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DEVELOPING DOUBLE SAMPLING PLANS FOR ATTRIBUTES TO MEET SAMPLE SIZE CRITERIA

Research Report No. 84-32

by

R.W. Rangarajan K.B. Beitler and R.S. Leavenworth

RESEARCH REPORT

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NAVAL RESEARCH LABORATORY

Industrial & Systems Engineering Department Liniversity of Florida Gainesule, FL 325U DEVELOPING DOUBLE SAMPLING PLANS FOR ATTRIBUTES TO MEET SAMPLE SIZE CRITERIA

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R.W. Rangarajan K.B. Beitler and R.S. Leavenworth

APROVED FOR PUBLIC RELEASE DIST. NLIMITED

September, 1984

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acceptable quality level ABSTRACT

This study reports on the development of a FORTRAN IV program to produce double sampling acceptance plans for attributes data. The plans must satisfy $operating characteristic of the application of the plane of the point (p_1, 1-q) and the plane of the sample for the$ $(p_2, A). Two models are given. MODEL I has an additional constraint that$ the maximum value of the ASN must be less than or equal to the sample size fora corresponding single sampling plan. MODEL II relieves this constraint. Ineither case, the resulting plan has a minimum ASN evaluated at the quality $level p_1 among all sampling plans satisfying the constraints.$

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INTRODUCTION

Frequently double sampling plans are employed in lieu of single sampling plans for lot acceptance by attributes especially when lot or batch sizes are large. A number of systems of sampling plans are available the most recognized of which at least in the United States, are the MIL-STD-105D AQL systems and the Dodge-Romig LTD and AQQL systems.

In addition methods have been reported for developing tailored double sampling plans usually based on the specification of two points on an OC curve (the likelihood function). Two such procedures, largely taken from a Chemical Corps Engineering Agency (1953) publication, are contained in Tables 8-2 and 8-3 of Duncan (1974). These tables, based on the Poisson distribution, allow the tailoring of a plan to a Producer's Risk (α) of 0.05 at a designated quality level p_1 and a Consumer's Risk of 0.10 at a designated level p_2 . Plans may be found for which n_2 , the second sample size equals $2n_1$ and where n_2 equals n_1 . In either case the rejection number (cummulative) on both samples is taken to be the acceptance number on the second sample (cummulative), c_2 , plus one.

In 1969 Guenther, following on some earlier work by Cameron (1952), developed what amounted to a brute-force algorithm for finding single sampling acceptance plans to satisfy two points on the likelihood function. The main difference between these two procedures is that while Cameron's procedure assured a plan with risk levels as close as possible to those designated for the quality levels p_1 and p_2 , Guenther's procedure assures risk levels at least as good as those stipulated. In addition, Guenther's procedure allows the use of the hypergeometric, binomial or Poisson distributions assuming adequate tables are available.

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In 1970 Guenther extended his work to double sampling plans. Again the algorithm is essentially brute force employing probability tables extensively. Unlike the Chemical Corps tables, however, the fixed relationship between n_1 and n_2 is not required nor does the Poisson have to be used. The disadvantage of his procedure is the laborious effort required if a plan is to be developed by hand calculation. The algorithms, however, are sufficiently simple to be programmed easily for the computer. Hailey (1980) provides an ANSI Standard FORTRAN program based on Guenther's algorithm which finds the minimum sample size single sampling plan.

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In this study, a program is developed for finding double sampling plans and characteristics of the average sample size functions of the plans found are compared. The basic algorithm follows that of Guenther. However, an objective function is introduced as described in the following paragraphs.

PROBLEM FORMULATION

When plotted as a function of p, the likelihood function L(p) of a sampling plan provides the operating characteristic (OC) curve for the plan. Using the binomial distribution as an example, L(p) for a double sampling plan is:

$$L(p) = \int_{d_1}^{c_1} (\begin{pmatrix} n_1 \\ d_1 \end{pmatrix} p^{d_1} (1-p)^{n_1-d_1} + \int_{d_1=c_1+1}^{c_1-1} (\int_{d_2=0}^{c_2-d_1} (\begin{pmatrix} n_1 \\ d_1 \end{pmatrix} (\int_{d_2}^{a_2} p^{d_1+d_2} (1-p)^{n_1+n_2-d_1-d_2}$$
(1)

where:

- p = incoming proportion of nonconforming units
- n₁ = first sample size
- n_2 = second sample size
- c_1 = acceptance number on first sample
- r_1 = rejection number on first sample

 c_2 = acceptance number on second sample (cumulative for

 n_1+n_2). The rejection number on the second sample, r_2 , is c_2+1 .

In this study, it is assumed that $r_1 = r_2 = c_2 + 1$. Thus the upper limit of the first summation, $(r_1 - 1)$ in the double summation portion of L(p) (the probability of acceptance on the second sample), may be replaced by c_2 . By so doing a double sampling acceptance plan is fully described by the plan parameters n_1 , n_2 , c_1 , and c_2 .

Discussion of the algorithm for finding double sampling plans will break L(p) into its two parts, the probability of acceptance on the first sample, $P_a(n_1)$, and the probability of acceptance on the second sample, $Pa(n_2)$. Thus:

 $L(p) = Pa(n_1) + Pa(n_2)$

where for the binomial case:

$$Pa(n_{1}) = B(c_{1} | n_{1}, p) = \sum_{d_{1}=0}^{c_{1}} {n_{1} \choose d_{1}} p^{d_{1}} (1-p)^{n_{1}-d_{1}}$$

$$Pa(n_{2}) = BB(c_{1}, c_{2} | n_{1}, n_{2}, p)$$

$$= \sum_{d_{1}=c_{1}+1}^{c_{2}} \sum_{d_{2}=0}^{c_{2}-d_{1}} {n_{1} \choose d_{1}} {n_{2} \choose d_{2}} p^{d_{1}+d_{2}} (1-p)^{n_{1}+n_{2}-d_{1}-d_{2}}$$

If two design points are designated on the likelihood function, ideally a single (n_1, n_2, c_1, c_2) set can be found yielding an OC curve which passes exactly through these points. This would suggest setting the likelihood function for each quality level equal to its respective probabilities and solving for a set (or several sets) which exactly satisfy the equations. However, since n_1 , n_2 , c_1 and c_2 all must be integer, it is doubtful that any set can be found which gives exact equality for both equations.

In recongnition of this fact the Cameron procedure for single sampling plans selects a plan which is as close as possible to the OC curve at the two points. The Guenther procedure rests on the formulation of inequality constraints that guarantee risk levels <u>at least</u> as good as those stipulated. It is the Guenther procedure which is used in this study.

The OC curve points selected are:

- p_1 = An Acceptable Quality Level (AQL), following the definition in MIL-STD-105D, which should be accepted with a probability of at least 1- α , α being the Type I error risk.
- $p_2 = A$ Rejectable (poor) Quality Level (RQL) which should be accepted with no more than a low probability β (Type II error risk).

These two design parameters therefore may be expressed as:

$$(p_1, 1-\alpha)$$
 and (p_2, β) .

The resulting constraint equations are:

$$L(p_1) > 1-\alpha \tag{2}$$

 $L(p_2) \leq \beta \tag{3}$

Actually an infinite number of double sampling plans may be found which will satisfy these two inequalities. Thus some measure of performance must be specified in order to choose among them. The measure chosen in this study was to minimize the ASN when the lots are at the AOL, p_1 .

The ASN function for a double sampling plan is:

$$ASN = n_1 + n_2 P(n_2)$$

where:

 $P(n_2)$ is the probability of taking the second sample. In binomial form,

ASN =
$$n_1 + n_2 \sum_{d_1 = c_1 + 1}^{c_2} {\binom{n_1}{d_1} p^{d_1} (1 - p)^{n_1 - d_1}}$$

= $n_1 + n_2 [B(c_2|n_1, p) - B(c_1|n_1, p)]$

Thus an objective function was introduced as follows:

Select $(n_1 n_2, c_1, c_2)$ to:

minimize
$$ASN(p_1) = n_1 + n_2 [B(c_2|n_1, p_1) - B(c_1|n_1, p_1)]$$
 (4)

Subject to equation (2) and (3).

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Two models were formulated on the basis of equations (2), (3) and (4). Model I introduced an additional constraint. along the lines of MIL-STD-105D, guaranteeing that the maximum value of the ASN of the double sampling plan, ASNMAX, does not exceed the sample size n_s of the minimum n and c for a single sampling plan satisfying equations (2) and (3). This value of n_s is designated n_s^* .

In order to study the effect that the ASNMAX constraint had on the value of the objective function, $ASN(p_1)$, the constraint was removed in Model II and Model II was run using the same design parameters. In this way one can assess, in terms of average number of items inspected, the penalty paid by requiring that the ASNMAX not exceed n_s^* .

DEVELOPMENT OF THE ALGORITHM

The first step in developing a double sampling plan is to find the minimum single sampling plan (n_s^*, c^*) satisfying equation (2) and (3) where, for example:

$$L(p) = \sum_{d=0}^{c} {n \choose d} p^{d} (1-p)^{n-d}$$
.

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For any fixed value of c, there is some maximum value of n which will satisfy equation (2). If n is increased beyond this value, designated n_u for an upper bound, equation (2) will be violated. Correspondingly, there is a minimum value of n, designated n_ℓ , which just satisfies equation (3). Any value of n less than n_ℓ will produce a value of L(p2) greater than β .

Beginning with c equals 0, n_{g} and n_{u} are found by incrementing n one unit at a time. If the resulting values of n_{u} is not greater than or equal to n_{g} , c is increased by one unit and the procedure repeated. Eventually at some value of c, $n_{u} > n_{g}$ is satisfied, the current value of c becomes c* and n_{s} * is set equal to n_{g} . Thus the minimum single sampling plan satisfying equations (2) and (3) is found. n_{s} * becomes an upper bound on n_{1} for the double sampling plan and c* becomes a lower bound on c_{2} where c_{1} must be less than c_{2} .

The logic behind these limits is that, as c_1 approaches c^* . n_1 will approach n_s^* . If c_1 equals c^* . n_1 will equal n_s^* . n_2 will approach zero and c_2 will equal c_1 . Thus any derived double sampling plan will degenerate to the corresponding single sampling plan.

The algorithm for deriving the double sampling plan employs equations (2) and (3) replacing the L(p) function with equation (1) or its Poisson or hypergeometric equivalents. This study concentrates on the binomial of equation (1) only. Initially c_2 is set equal to c^* and c_1 is varied from 0 to c_2 -1. On each iteration of c_1 , n_1 is incremented from c_1 +1 until the equation

$$B(c_1|n_1, p_2) < \beta$$

is just satisfied. This value of n_1 becomes $n_{1\ell}$, the lower feasible limit on n_1 .

The algorithm then switches to the complete likelihood function, equation (1), searching to satisfy

$$L(p_1) > 1-\alpha$$
$$L(p_2) < \beta.$$

The computational procedure incorporates an integer bisection method in order increase efficiency.

An upper bound on n_1 , n_{1u} , is obtained for the current value of c_1 from:

with the constraint:

n₁e < n₁u.

The algorithm then switches to the complete likelihood function, equation (1), searching to satisfy:

$$n_{2\ell}: B(c_1|n_1, p_2) + BB(c_1, c_2|n_1, n_2, p_2) \le R$$

$$n_{2u}: B(c_1|n_1, p_1) + BB(c_1, c_2|n_1, n_2, p_1) > 1-\alpha$$
with $n_{2\ell} \le n_2 \le n_{2u}$.

The lowest possible value for n_2 is $n_5^* - n_1$. Since it is necessary to set an upper bound on n_2 , n_{2u} , in order to use a bisection search procedure, an arbitrary value of 1.5 times n_1 was used. This value worked successfully in all cases examined herein.

Once the bounds for n_2 have been set, the values of the maximum ASN (ASNMAX) and of the ASN evaluated at p_1 (ASN(p_1)) are calcuated for each feasible value of n_2 . These values along with n_1 , n_2 s and n_{2l} form the output of the computer program.

ANALYSIS AND CONDITIONS FOR OPTIMALITY

The double sampling plan parameters which may be varied are n_1 , n_2 , c_1 , c_2 , and r_1 , five in all. Once minimum values for c_1 and c_2 have been established minimally satisfying the $(p_1, 1-\alpha)$ and (p_2, β) points on the OC curve, a number of conbinations of n_1 and n_2 values also will satisfy these constraints. Furthermore, it is possible to find feasible ranges of n_1 and n_2 conbinations for any values of c_1 and c_2 greater than the minimum (c_1, c_2) combination. Thus an infinite number of plans may be found satisfying the two OC curve points. They repesent, in effect, the solution of a pair of algebraic equations in five unknowns.

By introducing the MAXASN constraint, the number of feasible solutions is reduced as it is if the Dodge-Romig scheme of setting $r_1 = r_2 = c_2 + 1$ is employed. It is the introduction of the objective of minimizing the ASN at the AQL, designated p_1 , that makes it possible to select a single plan and allow the computer program to terminate. In order to accomplish this, the performance of the ASN (p_1) was analyzed by varying individually each of the main plan parameters n_1 , n_2 , c_1 , and c_2 with r_1 set equal to $r_2 = c_2 + 1$.

Examples of this procedure are presented in Appendix I, SAMPLE PROGRAM RUNS. Values entered in each block (cell) are $ASN(p_1)$, ASNMAX, n_1 and n_2 . Each block represents the results for a c_1 , c_2 combination in the feasible region.

Some General Constraints

The value of c_1 does not exceed the c* of the single sampling plan with the same OC curve. If c_1 equals c*, the double sampling plan will degenerate to the single sampling plan with $n_1 = ns^*$, $n_2=0$, and $c_2=c_1$. Additionally, it is obvious from the ASN formula that, if c_1 is greater than c*, the ASN(p_1) will be greater than ns* as will the ASNMAX. Hence the search on c_1 may be truncated at c*-1.

Relationship of α , β , p_1 , and p_2

For fixed α and β , the acceptance number of a single sampling plan will remain constant for a constant discrimination ratio, D, defined as the ratio p2/p1. As p1 increases, the sample size of a single sampling plan decreases considerably. This can be explained by the fact that it is the absolute difference (p2-p1) that influences n, not the discrimination ratio. For example, for p1=0.02 and p2=0.10 the single sample size is approximately 38

with an acceptance number of 3. For p1=0.001 and p2=0.005, the same discrimination ratio, the single sample size is approximately 1,350 with an acceptance number of 3. This confirms that if (p2-p1) is small, n will be large and if (p2-p1) is large n will be small. Furthermore as the discrimination ratio decreases to 1.5, the acceptance number increases rather dramatically to 52. This result is verified by use of the Poisson approximation to the binomial and the methodology used to solve for single sampling plans therewith. See Cameron (1952).

Larger values of c1 result in a smaller range between n1s and n11. As c1 increases for a specific value of c2, the lower bound on n1 increases. This in turn reduces the number of double sampling plans computed in each cell because the upper bound on n1, n1u, is dependent on c2, not on c1.

Effect of ASNMAX Constraint, MODEL I

When the constraint ASNMAX <ns* is imposed, the values of ASN(p1) in the region where c2 exceeds c* become infeasible. An advantage of this property is that the search routine to locate the global minimum does not need to search the region where c2 exceeds c*. However, MODEL II, in which the ASNMAX constraint is not imposed, requires the evaluation of columns for c2 > c*. It is practically important to study the difference in minimum ASN(p1) in each case until a global minimum has been found. Thus, MODEL II searches for column minimums and selects the global minimum from that group. MODEL I needs only to look at the values for c2=c*.

Behavior of $ASN(p_1)$ as a Function of n_1

Plotting of the values of $ASN(p_1)$ as a function of n_1 showed that it is a quasi-convex function of n_1 . (Integerization of n_1 and n_2 may explain why the

results were not purely convex) Thus a search for the minimum $ASN(p_1)$ needs only to continue one step beyond the point at which the minimum exists.

Behavior of ASN (p_1) as a Function of c_1 .

In general, the $ASN(p_1)$ proved to be a quasi-convex function of c_1 .

For a constant discrimination ratio, (D = p2/p1), MODEL I results showed that the minimum ASN(P1) occurs at the same value of c1 irrespective of the value of p1. Such is not the case with MODEL II.

The values of c_1 where the minimum ASN(p1) occurs is influenced by the ratio p2/p1. As the ratio decreases, the value of c1 increases. However this relationship also is affected by the magnitude of (p2-p1).

Table I.

| DESCRIMINATION RATIO (P2/P1) | VALUE of c1 AT min.ASN(p1) of column.(c2=c*) |
|---------------------------------|--|
| 25 | 0 |
| 10 | 0 or 1 |
| 5 | 0 or 1 |
| { 4 | 1 |
| 3 | 2 |
| 2.5 | 2 or 3 |

Table I indicates that as the ratio, D, decrease. the minimum ASN(p1) tends to increase. That is, the corresponding value of cl becomes larger. As the ratio increases curves of $ASN(p_1)$ shift and become truncated constantly increasing from the first feasible solution rather than moving downward to a minimum before sweeping upward. In MODEL II, as c2 increases, the value cl for the minimum ASN(p1) of each column may vary.

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Behavior of $ASN(p_1)$ as a Function of c_2

Plotting of the $ASN(p_1)$ values as a function of c_2 yielded quasi-convex results as well. However, there were substantially different results under the two models. When MODEL I (constrained ASNMAX) was employed, the minimum $ASN(p_1)$ value occured always for values of c_2 equal to c*. Under MODEL II, this was the case occassionally but not always.

Impact of the ASNMAX Constraint

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The main objective of the analysis of MODEL II was to evaluate the effect of the constraint ASNMAX<ns* on the objective function.

Table II shows the minimum ASN(p1) obtained from both models together with their ASNMAX values for some representative cases.

| VALUE OF p1 | VALUE OF p2 | RATIO OF p2/p1 | MODEL ASN(P1) | I ASN MAX | MODEL ASN(P1) | II ASN MAX | ℃ OF REDUCTION |
|-------------------|-------------------|----------------------|------------------|-----------------|------------------|------------------|----------------------|
| 0.001 | 0.04 | 4 | 179.8 | 185. | 149.1 | 232. | 17.1 |
| 0.02 | 0.8 | 4 | 75.4 | 86.5 | 73.2 | 115. | 2.23 |
| 0.005 | 0.125 | 25 | 19.9 | 27.1 | 19.9 | 27.1 | 0 |
| 0.005 | 0.05 | 10 | 67.1 | 98.9 | 67.1 | 98.9 | 0 |
| 0.04 | 0.2 | 5 | 21.6 | 31.5 | 21.6 | 31.5 | 0 |

TABLE II

As indicated, a reduction in ASN(p1) generally is obtained only when the D ratio is very low and/or the difference between p2 and p1 is small. For a D ratio of 4 and (p2-p1) equals 0.03, a reduction in minimum ASN(p1) of 17.1% is obtained by dropping the constraint. However, when the difference between p2 and p1 is increased to 0.6 with the same D, a reduction of only 2.9% in

ASN(p1) is seen, but seen at a cost of a substantial increase in ASNMAX. In the rest of the cases illustrated, no reduction in minimum ASN(p1) is obtained by eliminating the ASNMAX constraint. For these cases the D ratio was 5 or greater and (p2-p1) was 0.045 or greater. These results indicate that both the D ratio and the difference (p2-p1) affect ASNMAX but only when both are small.

Additionally, whenever a reduction in minimum ASN(p1) is obtained by dropping the constraint, the increase in corresponding ASNMAX value may be substantial. However the increase in ASNMAX may become smaller as the difference between p2 and p1 increases. In other words, as the difference between p2 and p1 decreases, the price to be paid for the protection against high ASNMAX's will start to increase.

SUMMARY OF THE ALGORITHMS

MODEL I

| STEP | 1. | Compute | smallest | single | sampling | plan. | (ns*. | . c*) | |
|------|----|---------|-----------|--------|----------|-------|-------|-------|--|
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STEP 2. Set $c^2 = c^*$.

STEP 3. Incrementing on c1(0, 1, 2,..., c*-1):

3a. Compute feasible bounds on n1; i.e., n11 and n1u.

3b. Incrementing on n1(n11, n11+1,..., n1u) compute bounds on n2: i.e., n21 and n2u.

3c. Incrementing on n2(n21, n21+1,...,n2u) compute ASNMAX and ASN(p1).

Condition for Optimality

Feasible values are those for which the likelihood constraints, $(p1,1-\alpha)$ and $(p2,\beta)$, and the ASNMAX constraints are satisfied. At each calculation in the feasible region, ASN(p1) is calculated and the optimal double sampling

plan is the one with min ASN(p1). Because of convexity of ASN(p1), the algorithm shifts from cell to cell (value of c_1) whenever the current calculation (ASN(p1)'s) exceeds that for the previous calculation.

MODEL II

| STEP | 1. | Compute | the | smallest | single | sampling | plan | (ns*, | c*) | • |
|------|----|---------|-----|----------|--------|----------|------|-------|-----|---|
| | | | | | | | | , | | |

STEP 2 Incrementing on c2(c*, c*+1, c*+2,...):

STEP 3 Then incrementing on $c1(0, 1, 2, ..., c^{*}-1)$:

- 3a. Compute feasible bounds on n1; i.e; n11 and n1u.
- 3b. Incrementing on n1(n11, n11+1,..., n1u) compute bounds on n2; i.e; n21 and n2u.
- 3c. Incrementing on n2(n21, n21+1, n21+2,...,n2u) compute ASN(p1).

Condition for Optimality

Feasible values are those for which the likelihood constraints are satisfied. At each calculation in the feasible region, ASN(p1) is calculated and the optimal double sampling plan is the one with min ASN(p1). Because of convexity of ASN(p1) the algorithm shifts from cell to cell (value of c_1) until the current calculation (ASN(p1)'s) exceeds that for the previous cell calculation. Similarily the algorithm shifts column to column (value of c_2) until the minimum ASN(p1) of the current column exceeds that for the previous column.

COMPUTER CODE

The program originally was written in FORTRAN IV to run on a PDP 11-34 computer. Later it was modified to allow r_1 to be entered externally (rather than set at $r_2 = c_2+1$) and to run on a VAX 11-750 computer. The complete code is included in APPENDIX II.

As stated previously, the single sampling plan is computed using a brute force method; i.e: the search starts with an acceptance number of zero and the sample size is incremented by one at each iteration until L(p1) and L(p2) satisfy their respective inequalities. If the solved value of nl exceeds nu, no feasible solution exists for that value of c, c is incremented by one, and the search process for nl and nu starts anew. Depending on the input parameters (α , β , pl & p2), the single sample size may become very large thus requiring a large number of iterations to reach the first feasible solution. To reduce unnecessary computations, the user may input a "seed" number as a starting value of the single sample size. The closer the seed is to the true solution, the lesser the number of iterations required. However, the user must be very careful in entering a seed value. If a higher value of the seed than the true solution minimum ns is entered, the algorithm will converge to a single sampling plan with an acceptance number higher than that of minimum single sampling plan (the desired solution).

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APPENDIX I

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SAMPLE PROGRAM RUNS

a har we ar we we get a second a second a second a second second second second a second second as a second second and the second state of th DEPT. OF ISE UNIVERSITY OF FLORIDA *****DOUBLE SAMPLING SYSTEM***** ALPHA =0.0500 P0 =0.0100 BETA =0. 1000 P1 =0. 0400 REJECTION NO. OF FIRST SAMPLE (R1) = C2+(1) S--ACCEPTANCE ND. (C) = 4 LOWER BOUND ON N (NS) = 198 UPPER BOUND ON N (NL) = 198 DOUBLE SAMPLING PLANS FOR C1= 0 C2= 4 136 62 61 62 61 186. 8584 187. 0367 181.4362 DOUBLE SAMPLING PLANS FOR C1 = 1 C2 = 4164 165 185. 1312 185. 5088 34 33 179. 7643 180. 3866 34 33 DOUBLE SAMPLING PLANS FOR C1=2 C2=4183 184 15 14 189. 1240 189. 7155 15 14 186. 5909 187. 3789 DOUBLE SAMPLING PLANS FOR $C1 = 3 \quad C2 = 4$ 188 194 11 4 190. 1723 194. 7896 188.8725 11

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S-ACCEPTANCE NO. (C) = 5 LOWER BOUND ON N (NS) = 230 UPPER BOUND ON N (NL) = 262 DOUBLE SAMPLING PLANS FOR C1 = 0 C2 = 5162. 9084 162. 1922 162. 3609 213 212 210 251. 3808 247. 0489 244. 4871 213 207 63 64 203 65 DOUBLE SAMPLING PLANS FOR C1 = 1 C2 = 5149. 1270 149. 1608 183 181 232.8200 229.4124 101 102 180 174 DOUBLE SAMPLING PLANS FOR C1= 2 C2= 5 134 135 136 223. 2362 218. 6858 215. 2449 157.9882 157.8415 157.9577 161 151 143 162 140 158 DOUBLE SAMPLING PLANS FOR $C1 = 3 \quad C2 = 5$ 220. 4352 213. 3944 209. 6420 166 167 168 149 127 114 177.8430 177.2542 177.3490 155 151 148 DOUBLE SAMPLING PLANS FOR C1 = 4 C2 = 5221. 8151 214. 4609 212. 7945 149. 16 198 199 200 262 262 262 202. 6424 202. 0609 202. 5721 134 87 GLOBAL MÍNIMUM ASN(PO)= 102 CORRESPONDING N1 CORRESPONDING N2S 174 CORRESPONDING C1 1 CORRESPONDING C2 = 5

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UNIVERSITY OF FLORIDA ****DOUBLE SAMPLING SYSTEM**** ALPHA =0.0500 P0 =0.0200 BETA =0. 1000 P1 =0. 0800 REJECTION NO. OF FIRST SAMPLE (R1) = C2+(1) S----ACCEPTANCE NO. (C) = 4 LOWER BOUND ON N (NS) = UPPER BOUND ON N (NL) = 98 99 DOUBLE SAMPLING PLANS FOR C1 = 0 C2 = 454 53 90.6816 90.8418 78.5600 46 47 54 53 DOUBLE SAMPLING PLANS FOR C1 = 1 C2 = 486. 5216 86. 8856 62 63 39 38 39 38 75. 4389 76. 3481 DOUBLE SAMPLING PLANS FOR C1= 2 C2= 4 27 24 86. 1676 86. 9210 79. 6567 81. 3386 75 77 27 24 DOUBLE SAMPLING PLANS FOR C1= 3 C2= 4 90. 2002 90. 7994 88.0654 88.9571 16 15 87 16 14 88

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S--ACCEPTANCE ND. (C) = 5 LOWER BOUND ON N (NS) = 114 UPPER BOUND ON N (NL) = 131 DOUBLE SAMPLING PLANS C1= 0 C2= 5 FOR 124.7199 121.2064 118.5924 31 32 33 79.8665 105 108 79. 6080 79. 7088 106 105 100 **9**6 DOUBLE SAMPLING PLANS C1= 1 C2= 5 FOR 93 91 115.1913 111.7191 73.2101 50 51 88 82 DOUBLE SAMPLING PLANS FOR C1= 2 C2= 5 111. 4853 107. 4158 77. 6491 77. 6818 81 72 66 67 84 81 DOUBLE SAMPLING PLANS C1= 3 C2= 5 FOR 109.7731 104.1007 102.1335 87.7234 87.4925 87.9863 75 57 49 83 79 75 82 83 84 DOUBLE SAMPLING PLANS C1= 4 C2= 5 FOR 109. 3472 105. 3023 73. 23 100.0915 100.1993 131 131 98 63 ģģ 35 GLOBAL MĪNIMUM ĀŠN(PO)= 51 CORRESPONDING N1 82 CORRESPONDING N2S = CORRESPONDING C1 = 1 CORRESPONDING C2 = 5

ጟ፟ዹጚዹዄኯዸቒጚዄጟቒዄኯፘቒጛቒዸቒጟጜጟፘኯዸዹ፟ጛዸፘዸጞዹኯዄጚኯኯዸኯ

DEPT. OF ISE UNIVERSITY OF FLORIDA *****DOUBLE SAMPLING SYSTEM**** ALPHA =0. 0500 BETA =0. 1000 PO = 0.0200P1 = 0.0600REJECTION NO. OF FIRST SAMPLE (R1) = C2+(1) S---ACCEPTANCE NO. (C) = 7 LOWER BOUND ON N (NS) = 194 UPPER BOUND ON N (NL) = 200 DOUBLE SAMPLING PLANS FOR C1= 0 C2= 7 149 148 195. 7031 194. 7807 152.9495 54 55 149 147 DOUBLE SAMPLING PLANS FOR C1= 1 C2= 7 187. 7091 186. 9561 139.1258 139.9612 124 123 80 124 122 81 DOUBLE SAMPLING PLANS FOR C1= 2 C2= 7 101 102 105 103 180. 3470 179. 8239 135. 4178 136. 3186 105 103 DOUBLE SAMPLING PLANS FOR C1= 3 C2= 7 121 122 86 85 174. 2581 174. 0120 140.0145 86 84 DOUBLE SAMPLING PLANS FOR C1= 4 C2= 7 139 140 172.2300 148.9510 149.8648 71 69 71 69 DOUBLE SAMPLING PLANS FOR C1= 5 C2= 7 157 158 57 53 57 55 174.6661 174.4244 161.6958 162.4568 DOUBLE SAMPLING PLANS C1= 6 C2= 7 FOR 181.5391 181.4739 43 36 176.6320 51 47 175 176

21

S ACCEPTANCE NO. (C) = B LUWER BOUND UN N (NS) = 215UPPER BOUND ON N (NL) = 236DOUBLE SAMPLING PLANS FOR C1= 0 C2= 8 44 45 197 193 199 234. 9933 232. 0821 160.0134 160.2446 198 DOUBLE SAMPLING PLANS FOR C1= 1 C2= 8 70 71 173 169 227. 3926 224. 7274 176 174 140.8520 DOUBLE SAMPLING PLANS FOR C1= 2 C2= 8 92 93 156 151 220. 2482 217. 1164 135.6124 157 156 DOUBLE SAMPLING PLANS FOR C1= 3 C2= 8 113 114 140 134 141 139 212. 4149 209. 1393 139.6817 140.0779 DOUBLE SAMPLING PLANS FOR C1= 4 C2= 8 134 135 203. 5589 200. 4914 125 123 120 113 149.6677 DOUBLE SAMPLING PLANS FOR C1= 5 C2= 8 154 155 106 116 113 200. 3735 196. 5552 163. 1247 DOUBLE SAMPLING PLANS FOR $C1 = 6 \quad C2 = 8$ 174 175 125 200. 3812 196. 7400 91 75 178.8247 118 179.0665 DOUBLE SAMPLING PLANS FOR C1= 7 C2= 8 194 195 236 236 204. 8341 201. 9844 76 49 195.9521 196.2862

8-ACCEPTANCE NO. (C) = 9 LOWER BOUND ON N (NS) = 235 UPPER BOUND ON N (NL) = 273 DOUBLE SAMPLING PLANS C2= 9 FOR C1 = 041 236 242 272. 4937 173.9185 229 224 240 239 266. 5910 262. 6524 172 9765 42 43 DOUBLE SAMPLING PLANS C1= 1 C2= 9 FOR 150.2853 149.3582 149.0869 149.5144 221 212 222 220 273. 3678 265. 8895 260. 2918 66 67 205 218 68 217 256. 5717 69 DOUBLE SAMPLING PLANS FOR C1 = 2 C2 = 989 90 91 92 204 193 185 179 266. 7397 258. 1279 252. 1336 247. 8829 142. 7357 141. 8807 141. 7317 142. 0573 204 202 200 198 DOUBLE SAMPLING PLANS FOR C1 = 3C2= 9 186 173 164 157 256. 2776 247. 1019 241. 0539 236. 5696 145. 0258 144. 3393 144. 3170 144. 6165 111 112 113 188 186 184 182 114 DOUBLE SAMPLING PLANS FOR C1 = 4C2= 9 132 133 134 175 156 145 176 174 171 249. 4611 237. 6933 231. 2966 154.0411 153.0990 153.1052 DOUBLE SAMPLING PLANS FOR C1 = 5C2 = 9166. 3868 165. 7449 166. 0171 153 154 155 154 132 121 237.1488 226.1181 221.0988 171 167 164

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DOUBLE SAMPLING PLANS FOR C1 = 6 C2 = 9174 175 120 103 187 181 223. 4432 217. 4335 181.0777 181.2191 DOUBLE SAMPLING PLANS FOR C1= 7 C2= 9 194 195 196 112 81 70 273 273 273 224. 6025 217. 1298 215. 1222 198. 0900 198. 0276 198. 6775 DOUBLE SAMPLING PLANS FOR C1= 8 C2= 9 215 53 273 216 41 273 GLOBAL MINIMUM ASN(PO)= 222. 1338 221. 5181 135. 42 215. 9652 216. 7635 CORRESPONDING N1 101 CORRESPONDING N2S 105 = CORRESPONDING C1 = 2 CORRESPONDING C2 = 7

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DEPT. OF ISE UNIVERSITY OF FLORIDA *****DOUBLE SAMPLING SYSTEM***** ALPHA =0. 0500 BETA =0. 1000 P1 =0. 0450 PO =0.0150 REJECTION NO. OF FIRST SAMPLE (R1) = C2+(1) S--ACCEPTANCE ND. (C) = 7 LOWER BOUND ON N (NS) = 260 UPPER BOUND ON N (NL) = 266 DOUBLE SAMPLING PLANS FOR C1= 0 C2= 7 260.0430 259.1314 207.2268 75 76 195 194 195 193 DOUBLE SAMPLING PLANS FOR C1 = 1 C2 = 7250. 0084 249. 2654 108 109 187.1743 187.9875 164 162 164 162 DOUBLE SAMPLING PLANS FOR C1= 2 C2= 7 239. 5929 239. 0794 184.0606 184.9139 138 139 135 135 133 134 DOUBLE SAMPLING PLANS FOR C1= 3 C2= 7 232. 4163 232. 1787 189. 4726 164 111 111 165 109 190. 3486 109 DOUBLE SAMPLING PLANS FOR C1 = 4 C2 = 7229. 9069 229. 5068 90 87 201.1494 201.8962 188 90 189 88 DOUBLE SAMPLING PLANS FOR $C1 = 5 \quad C2 = 7$ 211 212 72 70 233. 1982 232. 9633 217.1050 217.8536 72 68 DOUBLE SAMPLING PLANS FOR C1= 6 C2= 7 234 235 59 51 242. 9257 242. 7148 236.2712 64 60

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S---ACCEPTANCE NO. (C) = 8LUWER BOUND UN N (NS) = 28/UPPER BOUND ON N (NL) = 314 DOUBLE SAMPLING PLANS FOR C1= 0 C2= 8 59 60 314. 4307 311. 5352 214.7717 215.0103 264 260 265 263 DOUBLE SAMPLING PLANS C1= 1 C2= 8 FOR 93 94 306. 1528 303. 5039 188.7529 189.3262 235 231 235 233 DOUBLE SAMPLING PLANS FOR C1=2 C2=8205 200 207 206 291. 8433 288. 7336 182. 4429 182. 8303 124 125 DOUBLE SAMPLING PLANS FOR C1 = 3 C2 = 8281. 3353 278. 0829 187. 6786 188. 0451 *185* 183 152 153 183 DOUBLE SAMPLING PLANS FOR C1= 4 C2= 8 179 180 164 156 166 164 273. 5826 269. 9618 200. 6032 200. 8872 DOUBLE SAMPLING PLANS FOR C1= 5 C2= 8 268.2267 218. 5081 218. 6513 206 143 131 151 148 DOUBLE SAMPLING PLANS C2= 8 FOR C1 = 6277. 2705 266. 7340 263. 1187 240. 3717 239. 3440 239. 5680 163 157 151 232 233 234 157 117 101 DOUBLE SAMPLING PLANS C1= 7 C2= 8 FOR 262.1125 80 63 314 314 271.3427 269.9318 260 261

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CERTENENT CERTENETIES CENTRE CENTRE CENTRE SERVICES CENTRE EN CENTRE EN CONTRE DE CONTRE DE CONTRE CENTRE CENTR S-ACCEPTANCE NO. (C) = 9 LOWER BOUND ON N (NS) = 314 UPPER BOUND ON N (NL) = 363 DOUBLE SAMPLING PLANS FOR C1 = 0C2 = 9364 4188 358. 5363 353. 6355 349. 7147 322 320 319 317 55 316 309 233. 3808 232. 4478 231. 9713 231. 9728 56 303 57 298 58 DOUBLE SAMPLING PLANS FOR C1 = 1C2= 9 291 283 276 270 361. 2587 354. 7504 349. 1775 344. 5428 294 292 290 201. 4461 200. 8535 200. 5675 89 9Ò 91 92 289 200. 6042 DOUBLE SAMPLING PLANS FOR C1 = 2C2≈ 9 120 121 122 123 264 255 247 349. 1373 342. 3050 336, 3438 332. 1181 191.0035 269 190. 6154 190. 4330 190. 7493 267 265 263 241 DOUBLE SAMPLING PLANS FOR C1 = 3C2= 9 337. 9465 329. 6002 323. 5883 319. 1306 194. 2755 193. 7339 193. 7005 194. 0067 243 231 149 248 150 151 152 246 222 215 244 242 DOUBLE SAMPLING PLANS FOR C2= 9 C1 = 4327. 9336 317. 5687 310. 5440 306. 1929 206.0636 205.3329 205.1960 205.5502 177 178 179 226 209 197 232 229 227 ĨĖÒ 189 224 DOUBLE SAMPLING PLANS FOR C1 = 5C2 = 9312. 6136 302. 7364 297. 2081 222. 6100 222. 1069 222. 2803 205 206 207 198 223 178 220 166 216 DOUBLE SAMPLING PLANS FOR C1 = 6C2 = 9313. 9585 296. 9233 290. 9528 243. 8767 242. 4273 242. 5471 232 233 234 200 248 156 139 241 233

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DOUBLE SAMPLING PLANS

FOR C1= 7 C2= 9

TRANSPORT MARKETS TRANSPORT

260126363294.2362264.7458261106363289.8002265.0618DOUBLE SAMPLING PLANS

FOR C1= 8 C2= 9

| 287 288 GLOBAL | 82 63 MINIMU | 363 363 M ASN(PO) | 2° 2° | 77.9774 76.4333 184.06 | 288. 5138 289. 1825 |
|----------------------|--------------------|-------------------------|------------|------------------------------|------------------------|
| CORRESP | UNDING | NI | = | 138 | |
| CORRESP | ONDING | N2S | = | 135 | |
| CORRESP | DNDING | C1 | = ; | 2 | |
| CORRESP | DNDING | C2 | * ' | 7 | |

APPENDIX II

and the second territor because shorted and the second memory memory was an

FORTRAN IV PROGRAM

0001 C C GUALITY CONTROL DOUBLE SAMPLING PROGRAM TO ANALYSE DOUBLE SAMPLING PLANS, ASN(PO) AND ASNMAX. BINOMIAL AND POISON PROBALITY DISTRIBUTIONS USED. 0002 0003 CCCCCCC 0004 PROGRAMED BY R. WAREN RANGARAJAN INDUSTRIAL AND SYSTEMS ENGINEERING DEPARTMENT UNIVERSITY OF FLORIDA 0005 0000 0007 . 0008 GAINESVILLE, FLORIDA 32611 0009 ē . DOUBLE PRECISION SUMLOG INTEGER C, C1, C2, C1MIN, C2MIN, R1, R11 BYTE STING(8) 0010 0011 0012 BYTE STING(8) COMMON/BLK1/N2S, N2L COMMON/BLK2/PS, PL COMMON/BLK3/N1 COMMON/BLK3/N1 COMMON/BLK5/PO, P1 COMMON/BLK11/ASN, ASNMAX <u>0</u>013 ÕÕ14 0015 <u>0016</u> 0017 0018 0019 0021 0022 0023 0024 C C 0025 WRITE(5, *) ' NAME OF OUTPUT FILE?' READ(5, 1) STING 0027 0028 0029 1 FORMAT(10A1) С 0030 0031 0032 CALL ASSIGN (1, STING) CCC BEGINNING INITIALIZATION 0033 N=0 0035 C2 = 10000036 ASNMIN=15000. C = -10038 C C 0039 INPUT FORMAT **0**040 15 WRITE (5,16) 16 FORMAT (///' CODES FOR SELECTING APPR. PROB. DIST.'// 115%, 'BINOMIAL', 12%, '=1', 2/15%, 'POISSON', 13%, '=2') **0041 0042 0043** 0044 2/15%, 'PUISSUR', ISA, -2 , READ (5, +) K IF(K.GT.2.OR.K.LT.1) GOTO 15 22 WRITE(5, 21) 21 FORMAT(10%, 'SELECT'/16%, 'SAMPLE PLANS ONLY =1' 1/16%, 'ASN VALUES ONLY =2' 2/16%, 'OR BOTH =3') 0045 0046 0048 0049 0050 READ(5, *) KOPT IF(KOPT.GT.3.OR.KOPT.LT.1) GOTO 22 0051 ōō52 WRITE (5,17) FORMAT(10X,'INPUT ALPHA ') READ (5,*) ALPHA WRITE (5,18) 0053 0054 0055 17 **Ö**056 0057 FORMAT(10X, 'INPUT BETA ') 18

| D57\$M4 | IN |
|----------------------------------|--|
| 54787012345578701234557870123455 | <pre>WRITE (5,90) WRITE(1,90) 90 FORMAT(//10X, 'SINGLE SAMPLING PLAN '/'+',9X,21('_')) WRITE(5,91) C,NS,NL WRITE(1,91) C,NS,NL 91 FORMAT(10X, 'ACCEPTANCE ND.(C) =',I2 1,/10X, 'LOWER BOUND ON N (NS) =',I4 2,/10X, 'UPPER BOUND ON N (NL) =',I4) COMPUTATION OF DOUBLE SAMPLING PLAN BEGINS: FOR EACH VALUE O IF(C.LT.C2) MC=C+MC1-1 C2=C R1=C2+R11 DO 100 K1=1.C2 C1=K1-1 CALL SUBROUTINE TO COMBUTE THE FIRST SAMPLE NUMBER CALL TRY1(NTRY,C1,P1,NS,BETA,K) N1=NTRY IF(NTRY.GT.NS) GOTO 600 WRITE(5,161) WRITE(5,161) WRITE(5,160) C1,C2 WRITE(5,160) C1,C2 WR</pre> |
| C CCC | <pre>160 FORMAT(/10X, 'FOR C1=', I2, 2X, 'C2=', I2, //) NTEMP=N1 IF(KOPT.EQ.1) WRITE(5, 170) IF(KOPT.EQ.3) WRITE(5, 175) 175 FORMAT(10X, '(N1)', 3X, '(N2S)', 3X, '(N2L)', 4X, 'ASNMAX', 5X, 'A 1 /) 170 FORMAT(11X, '(N1)', 10X, '(N2S) (N2 (N2L)', 4X, 'ASNMAX', 5X, 'A 1 10X, 'PL'//) NTEMP1=NS COMPUTATION OF SECOND SAMPLE FOR EACH VALUE OF FIRST SAMPLE ASN=FLOAT(NS)+10 DD 190 IZ=NTEMP.NTEMP1</pre> |
| COC | I=IZ CALL SUBROUTINE TO COMPUTE SECOND SAMPLE CALL TRY2(NS, NL, K, I, R1) IF(KOPT.NE.1) GOTO 500 WRITE (5, 185) I, N2S, N2L, PS, PL WRITE (1, 185) I, N2S, N2L, PS, PL WRITE(1, 185) I, N2S, N2L, PS, PL WRITE(1, 185) I, N2S, N2L, PS, PL WRITE(1, 185) I, N2S, N2L, PS, PL 185 FORMAT(10X, I4, 13X, I4, ' (N2 (', I4, 4X, FB.6, 8X, GOTO 190) |

| 0172 0173 0174 0175 0176 0177 0177 0177 0177 0177 0177 0177 0177 0177 0177 0177 0177 0177 0178 0180 0180 0180 0180 0180 0181 0182 0183 0183 0184 0184 0184 0184 0184 0185 0185 0186 0186 0187 0186 0187 0190 000 CONTINUE 0190 0197 000 CONTINUE 0190 0197 000 CONTINUE 0190 0197 000 CONTINUE 0190 0197 000 CONTINUE 0190 0197 0100 CONTINUE 0190 0197 0100 CONTINUE 0190 0197 0100 CONTINUE 0190 0197 0100 CONTINUE 0197 0100 CONTINUE 0198 000 CONTINUE 0198 0198 0198 0198 0198 0198 0198 0198 0198 0198 0198 0198 0198 0198 0198 0198 0198 0198 |
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| 0001 0002 0003 | CCC | |
|---------------------------------|------------------|--|
| 0004 0005 | C | SUBROUTINE TRY1 (NTRY, C1, P, NL, BETA, K) |
| 0003 0007 0008 | L C C C | THIS SUBROUTINE COMPUTES FIRST SAMPLE NUMBER OF DOUBLE SAMPLING PLAN BY AN INTEGER FORM OF BISECTION METHOD |
| 0009 | c | INTEGER C1 |
| 012 | č | · NLARGE≠NL |
| 0014 0015 | С | NSMALL=0 |
| 017 018 019 020 021 | С | CALL APPROPRIATE PROBAILITY SUBROUTINE FOR PROB. CALCULATIONS 10 IF(K.EG.1) CALL PROBS1(NTRY, P, C1, BXLEC) IF(K.EG.2) CALL PROBS2(NTRY, P, C1, BXLEC) IF(BXLEC.LE.BETA) GOTO 50 NSMALL=NTRY |
| 022 | | GDTD 25 50 NLARGE=NTRY |
|)024)025)026 | | 25 IF(NSMALL.NE.(NLARGE-1)) GOTO 5 NTRY=NLARGE RETURN |
| 0027 | | END |
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| 0002 | COCC | |
|----------------------|--------|---|
| 0005 | C C | SUBROUTINE PROBD1(N1, N2, P, DPROB, K, R1) |
| 0007 | CCC | THIS SUBROUTINE COMPUTES DOUBLE PROBABILITIES FOR COMPUTING SECOND SAMPLE NUMBER OF DOUBLE SAMPLING NUMBER |
| 0010 | C | COMMON/BLK6/C1,C2 INTEGER C1,C2,R1 |
| 0012 0013 0014 | C C | IF (K. FQ. 1) CALL PROBSI (NI. P. CI. RYLEC) |
| 0015 | | IF(K.EQ.2) CALL PROBS2(N1, P, C1, BXLEC) DPROB=BXLEC |
| 0018 0019 | | NTEMP=BXLEC NTEMP=C1+1 KTEMP=R1-1 |
| 0020 0021 0022 | | DO 10 IX=NTEMP, KTEMP I=IX J=C2-I |
| 0023 0024 | | IF(K.EQ.1) CALL PROBS1(N1, P, I, BXLEC) IF(K.EQ.2) CALL PROBS2(N1, P, I, BXLEC) |
| 0025 | | PROBI=BXLEC-TEMP TEMP=BXLEC IF(K.EQ.1) CALL PROBSI(N2,P,J,BXLEC) |
| 0028 0029 0030 | 10 | IF(K.EQ.2) CALL PROBS2(N2,P,J,BXLEC) DPROB=DPROB+(PROB1*BXLEC) CONTINUE |
| 0031 | c Ì | RETURN |
| 0033 | | END |
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| | 0001 | c | |
| ¢ ل | 0002 | č | |
| - | 0003 | č | |
| | 0004 | С | |
| | 0005 | C | SUBROUTINE TRY2(NS, NL, K, J, R1) |
| | 0007 | č | THIS SUBROUTINE COMPUTES THE SECOND SAMPLE NUMBER OF |
| - | 0008 | č | THE DOUBLE SAMPLING NUMBER BY AN INTEGER BISECTION |
| • | 0009 | Ç | METHOD. SEVERAL TESTS ARE DONE TO LOCATE THE PARAMETER |
| ę . | 0010 | ç | AT ITS TRUE POSITION. |
| ر د | 0012 | C C | INTEGER C1.C2.R1 |
| | 0013 | С | |
| | 0014 | | COMMON/BLK1/N2S, N2L |
| | 0015 | | |
| | 0017 | | COMMON/BLK4/ALPHA, BETA |
| | 0018 | | COMMON/BLK5/PO, P1 |
| 1 | 0019 | r | CUMMON/BLK6/C1, C2 |
| 1 | 0021 | C | K1=C1+1 |
| l | ōō22 | C | |
| | 0023 | ç | SET LIMITS FOR COMPUTING N2S |
| | 0024 | G | NGMALL -NG-L |
| • | 0026 | | NLARGE=NSMALL |
| | 0027 | C | |
| • | 0028 | č | INDEXING TO SPECIFY WHAT BOUND (N2S OR N2L) IS BEING |
| | 0030 | č | |
| | 0031 | v | I=1 |
| 1 | 0032 | Ç | |
| ļ | 0033 | C C | INITIAL TEST AT EACH LIMIT |
| | 0035 | v | CALL PROBD1(J, NSMALL, P1, DPROB, K, R1) |
| ξ. | 0036 | _ | IF (DPROB.LE.BETA) GOTO 55 |
| 1 | 0037 | ç | RIGECTION METHOD |
| • | 0039 | č | BISECTION METHOD |
| | 0040 | - | NLARGE=NL |
| • | 0041 | | 5 NTRY=(NSMALL+NLARGE)/2.0 |
| | 0042 | c | GUIU (10,20),I |
| | 0044 | <u> </u> | O CALL PROBDI (JUNTRY, P1, DPROB, K, R1) |
| - | 0045 | - | IF(DPROB.LE.BETA) GOTO 50 |
| 1 | 0046 | ~ | GOTO 15 |
| 1 | 0047 | 2 | COCALL PROBUL(J, NIRY, PO, DPROB, K, R1) |
| | 0049 | 1 | 5 NSMALL=NTRY |
| | 0050 | - | GOTO 25 |
| | 0051 | 5 | O NLARGE=NTRY |
| 1 | 0052 | с ² | JITUNLAKGE-NƏMALLJ.GI.I) GUIU D |
| | 0054 | č | CHECK THE INDEX TO FIND WHERE THE PROCESS IS |
| i L | 0055 | С | GOTO (55.60), I |
| | 0057 | С | |
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| | TRY2 | | |
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| | 0058 0059 0060 0061 0062 0063 0064 0065 0064 0065 0066 0067 0068 0069 0070 0071 0072 0073 0074 0075 0076 0077 0078 0078 0079 0080 0081 | C CHANGE THE INDEX AFTER N2S COMPUTATION 55 I=I+1 C TESTING EACH POSSIBLE CASES TO LOCATE THE LOWER BOUND AT ITS TRUE POSITION N2S=MAXO(O, NLARGE) CALL PROBD1(J, NLARGE, P1, DPROB, K, R1) PS=DPROB MTEMP=NLARGE-5 NSMALL=MAXO(O, MTEMP) NLARGE=NL GOTO 5 60 N2L=NSMALL CALL PROBD1(J, NSMALL, P0, DPROB, K, R1) PL=DPROB CALL PROBD1(J, NLARGE, P0, DPROB, K, R1) IF(DPROB.GE.(1-ALPHA)) N2L=NLARGE IF(DPROB.GE.(1-ALPHA)) PL=DPROB 110 RETURN END | |
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| c | SUBROUTINE PROB | 51 (NN, P, C, BXLEC) |
| | THIS SUBROUTINE PROBABILITIES | COMPUTES CUMULATIVE BINOMIAL |
| - | INTEGER C | |
| ~ | COMMON/BLK7/SUML COMMON/BLK7/SUML | - GG(1500) |
| č | | |
| С | Q=1P | |
| č | BINOMIAL PRO | B. WHEN C=O |
| U. | CSUMS=Q**NN | |
| С | WRITE(6,500) CSU IF (C.EQ.0) GOT(| IMS 1 333 |
| C C | AVOID RECOMPUTI | NG SUMLOG(I)'S ALREADY IN MEMORY |
| С | IF (N-NN) 100.21 | 1. 211 |
| ~ | 100 M=N+1 | / |
| | COMPUTE N S | SUMLOGS-EQUIVALENT TO N-FACTORIAL |
| v | IF (M.GT.1) GOTO | 110 |
| | IF(NN.LE.1) GOTO | 211 |
| | M=2 | |
| | SUMLOG(I)=DLC | G10(DFLOAT(I))+SUMLOG(I-1) |
| C | | |
| C | COMPUTE C S | UMS-EQUIVALENT TO SSUM OF PROB.COMPIN. TIVE RINGMIAL DISTRIBUTION COMPLITATION |
| č | | |
| Č | LI IF (NN.GI.N) N=NN | · . |
| CCC | DETERMINE B | EST NUMBER HANDLING LOOP |
| - | IF (NN.GT.300) G | OTO 300 |
| | CSUMS=10. **(S | UMLOG(NN)-SUMLOG(NN-K)-SUMLOG(K)) |
| | 1 22 CONTINUE | *P**K*Q**(NN-K)+CSUMS |
| ç | WRITE(6,501) CSU | MS |
| | GOTO 333 | |
| CC | LOOP FOR LA | RGE EXPONENTS |
| С | 100 DC 322 K=1.C | |
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| • | |
| PROBS1 | |
| 0058 0059 0040 0041 0042 0043 0044 0045 0045 | CSUMS=10.**(SUMLOG(NN)-SUMLOG(NN-K)-SUMLOG(K) 1 +K*DLOG10(DBLE(P))+(NN-K)*DLOG10(DBLE(G)))+CSUMS C WRITE(6,501) CSUMS C 500 FORMAT(10X,FB.6) 322 CONTINUE C 333 BXLEC = CSUMS RETURN END |

| 法实际上的关系是不是的人的人的人,人们的发生人,这些人的人的是他是他是他是他是他是他是他是他是他是他是他是他是他是他是他是他是他是他 | |
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| 医马克斯基甲基斯酸 医副子宫 网络黄色 网络黄色 网络黄色黄色黄色黄色黄色 网络黄色黄色 黄色 黄色 网络古拉古拉 古圣子儿 不久 人名贝尔 医无关系的 化乙烯乙烯乙烯乙烯乙烯乙烯乙烯乙烯乙烯乙烯乙烯乙烯乙烯乙烯乙烯乙烯乙烯 | |
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| • | 0001 0002 0003 0004 | CCCC | | |
|---|--------------------------------------|-------------|-----|---|
| • | 0005 | c | | SUBROUTINE PROBS2(NN, P, C, BXLEC) |
| - | 0008 0008 | č c c | | THIS SUBROUTINE COMPUTES CUMULATIVE POISON PROBABILITIES |
| | 0010 0011 0012 0013 | c | | INTEGER C PP=P*NN TERM=1.0 SUM=TERM |
| | 0015 0016 0017 0018 0019 | | 100 | IF(C.EQ.O) GDTD 110 DD 100 I=1,C TERM=TERM*PP/I SUM=SUM+TERM CONTINUE |
| | 0021 | | 110 | BXLEC=SUM/EXP(PP) |
| | 0023 0024 | L | | RETURN END |

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| 0001 0002 | C C | |
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| 0003 | Ċ C | |
| 0006 0006 0007 | C C | THIS SUBROUTINE COMPUTES ASN(PO) VALUES AND ASNMAX |
| 0008 | C C | VALUES. |
| 0011 0012 | | DUUBLE PRECISION SUMLUG INTEGER CIMIN, C2MIN COMMON/BLKI/N25, N2L |
| 0013 0014 | | COMMON/BLK3/N1 COMMON/BLK4/ALPHA, BETA |
| 0016 0017 | | COMMON/BLKS/PO,P1 COMMON/BLK6/I1,I2 COMMON/BLK7/SUMLOG(2500) |
| 0018 0019 0020 | c | COMMON/BLKB/N COMMON/BLK11/ASN, ASNMAX |
| 0021 0022 0023 | 0000 | INITIALIZATION COMPUTE P* (MAXIMUM PROB. FOR ASNMAX) |
| 0024 | | J=I1+1 XXX=0.0 |
| 0026 0027 0028 0029 | | IF(I1.GT.O) XXX=SUMLUG(I1) AKONST=10.**(SUMLOG(I2)+SUMLOG(N11-I2-1)-XXX-SUMLOG(N11-I1-1 TEMP=1.0/FLOAT(I2-I1) AKONST=AKONST**TEMP |
| 0030 0031 0032 | | PSTAR=AKONST/(1.0+AKONST) IF(K.EQ.1) CALL PROBS1(N11,PSTAR, I2, BXLEC) IF(K.EQ.2) CALL PROBS2(N11 PSTAR, I2, BXLEC) |
| 0033 0034 | | TEMP=BXLEC IF(K.EQ.1) CALL PROBS1(N11, PSTAR, I1, BXLEC) |
| 0035 0036 0037 | | IF(K.EQ.2) CALL PROBS2(N11, PSTAR, I1, BXLEC) TEMP1=TEMP-BXLEC ASNMAX=FLDAT(N11)+N2S#TEMP1 |
| 0038 | С | IF (K. EQ. 1) CALL PROBSI (N11, PO, I2, BXLEC) |
| 0041 0042 | | TEMP=BXLEC IF(K.EG.1) CALL PROBS1(N11, PO, I1, BXLEC) |
| 0043 0044 0045 | | IF(K.EQ.2) CALL PROBS2(N11,PO,I1,BXLEC) TEMP2=TEMP-BXLEC ASN=FLOAT(N11)+N2S+TEMP2 |
| 0046 | | IF (ASNMAX.GT.NS.OR.ASN.GT.ASNMIN) GOTO 100 ASNMIN=ASN |
| 0048 0049 0050 | | NIMIN=N11 N2MIN=N2S C1MIN=I1 |
| 0051 0052 | C 100 | C2MIN≠I2 |
| 0054 0055 | 110 | WRITE(5,110) N11, N2S, TEMP1, ASNMAX, TEMP2, ASN FORMAT(//10X,2(I3,3X),2(F6.4,3X,F8.4,3X)) |
| 0056 0057 | C C | |
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| | 0001 0002 0003 0004 0005 0007 0007 00012 00012 0012 0012 0012 00 | 0001 C 0003 C 0004 C 0005 C 0007 C 0008 C 0010 C 0011 C 0012 C 0013 C 0014 C 0015 C 0014 C 0015 C 0016 C 0017 C 0018 C 0019 C 0020 C 0021 C 00220 C 00217 C 00222 C 00233 C 0024 C 0035 C 0036 C 0037 C 0038 C 0044 C 00450 C 0051 C 0052 C 0053 100 0054 C 00555 C |

