Estimation in Censored Sample Using Linex Loss Function

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Abstract: - In life testing experiment, it is common practice to terminate the experiment when certain numbers of items failed or a stipulated time has elapsed. In order to overcome these situations terminate most of the life tests before all the items fail. Pandey (1977) considered an exponential distribution which is generally used in life testing experiment. Epstein & Sobel (1953) and Bhattacharya & Srivastava (1974) have also considered the censoring procedure in life testing problem. Pandey and Malik (1994) proposed an improve estimator for variance in exponential distribution. In this paper proposes an estimator for variance in case of exponential distribution and studies its properties under Linex loss function.

Keywords: - Linex Loss Function, Censoring, Mean Square Error

I. INTRODUCTION

In life testing experiment, it is common practice to terminate the experiment when certain numbers of items failed or a stipulated time has elapsed. In order to overcome these situations terminate most of the life tests before all the items fail [Sinha (1980, 1986)]. To avoid this, many life tests are terminated before all the items fail. Let us consider the first n ordered statistics $x_{(1)}, x_{(2)}, \ldots, x_{(n)}$ in a sample of size (r), in this case the sample is censored on the right which is also called as Type II censoring. Similarly censoring on the left is called Type I censoring. For example, in many biological experiments, r samples from each living things are tested for antibodies after a certain period of time, only ‘n’ of these samples contain measurable amounts while remaining (n-r) of the animal develop the antigen at a level too low for measurements. Cohen (1965) and Srivastava (1976) have obtained the likelihood estimates of the parameters in case of continuous distributions such as exponential, normal, log-normal and logistic when samples are progressively censored under the assumption that the parameters of the distribution under consideration might change at each stage of censoring.

In such experiments after first stage, the experiment is continued with the remaining surviving items. The idea of removing some items at every stage of censoring stamps from the fact that these items might be required for use somewhere else for related experimentation. Bhattacharya and Srivastava (1974) have also proposed type I and type II error for censoring process.

Let $x_1, x_2, \ldots, x_n$ be a random sample of size n from normal distribution with mean $\theta$ and variance $\sigma^2$. In some practical situations, prior value of mean and variance are $\theta_0$ and $\sigma_0^2$ respectively. The preliminary estimator for the mean was suggested by Bancroft (1944) as:

If $\theta_0$ is the estimate of $\theta$ if $H_0: \theta = \theta_0$ is accepted, otherwise consider usual estimator $\bar{x}$ or in other words

$$\theta_{PT} = \begin{cases} \theta_0 & \text{if } H_0: \theta = \theta_0 \text{ is accepted} \\ \bar{x}, & \text{otherwise} \end{cases}$$

(1.1)

According to mean square error criterion, $MSE(\theta_0) = E(\theta_0 - \theta)^2$ and $MSE(\bar{x}) = \frac{\sigma^2}{n}$

Katti (1962) suggested that $MSE(\bar{x}) = \frac{\sigma^2}{n} \leq MSE(\theta_0) = E(\theta_0 - \theta)^2$ if $\bar{x} \in R$ where R is determined by the testing of hypothesis of $\theta = \theta_0$.

Also,

$$P \left[ \frac{E_1 - \theta_0}{\frac{\sigma}{\sqrt{n}}} \leq \frac{\bar{x} - \theta_0}{\frac{\sigma}{\sqrt{n}}} \leq E_1 \right] = 1 - \alpha$$

(1.2)
where $\alpha$ is the level of significance. Now equation (1.2) can be re-written as

$$P\left[\theta_0 - \frac{\sigma}{\sqrt{n}} \leq \bar{x} \leq \theta_0 + \frac{\sigma}{\sqrt{n}} \epsilon_1\right] = 1 - \alpha$$

(1.3)

Here distribution of $\bar{x}$ is known.

Pandey (1977) considered exponential distribution which is generally used in life testing distributions having mean $\theta$ and variance $\theta^2$. The improve estimator for $\theta$ is

$$P_1 = k\bar{x} + (1-k)\theta_0, 0 \leq k \leq 1$$

The value of $k$ for which risk will be minimum is

$$k_{\text{min}} = \frac{(\theta - \theta_0)^2}{(\theta - \theta_0)^2 + \frac{\theta^2}{n}} = \left(1 - \frac{\theta_0}{\theta}\right)^2 + \frac{1}{n}$$

which is less than one.

Here the value of $k$ depends on $n$ and $\frac{\theta_0}{\theta}$. If $\frac{\theta_0}{\theta} = 1$ then $k_{\text{min}} = 0$ and estimator is $\theta_0$.

The magnitude of relative efficiency will be maximum at $\frac{\theta_0}{\theta} = 1$ for smaller level of significance ($\alpha$) and smaller value of $n$.

Epstein and Sobel (1953) and Bhattacharya and Srivastava (1974) have also discussed censoring procedure in life testing problem and proposed an estimator as

$$H_{1PT}^\wedge = \begin{cases} \theta_0 & \text{if } H_0 : \theta = \theta_0 \text{ is accepted} \\ cT_r, & \text{otherwise} \end{cases}$$

Since $\frac{2n^2}{\theta_0}$ follows chi-square distribution with $2n$ degree of freedom. Also if $x_{(1)} \leq x_{(2)} \leq \cdots \leq x_{(r)}$ is the censored data of type II censoring then,

$$T_r = \sum_{i=1}^{r} x_{(i)} + (n-r)x_{(r)}$$

(1.4)

Since $\frac{2nT_r}{\theta}$ follows chi-square distribution with $2r$ degree of freedom i.e.

$$E\left[\frac{2T_r}{\theta}\right] = 2r, V\left[\frac{2T_r}{\theta}\right] = 4r, E\left[\frac{T_r}{r}\right] = \theta, V\left[T_r\right] = \theta^2$$

If case of two populations with common mean, the unbiased estimator for common mean in exponential distribution is

$$T_1 = \frac{n_1\bar{x}_1 + n_2\bar{x}_2}{n_1 + n_2}$$

$$T_2 = \frac{n_1\bar{x} + n_2\bar{y}}{(n_1 + n_2)^2 + 5(n_1 + n_2) + 6}$$

Pandey and Malik (1994) have considered the improve estimator for $\theta^2$ in exponential distribution as
which is in the form of improve estimator of $\theta^2$ in exponential distribution as suggested by Pandey and Singh (1977). Thus

$$T_i = \frac{n^2 x^2}{(n+2)(n+3)}$$

Let us consider displaced exponential distribution having density function

$$f(x, A, \theta) = \frac{1}{\theta} e^{-(x-A)/\theta}, \quad x > A, \; \theta > 0$$

Here A is location and $\theta$ is scale parameter. The maximum likelihood estimators for $\theta$ and A are $x_{(1)}$ and $x_{(1)}$ respectively.

Also $\frac{2n(\pi - x_{(1)})}{\theta}$ follows chi-square distribution with 2(n-1) degrees of freedom and other researchers have also considered preliminary estimators/adaptive estimators/ conditional specifications [Han et al. (1988), Hogg (1974), Pandey and Singh (1977), Hirano (1973)].

In section 2, proposes an estimator for variance ($\theta^2$) in case of exponential distribution as

$$Y_2 = c \left[ \frac{T_h + T_{r_1}}{r_1 + r} \right]^2$$

which provide $Y_3 = w_1 T_h^2 + w_2 T_{r_1}^2 + w_3 T_h T_{r_1}$ and studies its properties under Linex loss function.

II. ESTIMATION OF VARIANCE IN EXPONENTIAL DISTRIBUTION UNDER LINEX LOSS FUNCTION

Pandey and Singh (1977) have proposed improve estimator for $\theta$ as

$$\hat{\theta} = \begin{cases} \theta_0 & \text{if } H_0 : \theta = \theta_0 \text{ is accepted} \\ \frac{1}{a} \left( 1 - e^{-\frac{a}{r_{r_1}}} \right) T_r, & \text{otherwise} \end{cases}$$

(2.1)

Similarly if considered the improve estimator for $\theta^2$ as

$$\hat{\theta}^2 = \left[ \frac{T_h + T_{r_1}}{r_1 + r} \right]^2$$

Pandey and Malik (1994) have also proposed improve estimator for $\theta^2$ as

$$Y_2 = w_1 T_h^2 + w_2 T_{r_1}^2 + w_3 T_h T_{r_1}$$

(2.2)

The MSE criterion will provide

$$MSE(Y_2) = \theta^4$$

Let $K = r_1 (1 + r_1) (r_1 + 2) (r_1 + 3), P = r_2 (r_2 + 1) (r_2 + 2) (r_2 + 3), Q = r_1 (1 + r_1) r_2 (r_2 + 1)$

$$T = r_1 r_2 (r_2 + 2), S = r_1 (1 + r_1) r_2 (r_1 + 2), V = r_1 (1 + r_1), U = r_2 (r_2 + 1), M = r_1 r_2$$

Thus,

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\[ \text{MSE}(Y_2) = \left[ w_1^2 K + 2w_1w_2Q + 2w_1w_3S + 2w_3w_2T + w_3^2Q + w_1^2P - 2[w_1V + w_2U + w_3M] + 1 \right] \]

The value of \( w_1, w_2 \) and \( w_3 \) for which MSE \((Y_2)\) will be minimum, thus

\[
w_1 = \frac{D_1}{D}, w_2 = \frac{D_2}{D} \quad \text{and} \quad w_3 = \frac{D_3}{D}
\]

where

\[
D = K(PQ-T^2) - Q(Q^2-ST) + S(QT-SP), D_1 = V(PQ-T^2) - Q(UQ-MT) + S(UT-MP)
\]

\[
D_2 = K(UQ-MT) - V(Q^2-ST) + S(QM-SU) \quad \text{and} \quad D_3 = K(PM-TU) - Q(QM-SU) + V(QT-SP)
\]

In some cases, the over (under) estimation may exist and Varian (1975) proposed Linex (linear-exponential) loss function which may be appropriate. The Linex loss function which rises exponentially on one side of zero and almost linearly on the other side of zero. This loss function reduces to square error loss for value of a near to zero. The Linex loss function may be

\[
L(a, \Delta) = (e^{a\Delta} - a\Delta - 1), \quad a \neq 0
\]

Here \( a \) is shape parameter. If \( |a| \to 0 \), the Linex loss reduces to square error.

Using Linex loss function for the estimator \( Y'_1 = w_1T_{r_1}^2 + w_2T_{r_1}^2 + w_3T_{r_1}T_{r_2} \)

Thus,

\[
L(a, \Delta) = (e^{a\Delta} - a\Delta - 1), \quad a \neq 0
\]

\[
L(a, \Delta) = e \left[ \frac{w_1T_{r_1}^2 + w_2T_{r_2}^2 + w_3T_{r_1}T_{r_2}}{a^2} - 1 \right] - \left[ \frac{w_1T_{r_1}^2 + w_2T_{r_2}^2 + w_3T_{r_1}T_{r_2}}{a^2} - 1 \right] - 1
\]

which has

\[
\frac{2}{a^2} R(a, \Delta) = \left[ (1-a)\left( w_1^2 K + 2w_1w_2Q + 2w_1w_3S + 2w_3w_2T + w_3^2Q + w_1^2P \right) \right] - \left[ (a-2)\left( w_1V + w_2U + w_3M \right) \right] + 1 - a
\]

The value of \( w_1, w_2 \) and \( w_3 \) for which MSE \((Y'_1)\) will be minimum, then

\[
w_1 = \frac{D'_1}{D'}, w_2 = \frac{D'_2}{D'} \quad \text{and} \quad w_3 = \frac{D'_3}{D'}
\]

where

\[
D = K(PQ-T^2) - Q(Q^2-ST) + S(QT-SP), D_1 = \left( 1 - \frac{a}{2} \right) \left( V(PQ-T^2) - Q(UQ-MT) + S(UT-MP) \right)
\]

\[
D_2 = \left( \frac{1-a}{2} \right) K(UQ-MT) - V(Q^2-ST) + S(QM-SU)
\]

and \( D_3 = \left( \frac{1-a}{2} \right) K(PM-TU) - Q(QM-SU) + V(QT-SP) \)

Table 2.1 to 2.4, represents the relative efficiency of the estimator \( Y'_1 \) with respect to \( Y_2 \) for \( a = .1(.1).4, r_1 = 3,4,5,6, \) \( r_2 = 3, 4, 5, 6. \) The figure 2.1 to 2.4 shows that for small values of \( a \) and for smaller values of \( r_1, r_2 \) the proposed estimator performs better. The preliminary testimator may be also proposed and properties can be studied in future.
REFERENCES


APPENDICES

Figure 2.1 Relative Efficiency of Testimator $Y^1$ w.r.to $Y^2$ when $a = 0.1$

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- $r1=3$
- $r1=4$
- $r1=5$
- $r1=6$
Figure 2.2 Relative Efficiency of Testimator $Y_3$ w.r.to $Y_2$ when $a = 0.2$

Figure 2.3 Relative Efficiency of Testimator $Y_3$ w.r.to $Y_2$ when $a = 0.3$

Figure 2.4 Relative Efficiency of Testimator $Y_3$ w.r.to $Y_2$ when $a = 0.4$