# Goodness-of-fit tests based on Lorenz curve for progressive censored data from a location-scale distribution

# **Author and Position**

☐ Hyein Koo Graduate student, Department of Statistics, Daegu University

☐ Kyeongjun Lee Assistant professor, Division of Mathematics and Big Data Science,

Daegu University

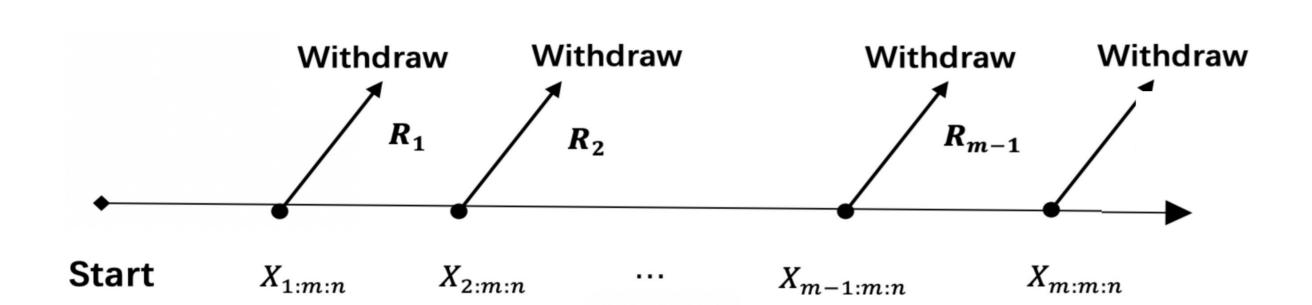
☐ Namjin Beak Undergraduate student, Division of Mathematics and Big Data Science,

Daegu University

# **Text**

#### 1. Introduction

- \* The progressively Type II censored (PC) have become fairly common in a reliability and lifetime-testing experiment as follows. R<sub>1</sub> surviving items are removed from the test at random after 1st observed failure time; furthermore, R<sub>2</sub> surviving items are then removed from the test at random after the 2<sup>nd</sup> observed failure time. Keep repeating this process until all the remaining  $R_m = n-R_1 - \cdots - R_{m-1} - m$  items are removed from the test immediately after the following mth observed failure time. In test, the PC scheme  $R = (R_1, R_2, ..., R_m)$  is pre-fixed. For this reason the m ordered observed failure times, which we denote by  $X_{1:m:n}$ ,  $X_{2:m:n}$ , . . . ,  $X_{m:m:n}$ , are referred to as PC.
- \* When the observed failure time data are PC data, the goodness-of-fit tests for perfect data can no longer be used. In this motive, the goodness-of-fit test under PC has received the attention from numerous authors.



### 2.1 Lorenz curve

- ❖ In this section explain to Lorenz curve. (LC), which is a standard Lorenz curve scaled up by the mean.
- \* Lorenz curve provides the means to evaluate income or wealth disparity between two distributions. Let F denotes the CDF of income or wealth distribution. Then the income or wealth is assumed to be non-negative. For a given percentile p, let;

$$\mathcal{F}^{-1}(p)=\inf\{y\,|\mathcal{F}(y)\geq p\}, 0\leq p\leq 1\quad,$$

denotes the inverse CDF corresponding to 3 .

It shall be assumed all through that F is continuous CDF with finite support.

The Lorenz curves corresponding to the distributions with F is defined as;

$$L(p) = \frac{1}{\mu} \int_0^{\mathscr{F}^{-1}} x \, d\mathscr{F}(x) ,$$

where  $\mu$  denotes the mean of the distribution with  $\pmb{\mathcal{F}}$ .

With these process the Lorenz curve corresponding to F is the Lorenz curve scaled up by the mean  $\mu$  .

## 2.2 LC with a location-scale distribution

❖ To test whether the PC data comes from a location-scale distribution, let above the PC data with PC scheme from a location-scale distribution. Also, the PC data have a location-scale distribution with a probability density function (PDF).

$$f(x; \mu, \sigma) = \frac{1}{\sigma} g\left(\frac{x - \mu}{\sigma}\right)$$
,

where  $g(\cdot)$  is the known function, but each  $\mu$  and  $\sigma$  is the unknown location scale parameter.

$$H_0: \mathscr{F} \in f_\theta \ for \ some \ \theta \in \Theta = \{(\mu, \sigma) | -\infty < \mu < \infty, \sigma > 0\}$$

where  $\mathcal{F} = \mathcal{F}$  (x;  $\mu$ ,  $\sigma$ ) denote the distribution function.

 $\bullet$  If  $U_{i:m:n} = \mathscr{F}(X_{i:m:n}; \mu, \sigma)$ , then  $p_{i:m:n} = E(U_{i:m:n})$  denote the expected value of the i<sup>th</sup> PC order statistics from the standard uniform distribution, which is given by;

$$p_{i:m:n} = 1 - \prod_{j=m-i+1}^{m} \left\{ \frac{j + R_{m-j+1} + \dots + R_m}{j + 1 + R_{m-j+1} + \dots + R_m} \right\} .$$

Since a Lorenz curve cannot show the characteristics of the skewed distribution, the above result is multiplied by  $(1 - p_{i:m:n})$ . Then mLC is obtained as;

$$mLC(p_{jm:n}) = \frac{\sum_{i=1}^{j} X_{i:m:n} - X_{1:m:n}}{\sum_{i=1}^{m} X_{i:m:n} - X_{1:m:n}} - p_{j:m:n} + 1 .$$

**❖** Let **F** denote the CDF of location-scale distribution, an nLC<sup>+</sup> and nLC<sup>−</sup> are obtained as;

$$nLC^{+}\left(p_{j:m:n}\right) = \frac{mLC\left(p_{j:m:n}\right)}{mLC_{f}\left(p_{j:m:n}\right)},$$

$$nLC^{-}\left(p_{j:m:n}\right) = \frac{mLC_{f}\left(p_{j:m:n}\right)}{mLC_{f}\left(p_{j:m:n}\right)} ,$$

where

$$mLC_f(p_{j:m:n}) = \frac{\sum_{i=1}^{j} F^{-1}(p_{i:m:n}) - F^{-1}(p_{1:m:n})}{\sum_{i=1}^{m} F^{-1}(p_{i:m:n}) - F^{-1}(p_{1:m:n})} - p_{j:m:n} + 1.$$

Here, the mLC, nLC<sup>+</sup>, and nLC<sup>-</sup> are clearly location-scale invariant.

$$G_{m:n}^{+} = \max_{1 \le i \le m} [|1 - nLC^{+}(p_{j:m:n})|]$$
,  $G_{m:n}^{-} = \max_{1 \le i \le m} [|1 - nLC^{-}(p_{j:m:n})|]$ ,

and  $G_{m:n}^+ = G_{m:n}^+ + G_{m:n}^-$ .

- $\bullet$  If the data accurately follows a location-scale distribution we expect the  $G_{m:n}^+$ ,  $G_{m:n}^$ and  $G_{m\cdot n}$  test statistics to be zero.
- **❖** Suggest new plot methonds using nLC<sup>+</sup> and nLC<sup>−</sup>;

$$g^{+}(p_{j:m:n}) = |1 - nLC^{+}(p_{j:m:n})|$$
,  $g^{-}(p_{j:m:n}) = |1 - nLC^{-}(p_{j:m:n})|$ .

- 3. Comparison of the simulated power values
- **❖** We generated 10,000 samples for different choices of sample sizes and PC schemes.

First, a normal distribution with the parent distribution with the PDF;

$$f(x; \mu, \sigma^2) = \frac{1}{\sigma \sqrt{2\pi}} \exp -\left\{\frac{(x-\mu)^2}{2\sigma^2}\right\}, -\infty < \mu < \infty, \sigma > 0$$

as the parent distribution.

The normal distribution, the alternative distribution is considered t distribution with PDF;

$$f(\mathbf{x}; \gamma) = \frac{\tau\left[\frac{\nu+1}{2}\right]}{\sqrt{\nu\pi}\tau\left(\frac{\gamma}{2}\right)} \left(1 + \frac{\mathbf{x}^2}{\gamma}\right), -\infty < \mathbf{x} < \infty, \gamma > .$$

Next, a Gumbel distribution with the parent distribution with the PDF;

$$f(x;\mu,\sigma^2) = \frac{1}{\sigma} \exp\left\{\frac{x-\mu}{\sigma}\right\} \exp\left[-\exp\left[\frac{x-\mu}{\sigma}\right]\right], -\infty < x < \infty, -\infty < \mu < \infty, \sigma > 0, \sigma > 0$$

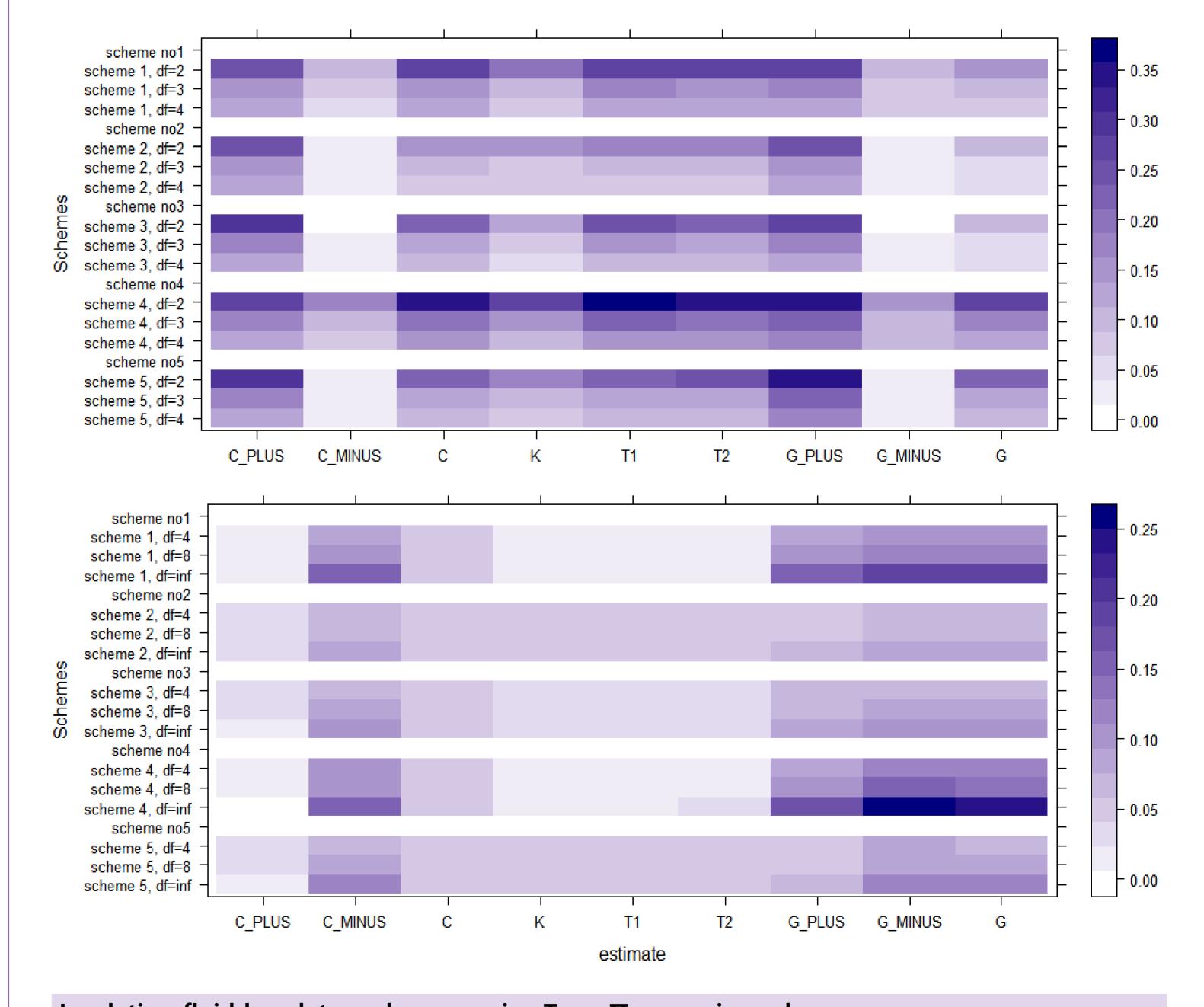
as the parent distribution.

For testing the normal distribution, the alternative distribution is considered log-gamma distribution with PDF;

$$f(\mathbf{x};\mathbf{k}) = \frac{\mathbf{k}^{\mathbf{k} - \frac{1}{2}}}{\tau(\mathbf{k})} \exp\left[\sqrt{\mathbf{k}\mathbf{x} - \mathbf{k} \exp\left\{\frac{\mathbf{x}}{\sqrt{\mathbf{k}}}\right\}}\right], -\infty < \mathbf{x} < \infty, \mathbf{k} > 0.$$

## 4. Simulation Study

\* The mean squared error (MSE) of the estimators are simulated by Monte Carlo method based on 10,000 runs sample 5 type scheme and 3 different df. We compare the estimators in the sense of the estimation for different censored schemes.



Insulating fluid log data and progressive Type II censoring scheme												
i	1	2	3	4	5	6	7	8				
$X_{i:m:n}$	-1.661	-0.249	-0.041	0.270	1.022	1.579	2.872	1.99				
$R_i$	0	0	3	0	3	0	0	5				

						3.1.3.1.3.1.1. <b>9</b>			
Criterion	C <sub>m:n</sub>	C <sub>m:n</sub>	C <sub>m:n</sub>	$K_{m:n}$	$T_{m:n}^{(1)}$	$T_{m:n}^{(2)}$	G <sup>+</sup> <sub>m:n</sub>	G <sub>m:n</sub>	G <sub>m:n</sub>
Test statistic	0.080	0.068	0.080	0.148	0.001	0.042	0.076	0.071	0.147
p – value	0.563	0.450	0.648	0.564	0.535	0.516	0.794	0.813	0.804

Test statistics and the corresponding p-values for the insulating fluid log data