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A SEQUENTIAL SAMPLING PLAN FOR A LOT WITH CLASSIFIED DEFECTIVES

By

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Abstract

We deal with a lot consisting of effectives, and defectives classified into several categories according to their inferiorities. A sequential sampling plan for the lot based on the sequential probability ratio test proposed by Wald [7] is investigated.

The primary object is to develop exact formulae for operating characteristic, average sample size and the moment generating function of the number of observations for the sequential plan. For this purpose, we use the combinatorial method for counting all the possible sampling processes fallen into acceptance or rejection region. Furthermore, we propose a way of determining constants which involve two equalities giving decision boundaries of acceptance or rejection of the lot.

1. Introduction

Let us consider lots containing a large number of units (products). Each unit in a lot is classified as effective or defective and furthermore, each of defectives is classified into either one of k categories according to its degree of inferiority such as minor defective, major defective and so on, where k is a positive integer. Suppose that C_0 is the category of effectives and C_i ($i=1, \dots, k$) is that of defectives with the i -th degree of inferiority. Let p_i be the proportion of units belonging to C_i in the lot for $i=0, \dots, k$, where $\sum_{i=0}^k p_i=1$.

Let us consider sampling inspections for the lot satisfying the conditions that the probability of rejecting the lot does not exceed α whenever $p_i \leq p_{0i}$ for $i=1, \dots, k$ and that of accepting the lot does not exceed β whenever $p_i \geq p_{1i}$ for $i=1, \dots, k$, where α is producer's risk, β is a consumer's risk, and $\{p_{0i}\}$ and $\{p_{1i}\}$ are preassigned probability distributions such that

$$0 < p_{0i} < p_{1i} \quad \text{for } i=1, \dots, k \quad \text{and} \quad \min_{1 \leq i \leq k} \frac{p_{1i}}{p_{0i}} \geq \frac{p_{00}}{p_{10}}. \quad (1.1)$$

Such sampling inspections for the lot occur often in practice but analysis for in-

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spection plans having desired Operating Characteristics are not yet sufficiently developed, except for cases of $k=1$ (see for instance, [1]~[3], [7]) and $k=2$ ([6]).

In this paper, we consider a sequential sampling plan based on a sequential probability ratio test proposed by A. Wald [7].

Firstly, we represent a sequential sampling process as a lattice path in $(k+1)$ -dimensional space, so that decision boundaries of acceptance or rejection of the lot can be given by two parallel hyperplanes in the space. It turns out that the probability distribution of the number of observations required by the sequential plan can be obtained by a combinatorial method of counting all the possible lattice paths fallen into acceptance or rejection region.

On the other hand, Sato (one of the present authors) and Sado [4] have dealt with the number of lattice paths restricted by two parallel hyperplanes in $(k+1)$ -dimensional space and given explicit formulae for the generating functions of those numbers, provided that coefficients (inclines) of hyperplanes are all integral. In terms of the results, we give explicit formulae for various characteristics such as the OC, the ASN and the moment generating functions, and the deviation of the number of observations. Furthermore we propose a method for determining the values of coefficients of the hyperplanes and illustrate the method with an example. In the example, the values of ASN can be saved probably 50% comparing with the sample size for the corresponding single sampling plan.

For a special case of $k=1$, the theoretical and numerical results are given in the other paper [5].

2. Sequential probability ratio test

The sequential sampling plan based on the probability ratio test satisfying the condition described in the previous section is given as follows: at t -th trial for any integral value t , let n_i ($i=1, \dots, k$) be the number of units belonging to C_i in sample inspected successively, where $\sum_{i=1}^k n_i = t$. Then

$$\prod_{i=0}^k (p_{1i}/p_{0i})^{n_i} < C \quad (\text{The lot is accepted})$$

$$C \leq \prod_{i=0}^k (p_{1i}/p_{0i})^{n_i} \leq B \quad (\text{Inspection is continued})$$

$$\prod_{i=0}^k (p_{1i}/p_{0i})^{n_i} > B \quad (\text{The lot is rejected})$$

where constants B and C ($0 < C < B$) are determined to be such that the probabilities of the first and second kinds will take the values α and β .

The above inequalities can be rewritten as follows:

$$n_0 > \sum_{i=1}^k d_i n_i + c \quad (\text{The lot is accepted}) \quad (2.1)$$

$$\sum_{i=1}^k d_i n_i - b \leq n_0 \leq \sum_{i=1}^k d_i n_i + c \quad (\text{Inspection is continued}) \quad (2.2)$$

$$n_0 < \sum_{i=1}^k d_i n_i - b \quad (\text{The lot is rejected}) \quad (2.3)$$

where

$$d_i = \frac{\log p_{1i}/p_{0i}}{\log p_{00}/p_{10}}, \quad (i=1, \dots, k) \quad (2.4)$$

$$b = \frac{\log B}{\log p_{00}/p_{10}}, \quad \text{and} \quad c = \frac{\log 1/C}{\log p_{00}/p_{10}}. \quad (2.5)$$

Note that the equalities in (2.2) are contained in (2.1) and (2.3) in the procedure described by Wald [7].

The parameters $\{d_i\}$ are determined only by the preassigned probability distributions $\{p_{0i}\}$ and $\{p_{1i}\}$ as shown in (2.4). However, concerning with the parameters b and c , the explicit formula for the OC function is required to be evaluated the exact values of them.

3. Lattice Path Representation of a Sampling Process

We denote by $\{z_1, \dots, z_k, z_0\}$ the set of axes in $(k+1)$ -dimensional space. A (minimal) lattice path in the orthant of the space, abbreviated by LP, means a path passing through lattice points and moving one unit to positive direction along either one of $(k+1)$ -axes.

Let us consider one to one correspondence between a LP and a sampling process as follows: let be correspond an occurrence of unit belonging to C_i in the sampling process to a movement of one unit to the positive direction along an axis z_i , where the initial point of LP is assumed to be origin. Thus a sampling process can be presented graphically as a LP starting from the origin. The criteria boundaries of the sampling plan given by (2.1)–(2.3) can be given by the following two parallel hyperplanes:

$$\begin{cases} h_1: z_0 = \sum_{i=1}^k d_i z_i - b \\ h_2: z_0 = \sum_{i=1}^k d_i z_i + c \end{cases}$$

That is, the inspection is continued as long as the corresponding LP lies between two hyperplanes h_1 and h_2 , and is terminated the first time that the path crosses either one of both hyperplanes. If the path crosses $h_2(h_1)$, then the lot is accepted (rejected) (See Fig. 1). Let be denote by (\mathbf{n}, n_0) the co-ordinates of a lattice point in the space, where $\mathbf{n} = (n_1, \dots, n_k)$. In Fig. 1, the number n_0 of effectives is measured along the vertical axis z_0 and the numbers \mathbf{n} of the defectives in (C_1, \dots, C_k) along the horizontal axis $\mathbf{z} = (z_1, \dots, z_k)$, for convenience.

The parameter d_i defined by (2.4) is a real number larger than or equal to one, because of (1.1). Hereafter, we assume that $d_i, i=1, \dots, k$, are all integral. Under the assumption, we can take b and c as also integral without varying the criteria regions, since each LP can pass through only lattice points.

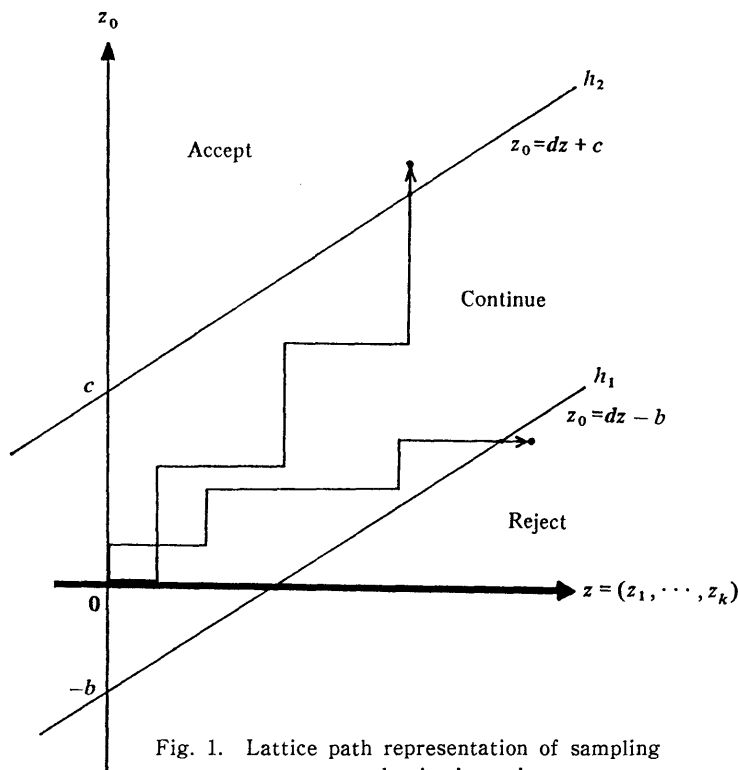


Fig. 1. Lattice path representation of sampling processes and criteria regions.

4. OC Function

Let be put $\mathbf{d}=(d_1, \dots, d_k)$ and $\mathbf{p}=(p_0, \dots, p_k)$.

For the sampling plan with parameters (\mathbf{d}, b, c) defined by (2.1)–(2.3), let us denote by $L(\mathbf{p}; \mathbf{d}, b, c)$ the OC function which is the probability of accepting the lot when the true value of the proportion of C_i ($i=0, \dots, k$) in the lot is p_i . Also we denote by $\bar{L}(\mathbf{p}; \mathbf{d}, b, c)$ the probability of rejecting the lot when the true proportion of C_i ($i=0, \dots, k$) is p_i . We show that these probabilities can be evaluated by counting LP's crossing either one of hyperplanes, respectively.

For non-negative integers a, b and c , let denote by $T(\mathbf{n}, a, b, c)$ the number of LP's from the origin to a lattice point $(\mathbf{n}, \mathbf{d}\mathbf{n}+a-b)$ without crossing any of h_1 and h_2 where $\mathbf{d}\mathbf{n}=\sum_{i=1}^k d_i n_i$, and introduce the generating function of these numbers:

$$T(\mathbf{x}, a, b, c) = \sum_{\substack{\mathbf{d}\mathbf{n}+a \geq b \\ \mathbf{n} \geq 0}} T(\mathbf{n}, a, b, c) \mathbf{x}^{\mathbf{n}},$$

where $\mathbf{x}^{\mathbf{n}} = x_1^{n_1} \dots x_k^{n_k}$ for $\mathbf{x}=(x_1, \dots, x_k)$ and $\mathbf{n}=(n_1, \dots, n_k)$ and the sum is extending over all non-negative n_i ($i=1, \dots, k$) such that $\mathbf{d}\mathbf{n}+a \geq b$. In a previous paper [4], the following expression for $T(\mathbf{x}, a, b, c)$ was given:

$$T(\mathbf{x}, a, b, c) = \frac{\varphi(\mathbf{x}, b)\varphi(\mathbf{x}, b+c-a)}{\varphi(\mathbf{x}, b+c+1)} - \varphi(\mathbf{x}, b-a-1), \quad (4.1)$$

where $\varphi(\mathbf{x}, n)$ is a polynomial of x_1, \dots, x_k such that

$$\varphi(\mathbf{x}, n) = \begin{cases} \sum_{0 \leq \mathbf{d} \leq n} \binom{n - \mathbf{d}}{\mathbf{l}} (-\mathbf{x})^{\mathbf{l}}, & n \geq 0 \\ 0, & n < 0 \end{cases} \quad (4.2)$$

where $(-\mathbf{x})^{\mathbf{l}} = (-x_1)^{l_1} \dots (-x_k)^{l_k}$ for $\mathbf{l} = (l_1, \dots, l_k)$, and $\binom{n}{\mathbf{l}}$ means the multinomial coefficient.

In terms of the above result, we have the following two Theorems.

THEOREM 1.

$$L(\mathbf{p}; \mathbf{d}, b, c) = p_0^{c+1} \phi(\mathbf{x}_{\mathbf{p}}, b, c) \quad (4.3)$$

where

$$\phi(\mathbf{x}, b, c) = \frac{\varphi(\mathbf{x}, b)}{\varphi(\mathbf{x}, b+c+1)} \quad (4.4)$$

and

$$\mathbf{x}_{\mathbf{p}} = (p_1 p_0^{d_1}, \dots, p_k p_0^{d_k}) \quad (4.5)$$

for $\mathbf{p} = (p_0, \dots, p_k)$.

PROOF. Let us consider all the possible sampling processes terminated with the acceptance of the lot, in each of which defectives belonging to C_i occur n_i times for $i=1, \dots, k$. Each of these processes corresponds to each of LP's starting from the origin to $(\mathbf{n}, \mathbf{d}\mathbf{n}+c)$ without crossing any of two hyperplanes and moving to $(\mathbf{n}, \mathbf{d}\mathbf{n}+c+1)$ after crossing the hyperplane h_2 . The number of such LP's is given by $T(\mathbf{n}, b+c, b, c)$. Since the probability of each of these LP's is given by $\left(\prod_{r=1}^k p_r^{n_r}\right) p_0^{d\mathbf{n}+c+1}$, the probability of acceptance of the lot for processes considered is given by

$$T(\mathbf{n}, b+c, b, c) \left(\prod_{r=1}^k p_r^{n_r}\right) p_0^{d\mathbf{n}+c+1} \quad (4.6)$$

which implies from (4.1),

$$\begin{aligned} L(\mathbf{p}; \mathbf{d}, b, c) &= \sum_{\mathbf{n} \geq 0} T(\mathbf{n}, b+c, b, c) \left(\prod_{r=1}^k p_r^{n_r}\right) p_0^{d\mathbf{n}+c+1} \\ &= p_0^{c+1} T(\mathbf{x}_{\mathbf{p}}, b+c, b, c) = p_0^{c+1} \phi(\mathbf{x}_{\mathbf{p}}, b, c). \quad \square \end{aligned}$$

If $p_i, i=1, \dots, k$, are sufficiently small, $L(\mathbf{p}; \mathbf{d}, b, c)$ is approximately given from (4.3) by the expansion of the rational function $\phi(\mathbf{x}, b, c)$ as follows:

COROLLARY 1.

$$L(\mathbf{p}; \mathbf{d}, b, c) \sim p_0^{c+1} \left\{ 1 + (c+1) \sum_{i=1}^k p_i p_0^{d_i} \right\} \quad (4.7)$$

THEOREM 2.

$$\bar{L}(\mathbf{p}; \mathbf{d}, b, c) = p_0^{-b} [R(\mathbf{p}, \mathbf{x}_{\mathbf{p}}, b) - R(\mathbf{p}, \mathbf{x}_{\mathbf{p}}, b+c+1) \phi(\mathbf{x}_{\mathbf{p}}, b, c)] \quad (4.8)$$

where $\phi(\mathbf{x}, b, c)$ and $\mathbf{x}_{\mathbf{p}}$ are defined by (4.4) and (4.5), and

$$R(\mathbf{p}, \mathbf{x}, n) = \varphi(\mathbf{x}, n) - \sum_{i=1}^k p_i \sum_{j=0}^{d_i-1} p_0^j \varphi(\mathbf{x}, n-1-j) \quad (4.9)$$

for $\mathbf{p} = (p_0, \dots, p_k)$.

PROOF. Similarly to the proof of Theorem 1, the probability of rejecting the lot by occurrences of n_i ($i=1, \dots, k$) defectives belonging to C_i is given as

$$\sum_{i=1}^k p_i \sum_{j=0}^{d_i-1} T(n, j, b, c) \left(\prod_{r=1}^k p_r^{n_r} \right) p_0^{dn-b+j}. \quad (4.10)$$

Therefore we have using (4.1)

$$\begin{aligned} \bar{L}(\mathbf{p}; \mathbf{d}, b, c) &= \sum_{i=1}^k p_i \sum_{j=0}^{d_i-1} \sum_{\mathbf{d}, \mathbf{n}+j \geq \mathbf{b}} T(n, j, b, c) \left(\prod_{r=1}^k p_r^{n_r} \right) p_0^{dn-b+j} \\ &= \sum_{i=1}^k p_i \sum_{j=0}^{d_i-1} p_0^j \left[\frac{\varphi(\mathbf{x}_{\mathbf{p}}, b)}{\varphi(\mathbf{x}_{\mathbf{p}}, b+c+1)} \varphi(\mathbf{x}_{\mathbf{p}}, b+c-j) - \varphi(\mathbf{x}_{\mathbf{p}}, b-j-1) \right], \end{aligned}$$

which implies (4.8) from (4.9). \square

5. The Properties of Polynomial $\varphi(\mathbf{x}, n)$ and $R(\mathbf{p}, \mathbf{x}, n)$

In this paragraph, we discuss about equations concerning with the polynomial $\varphi(\mathbf{x}, n)$ or $R(\mathbf{p}, \mathbf{x}, n)$, which are used later in this paper.

Firstly, we give the generating function for $\{\varphi(\mathbf{x}, n), n \geq 0\}$ and a difference equation of $\varphi(\mathbf{x}, n)$ which is useful for numerical calculations of $\varphi(\mathbf{x}_{\mathbf{p}}, n)$.

LEMMA 1.

$$(i) \quad \sum_{n=0}^{\infty} \varphi(\mathbf{x}, n) z^n = \frac{1}{A(\mathbf{x}, z)}, \quad (5.1)$$

where

$$A(\mathbf{x}, z) = 1 - z + \sum_{i=1}^k x_i z^{d_i+1}. \quad (5.2)$$

for $\mathbf{x} = (x_1, \dots, x_k)$.

(ii) The following difference equation is valid:

$$\varphi(\mathbf{x}, n) = \begin{cases} 1, & n=0 \\ \varphi(\mathbf{x}, n-1) - \sum_{i=1}^k x_i \varphi(\mathbf{x}, n-d_i-1), & n \geq 1. \end{cases} \quad (5.3)$$

PROOF. (i) From the definition of $\varphi(\mathbf{x}, n)$, we have

$$\begin{aligned} \sum_{n=0}^{\infty} \varphi(\mathbf{x}, n) z^n &= \sum_{n=0}^{\infty} \sum_{\mathbf{d}, \mathbf{l} \leq \mathbf{n}} \binom{n-\mathbf{d}\mathbf{l}}{\mathbf{l}} (-\mathbf{x})^{\mathbf{l}} z^n \\ &= \sum_{n=0}^{\infty} \sum_{\mathbf{l} \geq \mathbf{0}} \binom{n}{\mathbf{l}} (-x_1 z^{d_1})^{l_1} \dots (-x_k z^{d_k})^{l_k} z^n \\ &= \sum_{n=0}^{\infty} \left(z - \sum_{i=1}^k x_i z^{d_i+1} \right)^n, \end{aligned}$$

which is equal to the right hand side of (5.1).

(ii) It is clear since (i). \square

LEMMA 2.

$$\sum_{j=1}^k (d_j+1) x_j \varphi^{(j)}(\mathbf{x}, n) = \varphi^{(1)}(\mathbf{x}, n+d_1) + n \varphi(\mathbf{x}, n) \quad (5.4)$$

where $\varphi^{(j)}(\mathbf{x}, n) = \frac{\partial}{\partial x_j} \varphi(\mathbf{x}, n)$, $j=1, \dots, k$.

PROOF. Partial differentiating the both side of (5.1) with respect to $x_j (j=1, \dots, k)$ and z , we have

$$\sum_{n=0}^{\infty} \varphi^{(j)}(\mathbf{x}, n) z^n = \frac{-z^{d_j+1}}{A(\mathbf{x}, z)^2}, \quad j=1, \dots, k$$

and

$$\sum_{n=0}^{\infty} n \varphi(\mathbf{x}, n) z^n = \frac{z - \sum_{i=1}^k (d_i + 1) x_i z^{d_i+1}}{A(\mathbf{x}, z)^2},$$

which give

$$\sum_{n=0}^{\infty} \sum_{j=1}^k (d_j + 1) x_j \varphi^{(j)}(\mathbf{x}, n) z^n = \sum_{n=0}^{\infty} \varphi^{(1)}(\mathbf{x}, n + d_1) z^{n-d_1} + \sum_{n=0}^{\infty} n \varphi(\mathbf{x}, n) z^n.$$

Comparing with the coefficients of z^n in the both side of the above equation, we can get (5.3). \square

Hereafter, we use the notation of dash (') or two-dash (") instead of the partial derivative $\frac{\partial}{\partial x_1}$ or $\frac{\partial^2}{\partial x_1^2}$.

LEMMA 3. Let ω be a real variable.

$$(i) \quad R(\mathbf{p}_\omega, \mathbf{x}_{\mathbf{p}_\omega}, n) = (p_0 e^\omega)^n + (1 - e^\omega) S(\mathbf{p}_\omega, \mathbf{x}_{\mathbf{p}_\omega}, n-1) \quad (5.5)$$

$$(ii) \quad R(\mathbf{p}, \mathbf{x}_{\mathbf{p}}, n) = p_0^n, \quad (5.6)$$

where $\mathbf{x}_{\mathbf{p}}$ and $R(\mathbf{p}, \mathbf{x}, n)$ are defined by (4.5) and (4.9),

$$S(\mathbf{p}, \mathbf{x}, n) = \sum_{l=0}^n p_0^l \varphi(\mathbf{x}, n-l), \quad (5.7)$$

and

$$\mathbf{p}_\omega = e^\omega \mathbf{p}. \quad (5.8)$$

PROOF. (i) The generating functions for $R(\mathbf{p}, \mathbf{x}, n)$ and $S(\mathbf{p}, \mathbf{x}, n)$ defined by (4.9) and (5.7) may be derived from Lemma 1(i) as follows:

$$\sum_{n=0}^{\infty} R(\mathbf{p}, \mathbf{x}, n) z^n = \frac{B(\mathbf{p}, z)}{A(\mathbf{x}, z)} \quad (5.9)$$

and

$$\sum_{n=0}^{\infty} S(\mathbf{p}, \mathbf{x}, n) z^n = \frac{1}{(1 - p_0 z) A(\mathbf{x}, z)}, \quad (5.10)$$

respectively, where $A(\mathbf{x}, z)$ is defined by (5.2) and

$$B(\mathbf{p}, z) = 1 - \sum_{i=1}^k p_i \sum_{j=0}^{d_i-1} p_0^j z^{j+1}. \quad (5.11)$$

Since

$$A(\mathbf{x}_{\mathbf{p}_\omega}, z) = (1 - p_0 z e^\omega) B(\mathbf{p}_\omega, z) - (1 - e^\omega) z, \quad (5.12)$$

it turns out from (5.9) and (5.10) that

$$\begin{aligned} \sum_{n=0}^{\infty} R(\mathbf{p}_\omega, \mathbf{x}_{\mathbf{p}_\omega}, n) z^n &= \frac{1}{1 - p_0 z e^\omega} \left\{ 1 + \frac{(1 - e^\omega) z}{A(\mathbf{x}_{\mathbf{p}_\omega}, z)} \right\} \\ &= \frac{1}{1 - p_0 z e^\omega} + (1 - e^\omega) \sum_{n=0}^{\infty} S(\mathbf{x}_{\mathbf{p}_\omega}, n) z^{n+1}. \end{aligned}$$

Comparing with coefficients of z^n in the both side of the above equation, we get (5.5).

(ii) Since (5.5) for $\omega=0$, (5.6) is valid. \square

LEMMA 4.

$$(i) \quad S(\mathbf{p}, \mathbf{x}_p, n) = -R'(\mathbf{p}, \mathbf{x}_p, n+d_1+1) \quad (5.13)$$

$$(ii) \quad \left[\frac{\partial}{\partial \omega} S(\mathbf{p}_\omega, \mathbf{x}_{p_\omega}, n) \right]_{\omega=0} = -\frac{1}{2} [R''(\mathbf{p}, \mathbf{x}_p, n+2d_1+1) + 2nR'(\mathbf{p}, \mathbf{x}_p, n+d_1+1)] \quad (5.14)$$

where \mathbf{x}_p , $R(\mathbf{p}, \mathbf{x}, n)$, $S(\mathbf{p}, \mathbf{x}, n)$ and \mathbf{p}_ω are defined by (4.5), (4.9), (5.7) and (5.8).

PROOF. (i) From (5.12) for $\omega=0$, it follows that

$$A(\mathbf{x}_p, z) = (1-p_0z)B(\mathbf{p}, z). \quad (5.15)$$

After partial differentiating the both side of (5.9) w.r.t. x_1 we put $\mathbf{x}=\mathbf{x}_p$, it turns out (5.10) and (5.15) that

$$\begin{aligned} \sum_{n=0}^{\infty} R'(\mathbf{p}, \mathbf{x}_p, n)z^n &= \frac{-B(\mathbf{p}, z)z^{d_1+1}}{A(\mathbf{x}_p, z)^2} = \frac{-z^{d_1+1}}{(1-p_0z)A(\mathbf{x}_p, z)} \\ &= -z^{d_1+1} \sum_{n=1}^{\infty} S(\mathbf{p}, \mathbf{x}_p, n)z^n, \end{aligned}$$

which gives (5.12).

(ii) Similarly to (i), it follows that

$$\sum_{n=0}^{\infty} R''(\mathbf{p}, \mathbf{x}_p, n)z^n = \frac{z^{2(d_1+1)}}{2(1-p_0z)A(\mathbf{x}_p, z)^2}.$$

Hence we have from (5.10)

$$\begin{aligned} \left[\frac{\partial}{\partial \omega} \sum_{n=0}^{\infty} S(\mathbf{p}_\omega, \mathbf{x}_{p_\omega}, n)z^n \right]_{\omega=0} &= z \left(\frac{\partial}{\partial z} \sum_{n=0}^{\infty} S(\mathbf{p}, \mathbf{x}_p, n)z^n \right) - \frac{z}{(1-p_0z)A(\mathbf{x}_p, z)^2} \\ &= \sum_{n=0}^{\infty} nS(\mathbf{p}, \mathbf{x}_p, n)z^n - \frac{1}{2} z^{-(2d_1+1)} \sum_{n=0}^{\infty} R''(\mathbf{p}, \mathbf{x}_p, n)z^n, \end{aligned}$$

which together with (5.13), implies (5.14).

5. The Moment Generating Function for Sample Size

Let us denote by T the number of observations required by the sequential sampling plan. Then T is a random variable.

In this section, we derive a formula for the moment generating function of T , denoted by $E_p[e^{\omega T}]$.

It has been already shown that the probability that the sequential sampling procedure will eventually terminate is one ([5]). Here we show that it can be proved easily using Theorem 1, Theorem 2 and Lemma 3(ii).

THEOREM 3.

$$L(\mathbf{p}; \mathbf{d}, b, c) + \bar{L}(\mathbf{p}; \mathbf{d}, b, c) = 1. \quad (6.1)$$

PROOF. From Theorem 1, Theorem 2 and Lemma 3(ii), we have

$$\begin{aligned}\bar{L}(\mathbf{p}; \mathbf{d}, b, c) &= p_0^{-b} [p_0^b - p_0^{b+c+1} \cdot \phi(\mathbf{x}_p, b, c)] \\ &= 1 - L(\mathbf{p}; \mathbf{d}, b, c). \quad \square\end{aligned}$$

THEOREM 4.

$$(i) \quad E_p[e^{\omega T}] = (p_0 e^{\omega})^{c+b+1} \cdot \phi(\mathbf{x}_{p_\omega}, b, c) + (p_0 e^{\omega})^{-b} \cdot \{R(\mathbf{p}_\omega, \mathbf{x}_{p_\omega}, b) - R(\mathbf{p}_\omega, \mathbf{x}_{p_\omega}, b+c+1) \cdot \phi(\mathbf{x}_{p_\omega}, b, c)\} \quad (6.2)$$

$$(ii) \quad E_p[e^{\omega T}] = 1 + (1 - e^{\omega})(p_0 e^{\omega})^{-b} [S(\mathbf{p}_\omega, \mathbf{x}_{p_\omega}, b-1) - S(\mathbf{p}_\omega, \mathbf{x}_{p_\omega}, b+c) \cdot \phi(\mathbf{x}_{p_\omega}, b, c)] \quad (6.3)$$

where $\phi(\mathbf{x}, b, c)$, \mathbf{x}_p , $R(\mathbf{p}, \mathbf{x}, n)$, $S(\mathbf{p}, \mathbf{x}, n)$ and \mathbf{p}_ω are defined by (4.4), (4.5), (4.9), (5.7) and (5.8).

PROOF. (i) Let us consider the sampling processes dealt in the proof of Theorem 1 or Theorem 2. Clearly, the sample size of each of these processes is equal to the sum of powers of p_i ($i=0, \dots, k$) appearing in the expression ((4.6) or (4.10)) of the probability of its occurrence, for instance that for a sampling process with probability $\left(\prod_{r=1}^k p_r^{n_r}\right) p_0^{d_{n+c+1}}$ (see (4.6)) the size is $\sum_{r=1}^k n_r + d_{n+c+1}$. It means that $E_p[e^{\omega T}]$ is equal to the sum of the right hand sides of (4.3) and (4.8) substituting $p_i e^{\omega}$ instead of p_i for $i=0, 1, \dots, k$. Thus we find (6.2).

(ii) From (6.2) and Lemma 3(i), (6.3) is valid. \square

6. The ASN Function and the Second Moment of T

In this section, we show explicit formulae for the ASN and the second moment of T , denoted by $E_p[T]$ and $E_p[T^2]$, respectively.

THEOREM 5.

$$E_p[T] = p_0^{-b} [R'(\mathbf{p}, \mathbf{x}_p, b+d_1) - R'(\mathbf{p}, \mathbf{x}_p, b+c+1+d_1) \cdot \phi(\mathbf{x}_p, b, c)], \quad (7.1)$$

where $\phi(\mathbf{x}, b, c)$, \mathbf{x}_p and $R(\mathbf{p}, \mathbf{x}, n)$ are defined by (4.4), (4.5) and (4.9).

PROOF. It follows from Theorem 4(ii) and Lemma 5(i) that

$$\begin{aligned}E_p[T] &= \left[\frac{\partial}{\partial \omega} E_p[e^{\omega T}] \right]_{\omega=0} \\ &= p_0^{-b} [S(\mathbf{p}, \mathbf{x}_p, b+c) \cdot \phi(\mathbf{x}_p, b, c) - S(\mathbf{p}, \mathbf{x}_p, b-1)] \\ &= p_0^{-b} [R'(\mathbf{p}, \mathbf{x}_p, b+d_1) - R'(\mathbf{p}, \mathbf{x}_p, b+c+1+d_1) \cdot \phi(\mathbf{x}_p, b, c)]. \quad \square\end{aligned}$$

THEOREM 6.

$$\begin{aligned}E_p[T^2] &= p_0^{-b} [R''(\mathbf{p}, \mathbf{x}_p, b+2d_1) - R''(\mathbf{p}, \mathbf{x}_p, b+c+1+2d_1) \cdot \phi(\mathbf{x}_p, b, c) \\ &\quad + \frac{2R'(\mathbf{p}, \mathbf{x}_p, b+c+1+d_1)}{\phi(\mathbf{x}_p, b+c+1)} \{\varphi'(\mathbf{x}_p, b+c+1+d_1) \cdot \phi(\mathbf{x}_p, b, c) \\ &\quad - \varphi'(\mathbf{x}_p, b+d_1)\}] - E_p[T], \quad (7.2)\end{aligned}$$

where $\varphi(\mathbf{x}, n)$, $\phi(\mathbf{x}, b, c)$, \mathbf{x}_p and $R(\mathbf{p}, \mathbf{x}, n)$ are defined by (4.2), (4.4), (4.5) and (4.9).

PROOF. It follows from Theorem 4(ii) and Theorem 5 that

$$E_p[T^2] = (1-2b)E_p[T] + 2p^{-b} \frac{\partial}{\partial \omega} [S(\mathbf{p}_\omega, \mathbf{x}_{\mathbf{p}_\omega}, b+c) \cdot \phi(\mathbf{x}_{\mathbf{p}_\omega}, b, c) - S(\mathbf{p}_\omega, \mathbf{x}_{\mathbf{p}_\omega}, b-1)]_{\omega=0} \quad (7.3)$$

From Lemma 5(ii), we get

$$\left[\frac{\partial}{\partial \omega} S(\mathbf{p}, \mathbf{x}_{\mathbf{p}_\omega}, b+c) \right]_{\omega=0} = -\frac{1}{2} [R''(\mathbf{p}, \mathbf{x}_{\mathbf{p}}, b+c+2d_1+1) + 2(b+c)R'(\mathbf{p}, \mathbf{x}_{\mathbf{p}}, b+c+d_1+1)] \quad (7.4)$$

and

$$\left[\frac{\partial}{\partial \omega} S(\mathbf{p}, \mathbf{x}_{\mathbf{p}_\omega}, b-1) \right]_{\omega=0} = -\frac{1}{2} [R''(\mathbf{p}, \mathbf{x}_{\mathbf{p}}, b+2d_1) + 2(b-1)R'(\mathbf{p}, \mathbf{x}_{\mathbf{p}}, b+d_1)]. \quad (7.5)$$

On the other hand, we have from Lemma 2

$$\left[\frac{\partial}{\partial \omega} \varphi(\mathbf{x}_{\mathbf{p}_\omega}, n) \right]_{\omega=0} = \sum_{j=1}^k (d_j+1) p_j p_0^{d_j} \varphi^{(j)}(\mathbf{x}_{\mathbf{p}}, n) = \varphi'(\mathbf{x}_{\mathbf{p}}, n+d_1) + n\varphi(\mathbf{x}_{\mathbf{p}}, n),$$

which implies

$$\left[\frac{\partial}{\partial \omega} \phi(\mathbf{x}_{\mathbf{p}_\omega}, b, c) \right]_{\omega=0} = \frac{\varphi'(\mathbf{x}_{\mathbf{p}}, b+d_1) - \phi(\mathbf{x}_{\mathbf{p}}, b, c)\varphi'(\mathbf{x}_{\mathbf{p}}, b+c+1+d_1)}{\varphi(\mathbf{x}_{\mathbf{p}}, b+c+1)} - (c+1)\phi(\mathbf{x}_{\mathbf{p}}, b, c) \quad (7.6)$$

Substituting (7.4)-(7.6) to (7.3), we get (7.2). \square

8. Determination of the Values of Parameters

In this section, we propose a method for determining the values of the parameters (\mathbf{d}, b, c) and illustrate it with an example. After determining the values of parameters, we evaluate the values of the ASN and the deviation of T using the formulae obtained in the previous sections, and furthermore compare with the sample size for the corresponding single sampling plan.

In the sequential probability ratio test, the values of d_i ($i=1, \dots, k$) are determined by the preassigned probabilities $\{p_{0i}\}$ and $\{p_{1i}\}$ as defined by (2.4) and those of b and c are determined to be such that

$$\alpha(\mathbf{d}, b, c) = \alpha \quad \text{and} \quad \beta(\mathbf{d}, b, c) = \beta, \quad (8.1)$$

where

$$\alpha(\mathbf{d}, b, c) = 1 - L(\mathbf{p}_0; \mathbf{d}, b, c) \\ \beta(\mathbf{d}, b, c) = L(\mathbf{p}_1; \mathbf{d}, b, c),$$

for $\mathbf{p}_i = (p_{i0}, \dots, p_{ik})$ ($i=0, 1$) and the preassigned values α and β . But, the value of d_i given by (2.4) is not necessarily integral, and the equalities of (8.1) are not generally valid because the number of observations can take only integral values. Hence we shall determine the values of \mathbf{d}, b and c as follows: Firstly, let us introduce a set

$$\mathcal{D} = \{\mathbf{d} : d_i = [d_i^*] \text{ or } [d_i^*] + 1, i=1, \dots, k\}, \quad (8.2)$$

where $[d_i^*]$ means the largest integer not exceeding d_i^* and

$$d_i^* = \frac{\log p_{1i}/p_{0i}}{\log p_{00}/p_{10}}, \quad i=1, \dots, k. \quad (8.3)$$

We choose the optimal \mathbf{d} among the set \mathcal{D} consisting of 2^k possible \mathbf{d} 's, which is determined together with the values of b and c such that

$$\{\alpha + \beta - \alpha(\mathbf{d}, b, c) - \beta(\mathbf{d}, b, c)\} \quad (8.4)$$

is a minimum, under the conditions

$$\alpha(\mathbf{d}, b, c) \leq \alpha \quad \text{and} \quad \beta(\mathbf{d}, b, c) \leq \beta. \quad (8.5)$$

At the time finding in practice for the optimal values satisfying above condition, the following two Lemmas are useful.

LEMMA 5. (i) $L(\mathbf{p}; \mathbf{d}, b, c)$ is a increasing function of b and a decreasing function of c and d_i ($i=1, \dots, k$).

(ii) $E_{\mathbf{p}}[T]$ is a increasing function of b and c .

PROOF. It is clear because of the consideration of the regions of acceptance or rejection of the lot. \square

LEMMA 6. If $d_i = d_i^*$ for $i=1, \dots, k$ and (8.5) are satisfied, then

$$c \geq c_w \quad (8.6)$$

is valid, where

$$c_w = \frac{\log(1-\alpha)/\beta}{\log p_{00}/p_{10}} - 1 \quad (8.7)$$

PROOF. If $d_i = d_i^*$ for $i=1, \dots, k$, then from (4.5) we have $\mathbf{x}_{\mathbf{p}_0} = \mathbf{x}_{\mathbf{p}_1}$, which means

$$\varphi(\mathbf{x}_{\mathbf{p}_0}, n) = \varphi(\mathbf{x}_{\mathbf{p}_1}, n), \quad n \geq 0,$$

so that

$$\phi(\mathbf{x}_{\mathbf{p}_0}, b, c) = \phi(\mathbf{x}_{\mathbf{p}_1}, b, c).$$

Consequently, it follows from Theorem 1 and (8.5) that

$$(p_{00}/p_{10})^{c+1} \geq \frac{1-\alpha}{\beta}.$$

which give (8.6). \square

Note that the equality of (8.6) is hold for a case of (8.1) and c_w is the same as Wald's approximate formula for c ([7]).

In practical calculation, we adopt b_w and c_w as the initial values of b and c , where b_w is Wald's approximate formula for b as defined by

$$b_w = \frac{\log(1-\beta)/\alpha}{\log p_{00}/p_{10}} - 1,$$

and find the optimal values of (\mathbf{d}, b, c) , taking Lemma 4(i) into consideration.

As an illustration, consider the following example:

EXAMPLE. (A case of $k=2$) Let us consider a lot with three categories such that $C_0 = \{\text{effectives}\}$, $C_1 = \{\text{minor defectives}\}$ and $C_2 = \{\text{major defectives}\}$. Let

$$p_{01}=0.005, \quad p_{02}=0.015$$

$$p_{11}=0.02, \quad p_{12}=0.06$$

$$\alpha=0.05, \quad \beta=0.1$$

Then we have

$$d_1^*=21.94 \quad d_2^*=21.94 \quad b_w=34.63 \quad c_w=44.75.$$

After numerical calculations, we find the optimal values:

$$d_1=21 \quad d_2=22 \quad b=38 \quad c=35,$$

where

$$\alpha(\mathbf{d}, b, c)=0.0478 \quad \beta(\mathbf{d}, b, c)=0.0993.$$

For these optimal values, we get

$$E_{\mathbf{p}_0}[T]=58.968 \quad E_{\mathbf{p}_1}[T]=46.929$$

$$D_{\mathbf{p}_0}[T]=34.883 \quad D_{\mathbf{p}_1}[T]=38.072,$$

where $D_{\mathbf{p}}[T]$ means the deviation of T .

Now we compare the above results with those for single sampling plan. Let denote c_i ($i=1, 2$) for the acceptance number of defectives belonging to C_i and t for the sample size. Then we have the following results:

$$c_1=2 \quad c_2=4 \quad t=115,$$

where the first and second kinds of errors are given and denoted by

$$\alpha(t, c_1, c_2)=0.0499 \quad \beta(t, c_1, c_2)=0.0999.$$

In Table 1, the values of the OC functions (denoted by $L(\mathbf{p}; t, c_1, c_2)$) for the single sampling plans, the values of $L(\mathbf{p}; \mathbf{d}, b, c)$ and $E_{\mathbf{p}}[T]$ considered above are given in the upper, middle and lower rows of arrays, respectively, for several values of p_1 and

Table 1. The values of $L(\mathbf{p}; t, c_1, c_2)$, $L(\mathbf{p}; \mathbf{d}, b, c)$ and $E_{\mathbf{p}}(T)$.

$p_1 \backslash p_2$	0.010	0.015	0.030	0.045	0.060
0.001	.9936 .9917 47.239	.9697 .9747 53.647	.7363 .8128 72.032	.4057 .5090 74.945	.1738 .2557 63.002
0.005	.9735 .9807 52.191	.9501 .9522 58.968	.7211 .7428 75.035	.3971 .4328 72.948	.1700 .2104 59.416
0.010	.8855 .9552 58.864	.8640 .9079 65.586	.6548 .6444 76.877	.3597 .3470 69.345	.1535 .1641 54.995
0.015	.7464 .9119 65.574	.7281 .8429 71.454	.5506 .5418 76.513	.3012 .2743 65.022	.1279 .1277 50.803
0.020	.5914 .8477 71.563	.5767 .7582 75.764	.4349 .4435 74.243	.2367 .2148 60.430	.0999 .0993 46.929

p_2 . So far as the table shows, it may be considered that two OC curves are almost similar. It comes to the conclusion that the values of the ASN for the sequential plan can be saved as much as 50% comparing with the sample size for corresponding single sampling plan.

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