

## **Uniform Location Model: Equivariant Estimation based on Progressively Censored Samples**

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### **ABSTRACT**

In this paper, we consider uniform model and obtain minimum risk Equivariant estimators of the parameters based on type-II progressively censored samples under Standardized quadratic loss function,

Absolute error loss function and Linex loss function. These generalize the corresponding results for Type-II censored samples. Leo Alexander (2000).

**Keywords:** Equivariant estimation, Location model, Optimal estimation, Progressive censored sample and Uniform model.

### **1. INTRODUCTION**

In life-testing experiments, the common practice is to terminate the experiment when certain number of items have failed (Type-II censoring) or certain stipulated time has elapsed (Type-I censoring). Type-II progressive censoring involves removing certain fixed number of surviving units at each failure which is an extended version of Type-II censoring scheme. As pointed out by Balakrishnan and Sandhu, the scheme of progressive censoring is an attractive feature as it saves both cost and time for the experimenter. Progressively censored samples have previously been considered by Herd, Roberts, Cohen and others. This

developments have been summarized in Cohen, Balakrishnan and Cohen, and Balakrishnan and Aggarwala.

Lehmann and Casella discussed the marginal Equivariant estimation of the parameters of location, scale, and location-scale models. Edwin Prabakaran and Chandrasekar developed simultaneous Equivariant estimation approach and illustrated the method with suitable examples. Leo Alexander and Chandrasekar obtained minimum risk Equivariant estimators for the parameters of Exponential models based on Type-II censored samples. Viveros and Balakrishnan developed exact conditional inference based

on progressively Type-II censored samples. Tse and Tso compared the expected times under Type-I censoring, Type-II censoring and complete sampling plans in the exponential case. Also, Aggarwala and Balakrishnan established some properties of progressively Type-II censored order statistics from arbitrary continuous distributions. In his development and discussion on generalized order statistics, Kamps has proved some general properties of progressively Type-II censored order statistics as a special case of generalized order statistics. Further, the maximum likelihood estimation of the parameters of an exponential distribution based on progressively Type-II censored samples has been discussed recently by Cramer and Kamps and Balakrishnan et.al. All these developments on progressively Type-II censored order statistics have been synthesized in a recent book by Balakrishnan and Aggarwala.

In this paper, assuming that the sample is obtained through Type-II progressive censoring scheme, we obtain minimum risk Equivariant estimator(s) for the parameter(s) of the uniform model. The paper is organized as follows: Section 3

deals with the problem of Equivariant estimation for the uniform location model.

## 2. PRELIMINARIES

Let  $N$  denote the total number randomly selected items put to test simultaneously and  $n$  designate the number of samples specimens which fail. Thus the number of completely determined life spans is  $n$ . At the time of the  $i$ -th failure,  $r_i$  surviving units are randomly withdrawn from the test,  $i=1,2,\dots,n$ . Clearly,

$$r_n = N - n - \sum_{i=1}^{n-1} r_i . \text{ Let } X_{i:N}, i = 1, 2, \dots, n ,$$

denote the failure times of the completely observed times. Then, the joint probability density function (pdf) of  $(X_{1:N}, X_{2:N}, \dots, X_{n:N})$  is

$$g_{\theta}(x_1, x_2, \dots, x_n) = \prod_{i=1}^n (N - \sum_{j=1}^{i-1} r_j - i + 1) \times f_{\theta}(x_i) \{1 - F_{\theta}(x_i)\}^{r_i} \dots (2.1)$$

Here,  $f_{\theta}$  and  $F_{\theta}$  denote the common pdf and distribution function of the  $N$  items under life-test. Further,  $r_1, r_2, \dots, r_n$  are assumed to be pre-fixed by the experimenter.

3. LOCATION MODEL

In this case the pdf is taken to be

$$f_{\xi}(x) = \begin{cases} 1, & \xi \leq x \leq \xi + 1; \xi \in R \\ 0, & \text{otherwise} \end{cases}$$

Thus (2.1) reduces to

$$g_{\xi}(x_1, \dots, x_n) = \left\{ \prod_{i=1}^n (N - \sum_{j=1}^{i-1} r_j - i + 1) \right\} \prod_{i=1}^n \{1 - x_{i:N} + \xi\}^{\pm} (X_{i:N} - X_{1:N}), \quad i = 2, 3, \dots, n.$$

$$\xi < x_{1:N} < \dots < x_{n:N} < \xi + 1; \xi \in R.$$

Thus the joint distribution of  $\mathbf{X} = (X_{1:N}, \dots, X_{n:N})$  belongs to a location family with the location parameter  $\xi$ . We are interested in deriving MRE estimator of  $\xi$  by considering three loss functions. Following Lehmann and Casella (1998), the MRE estimator of  $\xi$  is given by

$$\delta^*(\mathbf{X}) = \delta_0(\mathbf{X}) - v^*(\mathbf{Y}),$$

where  $\delta_0$  is a location equivariant estimator,  $v(\mathbf{y}) = v^*(\mathbf{y})$  minimizes

$$E_0[\rho\{\delta_0(\mathbf{X}) - v(\mathbf{y})\} | \mathbf{y}]$$

and  $E_0$  denotes  $E_{\xi}$  when  $\xi = 0$ .

Here

$$Y_i = X_{i:N} - X_{1:N}, \quad i = 2, 3, \dots, n,$$

$$Y = (Y_2, Y_3, \dots, Y_n)$$

and  $\rho$  is an invariant loss function.

**Case (i):** If the loss is squared error, then

$$v^* = E_0(\delta_0 | \mathbf{y}).$$

Take  $\delta_0(\mathbf{X}) = (X_{1:N} + X_{n:N})/2$ . Clearly  $\delta_0$  is an equivariant estimator but not complete sufficient. In order to find  $v^*$ , assume that  $\xi = 0$  and consider the transformation

$$Y_1 = (X_{1:N} + X_{n:N})/2 \quad \text{and}$$

$$\text{Then } X_{1:N} = Y_1 - Y_n/2 \quad \text{and}$$

$$X_{i:N} = Y_i + Y_1 - Y_n/2, \quad i = 2, 3, \dots, n$$

and the Jacobian of the transformation is given by  $J = 1$ . Thus the joint pdf of  $(Y_1, Y_2, \dots, Y_n)$  is given by

$$h(y_1, \dots, y_n) = \left\{ \prod_{i=1}^n (N - \sum_{j=1}^{i-1} r_j - i + 1) \right\} \{1 - (y_1 - y_n/2)\}^n \times \prod_{i=2}^n \{[1 - (y_i + y_1 - y_n/2)]^i\},$$

$$0 < y_2 < y_3 < \dots < y_n,$$

$$y_n/2 < y_1 < 1 - (y_n/2), \quad 0 < y_n < 1.$$

Also, the joint pdf of  $(Y_2, \dots, Y_n)$  is given by

$$h_1(y_2, \dots, y_n) = \left\{ \prod_{i=1}^n (N - \sum_{j=1}^{i-1} r_j - i + 1) \right\} \times \int_{y_n/2}^{1-y_n/2} \{1 - (y_1 - y_n/2)\}^n \times \prod_{i=2}^n \{[1 - (y_i + y_1 - y_n/2)]^i\} dy_1, \dots \quad (3.1)$$

$$0 < y_2 < y_3 < \dots < y_n < 1.$$

Thus the conditional pdf of  $\delta_0 = Y_1$  given  $(Y_2, \dots, Y_n)$  is given by

$h_2(y_1 | y_2, \dots, y_n)$

$$= \left[ \{1 - (y_1 - y_n/2)\}^n \prod_{i=2}^n \{[1 - (y_i + y_1 - y_n/2)]^{r_i}\} \right] / \int_{y_n/2}^{1-y_n/2} \{1 - (y_1 - y_n/2)\}^n \prod_{i=2}^n \{[1 - (y_i + y_1 - y_n/2)]^{r_i}\} dy_1$$

$$v^* = \frac{\int_{(x_{2:3}-x_{1:3})/2}^{1-(x_{2:3}-x_{1:3})/2} y_1 \{1 - (x_{1:3} - x_{2:3})/2 - y_1\} dy_1}{\int_{(x_{2:3}-x_{1:3})/2}^{1-(x_{2:3}-x_{1:3})/2} \{1 - (x_{1:3} - x_{2:3})/2 - y_1\} dy_1}$$

...,  $y_n/2 < y_1 < 1 - (y_n/2)$ .

Now

$$v^* = E_0(\delta_0 | \mathbf{y}) = \frac{\int_{y_n/2}^{1-y_n/2} y_1 \{1 - (y_1 - y_n/2)\}^n \prod_{i=2}^n \{[1 - (y_i + y_1 - y_n/2)]^{r_i}\} dy_1}{\int_{y_n/2}^{1-y_n/2} \{1 - (y_1 - y_n/2)\}^n \prod_{i=2}^n \{[1 - (y_i + y_1 - y_n/2)]^{r_i}\} dy_1}$$

$$= \frac{\int_{(x_{2:3}-x_{1:3})/2}^{1-(x_{2:3}-x_{1:3})/2} y_1 dy_1 - \int_{(x_{2:3}-x_{1:3})/2}^{1-(x_{2:3}-x_{1:3})/2} y_1^2 dy_1}{\left[ \int_{(x_{2:3}-x_{1:3})/2}^{1-(x_{2:3}-x_{1:3})/2} dy_1 - \int_{(x_{2:3}-x_{1:3})/2}^{1-(x_{2:3}-x_{1:3})/2} y_1 dy_1 \right]}$$

in view of (3.2).

$$v^* = \frac{\int_{(x_{n:N}-x_{1:N})/2}^{1-(x_{n:N}-x_{1:N})/2} y_1 \prod_{i=1}^n \{1 - x_{i:N} + (x_{1:N} + x_{n:N})/2 - y_1\}^{r_i} dy_1}{\int_{(x_{n:N}-x_{1:N})/2}^{1-(x_{n:N}-x_{1:N})/2} \prod_{i=1}^n \{1 - x_{i:N} + (x_{1:N} + x_{n:N})/2 - y_1\}^{r_i} dy_1}$$

$$= \frac{[3\{1 - (x_{1:3} - x_{2:3})/2\}(1 - x_{1:3} + x_{2:3}) - 2\{[1 - (x_{2:3} - x_{1:3})/2]^3 - [(x_{2:3} - x_{1:3})/2]^3\}]}{3\{1 - (x_{1:3} - x_{2:3})\}^2}$$

Therefore the MRE estimator of  $\xi$  is given by

Therefore the MRE estimator of  $\xi$  is given by

$$\delta^*(\mathbf{X}) = \delta_0(\mathbf{X}) - E_0(\delta | \mathbf{y}) = (X_{1:N} + X_{n:N})/2 - v^* \quad \dots(3.4)$$

where  $v^*$  is as given in(3.3).

If  $N = 3, n = 2, r_1 = 1, r_2 = 0$ , then from (3.3) we obtain

$$\delta^*(\mathbf{X}) = (X_{1:3} + X_{2:3})/2 - \frac{[3\{1 - (x_{1:3} - x_{2:3})/2\}(1 - x_{1:3} + x_{2:3}) - 2\{[1 - (x_{2:3} - x_{1:3})/2]^3 - [(x_{2:3} - x_{1:3})/2]^3\}]}{3\{1 - (x_{1:3} - x_{2:3})\}^2}$$

Moreover, when the loss is squared error, the MRE estimator  $\delta^*(\mathbf{X})$  can be evaluated more explicitly by the Pitman form (Lehmann 1983 p.160).

Therefore the Pitman estimation of  $\xi$  is given by

$$\delta^*(\mathbf{X}) = \frac{\int_{x_{n:N}-1}^{x_{1:N}} u \prod_{i=1}^n (1-x_{i:N}+u)^{r_i} du}{\int_{x_{n:N}-1}^{x_{1:N}} \prod_{i=1}^n (1-x_{i:N}+u)^{r_i} du} \dots(3.5)$$

Taking  $u = \{(x_{1:N} + x_{n:N})/2 - y_1\}$  in (3.5),  $\delta^*(\mathbf{X})$  reduces to the following

$$\delta^*(\mathbf{X}) = \frac{\int_{(x_{n:N}-x_{1:N})/2}^{1-(x_{n:N}-x_{1:N})/2} \{(x_{1:N} + x_{n:N})/2 - y_1\} \times \prod_{i=1}^n \{1-x_{i:N} + (x_{1:N} + x_{n:N})/2 - y_1\}^{r_i} dy_1}{\int_{(x_{n:N}-x_{1:N})/2}^{1-(x_{n:N}-x_{1:N})/2} \prod_{i=1}^n \{1-x_{i:N} + (x_{1:N} + x_{n:N})/2 - y_1\}^{r_i} dy_1} \dots(3.6)$$

which coincides with the one given in (3.4).

**Remark 3.1:**

If  $r_k = 0, k = 1, 2, \dots, n-1$  and  $r_n = N - n$ , then the above estimator in (3.6) reduces to

$$\delta^*(\mathbf{X}) = \frac{(N - n + 1)X_{1:N} + X_{n:N} - 1}{(N - n + 2)},$$

which is same as the one obtain for type II right censored case given in (Leo Alexander, 2000).

**Case (ii) :** If the loss is absolute error, then the MRE estimator of  $\xi$  is

$$\delta^*(\mathbf{X}) = (X_{1:N} + X_{n:N})/2 - v_0,$$

where  $v_0$  is the median of the distribution with pdf given in (3.2).

**Case (iii):** Consider the location invariant Linex loss function (Varian, 1975)

$$L(\xi; \delta) = e^{a(\delta-\xi)} - a(\delta - \xi) - 1, \quad a \in R - \{0\}.$$

In order to find  $v^*$ , take

$$\begin{aligned} \delta_0(\mathbf{X}) &= (X_{1:N} + X_{n:N})/2, \text{ consider} \\ \delta_0(\mathbf{X}) &= E_0[e^{a(\delta_0 - v)} - a(\delta_0 - v) - 1 | \mathbf{y}] \\ &= e^{-av} E_0(e^{a\delta_0} | \mathbf{y}) - aE_0(\delta_0 | \mathbf{y}) + av - 1 \end{aligned}$$

$$\begin{aligned} &= e^{-av} \int_{(x_{n:N}-x_{1:N})/2}^{1-(x_{n:N}-x_{1:N})/2} e^{ay_1} \times \prod_{i=1}^n \{1-x_{i:N} + (x_{1:N} + x_{n:N})/2 - y_1\}^{r_i} dy_1 \\ &= e^{-av} \frac{\int_{(x_{n:N}-x_{1:N})/2}^{1-(x_{n:N}-x_{1:N})/2} e^{ay_1} \times \prod_{i=1}^n \{1-x_{i:N} + (x_{1:N} + x_{n:N})/2 - y_1\}^{r_i} dy_1}{\int_{(x_{n:N}-x_{1:N})/2}^{1-(x_{n:N}-x_{1:N})/2} \prod_{i=1}^n \{1-x_{i:N} + (x_{1:N} + x_{n:N})/2 - y_1\}^{r_i} dy_1} \end{aligned}$$

$$\begin{aligned}
 & \int_{(x_{n:N} - x_{1:N})/2}^{1 - (x_{n:N} - x_{1:N})/2} y_1 \times \\
 & \left\{ \prod_{i=1}^n \{1 - x_{i:N} + (x_{1:N} + x_{n:N})/2 - y_1\}^{r_i} dy_1 \right\} \\
 -a & \frac{\int_{(x_{n:N} - x_{1:N})/2}^{1 - (x_{n:N} - x_{1:N})/2} \left\{ \prod_{i=1}^n \{1 - x_{i:N} + (x_{1:N} + x_{n:N})/2 - y_1\}^{r_i} dy_1 \right\}}{\int_{(x_{n:N} - x_{1:N})/2}^{1 - (x_{n:N} - x_{1:N})/2} \left\{ \prod_{i=1}^n \{1 - x_{i:N} + (x_{1:N} + x_{n:N})/2 - y_1\}^{r_i} dy_1 \right\}} \\
 & + av - 1 \quad ,
 \end{aligned}$$

in view of (3.3). Thus  $v^*$  is to be obtain by minimizing  $R(\delta | \mathbf{y})$ . Hence the MRE estimator of  $\xi$  is given by

$$\delta^*(\mathbf{X}) = (X_{1:N} + X_{n:N})/2 - v^* .$$

**Remark 3.2:** If

$r_k = 0$ ,  $k = 1, 2, \dots, n-1$  and  $r_n = N - n$ , then the above estimator reduces to

$$\delta^*(\mathbf{X}) = (X_{1:N} + X_{n:N})/2 - (1/a) \{ \text{cgf of } (\delta_0 | \mathbf{y}) \text{ at } a \} ,$$

which is the one obtain under Linex loss function based on Type-II censored sample (Leo Alexander, 2000).

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### References

1. Aggarwala, R. Balakrishnan, N. Some Properties of Progressive Censored Order Statistics from Arbitrary and Uniform Distributions with Applications to Inference and Simulation. Journal of Statistical Planning and Inference 1998, 70, 35-49.
2. Balakrishnan, N. and Sandhu, R.A. (1996). Best linear unbiased and maximum likelihood estimation for exponential distributions under general progressive Type II censored samples, Sankhya, Series B, 58, 1-9.
3. Balakrishnan, N, Cohen, A.C. order Statistics and Inference: Estimation Methods; Academic Press: San Diego, 1991.
4. Balakrishnan, N. Aggarwala, R. Progressive Censoring: Theory, Methods And Applications; Birkhauser: Boston, 2000.
5. Balakrishnan, N. Cramer, E. Kamps, U. Schenk, N. Progressive Type-II Censored Order Statistics from Exponential distributions. Statistics 2001, 35, 537-556.
6. Basu, D. On Statistical Independent of complete Sufficient Statistic. Sankhya 1955, 15, 377-380.
7. Chandrasekar, B. Edwin Prabakaran, T. Equivariant Estimation for Vector Location Parameter; Proceedings of the XV ISPS Conference, 22-26, MS University, India, 1995.
8. Cohen, A.C, Progressively Censored Samples in Life Testing. Technometrics 1963, 5, 327-329.
9. Cohen, A.C, Truncated and Censored Samples; Marcel Dekker: New York, 1991.
10. Cramer, E. Kamps, U. Estimation with sequential order Statistics from Exponential Distributions. Annals of the Institute of statistical Mathematics 2001, 53, 307-324.
11. Edwin Prabakaran, T. Chandrasekar, B. Simultaneous Equivariant Estimation for Location-Scale Models. Journal of Statistical Planning and Inference 1994, 40, 51-59.
12. Edwin Prabakaran, T., Contributions to theory of simultaneous Equivariant estimation. Ph.D. thesis, University of Madras, 1995.
13. Lehmann, E.L. Casella, G. Theory of Point Estimation, Second Edn. Springer-Verlag: New York, 1998.
14. Herd, G.R. Estimation of Reliability from Incomplete Data, Proceedings of the sixth National Symposium on Reliability and Quality Control, 1960.
15. Kamps, U. A Concept of Generalized Order Statistics; Stuttgart, teubner, Germany, 1995.
16. Kamps, U. A Concept of Generalized Order Statistics; Journal of Statistical Planning and Inference 1995, 48, 1-23.
17. Leo Alexander, T. Contributions to Simultaneous Equivariant estimation based on censored samples. ( Ph.D. Thesis ).

18. Leo Alexander, T. Chandrasekar, B. Equivariant Estimation for the Parameters of an Exponential Models Based on Censored Sampling. Biometric Journal 1999,41,471-481.
19. Roberts, H.R. Some Results in Life Testing Based on Hypercensored Samples from an Exponential Distribution , Ph.D. dissertation , George Washington University, Washington Dc, 1962.
20. Roberts, H.R. Life Test Experiments Hypercensored Samples; Proceedings of the 18<sup>th</sup> annual Quality Control Conference, Rochester Society for Quality Control, Rochester, New York, 1962.
21. Tse, S.K. G. Efficiencies of Maximum Likelihood Estimators Under Censoring in Life Testing. Journal of Applied Statistics 1996, 23, 515-524.
22. Viveros, R. Balakrishnan,N. Interval Estimation of Life Characteristics from Progressively Censored Data. Technometrics 1994, 36, 84-91.
23. Zacks, S., The Theory of Statistical Inference. John Wiley and Sons, New York, 1971.