

# Comparisons of point estimation methods in 3-parameter Weibull and 3-parameter Gamma distributions in the presence of outliers

Eldesoky E. Afify

Dept. of Mathematics, Faculty of Eng., Menoufiya University, Shibeen El-Kom, Egypt

This paper is concerned with the estimation of parameters of Weibull and Gamma distributions. For each of these distributions, we estimate the parameter by moment equations. Also we compare between estimates in the case of outliers. We use the coefficient of skewness to find the approximate value of the shape parameter of Weibull distribution. Numerical examples are given in which the parameters are estimated from complete data.

في هذا البحث تم تقدير معالم كل من توزيع وايبيل وتوزيع جاما ذو ثلاث الوسيطات باستعمال معادلات العزوم. معامل الالتواء هو الذي استعمل لتقدير تلك الوسيطات الثلاث لكل من التوزيعين. وقد تم عمل مقارنة بين تقدير الوسيطات في وجود وعدم وجود نقط خارجية وتبين من الدراسة أنه في حالة وجود نقط خارجية فإنه يقل الخطأ النسبي ولذا فإنه يجب معالجة تلك النقط الخارجية. استعمال معامل التغير أو معامل الالتواء بيسهل الحسابات ويأخذ وقت قصير في استعمال الحاسب عن الطرق الأخرى وقد تم توضيح ذلك بمثال عددي ويمكن تحديد معامل وايبيل ذو الأربع وسيطات باستعمال معامل التفرطح.

**Keywords:** Weibull distribution, Gamma distribution, Moment method, Skewness, Outliers

## 1. Introduction

The Weibull distributions have in recent years assumed a position of importance in the field of reliability and life testing. Various problems associated with these distributions have been considered by numerous writers, among them are Harter and Moore [1] used the maximum likelihood to estimate the parameters of 3 parameter Gamma and 3 parameter Weibull distributions. Cohen and Whitten [2] used the maximum likelihood and modified maximum likelihood to estimate the parameter of 3 parameter Gamma distribution. and A.C. Cohen [3] used maximum likelihood for estimating the parameters of 2 parameter Weibull distribution, also he derived the coefficient of variation to find the approximate value of the shape parameter.

Rosaiah, et al. [4] used the maximum likelihood and modified maximum likelihood, defined by [4] and Batakishuan [5] to estimate the parameters of Gamma distribution.

This paper is concerned with moment estimation in the case of complete samples

from three parameter Weibull and three parameter Gamma distributions, We obtain the parameter estimates of Gamma distribution in terms of the sample moments. Also coefficient of skewness is used to find the approximate value of the shape parameter of Weibull distribution.

## 2. Gamma population - mathematical formulation

The probability density function of the random variable  $x$  having a Gamma distribution with location parameter  $\alpha \geq 0$ , scale parameter  $\delta$  and shape parameter  $p$  is given by:

$$f(x, \alpha, \delta, p) = \begin{cases} \frac{1}{\delta \Gamma(p)} \left( \frac{x - \alpha}{\delta} \right)^{p-1} \exp \left[ - \left( \frac{x - \alpha}{\delta} \right) \right] & \delta, p > 0, \quad \alpha < x < \infty \\ 0 & \text{Otherwise} \end{cases}$$

The  $r$ th moment of the distribution  $\mu_r^1$  can be obtained from the relation:

$$\mu_r^{-1} = E(x^r) = \int_{-\infty}^{\infty} x^r f(x) dx,$$

$$\mu_r^{-1} = \sum_{i=0}^r \binom{r}{i} \alpha^i \delta^{r-i} \frac{\Gamma(p+r-i)}{\Gamma(p)}.$$

From the last equation, the first three moments can be obtained as:

$$\mu_1^{-1} = \delta p + \alpha, \tag{1}$$

$$\mu_2^{-1} = \delta^2 p(p+1) + 2\alpha\delta p + \alpha^2, \tag{2}$$

$$\mu_3^{-1} = \alpha^3 p(p+1)(p+2) + 3\alpha\delta^2 p(p+1) + 3\alpha^2 p\delta + \alpha^3. \tag{3}$$

The first three moments of the sample  $m_1^{-1}$ ,  $m_2^{-1}$ , and  $m_3^{-1}$  can be obtained from the relation;

$$m_r^{-1} = \frac{1}{n} \sum_{i=1}^n x_i^r.$$

Where  $m_r^{-1}$  is the rth moment of the sample.

By equating the first three moments of the sample to the corresponding first three moments of the distribution and solving for  $\alpha$ ,  $p$  and  $\delta$  we have,

$$m_2^{-1} - m_1^{-2} = \delta^2 p, \tag{4}$$

$$m_3^{-1} - m_1^{-3} = 2\delta(\delta^2 p) + 3(\delta^2 p)(\delta p + \alpha), \tag{5}$$

$$\hat{\delta} = \frac{1}{2} \left[ \frac{m_3^{-1} - m_1^{-3} - 3m_1^{-1}(m_2^{-1} - m_1^{-2})}{m_2^{-1} - m_1^{-2}} \right], \tag{6}$$

$$\hat{p} = \frac{4(m_2^{-1} - m_1^{-2})^3}{(m_3^{-1} - 3m_1^{-1}m_2^{-1} + 2m_1^{-3})^2}, \tag{7}$$

$$\hat{\alpha} = m_1^{-1} - \frac{2(m_2^{-1} - m_1^{-2})^2}{m_3^{-1} - 3m_1^{-1}m_2^{-1} + 2m_1^{-3}}. \tag{8}$$

From eqs. (6), (7) and (8) the estimates of  $\hat{\delta}$ ,  $\hat{p}$  and  $\hat{\alpha}$  can be estimated directly, on the other hand, we can derive,

$$\mu_3 = \mu_3^1 - 3\mu_2^1\mu_1 + 2\mu_1^3,$$

$$\mu_3 = 2\delta^3 p, \tag{9}$$

$$\text{Var}(x) = \mu_2^1 - \mu_1^2 = \delta^2 p.$$

The coefficient of skewness (SK) of the distribution is given by:

$$\text{Skewness} = \frac{\mu_3}{\sigma^3} = \frac{2}{\sqrt{p}},$$

where  $\sigma$  is the standard deviation.

Equating skewness of the sample to  $\frac{2}{\sqrt{P}}$ ,

can obtain  $p$  then substitute in eq. (9) about  $p$ , we will have  $\delta$ , from eq. (1)  $\alpha$  can be obtained.

### 3. Weibull population, mathematical formulation

The probability density function of random variable  $x$  having a Weibull distribution with location parameter  $\gamma$ , scale parameter  $\beta$  and shape parameter  $c$  is given by :

$$f(x, \beta, \gamma, c) = \begin{cases} \frac{c(x-\gamma)^{c-1}}{\beta^c} e^{-\left(\frac{x-\gamma}{\beta}\right)^c} & \gamma < x < \infty, \beta, c > 0 \\ 0 & \text{otherwise} \end{cases}$$

The first three moments can be obtained as:

$$\mu_1 = \beta\Gamma(1+1/c) + \gamma,$$

$$\mu_2 = \beta^2\Gamma(1+2/c) + 2\gamma\beta\Gamma(1+1/c) + \gamma^2, \tag{10}$$

$$\mu_3 = \beta^3\Gamma(1+3/c) + 3\gamma\beta^2\Gamma(1+2/c)$$

$$+ 3\gamma^2\beta\Gamma(1+1/c) + \gamma^3,$$

$$\text{Var}(x) = \mu_2 - \mu_1^2,$$

$$= \beta^2 [\Gamma(1+2/c) - \Gamma^2(1+1/c)], \tag{11}$$

$$\begin{aligned} \delta^3 &= \beta^3 [\Gamma(1+2/c) - \Gamma^2(1+1/c)]^{3/2}, \\ \mu_3 &= \mu_3^1 - 3\mu_2^1 \mu_1^2 \mu_1^3, \\ &= \beta^3 [\Gamma(1+3/c) - 3\Gamma(1+2/c) \Gamma(1+1/c) \\ &\quad - 2\Gamma^3(1+1/c)], \end{aligned}$$

$$\begin{aligned} \text{skewness} &= \frac{\mu_3}{\delta^3} \\ &= \frac{[\Gamma(1+3/c) - 3\Gamma(1+2/c)\Gamma(1+1/c) + 2\Gamma^3(1+1/c)]}{[\Gamma(1+2/c) - \Gamma^2(1+1/c)]^{3/2}}, \\ \text{SK.} &= \frac{3c^2\Gamma(3/c) - 6c\Gamma(2/c)\Gamma(1/c) + 2\Gamma^3(1/c)}{[2c\Gamma(2/c) - \Gamma^2(1/c)]^{3/2}}. \quad (12) \end{aligned}$$

The coefficient of skewness given by eq. (12) is a function of the shape parameter  $c$  only, the values of skewness (SK) of the sample when there are no outliers and when we have an outlier are  $-0.257$  and  $2.528$ , respectively, by equating the right hand side of eq. (12) to skewness of the sample when there are no outliers and when we have an outlier and solve each of those equations by iterative method or by trial and error, we obtain the approximate value of the shape  $c$ , with tolerance of  $0.02$ , we can form the following table 2.

Table 1  
Parameter estimates of 3 parameter gamma distribution

N	$\theta$	True value	MEN	MEO	REN	REO
10	$\hat{\alpha}$	20	-365.25	-19.54	0.679	0.067
	$\hat{\delta}$	50	14.821	196.72	0.2693	0.142
	$\hat{P}$	30	29.77	0.659	0.376	0.127
20	$\hat{\alpha}$	20	-454.21	13.27	0.679	0.067
	$\hat{\delta}$	50	23.77	203.4	0.2693	0.142
	$\hat{P}$	30	23.69	0.554	0.376	0.127

$\theta$  is the vector of parameter estimates.  
 N is number of observations.  
 MEN (moment estimates without outliers)  
 MEO (moment estimates with outliers)  
 REN (relative error without outliers)  
 REO (relative error with outliers)

Table 2  
Approximation of the shape parameter  $c$  when  $N = 10$ ,  $SK = -0.257$  and  $SK = 2.528$

C	SK
0.7	2.16
0.75	2.19
0.8	2.8146
0.85	2.5801
0.9	2.725
1	2
2	0.631
3	0.137
3.5	-0.0442
4	-0.085
4.2	-0.2536
4.5	-0.856
4.7	-0.871
4.8	0.3262
5	0.833

From the above table 2, we find that the shape parameter  $c=4.2$  when there are no outliers and  $c=0.85$  when we have an outlier. When we substitute in eq. (11) we have the parameter estimate  $\beta$ , then substitute in eq. (10), we will obtain on the parameter estimate  $\gamma$ .

Table 3  
Parameter estimates of 3- Parameter Weibull distribution

N	$\theta$	True value	MEN	MEO	REN	REO
10	$\hat{\beta}$	100	270.384	87.82	0.17	0.012
	$\hat{\gamma}$	20	-180.882	-8.685	0.948	0.143
	$\hat{C}$	3	4.2	0.85	0.04	0.072

#### 4. Conclusions

1. Since A. C. Cohen used the coefficient of variation to find the approximate value of the shape parameter of 2- parameter Weibull distribution and in this paper we use the coefficient of skewness to obtain the approximate value of the shape parameter of 3- parameter Weibull distribution, one can use the coefficient of kurtosis to obtain the approximate value of the shape parameter of 4 parameter Weibull distribution.
2. Outliers reduce the values of the relative errors, and we have to remove these observations, which represent the outliers.
3. The use of coefficient of variation or coefficient of skewness is easy in computation and takes few time on computer than maximum likelihood calculations.

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From the above Table 2, we find that the shape parameter  $\alpha$  and  $\beta$  are not outliers here and  $\theta = 0.85$  when we have an outlier. When we substitute in eq. (1) we have the parameter estimate  $\hat{\theta}$ , then substitute in eq. (2) we will obtain on the parameter estimate  $\hat{\theta}$ .

Table 2  
 Parameter estimates of 3-parameter Weibull distribution

N	True value	MEM	MEMO	REM	REMO
10	0.10	0.100	0.100	0.10	0.10
20	0.10	0.100	0.100	0.10	0.10
30	0.10	0.100	0.100	0.10	0.10

4. Conclusions

1. Since A. C. Cohen used the coefficient of variation to find the approximate value of the shape parameter of 3-parameter Weibull distribution and in this paper we use the coefficient of skewness to obtain the approximate value of the shape parameter of 3-parameter Weibull distribution, one can use the coefficient of kurtosis to obtain the approximate value of the shape parameter of 4-parameter Weibull distribution.  
 2. Outliers reduce the values of the relative error and we have to remove these observations when treatment the outliers.  
 3. The use of coefficient of variation or coefficient of skewness is easy in computation and takes less time on computer than maximum likelihood estimators.

The coefficient of skewness given by eq. (1) is a function of the shape parameter  $\alpha$  only, the value of skewness (S) of the sample when there are no outliers and when we have an outlier are 0.107 and 0.818, respectively. By equating the right hand side of eq. (1) to skewness of the sample when there are no outliers and when we have an outlier and solve each of those equations by iterative method or by trial and error, we obtain the approximate value of the shape  $\alpha$  with tolerance of 0.02, we can form the following Table 2.

Table 1  
 Parameter estimates of 3-parameter Weibull distribution

N	True value	MEM	MEMO	REM	REMO
10	0.10	0.100	0.100	0.10	0.10
20	0.10	0.100	0.100	0.10	0.10
30	0.10	0.100	0.100	0.10	0.10

MEM is the vector of parameter estimates  
 MEMO is the vector of parameter estimates without outliers  
 REM is the vector of parameter estimates with outliers  
 REMO is the vector of parameter estimates with outliers