

# Hypothesis Testing and Interval Estimation of a Single Lognormal Mean

Let  $y_1, \dots, y_n$  be a sample of observations from a lognormal distribution with parameters  $\mu$  and  $\sigma^2$ . Let  $x_i = \ln(y_i)$ ,  $i = 1, 2, \dots, n$ . The sample mean and the variance of the  $x_i$ 's are respectively given by

$$\bar{x} = \frac{1}{n} \sum_{i=1}^n x_i \quad \text{and} \quad s^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2.$$

Note that  $\exp(\mu + \sigma^2/2)$  is the mean of the lognormal distribution, and testing the mean is equivalent to testing  $\eta = \mu + \sigma^2/2$ . Consider

$$H_0 : \eta \geq \eta_0 \quad \text{vs.} \quad H_a : \eta < \eta_0. \quad (1)$$

## Algorithm 1

For a given logged data set, compute the observed sample mean and variance, namely,  $\bar{x}$  and  $s^2$ , respectively.

For  $i = 1$  to  $m$

Generate a standard normal variate  $Z$

Generate a chi-square random variate  $V^2$  with degrees of freedom  $n - 1$

Set  $T_{2i} = \bar{x} - \frac{Z}{V/\sqrt{n-1}} \frac{s}{\sqrt{n}} + \frac{1}{2} \frac{s^2}{V^2/(n-1)}$

Set  $K_i = 1$  if  $T_{2i} > \eta_0$ , else  $K_i = 0$

(end  $i$  loop)

$\frac{1}{m} \sum_{i=1}^m K_i$  is the generalized p-value for testing the hypotheses in (1).

If the computed generalized p-value is less than the nominal level  $\alpha$ , then the null hypothesis in (1) will be rejected.

The  $100(1-\alpha)$ th percentile of  $T_{21}, \dots, T_{2m}$ , denoted by  $T_{2,1-\alpha}$ , is the  $100(1-\alpha)\%$  generalized upper confidence limit for  $\eta = \mu + \sigma^2/2$ . Furthermore,  $\exp(T_{2,1-\alpha})$  is the  $100(1-\alpha)\%$  generalized upper limit for the lognormal mean. Appropriate quantiles of  $T_{2i}$ 's can be used to construct two-sided limits for  $\eta$  or  $\exp(\eta)$ .

In order to get consistent results regardless of the initial seed used for random number generation, the number of iteration in the above algorithm (i.e., the value of  $m$ ) should be at least 100,000.