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Bayesian Group Chain Sampling Plan Based on Beta Binomial Distribution through Quality Region

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Abstract— **In this article, we introduced Bayesian Group Chain Sampling Plan (BGChSP) using different combination of parameters. In acceptance sampling plan, the random fluctuations can be describe in the selection of distribution Bayesian approach which is based on prior process history. We apply beta distribution as a suitable prior distribution. By considering consumer's and producer's risks, we consider Probabilistic and Indifference Quality Regions for the specified AQL and LQL. For the selection of parameters in BGChSP, Maximum Allowable Percent Defectives (MAPD) is also considered.**

Keywords— *acceptance sampling plan, binomial, beta distribution, quality region, consumer's risks, producer's risk, acceptance quality level, limiting quality level.*

1. Introduction

Bayesian acceptance sampling approach is based on the combination of lot information and the prior information for the selection of distribution. To describe the random variation, Bayesian approach required to specify from lot to lot a distribution of defectives. Prior distribution is the expected distribution of a lot quality, that is going for inspection. This distribution is formulated before taking the sample, so it is called prior. The empirical knowledge is based on sample under study is called sample distribution or data distribution. The combination of prior and empirical information's leads to take decision about lot.

For Bayesian sampling inspection a statistical model considers the following three components:

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- 1. The prior distribution must according to quality of the submitted lot.
- 2. Sampling inspection cost on acceptance and rejection.
- 3. On the base of mean rejection, a class is designed in the sampling plans, to give acceptance protection against a poor quality lot.

The sampling plans based on economic consider different factors to design a cost effect plan: like cost of inspection rejecting a conforming product and accepting a non-conforming product. The history of similar lots that already submitted for inspection are count in to the Bayesian sampling plan. Non-Bayesian sampling methodology does not based on past history.

There is tough competition in industry by rapidly increasing in the needs of statistical and analytical techniques towards the improvement of product quality. This study is related to BGChSP by using a novel approach called quality region or quality interval sampling (QIS). Instead of point this plan is based on quality range. This plan delivers decision rules of acceptance for both supplier and customer to meet the present quality condition of the product. Improvement in the technology is rapidly increasing with the passage of time and supplier needs high quality products with low defective fraction. Unfortunately, in some particular situation traditional methods can not detect out defect in the product. QIS was introduced to overcome such problems. By involving QIS, this article designs the parameters for the plan indexed with quality region.

For inspection Chain sampling plan was introduced by Dodge [1]. Under an assumption that cost is linear in p that is fraction of defective; Hald [2] provide a system attribute single sampling plan obtain by minimizing average cost. By using gamma

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prior a Bayesian chain sampling plan was discussed by Latha and Suresh [3] for the construction and performance measure. By using Beta prior, a Bayesian double sampling plan was discussed by Latha and Arivazhagan [4] for the construction and performance measure through quality region.

Designs the plan indexed parameters in this study are α, β, LQL, AQL, IQL, IQR and PQR for the specified value of s, r and g. Also, with the numerical illustrations of prior distribution parameters are provided.

2. Group Chain Sampling Plan (GChSP)

Aslam, Mughal, Ahmad and Yab [5] designed Group Acceptance Sampling Plan (GASP) for truncated life tests. Here the product life time was following pareto distribution of second kind. Later for a family of Pareto distribution, an efficient GASP was introduced by Mughal and Aslam [6], they assumed the total number of defectives as groups. The number of defectives in the proposed plan were recorded on the base of all groups instead of an individual group. Later, for a family of Pareto distribution Mughal and Ismail [7], proposed an economic reliability acceptance sampling plan by using an efficient group sampling technique. By satisfying design parameters for a given group and acceptance number a minimum termination time required for a proposed plan. Using group acceptance sampling plan for Pareto distribution of 2nd kind. An economic reliability group acceptance sampling plan was developed by Mughal, Zain and Aziz [8]. For the biased data theory to find the required design parameters they were used Poisson and weighted Poisson distributions. It was proved that the proposed plans required minimum testing time.

Mughal, Zain and Aziz [9], developed a GChSP plan for the lifetime of a product follow Pareto distribution of 2nd kind. To satisfying pre-assumed design parameters at several quality levels probability of lot acceptance was obtained. Mughal [10] extended and proposed a generalized GChSP on the base of sampling plan developed by Mughal, Zain and Aziz [9]. By considering several values of the proportion of defectives minimum sample size and probability of lot acceptance were found to satisfying pre-specified consumer's risk.

3. Methodology

3.1 Operational Procedure

The operating procedure of GChSP is as follow,

- Select minimum number of groups for each lot and each group contain *r* items, so that $n=r*g$ be the required sample size.
- The lot is accepted if *d=0* and rejected if *d > 1*
- If $d = 1$ accept the lot, if no defectives are found in immediately preceding *i* sample of size *g*r*

Binomial distribution is applied in order to achieve the probability of lot acceptance for zero and one defective products. Here the binomial distribution is applicable because the product fulfils all four properties of binomial experiment. This is applicable, when lot consist on identical and independent trails, the inspection outcomes are categorized into two mutually exclusive and independent outcomes. So, the probability of lot acceptance can be written as:

$$
L(p) = \sum_{c=0}^{1} {r * g \choose c} p^c (1-p)^{r * g - c}
$$
 (1)

where *p* is the proportion of defective.

By solving Equation (1) for zero and one defective product, each probability of lot acceptance is:

$$
P_0 = (1 - p)^{r \ast g} \tag{2}
$$

$$
P_1 = (r * g)p(1-p)^{r * g-1}
$$
\n(3)

The operating characteristic function of ChSP-1 has given by Dodge [1].

$$
P_a(p) = P_0 + P_1 P_0^i \tag{4}
$$

Mughal [10] suggest probability of lot acceptance for (GChSP) after plugin equation (2) and (3) in (4) is:

$$
L(p) = (1 - p)^{r * g}
$$

+ $(r * g)p(1 - p)^{r * g - 1}(1 - p)^{r * g * i}$ (5)

where p represents the probability of defective in the lot of *g*r*.

The general expression of probability of lot acceptance in GChSP by considering Binomial distribution we can rewrite (5) the binomial model of OC function as:

$$
L(p) = (1 - p)^{r * g} + (r * g)p(1 - p)^{r * g(1+i)-1}
$$
 (6)

Let the prior distribution of the process is beta distribution. That stat *p* process average follow beta prior distribution with *s* and *t* both are shape parameters and the probability distribution function PDF:

$$
f(p) = \frac{1}{\beta(s,t)} p^{s-1} (1-p)^{t-1}
$$
 (7)

 $0 < p < 1, s, t > 0$ and $q = 1 - p$ where $\mu = \frac{s}{s}$ $\frac{s}{s+t}$, under the proposed Sampling Plan.

On the base of beta binomial distribution in GChSP, probability of acceptance is as follows:

$$
\bar{P} = \int_{0}^{1} L(p)f(p) \, dp \tag{8}
$$

After replacing (6) and (7) in equation (8) we get:

$$
\bar{P} = \int_{0}^{1} \left((1-p)^{r*g} + (r*g)p(1-p)^{r*g(1+i)-1} \right)
$$

$$
* \frac{1}{\beta(s,t)} p^{s-1} (1-p)^{t-1} dp
$$
 (9)

$$
\bar{P} = \frac{1}{\beta(s,t)} [\beta(s,r * g + t) + (r * g)\beta(s+1,r * g(1+i) + t
$$
\n⁽¹⁰⁾\n⁽¹⁰⁾

$$
\bar{P} = \frac{\Gamma(s+t)\Gamma(r*g+t)}{\Gamma t \Gamma(s+r*g+t)} + (r*g) \n\frac{s\Gamma(s+t)\Gamma(r*g(1+i)+t-1)}{\Gamma t \Gamma(s+r*g(1+i)+t)}
$$
\n(11)

Equation (11) is the mixed distribution of beta binomial distribution.

The above equation can be rewritten as by reducing \overline{P} and μ_0 be the point of control:

For $s = 1$, after simplification Equation (11) is:

$$
\bar{P} = \frac{1 - \mu}{rg\mu + 1 - \mu} + \frac{rg\mu(1 - \mu)}{(rg\mu(1 + i) + 1 - \mu)(rg\mu(1 + i) + 1 - 2\mu)}
$$
(12)

For $s = 2$, after simplification Equation (11) is:

$$
\bar{P} = \frac{(2 - \mu)(2 - 2\mu)}{(rg\mu + 2 - \mu)(rg\mu + 2 - 2\mu)} + \frac{2rg\mu(2 - \mu)(2 - 2\mu)}{(rg\mu(1 + i) + 2 - \mu)(rg\mu(1 + i) + 2 - 2\mu)(rg\mu(1 + i) + 2 - 3\mu)}
$$
(13)

For $s = 3$, after simplification Equation (11) is:

$$
\bar{P} = \frac{(3 - \mu)(3 - 2\mu)(3 - 3\mu)}{(rg\mu + 3 - \mu)(rg\mu + 3 - 2\mu)(rg\mu + 3 - 3\mu)}
$$
\n
$$
+ \frac{3rg\mu(3 - \mu)(3 - 2\mu)(3 - 3\mu)}{(rg\mu(1 + i) + 3 - \mu)(rg\mu(1 + i) + 3 - 2\mu)(rg\mu(1 + i) + 3 - 3\mu)(rg\mu(1 + i) + 3 - 4\mu)}
$$
\n(14)

3.2 Constructing Plans for given AQL, LQL, α and β

For the selection of (BGChSP) table 1 and 2 are used for specified AQL, LQL, α and β by using following steps:

- 1. First calculate the operating ratio μ_1/μ_2 to construct a plan for the given AQL $(1-\alpha)$ and LQL $(1-\beta)$.
- 2. Locate table value of operating ratio which is equal or just less than the desired operating ratio in the column desired α, β for fixed value of r and g.
- 3. The value of s, r and g can be obtained for the corresponding located value of operating ratio.

Example 1: For $\bar{P} = 0.50$ s=1, g= 1 and r=3 the corresponding indifference Quality level (IQL) is $\mu_0 = 0.2980$ and for s=3, g= 3, r=3 the corresponding value of AQL is $\mu_1 = 0.0166$ and LQL value $\mu_2 = 0.3172$.

From the above equation in Table 1 average probability of acceptance for the given variation, by using newton's approximation the average product quality level μ is obtained. In Example 1 we can see that the average product quality decreased as the values of s and g increased.

Example 2: Let the values of μ_1 and μ_2 are assumed to be 0.008 and 0.90 respectively, then the operating ratio is 112.5. Now the value from table 2 is approximately equal to this calculated operating ratio is 111.6 and the corresponding parametric

values are $s = 1$, $g = 2$, $r = 3$, $i = 2$ and $\mu_1 = 0.0085$, $\mu_2 = 0.9486$ at $(\alpha = 0.01 \text{ and } \beta = 0.01)$.

Table 1. For specified values of $P(\mu)$ Certain μ values in BGChSP

3.3 Construction of Quality interval for (BGChSP)

Probabilistic Quality Region (PQR)

In this interval of quality product is accepted with minimum and maximum probabilities ($\mu_1 < \mu <$ μ_2) respectively 0.10 and 0.95. Here $d_2 = \mu_2 - \mu_1$ denotes the range of probability that derived from equation of the average probability of acceptance:

 $\bar{P}(\mu_1 < \mu < \mu_2) = \frac{\Gamma(s+t)\Gamma(r*g+t)}{\Gamma(t)(s+rs,g+t)}$ $\frac{\Gamma t \Gamma(s + r * g + t)}{\Gamma t \Gamma(s + r * g + t)} + (r * g)$ $s\Gamma(s + t)\Gamma(r * g(1 + i) + t - 1)$ $\Gamma t \Gamma(s + r * g(1 + i) + t)$

Beta distribution is the prior distribution so mean of beta $\mu = \frac{s}{s}$ $\frac{s}{s+t}$ be approximately mean value of the product quality.

Indifference Quality Region (IQR)

In this interval of quality product is accepted with minimum and maximum probabilities ($\mu_1 < \mu <$ μ_0) respectively 0.50 and 0.95. Here $d_0 = \mu_0 - \mu_1$ denotes the range of probability that derived from equation of the average probability of acceptance:

$$
\bar{P}(\mu_1 < \mu < \mu_0) = \frac{\Gamma(s+t)\Gamma(r * g + t)}{\Gamma t \Gamma(s + r * g + t)} + (r * g)
$$
\n
$$
\frac{s\Gamma(s+t)\Gamma(r * g(1+i) + t - 1)}{\Gamma t \Gamma(s + r * g(1+i) + t)}
$$

Where $\mu = \frac{s}{s}$ $\frac{s}{s+t}$ be the approximate mean of product quality.

S	g	r	i	μ_1	μ_0	μ_2	d_2	d_0	T	μ_2/μ_1
$\mathbf{1}$	$\mathbf{1}$	$\overline{2}$	$\mathbf{1}$	0.0843	0.4416	0.86	0.7757	0.3573	2.171	10.2017
		3	2	0.044	0.298	0.7741	0.7301	0.254	2.8744	17.5932
		4	3	0.0287	0.2257	0.708	0.6793	0.197	3.4482	24.669
	$\overline{2}$	2	1	0.0421	0.2751	0.7474	0.7053	0.233	3.027	17.753
		3	2	0.0222	0.1736	0.6298	0.6076	0.1514	4.0132	28.3694
		4	3	0.0144	0.1268	0.5474	0.533	0.1124	4.742	38.0139
	3	2	1	0.0281	0.2	0.6613	0.6332	0.1719	3.6835	23.5338
		3	2	0.0148	0.1225	0.5309	0.5161	0.1077	4.792	35.8716
		4	3	0.0096	0.0882	0.4462	0.4366	0.0786	5.5547	46.4792
	4	2	1	0.0211	0.1571	0.593	0.5719	0.136	4.2051	28.1043
		3	2	0.0111	0.0947	0.4589	0.4478	0.0836	5.3565	41.3423
		4	3	0.0072	0.0676	0.3766	0.3694	0.0604	6.1159	52.3056
$\overline{2}$	$\mathbf{1}$	$\overline{2}$	1	0.0928	0.4322	0.8281	0.7353	0.3394	2.1665	8.9235
		3	$\overline{2}$	0.0479	0.2748	0.69	0.6421	0.2269	2.8299	14.405
		4	3	0.0309	0.2017	0.5932	0.5623	0.1708	3.2922	19.1974
	$\overline{2}$	2	1	0.046	0.2527	0.6433	0.5973	0.2067	2.8897	13.9848
		3	2	0.024	0.1525	0.4835	0.4595	0.1285	3.5759	20.1458
		4	3	0.0155	0.1089	0.3909	0.3754	0.0934	4.0193	25.2194
	3	2	1	0.0306	0.1784	0.5191	0.4885	0.1478	3.3051	16.9641
		3	2	0.016	0.1054	0.3697	0.3537	0.0894	3.9564	23.1063
		4	3	0.0104	0.0746	0.2903	0.2799	0.0642	4.3598	27.9135
	4	2	1	0.0229	0.1378	0.4337	0.4108	0.1149	3.5753	18.9389
		3	2	0.012	0.0806	0.2988	0.2868	0.0686	4.1808	24.9
		4	3	0.0078	0.0567	0.2307	0.2229	0.0489	4.5583	29.5769
3	$\mathbf{1}$	$\overline{2}$	$\mathbf{1}$	0.0966	0.4286	0.8071	0.7105	0.332	2.1401	8.3551
		3	2	0.0496	0.2666	0.6467	0.5971	0.217	2.7516	13.0383
		4	3	0.032	0.1934	0.5429	0.5109	0.1614	3.1654	16.9656

Table 3. For specified values of s, g, r and *i* values of PQR and IQR, μ_2/μ_1

3.4 Selection of the Sampling Plans

For difference values of s, g and r in Table 3 the operating ratio T is given, i.e. $T = \frac{\mu_2 - \mu_1}{\mu_2 - \mu_2}$ $\frac{\mu_2 - \mu_1}{\mu_0 - \mu_1} = \frac{d_2}{d_0}$ $rac{a_2}{d_0}$ where $d_2 = \mu_2 - \mu_1$ and $d_0 = \mu_0 - \mu_1$ is used to characterize the sampling plan. Operating ratio $T =$ d_2 $\frac{a_2}{d_0}$ can be find for any given values of PQR (d_2) and IQR (d_0) . Find the value that is equal to or just less than the specified ratio, "in Table 3 under the column of T" corresponding to s, g and r values are noted. For (BGChSP) the parameters can be determine from this operating ratio.

The above equation is used in the same way to the average probability of acceptance 0.95 AQL (μ_1) and 0.10 IQL(μ_2) are obtained μ_2/μ_1 in Table 3.

Example 3: Given that $\mu_1 = 0.01$ to compute the value of T, first compute PQR and IQR. From Table 3 select the respective value. The nearest values of PQR and IQR corresponding to $s = 3$, $g = 3$, $r =$ 4, $i = 3$ and $\mu_1 = 0.0107$ are that $d_2 = 0.2348$ and that $d_0 = 0.0597$, then that $T = 3.933$. Hence through quality interval the required plan has parameters $s = 3$, $q = 3$, $r = 4$ and $i = 3$.

4. Conclusion

Bayesian acceptance sampling deals with the procedure to make acceptance decision about process or lot based on the combination of sample information and past history lot. There are many options to determine an appropriate sampling plan. There are sampling plans that consider the risk of producers and consumers and some plans are based on non-economic requirements. In the situation of unavailability of a prior distribution cost function to reduce the inspection sample size, Bayesian sampling attribute plan can be used. This article presented work is mainly related technique to proposed BGChSP for acceptance quality level, indifference quality levels, indifference and probabilistic quality region, producer's and consumer's risks.

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