK UNIT NUMBER	
14. MONITORING AGENCY NAME & ADDRESS (if different from Controlling Office)		12. REPORT DATE January 1983.	
		13. NUMBER OF PAGES 17	
		15. SECURITY CLASS (of this report) UNCLASSIFIED	
		16. DECLASSIFICATION/DOWNGRADING SCHEDULE	
16. DISTRIBUTION STATEMENT (of this Report) Approved for Public Release--Distribution Unlimited			
17. DISTRIBUTION STATEMENT (of the Abstract entered in this report) (See instructions for details)			
18. SUPPLEMENTARY NOTES			
19. KEY WORDS (Continue on reverse side if necessary and identify by block number) Multivariate discrete distributions; Errors in inspection; Multistage sampling; Single-stage sampling			
20. ABSTRACT (Continue on reverse side if necessary and identify by block number) In the present paper we extend in two ways some results presented in Kotz and Johnson (Comm. in Statistics (1982) A11, 1997-2016) relating to the study of distributional aspects and applications of effects of errors in inspection sampling: 1) Multistage sampling with k successive samples involving the possibility of two types of errors in inspection; 2) Single-stage sampling considering several types of defects of which only one is tested on inspection.			

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