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## Design of Rectifying Inspection Plans by Variables

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Although the inspection is to control the fraction defective in the lot inspected, there often appears that the lot quality characteristic obeys a continuous distribution. Two kinds of rectifying inspection plans, called LTPD plans and AOQL plans, are considered for which the average total inspection is to be minimized subject to particular side conditions. In this paper we propose the rectifying inspection plans by variables with a sampling plan having sample size and acceptance coefficient under the condition that the lot quality characteristic obeys a normal distribution. It is illustrated that the average total inspection can be economized by using the proposed inspection plans.

### 1. Introduction

Two kinds of rectifying sampling plans, called LTPD (lot tolerance percent defective) and AOQL (average outgoing quality limit) plans which have been given extensive treatment by Dodge and Romig<sup>1)</sup>, are considered for which the average total inspection (ATI) is to be minimized subject to particular side conditions. The third kind of plans developed by Hall and Hassan<sup>2)</sup> are called outgoing quality (OQ) plans. All of those three types of plans are rectifying inspection plans by attributes with a sampling plan  $(n, c)$  having sample size  $n$  and acceptance number  $c$ .

Although the inspection is to control the fraction defective in the lot inspected, there often appears that the lot quality characteristic obeys a continuous distribution. When the lot quality characteristic obeys a normal distribution with unknown variance, Bender<sup>3)</sup> produced a table of variables sampling plans, fitted to the attribute plans of Table II-A of MIL-STD-105D, by means of an iterative computer program based on the noncentral  $t$ -distribution. Hamaker<sup>4)</sup> demonstrated that the OC curve for  $s$ -method plans could be adequately derived from a normal approximation and that the more complicated use of the noncentral  $t$ -distribution could be avoided. Recently, by using Hamaker's approximations, Govindaraju<sup>5)</sup> provided procedures and tables for the selection of variables sampling plans for given AQL (acceptable quality level) and AOQL, whenever rejected lots were 100% inspected for replace-

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ment of nonconforming units. However, the rectifying sampling plans provided by Govindaraju do not always satisfy the requirement of minimizing ATI. In this paper we propose procedures for determining variables sampling plans for the two kinds of rectifying inspection plans, called LTPD and AOQL plans, under the condition that the lot quality characteristic obeys a normal distribution. It is illustrated that the average total inspection can be economized by using the proposed inspection plans.

## 2. Variables Sampling Plans

Suppose that the quality characteristic  $x$  is measurable on a continuous scale and normally distributed with mean  $\mu$  and standard deviation  $\sigma$ . The following notations will be used:

- $N$  : lot size
- $S_u$  : upper specification limit
- $\bar{p}$  : process average
- $p_1$  : lot tolerance percent defective
- $\phi(\cdot)$  : p. d. f. of  $N(0, 1)$
- $\Phi(\cdot)$  : c. d. f. of  $N(0, 1)$
- $z_p$  : upper  $p$ -quantile value of  $N(0, 1)$
- $n$  : sample size
- $k$  : acceptance coefficient
- $L(p)$  : probability of acceptance
- ATI : average total inspection

where we put the ' $\sigma$ ' and ' $s$ ' on foot to indicate the cases in which the standard deviation  $\sigma$  is known ( $\sigma$ -method) and unknown ( $s$ -method), respectively.

The acceptance criterion for the  $\sigma$ -method plan is

$$\begin{cases} \text{if } \bar{x} + k_\sigma \sigma \leq S_u, \text{ then accept the lot,} \\ \text{otherwise reject the lot,} \end{cases} \quad (1)$$

where  $\bar{x}$  is the average quality characteristic derived from the sample. Under this criterion, the probability of acceptance will be

$$L_\sigma(p) = \int_{-\infty}^{\omega} \frac{1}{\sqrt{2\pi}} \exp(-z^2/2) dz = \Phi(\omega) \quad (2)$$

with

$$\omega = \sqrt{n_\sigma} (S_u - k_\sigma - \mu) / \sigma. \quad (3)$$

Notice that the fraction defective  $p$  in the lot is described as

$$p = \int_{(S_u - \mu)/\sigma}^{\infty} \phi(z) dz, \quad (4)$$

therefore, from Eqs. (3) and (4),  $\omega$  is found by

$$\omega = \sqrt{n_\sigma} (z_p - k_\sigma) \quad (5)$$

When the standard deviation  $\sigma$  is unknown, the acceptance criterion of Eq. (1) is replaced by

$$\begin{cases} \text{if } \bar{x} + k_s s \leq S_u, \text{ then accept the lot,} \\ \text{otherwise reject the lot,} \end{cases} \quad (6)$$

where  $s$  is the customary estimate of  $\sigma$ . Under this criterion, the probability of acceptance will be based on the noncentral  $t$ -distribution. To avoid the complication of using the noncentral  $t$ -distribution, we use the approximations given by Hamaker<sup>4</sup>. Since the relationships between parameters are

$$k_\sigma = k_s(4n_s - 5) / (4n_s - 4), \quad (7)$$

$$\frac{1}{n_\sigma} = \frac{1}{n_s} + \frac{k_s^2}{2(n_s - 1)}, \quad (8)$$

then we may expect that the  $\sigma$ -method and  $s$ -method plans will pass nearly identical OC curve as follows

$$L_s(p) \simeq L_\sigma(p). \quad (9)$$

### 3. Design of Rectifying Inspection Plans When the Variance is Known

When all nonconforming units found in the rejected lots are replaced by conforming units in a rectifying inspection plan, the average total inspection (ATI) is given by

$$ATI_\sigma = n_\sigma L_\sigma(\bar{p}) + N[1 - L_\sigma(\bar{p})] = N - (N - n_\sigma) L_\sigma(\bar{p}), \quad (10)$$

where  $\bar{p}$  is the process average.

#### 3.1 LTPD Plans

When standard deviation  $\sigma$  is known, the design problem of LTPD plans is to minimize the average total inspection ATI defined by Eq. (10), subject to

$$L_\sigma(p_1) \leq \beta, \quad (11)$$

where  $\beta$  is a given consumer's risk. From the economic viewpoint, the sampling plans should be designed under the condition of  $L_\sigma(p_1) = \beta$ . In this case, the auxiliary variable  $\omega_1$  corresponding to  $L_\sigma(p_1) = \beta$  will be fixed at  $\omega_1 = z_{1-\beta}$  by using Eq.(2). Hence, based on Eqs.(3) ~ (5) and given  $(p_1, \beta)$ , we can obtain the following relations:

$$\bar{\omega} = z_{1-\beta} + \sqrt{n_\sigma} (z_{\bar{p}} - z_{p_1}), \quad (12)$$

$$k_\sigma = z_{p_1} - z_{1-\beta} / \sqrt{n_\sigma} \quad (13)$$

where  $\bar{\omega}$  is the auxiliary variable corresponding to the value of  $L_\sigma(\bar{p})$ . Then from Eq.(12), it can be found that the probability of acceptance is a function of one variable of sample size  $n_\sigma$ , so that  $ATI_\sigma$  is a function of  $n_\sigma$  too. Hence we can find the optimal sample size  $n_\sigma$  minimizing  $ATI_\sigma$  in a simple way. In fact, when  $n_\sigma$  is considered as a real number in the region of  $0 < n_\sigma \leq N$ , if  $\beta \geq 0.023$ , that is  $z_{1-\beta}^2 \leq 4$  and logically  $\bar{p} < p_1 < 0.5$ , we find

$$\frac{dATI_\sigma}{dn_\sigma} = \Phi(\bar{\omega}) - \frac{(N - n_\sigma)}{2\sqrt{n_\sigma}} \phi(\bar{\omega}) (z_{\bar{p}} - z_{p_1}) \quad (14)$$

and

$$\frac{d^2ATI_\sigma}{dn_\sigma^2} = \frac{N - n_\sigma}{4\sqrt{n_\sigma}} \left\{ (z_{\bar{p}} - z_{p_1} + \frac{z_{1-\beta}}{2\sqrt{n_\sigma}})^2 + \frac{4 - z_{1-\beta}^2}{4n_\sigma} + \frac{4}{N - n_\sigma} \right\} (z_{\bar{p}} - z_{p_1}) \phi(\bar{\omega}) > 0 \quad (15)$$

so that  $ATI_\sigma$  is a convex function of  $n_\sigma$ . In another way, we have:

$$\lim_{n_\sigma \rightarrow +0} \frac{dATI_\sigma}{dn_\sigma} = -\infty$$

$$\left. \frac{dATI_\sigma}{dn_\sigma} \right|_{n=N} > 0$$

then there is only one  $n_\sigma$  satisfying  $\frac{dATI_\sigma}{dn_\sigma} = 0$  in region of  $0 < n_\sigma < N$ . Thus, the optimal  $n_\sigma$  can be found easily and further acceptance coefficient  $k_\sigma$  will be calculated by using Eq.(13).

### 3.2 APQL Plans

If the fraction defective of the accepted lot is  $p$  and all defectives found in the rejected lots are replaced by nondefectives, the average outgoing quality (AOQ) will be

$$AOQ_{\sigma}(p) = pL_{\sigma}(p) \quad (16)$$

where  $p$  is defined by Eq.(4). The problem is to minimize  $ATI_{\sigma}$  subject to

$$\sup_p AOQ_{\sigma}(p) \leq AOQL. \quad (17)$$

Here we also consider the case of the equal condition of Eq.(17). A value of  $p^*$  is assumed and  $AOQ_{\sigma}$  at  $p^*$  will be

$$p^*L_{\sigma}(p^*) = AOQL. \quad (18)$$

Therefore  $p^*$  must be satisfied by

$$\left. \frac{dAOQ_{\sigma}}{dp} \right|_{p=p^*} = \Phi(z_{p^*})\Phi(\omega^*)/\sigma - \sqrt{n_{\sigma}} \phi(\omega^*) \{1 - \Phi(z_{p^*})\}/\sigma = 0. \quad (19)$$

From Eqs. (18) and (19)

$$\sqrt{n_{\sigma}} \phi(\omega^*) (p^*)^2 / \phi(z_{p^*}) = AOQL \quad (20)$$

and

$$\omega^* = z_{1-AOQL/p^*} \quad (21)$$

where  $\omega^*$  is the auxiliary variable corresponding to value of  $L_{\sigma}(p^*)$ . By using Eqs. (20) and (21)

$$n_{\sigma} = \frac{(AOQL)^2}{(p^*)^4} \exp \{z_{1-AOQL/p^*}^2 - (z_{p^*})^2\} \quad (22)$$

$$k_{\sigma} = z_{p^*} - z_{1-AOQL/p^*} / \sqrt{n_{\sigma}} \quad (23)$$

So that we can obtain  $p^*$  by using Eq.(22) at fixed  $n_{\sigma}$  with numerical methods, then calculate  $k_{\sigma}$  with Eq.(23) and calculate  $ATI_{\sigma}$  with Eq.(10). Following this process, we can get the optimal solution  $(n_{\sigma}, k_{\sigma})$  by using iterative computer programs in which  $ATI_{\sigma}$  values are compared.

#### 4. Design of Rectifying Inspection Plans When the Variance is Unknown

When  $\sigma$  is unknown, under the relationship between  $(n_{\sigma}, k_{\sigma})$  and  $(n_s, k_s)$  satisfying Eqs.(7) and (8), the average total inspection can be calculated approximately as

$$ATI_s = N - (N - n_s) L_s(\bar{p}) \simeq N - (N - n_s) L_{\sigma}(\bar{p}). \quad (24)$$

#### 4.1 LTPD plans

The problem of a LTPD plan is

$$\begin{aligned} \text{ATI}_s &= N - (N - n_s) L_\sigma(\bar{p}) \rightarrow \min \\ \text{subject to } L_s(p_1) &= \beta. \end{aligned} \quad (25)$$

From Eqs. (7), (8) and (13), the relationship between  $n_\sigma$  and  $n_s$  will be

$$a(\sqrt{n_\sigma})^2 + b\sqrt{n_\sigma} + c = 0 \quad (26)$$

where

$$a = 2(n_s - 1)(4n_s - 5)^2 + n_s(4n_s - 4)^2 z_{p_1} \quad (27)$$

$$b = -2n_s(4n_s - 4)^2 z_{p_1, z_{1-\beta}} \quad (28)$$

$$c = n_s(4n_s - 4)^2 z_{1-\beta} - 2(n_s - 1)(4n_s - 5)^2. \quad (29)$$

Generally in the case of  $p_1 < 0.5$  and  $\beta = 10\%$ , the negative solution of Eq.(26) is ignored and only one  $n_\sigma$  can be obtained from equation Eq.(26) at fixed  $n_s$ . Hence we can calculate  $k_\sigma$  by using Eq.(13) and then calculate  $\text{ATI}_s$  with Eq.(24) at this fixed  $n_s$ , so that we can find the optimal sample size  $n_s$  by comparing  $\text{ATI}_s$  with a computer program and then find  $k_s$  by Eq.(7).

#### 4.2 AOQL plans

When  $\sigma$  is unknown, the problem of an AOQL plan is

$$\begin{aligned} \text{ATI}_s &= N - (N - n_s) L_\sigma(\bar{p}) \rightarrow \min \\ \text{subject to } \sup_p \text{AOQ}_s(p) &= \text{AOQL}. \end{aligned} \quad (30)$$

Dealing with Hamaker's formulae Eqs.(7), (8) and Eqs.(22), (23), the relationship between  $p^*$  and  $n_s$  can be drawn out by eliminating the  $k_s$ ,  $n_\sigma$ ,  $k_\sigma$ , so that we can proceed in the same way to find  $p^*$  at fixed  $n_s$  with numerical methods and then obtain  $n_\sigma$ ,  $p_\sigma$  from Eqs.(22) and (23). Therefore  $L_\sigma(\bar{p})$  can be found by using Eqs. (2) and (5), then  $\text{ATI}_s$  can be calculated by Eq.(24). The problem will be solved with the help of iterative computer programs in which  $\text{ATI}_s$  values are compared too.

### 5. Examples

Table 1 gives some examples of LTPD variable plans compared with corresponding attribute plans and Table 2 gives those examples of AOQL plans. By using the information of normal distribution, the average total inspection by using variable plans can be economized in comparison with using attribute plans.

Table 1 Single Samling LTPD Plans by Variables with  $p_1 = 1\%$  and  $\beta = 10\%$

$\bar{p}(\%)$	$\bar{p} = 0.05\%$				Attribute plans $n, c, ATI$	$\bar{p} = 0.1\%$				Attribute plans $n, c, ATI$
	$n_\sigma$ $n_s$	$k_\sigma$ $k_s$	$ATI_\sigma$ $ATI_s$	$L_\sigma(\bar{p})$ $L_s(\bar{p})$		$n_\sigma$ $n_s$	$k_\sigma$ $k_s$	$ATI_\sigma$ $ATI_s$	$L_\sigma(\bar{p})$ $L_s(\bar{p})$	
500	16	2.647	18.43	0.9950	180,0,207.54	23	2.594	27.12	0.9914	180,0,232.71
	53	2.725	65.07	0.9730		70	2.665	89.30	0.9551	
1000	18	2.629	20.45	0.9975	205,0,282.45	26	2.578	30.38	0.9955	205,0,352.36
	62	2.690	74.14	0.9871		85	2.629	104.00	0.9792	
5000	22	2.600	24.98	0.9994	385,1,460.29	33	2.550	37.72	0.9991	530,2,604.94
	82	2.636	93.60	0.9976		117	2.580	135.17	0.9963	
10000	24	2.588	26.89	0.9997	530,2,554.11	36	2.540	40.81	0.9995	530,2,688.76
	90	2.620	101.58	0.9988		130	2.565	147.84	0.9982	

Table 2 Single Samling AOQL Plans by Variables with AOQL = 0.5%

$\bar{p}(\%)$	$\bar{p} = 0.05\%$				Attribute plans $n, c, ATI$	$\bar{p} = 0.1\%$				Attribute plans $n, c, ATI$
	$n_\sigma$ $n_s$	$k_\sigma$ $k_s$	$ATI_\sigma$ $ATI_s$	$L_\sigma(\bar{p})$ $L_s(\bar{p})$		$n_\sigma$ $n_s$	$k_\sigma$ $k_s$	$ATI_\sigma$ $ATI_s$	$L_\sigma(\bar{p})$ $L_s(\bar{p})$	
500	8	2.332	9.65	0.9967	70,0,84.80	11	2.327	13.77	0.9943	70,0,99.07
	23	2.377	28.53	0.9884		29	2.356	38.09	0.9807	
1000	9	2.329	10.94	0.9980	70,0,63.36	13	2.328	15.94	0.9970	70,0,95.60
	27	2.361	32.90	0.9939		35	2.346	45.28	0.9893	
5000	12	2.327	14.10	0.9996	165,1,180.58	18	2.335	21.34	0.9993	165,1,224.01
	37	2.344	43.79	0.9986		52	2.340	63.79	0.9976	
10000	13	2.328	15.58	0.9997	165,1,196.68	20	2.338	23.82	0.9996	270,2,296.10
	41	2.341	48.80	0.9992		60	2.341	72.47	0.9987	

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