



AD

AD 687287

# TECHNICAL REPORT

WVT-6910

## CONFIDENCED NORMAL AND LOGNORMAL RELIABILITY FOR ANY SAMPLE SIZE

BY

ROYCE W. SOANES, JR.

APRIL 1969

DDC  
RECEIVED  
MAY 22 1969  
RECEIVED  
B

# BENET R&E LABORATORIES

## WATERVLIET ARSENAL

WATERVLIET-NEW YORK

AMCMS No. 4440.25.2228.1.13

DA Project No. 68724

Reproduced by the  
CLEARINGHOUSE  
for Federal Scientific & Technical  
Information Springfield Va. 22151

This document has been approved  
for public release and under its  
distribution is unlimited

27

DISPOSITION

Destroy this report when it is no longer needed. Do not return it to the originator.

DISCLAIMER

The findings in this report are not to be construed as an official Department of the Army position.

CONFIDENCED NORMAL AND LOGNORMAL RELIABILITY  
FOR ANY SAMPLE SIZE

ABSTRACT

For a given mission life ( $x$ ), an exact 100% lower confidence limit on population reliability is found.

The conditions and/or assumptions underlying this analysis are: (1) The population is either normal or lognormal, (2) A sample of two or more representative failures is available. Neyman's general classical method of confidence intervals is used. This is possible because the distribution of the reliability estimator is a one parameter distribution which is independent of both failure density parameters and mission life.

Cross-Reference Data

Reliability  
Confidence Interval  
Neyman  
Normal  
Lognormal  
Estimator  
Parameters  
Exact Sampling Distribution  
Independence  
Jacobian

TABLE OF CONTENTS

<u>SECTION</u>		<u>PAGE</u>
	Abstract	1
1	Introduction	3
2	Notation	4
3	Results	5
4	Derivation of Results	7
	References	23

1. INTRODUCTION

Important aspects of this analysis are:

- a. Two-parameter distributions are dealt with.
- b. Both parameters are estimated from the sample (i.e., no parameter is "assumed to be known")!
- c. A representative sample of size 2 or greater is acceptable data.

The results of the analysis will be presented first for the sake of brevity.

The derivations will be carried out for the normal population only, since the lognormal case is only trivially different.

The general procedure in the derivation of equation (1) will be:

1. Find the joint density of  $\hat{\mu}$  and  $\hat{\sigma}$ .
2. Obtain the density and distribution function of  $\hat{R}$  by making a change of variable.
3. Use Neyman's method to obtain a one-sided confidence interval for  $R$ .
4. Simplify the resulting expressions.

## 2. NOTATION

$n$  = sample size

$\mu, \sigma$  = normal population mean and standard deviation

$\hat{\mu}, \hat{\sigma}$  = estimators of  $\mu$  and  $\sigma$

$x$  = mission life

$R$  = Population reliability

$\hat{R}$  = estimator of  $R$

$\Phi(x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^x e^{-\frac{1}{2}t^2} dt$  = standard normal distribution function

$\Phi^{-1}$  = standard normal fractile function (inverse of distribution function)

$\mu^*, \sigma^*$  = estimates of  $\mu$  and  $\sigma$

$R^*$  = estimate of  $R$

$C$  = confidence level

$\mu_L, \sigma_L$  = parameters of the lognormal distribution

$\mu_L^*, \sigma_L^*$  = estimates of  $\mu_L, \sigma_L$  (mean and standard deviation of the logs of the data)

$R_c$  = confidence reliability

### 3. RESULTS

#### a. Normal population:

The equation which yields confidence reliability  $R_c$  is:

$$\frac{(1-c)\Gamma\left(\frac{n-1}{2}\right)}{2\left(\frac{n}{2}\right)^{\frac{n-1}{2}}} = \int_0^{\infty} x^{n-2} e^{-\frac{nx^2}{2}} \Phi\left[\sqrt{n}\left(x\left(\frac{x-\mu^*}{\sigma^*}\right) - Z_{1-R_c}\right)\right] dx \quad (1)$$

Equation (1) is solved for  $Z_{1-R_c}$  and then —

$$R_c = 1 - \Phi(Z_{1-R_c})$$

The inverse problem may just as easily be solved, i.e., if one stipulates desired confidence reliability, one can solve equation (1) for the necessary mission life  $x$ .

#### b. Lognormal population:

The equation which yields confidence reliability  $R_c$  is:

$$\frac{(1-c)\Gamma\left(\frac{n-1}{2}\right)}{2\left(\frac{n}{2}\right)^{\frac{n-1}{2}}} = \int_0^{\infty} x^{n-2} e^{-\frac{nx^2}{2}} \Phi\left[\sqrt{n}\left(x\left(\frac{\ln x - \mu_e^*}{\sigma_e^*}\right) - Z_{1-R_c}\right)\right] dx \quad (2)$$

The foregoing remarks apply equally to equation (2).  
In order to implement the numerical solution of  
these equations, the following programs are needed:

1. A program to evaluate the standard normal distribution function.
2. An integrating program (such as Romberg integration).
3. A univariate nonlinear equation solver (a reliable method being the bisection or midpoint method).

Needless to say, these equations are of very little value if one does not have a digital computer available.

4. DERIVATION OF RESULTS

Let  $x_i$  ( $i=1, 2, \dots, n$ ) be normally distributed with mean  $\mu$  and standard derivation  $\sigma$ .

If  $x$  is the desired mission life, population reliability is given by:

$$R = 1 - \Phi\left(\frac{x - \mu}{\sigma}\right) \quad (3)$$

and the reliability estimator is given by:

$$\hat{R} = 1 - \Phi\left(\frac{x - \hat{\mu}}{\hat{\sigma}}\right) \quad (4)$$

Equation (4) gives  $\hat{R}$  as a function of  $\hat{\mu}$  and  $\hat{\sigma}$ ; therefore the distribution of  $\hat{R}$  may be obtained if one first finds the joint distribution of  $\hat{\mu}$  and  $\hat{\sigma}$ .

Form a bivariate change of variable using the dummy variable  $u$ .

$$\hat{R} = 1 - \Phi\left(\frac{x - \hat{\mu}}{\hat{\sigma}}\right) \quad (4)$$

$$u = \hat{\sigma} \quad (5)$$

Finding the inverse of this relation:

$$\Phi\left(\frac{x - \hat{\mu}}{\hat{\sigma}}\right) = 1 - \hat{R}$$

$$\frac{x - \hat{\mu}}{\hat{\sigma}} = Z_{1-\hat{R}}$$

$$\hat{\mu} = x - \hat{\sigma} Z_{1-\hat{R}} \quad (6)$$

the inverse relation is:

$$\hat{\mu} = x - u Z_{1-\hat{R}} \quad (7)$$

$$\hat{\sigma} = u \quad (8)$$

now,

$$h(\hat{R}, u) = f(\hat{\mu}, \hat{\sigma}) \left| \frac{\partial(\hat{\mu}, \hat{\sigma})}{\partial(\hat{R}, u)} \right| \quad (9)$$

Where  $h$  and  $f$  are joint density functions.

The Jacobian is obtained first, using equations (7) and (8):

$$J = \frac{\partial(\hat{\mu}, \hat{\sigma})}{\partial(\hat{R}, u)} = \begin{vmatrix} \frac{\partial \hat{\mu}}{\partial \hat{R}} & \frac{\partial \hat{\mu}}{\partial u} \\ \frac{\partial \hat{\sigma}}{\partial \hat{R}} & \frac{\partial \hat{\sigma}}{\partial u} \end{vmatrix}$$

$$= \begin{vmatrix} \frac{\partial \hat{\mu}}{\partial \hat{R}} & \frac{\partial \hat{\mu}}{\partial u} \\ 0 & 1 \end{vmatrix} = \frac{\partial \hat{\mu}}{\partial \hat{R}}$$

$$= -u \frac{d}{dR} Z_{1-R} \quad (10)$$

but  $\Phi(Z_\alpha) = \alpha$  (11)

$$\therefore \frac{d\alpha}{d\sigma} = \Phi'(Z_\alpha) \frac{dZ_\alpha}{d\sigma}$$

$$\therefore \frac{dZ_\alpha}{d\sigma} = \frac{1}{\Phi'(Z_\alpha)} \frac{d\alpha}{d\sigma}$$

but

$$\Phi(x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^x e^{-\frac{1}{2}t^2} dt$$

$$\therefore \Phi'(x) = \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}x^2}$$

$$\begin{aligned} \frac{dZ_\alpha}{d\alpha} &= \frac{1}{\frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}Z_\alpha^2}} \frac{d\alpha}{d\alpha} \\ &= \sqrt{2\pi} e^{\frac{1}{2}Z_\alpha^2} \frac{d\alpha}{d\alpha} \end{aligned} \quad (12)$$

Using equations (10) and (12)

$$\begin{aligned} J &= -u \cdot \sqrt{2\pi} e^{\frac{1}{2}Z_{1-\hat{R}}^2} \cdot (-1) \\ &= \sqrt{2\pi} u e^{\frac{1}{2}Z_{1-\hat{R}}^2} \end{aligned} \quad (13)$$

Therefore from equations (9) and (13):

$$h(\hat{R}, u) = f(\hat{\mu}, \hat{\sigma}) \sqrt{2\pi} u e^{\frac{1}{2}Z_{1-\hat{R}}^2}$$

or

$$h(\hat{R}, \hat{\sigma}) = f(\hat{\mu}, \hat{\sigma}) \sqrt{2\pi} \hat{\sigma} e^{-\frac{1}{2} \frac{\hat{\sigma}^2}{1-\hat{R}}} \quad (14)$$

After  $f(\hat{\mu}, \hat{\sigma})$  is determined,  $\hat{\sigma}$  may be integrated out (in equation (14)) to obtain the density of  $\hat{R}$ .

Now,  $X_i$  is normal with mean  $\mu$  and standard deviation  $\sigma$ .

$\therefore y = \frac{n\hat{\sigma}^2}{\sigma^2}$  is chi-square distributed with  $n-1$  degrees of freedom:

$$\therefore g(y) = \frac{1}{\Gamma\left(\frac{n-1}{2}\right) 2^{\frac{n-1}{2}}} y^{\frac{n-1}{2}-1} e^{-\frac{y}{2}}$$

The density of  $\hat{\sigma}$  is therefore given by:

$$g(\hat{\sigma}) = g(y) \left| \frac{dy}{d\hat{\sigma}} \right|$$

but  $y = \frac{n\hat{\sigma}^2}{\sigma^2}$

$$\therefore \frac{dy}{d\hat{\sigma}} = \frac{2n\hat{\sigma}}{\sigma^2}$$

$$\therefore g(\hat{\sigma}^2) = \frac{1}{\Gamma\left(\frac{n-1}{2}\right) 2^{\frac{n-1}{2}} \left(\frac{n\hat{\sigma}^2}{\sigma^2}\right)^{\frac{n-3}{2}} e^{-\frac{1}{2} \frac{n\hat{\sigma}^2}{\sigma^2}} \cdot \frac{2n\hat{\sigma}^1}{\sigma^2}$$

\(\therefore\) The density of  $\hat{\sigma}^1$  is given by:

$$g(\hat{\sigma}^1) = \frac{2 \left(\frac{n}{2}\right)^{\frac{n-1}{2}}}{\sigma \Gamma\left(\frac{n-1}{2}\right)} \left(\frac{\hat{\sigma}^1}{\sigma}\right)^{n-2} e^{-\frac{n}{2} \left(\frac{\hat{\sigma}^1}{\sigma}\right)^2} \quad (15)$$

Now,  $\hat{\mu}^1$  is normally distributed with mean  $\mu$  and standard deviation  $\frac{\sigma}{\sqrt{n}}$ ; therefore, the density of  $\hat{\mu}^1$  is given by:

$$\begin{aligned} f(\hat{\mu}^1) &= \frac{1}{\sqrt{2\pi} \frac{\sigma}{\sqrt{n}}} e^{-\frac{1}{2} \left(\frac{\hat{\mu}^1 - \mu}{\frac{\sigma}{\sqrt{n}}}\right)^2} \\ &= \frac{1}{\sigma} \sqrt{\frac{n}{2\pi}} e^{-\frac{n}{2} \left(\frac{\hat{\mu}^1 - \mu}{\sigma}\right)^2} \quad (16) \end{aligned}$$

Now, since  $\hat{\mu}^1$  and  $\hat{\sigma}^1$  are independent random variables, equations (15) and (16) yield the joint density of

$\hat{\mu}$  and  $\hat{\sigma}$  :

$$\begin{aligned}
 f(\hat{\mu}, \hat{\sigma}) &= g(\hat{\sigma}) \times r(\hat{\mu}) \\
 &= \frac{2}{\sigma} \frac{\left(\frac{m}{2}\right)^{\frac{m-1}{2}}}{\Gamma\left(\frac{m-1}{2}\right)} \left(\frac{\hat{\sigma}}{\sigma}\right)^{m-2} e^{-\frac{m}{2} \left(\frac{\hat{\sigma}}{\sigma}\right)^2} \\
 &\quad \frac{1}{\sigma} \sqrt{\frac{m}{2\pi}} e^{-\frac{m}{2} \left(\frac{\hat{\mu} - \mu}{\sigma}\right)^2}
 \end{aligned}$$

$$\therefore f(\hat{\mu}, \hat{\sigma}) = \frac{2}{\sigma^2} \sqrt{\frac{m}{2\pi}} \frac{\left(\frac{m}{2}\right)^{\frac{m-1}{2}}}{\Gamma\left(\frac{m-1}{2}\right)} \left(\frac{\hat{\sigma}}{\sigma}\right)^{m-2} e^{-\frac{m}{2} \left[ \left(\frac{\hat{\sigma}}{\sigma}\right)^2 + \left(\frac{\hat{\mu} - \mu}{\sigma}\right)^2 \right]} \quad (17)$$

Substituting this expression into equation (14):

$$\begin{aligned}
 h(\hat{R}, \hat{\sigma}) &= \frac{2}{\sigma^2} \sqrt{\frac{m}{2\pi}} \frac{\left(\frac{m}{2}\right)^{\frac{m-1}{2}}}{\Gamma\left(\frac{m-1}{2}\right)} \left(\frac{\hat{\sigma}}{\sigma}\right)^{m-2} e^{-\frac{m}{2} \left[ \left(\frac{\hat{\sigma}}{\sigma}\right)^2 + \left(\frac{\hat{\mu} - \mu}{\sigma}\right)^2 \right]} \\
 &\quad \sqrt{2\pi} \hat{\sigma} e^{\frac{1}{2} \hat{\sigma}^2} \hat{R}
 \end{aligned}$$

$$h(\hat{R}, \hat{\sigma}) = \frac{2\sqrt{m} \left(\frac{m}{2}\right)^{\frac{m-1}{2}}}{\sigma \Gamma\left(\frac{m-1}{2}\right)} \left(\frac{\hat{\sigma}}{\sigma}\right)^{m-1} e^{-\frac{m}{2} \left[ \left(\frac{\hat{\sigma}}{\sigma}\right)^2 + \left(\frac{\hat{\mu} - \mu}{\sigma}\right)^2 \right]} + \frac{1}{2} \hat{Z}_{1-\hat{R}}^2 \quad (18)$$

but  $\hat{\mu} = \kappa - \hat{\sigma} \hat{Z}_{1-\hat{R}}$  Equation (6)

and  $\mu = \kappa - \sigma Z_{1-R}$

$$\begin{aligned} \therefore \frac{\hat{\mu} - \mu}{\sigma} &= \frac{\sigma Z_{1-R} - \hat{\sigma} \hat{Z}_{1-\hat{R}}}{\sigma} \\ &= Z_{1-R} - \frac{\hat{\sigma}}{\sigma} \hat{Z}_{1-\hat{R}} \end{aligned} \quad (19)$$

$$\text{let } K = \frac{2\sqrt{m} \left(\frac{m}{2}\right)^{\frac{m-1}{2}}}{\Gamma\left(\frac{m-1}{2}\right)} \quad (20)$$

Therefore, using equations (18), (19), and (20):

$$h(\hat{R}, \hat{\sigma}) = \frac{K}{\sigma} \left(\frac{\hat{\sigma}}{\sigma}\right)^{m-1} e^{-\frac{m}{2} \left[ \left(\frac{\hat{\sigma}}{\sigma}\right)^2 + \left( Z_{1-R} - \frac{\hat{\sigma}}{\sigma} \hat{Z}_{1-\hat{R}} \right)^2 \right]} + \frac{1}{2} \hat{Z}_{1-\hat{R}}^2 \quad (21)$$

Now,  $\hat{\sigma}$  is integrated out to obtain the density of  $\hat{R}$ :

$$h(\hat{R}) = \int_0^{\infty} \frac{\kappa}{\sigma} \left(\frac{\sigma^1}{\sigma}\right)^{m-1} e^{-\frac{m}{2} \left[ \left(\frac{\sigma^1}{\sigma}\right)^2 + \left(\beta_{1-R} - \frac{\sigma^1}{\sigma} \beta_{1-\hat{R}}\right)^2 \right] + \frac{1}{2} \beta_{1-\hat{R}}^2} d\sigma^1 \quad (22)$$

$$\text{let } u = \frac{\sigma^1}{\sigma}$$

$$\therefore d\sigma^1 = \sigma du$$

$\therefore$  The density of  $\hat{R}$  is given by:

$$h(\hat{R}) = \kappa \int_0^{\infty} u^{m-1} e^{-\frac{m}{2} \left[ u^2 + \left(\beta_{1-R} - u \beta_{1-\hat{R}}\right)^2 \right] + \frac{1}{2} \beta_{1-\hat{R}}^2} du \quad (23)$$

Since  $u$  is a dummy variable of integration and  $\hat{R}$  is the argument of  $h$ , the only numbers upon which the form of  $h$  is dependent are  $m$  and  $R$ . Therefore, the density of the reliability estimator is a one parameter ( $R$ ) density which is independent of the failure density population parameters and mission life.

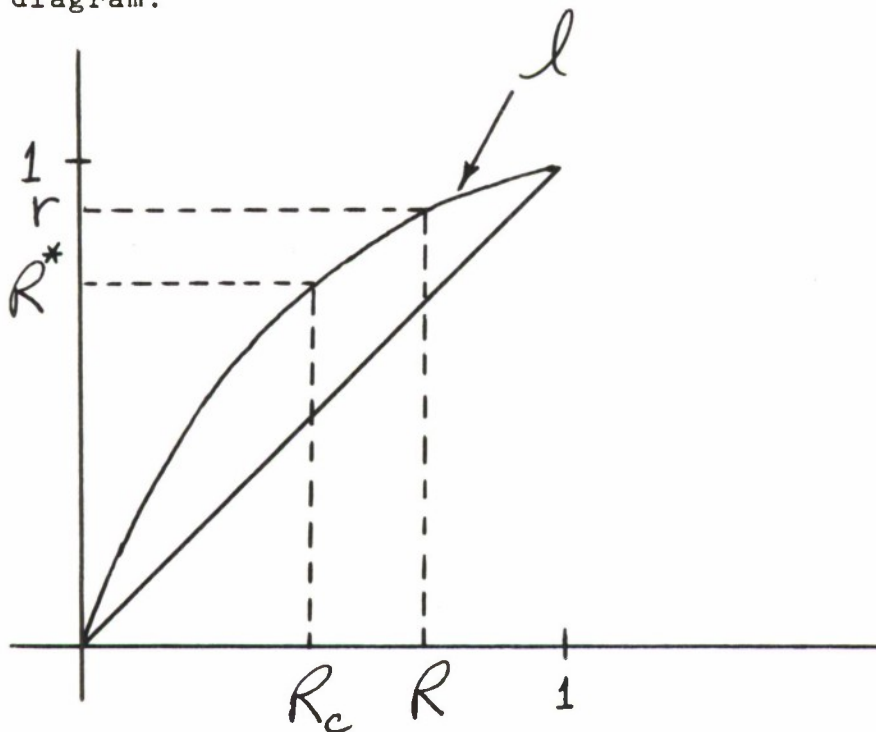
Changing the argument of  $h$  to avoid confusion and adding the subscript  $R$  to  $h$  to indicate its dependence on the population reliability  $R$ , the density of  $\hat{R}$  is:

$$h_R(v) = K \int_0^{\infty} s^{m-1} e^{-\frac{n}{2} [s^2 + (s\bar{z}_{i-v} - \bar{z}_{i-R})^2]} + \frac{1}{2} \bar{z}_{i-v}^2 ds \quad (24)$$

The distribution function of  $\hat{R}$  is therefore given by:

$$H_R(r) = K \int_0^r \int_0^{\infty} s^{m-1} e^{-\frac{n}{2} [s^2 + (s\bar{z}_{i-v} - \bar{z}_{i-R})^2]} + \frac{1}{2} \bar{z}_{i-v}^2 ds dv \quad (25)$$

The Neyman method of finding a one-sided confidence interval for  $R$  may be explained through the following diagram:



The curve  $l$  is determined by:

$$P(\hat{R} < r | R) = C$$

or

$$H_R(r) = C \quad (26)$$

The information in the failure data is introduced at this point:

$$R^* = 1 - \Phi\left(\frac{x - \mu^*}{\sigma^*}\right) \quad (27)$$

$R_c$  is then determined by solving equation (28)

$$H_{R_c}(R^*) = C \quad (28)$$

In order to solve equation (28), the distribution function of  $\hat{R}$  given by equation (25) should be simplified.

$$H_R(r) = K \int_0^{\infty} \int_0^{\infty} s^{m-1} e^{-\frac{m}{2} \left[ s^2 + (s \beta_{1-v} - \beta_{1-R})^2 \right] + \frac{1}{2} \beta_{1-v}^2} ds dv \quad (25)$$

Changing the order of integration:

$$\begin{aligned}
 H_R(r) &= K \int_0^{\infty} \int_0^r \alpha^{m-1} e^{-\frac{m}{2} \left[ \alpha^2 + (\alpha Z_{1-v} - Z_{1-R})^2 \right] + \frac{1}{2} Z_{1-v}^2} d\alpha dZ_{1-v} \\
 &= K \int_0^{\infty} \alpha^{m-1} e^{-\frac{m}{2} \alpha^2} \int_0^r e^{-\frac{m}{2} (\alpha Z_{1-v} - Z_{1-R})^2 + \frac{1}{2} Z_{1-v}^2} dZ_{1-v} d\alpha \quad (29)
 \end{aligned}$$

But from equation (12)

$$\begin{aligned}
 dZ_{\alpha} &= \sqrt{2\pi} e^{-\frac{1}{2} Z_{\alpha}^2} d\alpha \\
 \therefore dZ_{1-v} &= -\sqrt{2\pi} e^{-\frac{1}{2} Z_{1-v}^2} dZ_{1-v} \quad (30)
 \end{aligned}$$

Substituting (30) into (29):

$$H_R(r) = \frac{-K}{\sqrt{2\pi}} \int_0^{\infty} \alpha^{m-1} e^{-\frac{m}{2} \alpha^2} \int_0^r e^{-\frac{1}{2} (\sqrt{m} \alpha Z_{1-v} - \sqrt{m} Z_{1-R})^2} dZ_{1-v} d\alpha \quad (31)$$

Now let

$$u = \sqrt{m} \alpha Z_{1-v} \quad (32)$$

$$\therefore dZ_{1-v} = \frac{du}{\sqrt{m} \alpha} \quad (33)$$

Substituting (32) and (33) into (31)

$$H_R(r) = \frac{-K}{\sqrt{2\pi m}} \int_0^{\infty} s^{n-2} e^{-\frac{ms^2}{2}} \int_{-\infty}^{\sqrt{m}(\beta_{i-r} - r)} e^{-\frac{1}{2}(u - \sqrt{m}(\beta_{i-r} - r))^2} du ds$$

$$H_R(r) = \frac{K}{\sqrt{m}} \int_0^{\infty} s^{n-2} e^{-\frac{ms^2}{2}} \cdot \frac{1}{\sqrt{2\pi}} \int_{\sqrt{m}(\beta_{i-r} - r)}^{\infty} e^{-\frac{1}{2}(u - \sqrt{m}(\beta_{i-r} - r))^2} du ds$$

$$H_R(r) = \frac{K}{\sqrt{m}} \int_0^{\infty} s^{n-2} e^{-\frac{ms^2}{2}} \cdot \frac{1}{\sqrt{2\pi}} \int_{\sqrt{m}(\beta_{i-r} - \beta_{i-R})}^{\infty} e^{-\frac{1}{2}u^2} du ds$$

$$H_R(r) = \frac{K}{\sqrt{m}} \int_0^{\infty} s^{n-2} e^{-\frac{ms^2}{2}} \left\{ 1 - \Phi \left[ \frac{u}{\sqrt{m}(\beta_{i-r} - \beta_{i-R})} \right] \right\} ds \quad (34)$$

Separating (34) into two integrals:

$$H_R(r) = \frac{K}{\sqrt{m}} \int_0^{\infty} s^{m-2} e^{-\frac{ms^2}{2}} ds - \frac{K}{\sqrt{m}} \int_0^{\infty} s^{m-2} e^{-\frac{ms^2}{2}} \Phi \left[ \sqrt{m} \left( s \frac{z}{r} - \frac{z}{r-R} \right) \right] ds \quad (35)$$

Let 
$$I = \int_0^{\infty} s^{m-2} e^{-\frac{ms^2}{2}} ds \quad (36)$$

To evaluate I, make the change of variable:  $u = \frac{ms^2}{2}$

$$\therefore du = ms ds, \quad s^2 = \frac{2u}{m}, \quad s = \left( \frac{2u}{m} \right)^{1/2}$$

$$s^{m-2} = \left( \frac{2u}{m} \right)^{\frac{m-2}{2}}$$

$$\therefore I = \int_0^{\infty} \left( \frac{2u}{m} \right)^{\frac{m-2}{2}} e^{-u} \frac{du}{m \left( \frac{2u}{m} \right)^{1/2}}$$

$$= \int_0^{\infty} \frac{1}{m} \left( \frac{2u}{m} \right)^{\frac{m-3}{2}} e^{-u} du$$

$$= \frac{1}{m} \left( \frac{2}{m} \right)^{\frac{m-3}{2}} \int_0^{\infty} u^{\frac{m-1}{2}-1} e^{-u} du$$

$$= \frac{1}{m} \left( \frac{2}{m} \right)^{\frac{m-3}{2}} \Gamma \left( \frac{m-1}{2} \right) \quad (37)$$

but from (20),  $K = \frac{z \Gamma\left(\frac{n}{2}\right)^{\frac{n-1}{2}}}{\Gamma\left(\frac{n-1}{2}\right)}$

$$\therefore \frac{KI}{\Gamma n} = \frac{z \left(\frac{n}{2}\right)^{\frac{n-1}{2}}}{\Gamma\left(\frac{n-1}{2}\right)} \cdot \frac{1}{n} \left(\frac{z}{n}\right)^{\frac{n-3}{2}} \Gamma\left(\frac{n-1}{2}\right) = 1 \quad (38)$$

using (38) in (35)

$$H_R(r) = 1 - \frac{z \left(\frac{n}{2}\right)^{\frac{n-1}{2}}}{\Gamma\left(\frac{n-1}{2}\right)} \int_0^{\infty} s^{n-2} e^{-\frac{ms^2}{2}} \Phi \left[ \Gamma n \left( s \beta_{1-R} - \beta_{1-R} \right) \right] ds \quad (39)$$

Using equation (39), equation (28) now becomes:

$$C = 1 - \frac{z \left(\frac{n}{2}\right)^{\frac{n-1}{2}}}{\Gamma\left(\frac{n-1}{2}\right)} \int_0^{\infty} s^{n-2} e^{-\frac{ms^2}{2}} \Phi \left[ \Gamma n \left( s \beta_{1-R}^* - \beta_{1-R_c} \right) \right] ds$$

or

$$\frac{(1-C) \Gamma\left(\frac{n-1}{2}\right)}{z \left(\frac{n}{2}\right)^{\frac{n-1}{2}}} = \int_0^{\infty} s^{n-2} e^{-\frac{ms^2}{2}} \Phi \left[ \Gamma n \left( s \beta_{1-R}^* - \beta_{1-R_c} \right) \right] ds \quad (40)$$

But from equation (27),  $z_{1-R^*} = \frac{x - \mu^*}{\sigma^*}$

Therefore equation (1) is obtained:

$$\frac{(1-c)^{1/(m-1)}}{2 \left(\frac{n}{2}\right)^{\frac{m-1}{2}}} = \int_0^{\infty} s^{m-2} e^{-\frac{ms^2}{2}} \Phi \left[ \Gamma_m \left( s \left( \frac{x - \mu^*}{\sigma^*} \right) - z_{1-R_c} \right) \right] ds$$

Equation (2) is obtained analogously; one simply thinks of working entirely within the domain of the logarithms of the data points.

REFERENCES:

1. Cramer, H., Mathematical Methods of Statistics, Princeton, pp. 509-513, 378-390
2. Mood, A., and Graybill, F., Introduction to the Theory of Statistics, McGraw-Hill, pp. 256-260, 228-230

Unclassified

Security Classification

**DOCUMENT CONTROL DATA - R & D**

*(Security classification of title, body of abstract and indexing annotation must be entered when the overall report is classified)*

1. ORIGINATING ACTIVITY (Corporate author) Watervliet Arsenal Watervliet, N.Y. 12189		2a. REPORT SECURITY CLASSIFICATION Unclassified	
		2b. GROUP	
3. REPORT TITLE CONFIDENCED NORMAL AND LOGNORMAL RELIABILITY FOR ANY SAMPLE SIZE			
4. DESCRIPTIVE NOTES (Type of report and inclusive dates) Technical Report			
5. AUTHOR(S) (First name, middle initial, last name) Royce W. Soanes, Jr.			
6. REPORT DATE April 1969		7a. TOTAL NO. OF PAGES 25	7b. NO. OF REFS 2
8a. CONTRACT OR GRANT NO. AMCMS No. 4440.25.2226.1.13		8a. ORIGINATOR'S REPORT NUMBER(S) WVT-6910	
b. PROJECT NO. DA Project No. 66724		8b. OTHER REPORT NO(S) (Any other numbers that may be assigned this report)	
c.			
d.			
10. DISTRIBUTION STATEMENT This document has been approved for public release and sale; its distribution is unlimited.			
11. SUPPLEMENTARY NOTES		12. SPONSORING MILITARY ACTIVITY U. S. Army Weapons Command	
13. ABSTRACT For a given mission life (x), an exact 100% lower confidence limit on population reliability is found. The conditions and/or assumptions underlying this analysis are: (1) The population is either normal or lognormal, (2) A sample of two or more representative failures is available. Neyman's general classical method of confidence intervals is used. This is possible because the distribution of the reliability estimator is a one parameter distribution which is independent of both failure density parameters and mission life.			

14. KEY WORDS	LINK A		LINK B		LINK C	
	ROLE	WT	ROLE	WT	ROLE	WT
Reliability						
Confidence Interval						
Neyman						
Normal						
Lognormal						
Estimator						
Parameters						
Exact Sampling Distribution						
Independence						
Jacobian						

<p>AD <u>Accession No.</u> Benet Laboratories, Watervliet Arsenal, Watervliet, N.Y.</p> <p><b>CONFIDENCED NORMAL AND LOGNORMAL RELIABILITY FOR ANY SAMPLE SIZE</b> by Royce W. Soanes, Jr.</p> <p>Report No. WVT-6910, April 1969, 25 pages. AMOMS No. 4440.25.2226.1.13, DA Project No. 66724. Unclassified Report.</p> <p>For a given mission life (x), an exact 100% lower confidence limit on population reliability is found. The conditions and/or assumptions underlying this analysis are: (1) The population is either normal or lognormal, (2) A sample of two or more representative failures is available. Neyman's general classical method of confidence intervals is used. This is possible because the distribution of the reliability estimator is a one parameter distribution which is independent of both failure density parameters and mission life.</p>	<p>Reliability Confidence Interval</p> <p>Neyman Normal</p> <p>Lognormal Estimator</p> <p>Parameters</p> <p>Exact Sampling Distribution</p> <p>Independence</p> <p>Jacobian</p> <p>Distribution Unlimited</p>	<p>AD <u>Accession No.</u> Benet Laboratories, Watervliet Arsenal, Watervliet, N.Y.</p> <p><b>CONFIDENCED NORMAL AND LOGNORMAL RELIABILITY FOR ANY SAMPLE SIZE</b> by Royce W. Soanes, Jr.</p> <p>Report No. WVT-6910, April 1969, 25 pages. AMOMS No. 4440.25.2226.1.13, DA Project No. 66724. Unclassified Report.</p> <p>For a given mission life (x), an exact 100% lower confidence limit on population reliability is found. The conditions and/or assumptions underlying this analysis are: (1) The population is either normal or lognormal, (2) A sample of two or more representative failures is available. Neyman's general classical method of confidence intervals is used. This is possible because the distribution of the reliability estimator is a one parameter distribution which is independent of both failure density parameters and mission life.</p>	<p>Reliability Confidence Interval</p> <p>Neyman Normal</p> <p>Lognormal Estimator</p> <p>Parameters</p> <p>Exact Sampling Distribution</p> <p>Independence</p> <p>Jacobian</p> <p>Distribution Unlimited</p>
<p>AD <u>Accession No.</u> Benet Laboratories, Watervliet Arsenal, Watervliet, N.Y.</p> <p><b>CONFIDENCED NORMAL AND LOGNORMAL RELIABILITY FOR ANY SAMPLE SIZE</b> by Royce W. Soanes, Jr.</p> <p>Report No. WVT-6910, April 1969, 25 pages. AMOMS No. 4440.25.2226.1.13, DA Project No. 66724. Unclassified Report.</p> <p>For a given mission life (x), an exact 100% lower confidence limit on population reliability is found. The conditions and/or assumptions underlying this analysis are: (1) The population is either normal or lognormal, (2) A sample of two or more representative failures is available. Neyman's general classical method of confidence intervals is used. This is possible because the distribution of the reliability estimator is a one parameter distribution which is independent of both failure density parameters and mission life.</p>	<p>Reliability Confidence Interval</p> <p>Neyman Normal</p> <p>Lognormal Estimator</p> <p>Parameters</p> <p>Exact Sampling Distribution</p> <p>Independence</p> <p>Jacobian</p> <p>Distribution Unlimited</p>	<p>AD <u>Accession No.</u> Benet Laboratories, Watervliet Arsenal, Watervliet, N.Y.</p> <p><b>CONFIDENCED NORMAL AND LOGNORMAL RELIABILITY FOR ANY SAMPLE SIZE</b> by Royce W. Soanes, Jr.</p> <p>Report No. WVT-6910, April 1969, 25 pages. AMOMS No. 4440.25.2226.1.13, DA Project No. 66724. Unclassified Report.</p> <p>For a given mission life (x), an exact 100% lower confidence limit on population reliability is found. The conditions and/or assumptions underlying this analysis are: (1) The population is either normal or lognormal, (2) A sample of two or more representative failures is available. Neyman's general classical method of confidence intervals is used. This is possible because the distribution of the reliability estimator is a one parameter distribution which is independent of both failure density parameters and mission life.</p>	<p>Reliability Confidence Interval</p> <p>Neyman Normal</p> <p>Lognormal Estimator</p> <p>Parameters</p> <p>Exact Sampling Distribution</p> <p>Independence</p> <p>Jacobian</p> <p>Distribution Unlimited</p>

