STAT 423/523 Statistical Methods for Engineers and Scientists

Lecture 5: Hypothesis Testing II

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From Z-test to T-test

We discussed Z-test in the last lecture. The Z-test is based on the Z-statistic:

$$Z = \frac{\bar{X} - \mu_0}{\sigma / \sqrt{n}}$$

where σ is the population standard deviation assumed to be known.

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where σ is the population standard deviation assumed to be known.

In practice, we often do not know the population standard deviation.

T-Statistic

For a population with unknown standard deviation, we consider the **T-statistic**:

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

where S is the sample standard deviation:

$$S = \sqrt{\frac{1}{n-1} \sum_{i=1}^{n} (X_i - \bar{X})^2}.$$

T-Statistic

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The test based on the T-statistic is called the **T-test**, which is similar to the Z-test.

Two-sided T-Test

Let X_1, \ldots, X_n be a random sample from a normal population with unknown mean μ and unknown standard deviation. We want to test

$$H_0: \mu = \mu_0 \quad \text{v.s.} \quad H_a: \mu \neq \mu_0.$$

The test statistic is

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The decision rule is

reject null if
$$|T| > t_{\alpha/2,n-1}$$
,

where $t_{\alpha/2,n-1}$ is the $\alpha/2$ quantile of the t-distribution with n-1 degrees of freedom.

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The test statistic is

$$t = \frac{\bar{x} - 45}{s/\sqrt{n}} = 6.50$$

The quantile of the t-distribution is $t_{0.025,19} = 2.093$.

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The test statistic is

$$t = \frac{\bar{x} - 45}{s/\sqrt{n}} = 6.50$$

The quantile of the t-distribution is $t_{0.025,19}=2.093$. We reject the null hypothesis since |t|=6.50>2.093.

P-value of Two-sided T-Test

Similar to the Z-test, the p-value of the two-sided T-test is

$$p$$
-value = $2(1 - F_{t,n-1}(|T|))$,

where $F_{t,n-1}(\cdot)$ is the cdf of the t-distribution with n-1 degrees of freedom.

P-value of Two-sided T-Test

Similar to the Z-test, the p-value of the two-sided T-test is

$$ext{p-value} = 2(1 - F_{t,n-1}(|T|)),$$

where $F_{t,n-1}(\cdot)$ is the cdf of the t-distribution with n-1 degrees of freedom.

We reject the null hypothesis if the p-value is less than the significance level α .

Power of Two-sided T-Test

Consider $\mu_1 \neq \mu_0$ in the alternative hypothesis. The power $(1 - \beta(\mu_1))$ of the two-sided T-test is

$$\begin{split} \text{Power} &= P(\text{reject} \mid H_1) \\ &= P(|T| > t_{\alpha/2,n-1} | \mu = \mu_1) \\ &= F_{t,n-1} \left(-t_{\alpha/2,n-1} + \frac{\mu_1 - \mu_0}{s/\sqrt{n}} \right) + F_{t,n-1} \left(-t_{\alpha/2,n-1} - \frac{\mu_1 - \mu_0}{s/\sqrt{n}} \right) \end{split}$$

Sample Size Determination

In order to have a desired power $1 - \beta^*$, we need to determine the minimum sample size n such that $1 - \beta(\mu_1) \ge 1 - \beta^*$.

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Similar to the two-sided Z-test, the approximate formula for the sample size is

$$n^* = \left[\left(\frac{s(t_{\alpha/2, n-1} - t_{1-\beta^*, n-1})}{\mu_1 - \mu_0} \right)^2 \right]$$

Summary of Two-sided T-test

Hypothesis:

$$H_0:~\mu=\mu_0~$$
 v.s. $H_a:~\mu
eq\mu_0$

► Test statistic:

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

► Rejection region:

reject null if
$$|T|>t_{lpha/2,n-1}$$

► P-value:

p-value =
$$2(1 - F_{t,n-1}(|T|))$$

Power:

$$1 - \beta(\mu_1) = F_{t,n-1} \left(-t_{\alpha/2,n-1} - \frac{\mu_1 - \mu_0}{s/\sqrt{n}} \right) + F_{t,n-1} \left(-t_{\alpha/2,n-1} + \frac{\mu_1 - \mu_0}{s/\sqrt{n}} \right)$$

► Sample size determination: (approximation)

$$n^* = \left[\left(\frac{s(t_{\alpha/2, n-1} - t_{1-\beta^*, n-1})}{\mu_1 - \mu_0} \right)^2 \right]$$

One-sided T-Test (Summary)

► Hypothesis:

$$H_0: \ \mu = \mu_0 \quad \text{v.s.} \quad H_a: \ \mu > \mu_0$$

► Test statistic:

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

Rejection region:

reject null if
$$T > t_{\alpha,n-1}$$

P-value:

p-value =
$$1 - F_{t,n-1}(T)$$

Power:

$$1 - \beta(\mu_1) = 1 - F_{t,n-1} \left(t_{\alpha/2,n-1} - \frac{\mu_1 - \mu_0}{s/\sqrt{n}} \right)$$

Sample size determination:

$$n^* = \left[\left(\frac{s(t_{\alpha,n-1} - t_{1-\beta^*,n-1})}{\mu_1 - \mu_0} \right)^2 \right]$$

The failure stree for 19 carbon nanofibers are measured in MPa. The sample data is summarized as follows:

$$\bar{x} = 562.68, \quad s = 180.874$$

The researchers want to test whether the average failure stress is greater than 500 MPa with $\alpha=0.05$.

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$$T = \frac{\bar{x} - 500}{s/\sqrt{n}} = 1.51$$

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The p-value is $1 - F_{t,18}(1.51) = 0.074$.

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Test statistic:

$$T = \frac{\bar{x} - 500}{s / \sqrt{n}} = 1.51$$

The p-value is $1 - F_{t,18}(1.51) = 0.074$.

We failed to reject the null hypothesis since the p-value is greater than $\alpha=0.05$.