

Lecture 32

The prediction interval formulas for the next observation from a normal distribution when σ is unknown

1 Introduction

In this lecture we will derive the formulas for the symmetric two-sided prediction interval for the $n + 1$ -st observation and the upper-tailed prediction interval for the $n + 1$ -st observation from a normal distribution *when the variance σ^2 is unknown*. We will need the following theorem from probability theory that gives the distribution of the statistic $\bar{X} - X_{n+1}$.

Suppose that $X_1, X_2, \dots, X_n, X_{n+1}$ is a random sample from a normal distribution with mean μ and variance σ^2 .

Theorem 1. *The random variable $T = (\bar{X} - X_{n+1}) / (\sqrt{\frac{n+1}{n}} S)$ has t distribution with $n - 1$ degrees of freedom.*

2 The two-sided prediction interval formula

Now we can prove the theorem from statistics giving the required prediction interval for the next observation x_{n+1} in terms of n observations x_1, x_2, \dots, x_n . Note that it is symmetric around \bar{X} . This is one of the basic theorems that you have to learn how to prove. There are also asymmetric two-sided prediction intervals.

Theorem 2. *The random interval $(\bar{X} - t_{\alpha/2, n-1} \sqrt{\frac{n+1}{n}} S, \bar{X} + t_{\alpha/2, n-1} \sqrt{\frac{n+1}{n}} S)$ is a $100(1 - \alpha)\%$ -prediction interval for x_{n+1} .*

Proof. We are required to prove

$$P(X_{n+1} \in (\bar{X} - t_{\alpha/2, n-1} \sqrt{\frac{n+1}{n}} S, \bar{X} + t_{\alpha/2, n-1} \sqrt{\frac{n+1}{n}} S)) = 1 - \alpha.$$

We have

$$\begin{aligned} LHS &= P(\bar{X} - t_{\alpha/2, n-1} \sqrt{\frac{n+1}{n}} S < X_{n+1}, X_{n+1} < \bar{X} + t_{\alpha/2, n-1} \sqrt{\frac{n+1}{n}} S) = P(\bar{X} - X_{n+1} < t_{\alpha/2, n-1} \sqrt{\frac{n+1}{n}} S) \\ &= P(\bar{X} - X_{n+1} < t_{\alpha/2, n-1} \sqrt{\frac{n+1}{n}} S, \bar{X} - X_{n+1} > -t_{\alpha/2, n-1} \sqrt{\frac{n+1}{n}} S) \\ &= P((\bar{X} - X_{n+1}) / \sqrt{\frac{n+1}{n}} S < t_{\alpha/2, n-1}, (\bar{X} - X_{n+1}) / \sqrt{\frac{n+1}{n}} S > -t_{\alpha/2, n-1}) \\ &= P(T < t_{\alpha/2, n-1}, T > -t_{\alpha/2, n-1}) = P(-t_{\alpha/2, n-1} < T < t_{\alpha/2, n-1}) = 1 - \alpha \end{aligned}$$

To prove the last equality draw a picture. □

Once we have an actual sample x_1, x_2, \dots, x_n we obtain the the observed value $(\bar{x} - t_{\alpha/2, n-1} \sqrt{\frac{n+1}{n}} s, \bar{x} + t_{\alpha/2, n-1} \sqrt{\frac{n+1}{n}} s)$ for the prediction (random) interval $(\bar{X} - t_{\alpha/2, n-1} \sqrt{\frac{n+1}{n}} S, \bar{X} + t_{\alpha/2, n-1} \sqrt{\frac{n+1}{n}} S)$ The observed value of the prediction (random) interval is also called the two-sided $100(1 - \alpha)\%$ prediction interval for x_{n+1} .

3 The upper-tailed prediction interval

In this section we will give the formula for the upper-tailed prediction interval for the next observation x_{n+1} .

Theorem 3. *The random interval $(\bar{X} - t_{\alpha, n-1} \sqrt{\frac{n+1}{n}} S, \infty)$ is a $100(1 - \alpha)\%$ -prediction interval for the next observation x_{n+1} .*

Proof. We are required to prove

$$P(X_{n+1} \in (\bar{X} - t_{\alpha, n-1} \sqrt{\frac{n+1}{n}} S, \infty)) = 1 - \alpha.$$

We have

$$\begin{aligned}LHS &= P(\bar{X} - t_{\alpha, n-1} \sqrt{\frac{n+1}{n}} S < X_{n+1}) \\&= P(\bar{X} - X_{n+1} < t_{\alpha, n-1} \sqrt{\frac{n+1}{n}} S) \\&= P((\bar{X} - X_{n+1}) / \sqrt{\frac{n+1}{n}} S < t_{\alpha, n-1}) \\&= P(T < t_{\alpha, n-1}) \\&= 1 - \alpha\end{aligned}$$

To prove the last equality draw a picture - I want *you* to draw the picture on tests and the final. □

Once we have an actual sample x_1, x_2, \dots, x_n we obtain the observed value $(\bar{x} - t_{\alpha, n-1} \sqrt{\frac{n+1}{n}} s, \infty)$ of the upper-tailed prediction (random) interval $(\bar{X} - t_{\alpha, n-1} \sqrt{\frac{n+1}{n}} S, \infty)$. The observed value of the upper-tailed prediction (random) interval is also called the upper-tailed $100(1 - \alpha)\%$ prediction interval for x_{n+1} .

The number random variable $\bar{X} - t_{\alpha, n-1} \sqrt{\frac{n+1}{n}} S$ or its observed value $\bar{x} - t_{\alpha, n-1} \sqrt{\frac{n+1}{n}} s$ is often called a prediction *lower bound* for x_{n+1} because

$$P(\bar{X} - t_{\alpha, n-1} \sqrt{\frac{n+1}{n}} S < X_{n+1}) = 1 - \alpha.$$