

Slides 11: t-tests

- t-distributions
- Exact confidence interval for μ
- Exact confidence interval for σ
- One sample t-test
- Two sample t-test

**The risk to a quality
test result comes from
very small samples;
not from a sample
that's too large.**

Exact confidence interval for the mean

In this special case, when a random sample (x_1, \dots, x_n) is taken from a normal distribution $N(\mu, \sigma)$,

$$\frac{\bar{X} - \mu}{S/\sqrt{n}} \sim t_{n-1}$$

has the so-called t-distribution with $n - 1$ degrees of freedom. This implies an exact $100(1 - \alpha)\%$ confidence interval

$$I_\mu = \bar{x} \pm t_{n-1}\left(\frac{\alpha}{2}\right) \cdot \frac{s}{\sqrt{n}}$$

For example,

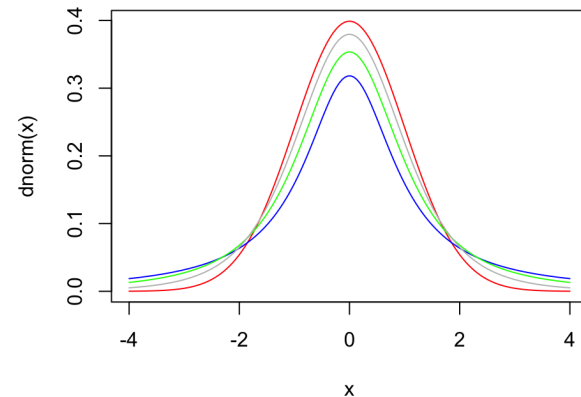
$$t_{10}(0.025) = 2.23,$$

$$t_{20}(0.025) = 2.09,$$

$$t_{30}(0.025) = 2.04$$

A t_k -distribution curve looks similar to $N(0,1)$ -curve being symmetric around zero.

If $k \geq 3$, then the variance is $\frac{k}{k-2}$.



t-distribution curves with $df = 1, 2, 5, \infty$

Exact confidence interval for σ

If Z, Z_1, \dots, Z_k are $N(0,1)$ and independent, then

$$\frac{Z}{\sqrt{(Z_1^2 + \dots + Z_k^2)/k}} \sim t_k.$$

Moreover, in the $N(\mu, \sigma)$ case we get access to an exact confidence interval formula for the variance thanks to the following result.

Exact distribution $\frac{(n-1)S^2}{\sigma^2} \sim \chi_{n-1}^2$

Exact $100(1 - \alpha)\%$ confidence interval

$$I_\sigma = \left(c\left(\frac{\alpha}{2}\right)s, c\left(1 - \frac{\alpha}{2}\right)s \right)$$

where $c^2(p) = \frac{n-1}{\chi_{n-1}^2(p)}$. Examples of 95% confidence intervals

$$I_\sigma = (0.69s, 1.82s) \text{ for } n = 10,$$

$$I_\sigma = (0.74s, 1.55s) \text{ for } n = 16,$$

$$I_\sigma = (0.78s, 1.39s) \text{ for } n = 25,$$

$$I_\sigma = (0.85s, 1.22s) \text{ for } n = 60.$$

For the normal model, $\text{Var}(S^2) = \frac{2\sigma^4}{n-1}$, standard error for s^2 is $\sqrt{\frac{2}{n-1}}s^2$.

One sample t-test

We wish to test $H_0: \mu = \mu_0$ against either the two-sided or a one-sided alternative. One-sample t-test is used for small n , under the assumption that the population distribution is normal. The t-test statistic

$$t = \frac{\bar{x} - \mu_0}{s/\sqrt{n}}, \quad T \stackrel{H_0}{\sim} t_{n-1}.$$

Example: Smoking and platelet aggregation

$n = 11$ paired observations (x_i, y_i) before and after smoking

(25, 27); (25, 29); (27, 37); (28, 43); (30, 46); (44, 56); (52, 61); (53, 57); (53, 80); (60, 59); (67, 82)

Assuming that the population distribution for the differences $d_i = x_i - y_i$ is normal, we test

$$H_0: \mu_1 - \mu_2 = 0 \text{ against } H_1: \mu_1 - \mu_2 \neq 0.$$

using the one-sample t-test. The observed test statistic value

$$t_{\text{obs}} = \frac{\bar{d}}{s_d/\sqrt{n}} = \frac{10.27}{2.40} = 4.28$$

gives two-sided p-value $2 * (1 - \text{pt}(4.28, 10)) = 0.0016$.

Two sample t-test

Two independent random samples (x_1, \dots, x_n) and (y_1, \dots, y_m) from two populations. The key assumption for the two-sample t-test:

Two normal population distributions $X \sim N(\mu_1, \sigma)$, $Y \sim N(\mu_2, \sigma)$.

In other words, there is a two level main factor plus noise $N(0, \sigma)$. The two levels of the main factor are quantified by μ_1 and μ_2 .

Define the pooled sample variance by

$$s_p^2 = \frac{\sum_{i=1}^n (x_i - \bar{x})^2 + \sum_{i=1}^m (y_i - \bar{y})^2}{n + m - 2}$$

Note that

$$s_p^2 = \frac{n-1}{n+m-2} \cdot s_1^2 + \frac{m-1}{n+m-2} \cdot s_2^2$$

The pooled sample variance is an unbiased estimate of the variance σ^2 :

$$E(S_p^2) = \frac{n-1}{n+m-2} E(S_1^2) + \frac{m-1}{n+m-2} E(S_2^2) = \sigma^2.$$

Question. There are $n + m$ terms in the numerator of s_p^2 but the denominator is $n + m - 2$. Why?

Two sample t-test

In the case of equal variances,

$$\text{Var}(\bar{X} - \bar{Y}) = \text{Var}(\bar{X}) + \text{Var}(\bar{Y}) = \frac{\sigma^2}{n} + \frac{\sigma^2}{m} = \sigma^2 \frac{n+m}{nm},$$

which yields the following expression for the standard error

$$s_{\bar{x}-\bar{y}} = s_p \sqrt{\frac{n+m}{nm}}.$$

Exact distribution $\frac{(\bar{X}-\bar{Y})-(\mu_1-\mu_2)}{s_p} \cdot \sqrt{\frac{nm}{n+m}} \sim t_{n+m-2}$

Exact confidence interval formula

$$I_{\mu_1-\mu_2} = \bar{x} - \bar{y} \pm t_{n+m-2}\left(\frac{\alpha}{2}\right) \cdot s_p \cdot \sqrt{\frac{n+m}{nm}}.$$

Two sample t-test uses the test statistic $t = \frac{\bar{x}-\bar{y}}{s_p} \cdot \sqrt{\frac{nm}{n+m}}$ for testing $H_0: \mu_1 = \mu_2$. The null distribution of the test statistic is

$$T \sim t_{n+m-2}$$

Example: iron retention

The data on percentage of iron retained by mice

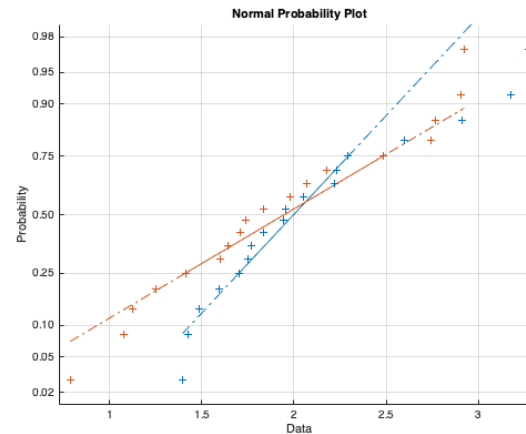
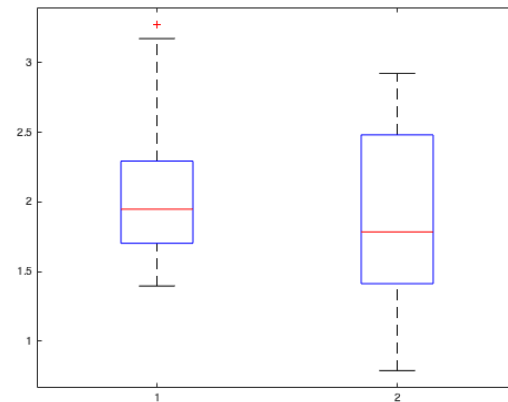
$$\text{Fe}^{2+}: n = 18, \bar{x} = 9.63, s_1 = 6.69, s_{\bar{x}} = 1.58$$

$$\text{Fe}^{3+}: m = 18, \bar{y} = 8.20, s_2 = 5.45, s_{\bar{y}} = 1.28$$

has "iron form" as the main factor with two levels Fe^{2+} and Fe^{3+} .

The boxplots and normal probability plots show that the distributions are not normal. After the log transformation the data look more like normally distributed

$$\begin{aligned} \bar{x}' &= 2.09, s_1' = 0.659, s_{\bar{x}'} = 0.155, \\ \bar{y}' &= 1.90, s_2' = 0.574, s_{\bar{y}'} = 0.135. \end{aligned}$$



For the log-transformed data we get $t_{\text{obs}} = 0.917$, $\text{df} = 34$, so that the two-sided p-value = 36.6%.