

Drill 5

In this exercise, we perform the two-sided hypothesis testing for $H_0 : \mu = \mu_0$ versus $H_1 : \mu \neq \mu_0$. We assume that X_1, X_2, \dots, X_n are from the normal with mean μ and variance σ^2 . This test is well known as z -test (when σ is known) or t -test (when σ is unknown) in the statistics literature. Note that the rejection region of the z -test and t -test are given by

$$Z = \frac{|\bar{X} - \mu_0|}{\sigma/\sqrt{n}} > z_{\alpha/2} \quad \text{and} \quad T = \frac{|\bar{X} - \mu_0|}{S/\sqrt{n}} > t_{\alpha/2}.$$

1. (a) When the variance known, obtain the theoretical power function of the z -test

When the variance σ^2 is known, the power function of the z -test is then given by

$$\begin{aligned} K_z(\mu) &= P\left[\frac{|\bar{X} - \mu_0|}{\sigma/\sqrt{n}} > z_{\alpha/2}\right] \\ &= 1 - P\left[\frac{|\bar{X} - \mu_0|}{\sigma/\sqrt{n}} \leq z_{\alpha/2}\right] \\ &= 1 - P\left[-z_{\alpha/2} \leq \frac{\bar{X} - \mu_0}{\sigma/\sqrt{n}} \leq z_{\alpha/2}\right]. \end{aligned} \tag{1}$$

Notice that X_i are from $N(\mu, \sigma^2)$, not from $N(\mu_0, \sigma^2)$. Thus, $(|\bar{X} - \mu|)/(\sigma/\sqrt{n})$ is distributed as the standard normal distribution, $N(0, 1)$, but $(|\bar{X} - \mu_0|)/(\sigma/\sqrt{n})$ is not.

We rewrite the power function in (1) by

$$\begin{aligned} K_z(\mu) &= 1 - P\left[-z_{\alpha/2} + \frac{\mu_0}{\sigma/\sqrt{n}} \leq \frac{\bar{X}}{\sigma/\sqrt{n}} \leq z_{\alpha/2} + \frac{\mu_0}{\sigma/\sqrt{n}}\right] \\ &= 1 - P\left[-z_{\alpha/2} + \frac{\mu_0 - \mu}{\sigma/\sqrt{n}} \leq \frac{\bar{X} - \mu}{\sigma/\sqrt{n}} \leq z_{\alpha/2} + \frac{\mu_0 - \mu}{\sigma/\sqrt{n}}\right] \\ &= 1 - P\left[-z_{\alpha/2} - \frac{\mu - \mu_0}{\sigma/\sqrt{n}} \leq \frac{\bar{X} - \mu}{\sigma/\sqrt{n}} \leq z_{\alpha/2} - \frac{\mu - \mu_0}{\sigma/\sqrt{n}}\right] \\ &= 1 - P\left[-z_{\alpha/2} - \frac{\mu - \mu_0}{\sigma/\sqrt{n}} \leq Z \leq z_{\alpha/2} - \frac{\mu - \mu_0}{\sigma/\sqrt{n}}\right], \end{aligned}$$

where $Z \sim N(0, 1)$. Thus, we have

$$K_z(\mu) = 1 - \left\{ \Phi\left(z_{\alpha/2} - \frac{\mu - \mu_0}{\sigma/\sqrt{n}}\right) - \Phi\left(-z_{\alpha/2} - \frac{\mu - \mu_0}{\sigma/\sqrt{n}}\right) \right\},$$

where $\Phi()$ is the cdf of the standard normal distribution. Using the relation $\Phi(-z) = 1 - \Phi(z)$, we can rewrite the above by

$$K_z(\mu) = 1 - \left\{ 1 - \Phi\left(-z_{\alpha/2} + \frac{\mu - \mu_0}{\sigma/\sqrt{n}}\right) - 1 + \Phi\left(z_{\alpha/2} + \frac{\mu - \mu_0}{\sigma/\sqrt{n}}\right) \right\}.$$

Therefore, the power function of the z -test is given by

$$\begin{aligned} K_z(\mu) &= 1 + \Phi\left(-z_{\alpha/2} + \frac{\mu - \mu_0}{\sigma/\sqrt{n}}\right) - \Phi\left(z_{\alpha/2} + \frac{\mu - \mu_0}{\sigma/\sqrt{n}}\right) \\ &= 1 - \Phi\left(z_{\alpha/2} + \frac{\mu - \mu_0}{\sigma/\sqrt{n}}\right) + \Phi\left(-z_{\alpha/2} + \frac{\mu - \mu_0}{\sigma/\sqrt{n}}\right). \end{aligned}$$

- (b) When the variance unknown, obtain the theoretical power function of the t -test

Theorem 1. Let X_1, X_2, \dots, X_n be a random sample from a normal distribution with μ and variance σ^2 . Let \bar{X} and S^2 denote the sample mean and variance, respectively. Then the following t -test statistic under the local alternative $H_1 : \mu = \mu_0 + \delta\sigma/\sqrt{n}$

$$\frac{\bar{X} - \mu_0}{S/\sqrt{n}}$$

has a non-central t -distribution with $n - 1$ degrees of freedom and non-centrality δ .

Proof. Recall the definition of a non-central t -distribution. Let $Z \sim N(0, 1)$ and V has a chi-square distribution with r degrees of freedom. Suppose that Z and V are independent. Then the quotient below has a non-central t -distribution with r degrees of freedom and non-centrality δ :

$$\frac{Z + \delta}{\sqrt{V/r}}.$$

Let $V = (n - 1)S^2/\sigma^2$ for convenience. Then V has a chi-square distribution with $n - 1$ degrees of freedom. See Theorem 5.3.1 of Casella and Berger (2002).

We have

$$\begin{aligned}\frac{\sqrt{n}(\bar{X} - \mu_0)}{S} &= \frac{\sqrt{n}(\bar{X} - \mu_0)/\sigma}{\sqrt{V/(n-1)}} \\ &= \frac{\sqrt{n}(\bar{X} - \mu)/\sigma + \sqrt{n}(\mu - \mu_0)/\sigma}{\sqrt{V/(n-1)}} \\ &= \frac{Z + \sqrt{n}(\mu - \mu_0)/\sigma}{\sqrt{V/(n-1)}},\end{aligned}$$

where $Z = (\bar{X} - \mu)/(\sigma/\sqrt{n})$ and $Z \sim N(0, 1)$. Thus, under the local $H_1 : \mu = \mu_0 + \delta\sigma/\sqrt{n}$, we have

$$\frac{\bar{X} - \mu_0}{S/\sqrt{n}} = \frac{Z + \delta}{\sqrt{V/(n-1)}}.$$

Since S^2 and \bar{X} are independent, V and Z are also independent. This completes the proof. \square

Next, we want to obtain the power function for testing $H_0 : \mu = \mu_0$ versus $H_1 : \mu \neq \mu_0$. Since $\mu = \mu_0 + \{\sqrt{n}(\mu - \mu_0)/\sigma\} \cdot \{\sigma/\sqrt{n}\}$, it is immediate upon using Theorem 1 that $(\bar{X} - \mu_0)/(S/\sqrt{n})$ under H_1 has the non-central t -distribution with $\nu = n - 1$ degrees of freedom and non-centrality $\delta = (\mu - \mu_0)/(\sigma/\sqrt{n})$.

The critical region for testing $H_0 : \mu = \mu_0$ versus $H_1 : \mu \neq \mu_0$ is given by

$$\frac{|\bar{X} - \mu_0|}{S/\sqrt{n}} > t_{\alpha/2}.$$

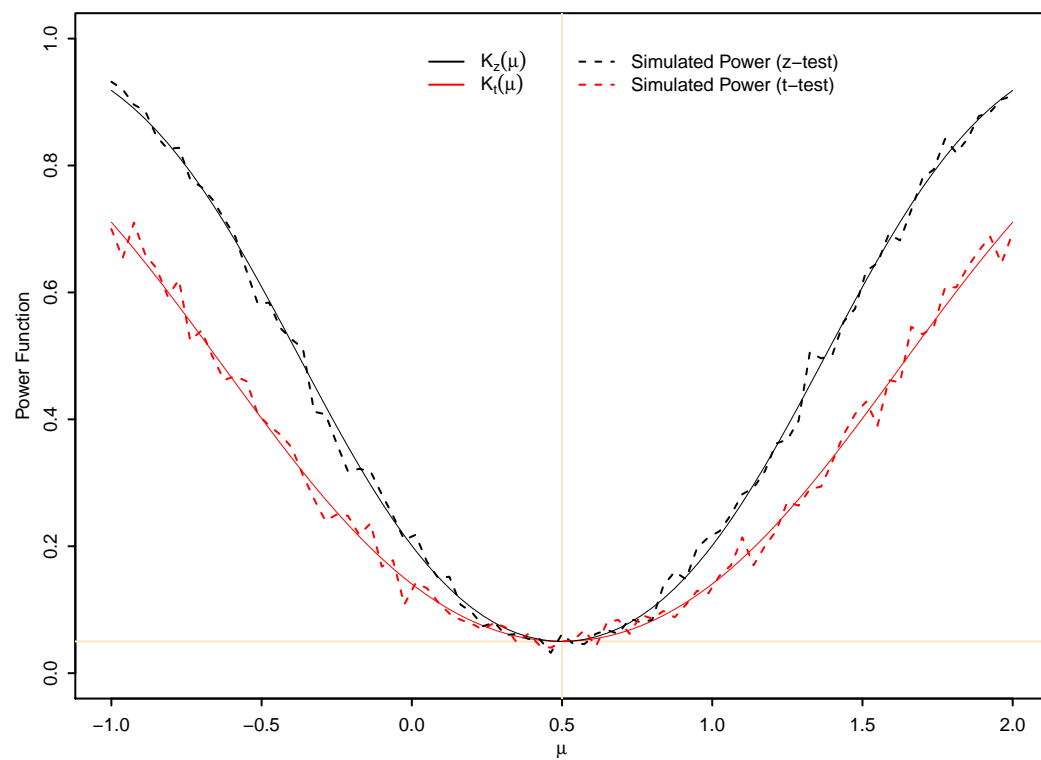
For convenience, we let $T_{n-1}(\delta) = (\bar{X} - \mu_0)/(S/\sqrt{n})$. Then the critical region can be rewritten as $|T_{n-1}(\delta)| > t_{\alpha/2}$. Then the power function is given by

$$\begin{aligned}K_t(\mu) &= P(|T_{n-1}(\delta)| > t_{\alpha/2}) \\ &= P(T_{n-1}(\delta) > t_{\alpha/2}) + P(T_{n-1}(\delta) < -t_{\alpha/2}) \\ &= 1 - \Phi_{\nu, \delta}(t_{\alpha/2}) + \Phi_{\nu, \delta}(-t_{\alpha/2}),\end{aligned}$$

where $\Phi_{\nu, \delta}(\cdot)$ is the cdf of the non-central t -distribution with $\nu = n - 1$ degrees of freedom and non-centrality $\delta = (\mu - \mu_0)/(\sigma/\sqrt{n})$.

2. Obtain the simulated power functions of the z -test and t -test for testing $H_0 : \mu = 1/2$ versus $H_1 : \mu \neq 1/2$ with the significance level $\alpha = 0.05$. Generate a sample of size $n = 5$ from the normal distribution with mean μ and $\sigma = 1$, where μ varies from -1 to 2 .

3. Compare the theoretical and simulated power functions of two tests. (The results should be similar to the following plot).



References

Casella, G. and Berger, R. L. (2002). *Statistical Inference*. Duxbury, Pacific Grove, CA, second edition.