## 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  $X_1, X_2, \ldots, X_n$  are from the normal with mean  $\mu$  and variance  $\sigma^2$ . This test is well known as z-test (when  $\sigma$  is known) or t-test (when  $\sigma$  is unknown) in the statistics literature. Note that the rejection region of the z-test and t-test are given by

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

1. (a) When the variance is known, obtain the theoretical power function of the z-test When the variance  $\sigma^2$  is known, the power function of the z-test is then given by

$$K_{z}(\mu) = P\left[\frac{|\bar{X} - \mu_{0}|}{\sigma/\sqrt{n}} > z_{\alpha/2}\right]$$

$$= 1 - P\left[\frac{|\bar{X} - \mu_{0}|}{\sigma/\sqrt{n}} \le z_{\alpha/2}\right]$$

$$= 1 - P\left[-z_{\alpha/2} \le \frac{\bar{X} - \mu_{0}}{\sigma/\sqrt{n}} \le z_{\alpha/2}\right]. \tag{1}$$

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

$$K_{z}(\mu) = 1 - P\left[-z_{\alpha/2} + \frac{\mu_0}{\sigma/\sqrt{n}} \le \frac{\bar{X}}{\sigma/\sqrt{n}} \le z_{\alpha/2} + \frac{\mu_0}{\sigma/\sqrt{n}}\right]$$

$$= 1 - P\left[-z_{\alpha/2} + \frac{\mu_0 - \mu}{\sigma/\sqrt{n}} \le \frac{\bar{X} - \mu}{\sigma/\sqrt{n}} \le z_{\alpha/2} + \frac{\mu_0 - \mu}{\sigma/\sqrt{n}}\right]$$

$$= 1 - P\left[-z_{\alpha/2} - \frac{\mu - \mu_0}{\sigma/\sqrt{n}} \le \frac{\bar{X} - \mu}{\sigma/\sqrt{n}} \le z_{\alpha/2} - \frac{\mu - \mu_0}{\sigma/\sqrt{n}}\right]$$

$$= 1 - P\left[-z_{\alpha/2} - \frac{\mu - \mu_0}{\sigma/\sqrt{n}} \le Z \le z_{\alpha/2} - \frac{\mu - \mu_0}{\sigma/\sqrt{n}}\right],$$

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

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

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

**Theorem 1.** Let  $X_1, X_2, ..., X_n$  be a random sample from a normal distribution with  $\mu$  and variance  $\sigma^2$ . 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  $\delta$ .

*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  $\delta$ :

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

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

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

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.

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

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

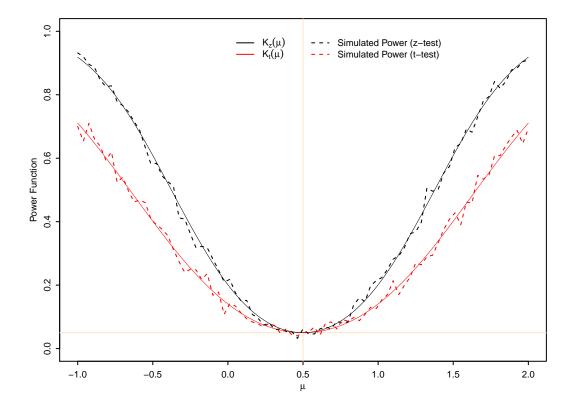
For convenience, we let  $T_{n-1}(\delta) = (\bar{X} - \mu_0)/(S/\sqrt{n})$ . Then the critical region can be rewritten as  $|T_{n-1}(\delta)| > t_{\alpha/2}$ . Then the power function is given by

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

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  $\mu$  and  $\sigma = 1$ , where  $\mu$  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.