

**POLITECNICO**  
MILANO 1863

DEPARTMENT OF MECHANICAL  
ENGINEERING

## EXERCISE CLASS 1 (part 2/2)

**Review of basic statistical concepts:  
assumptions check and hypothesis testing  
(2 samples)**

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# Hypotesis testing with 2 samples



- One sample tests:
  - Test for mean (known variance): one-sample z-test
  - Test for mean (unknown variance): one-sample t-test
  - Test for variance: chi-squared test (variance)
- Two samples tests
  - Test for mean difference (known var): two-sample z-test
  - Test for mean difference (unknown var): two-sample t-test
  - Test for mean of paired data (unknown var): paired t-test
  - Test for equality of variances: F-test (variances)

# Paired t-test

## Assumptions:

- $X_{11}, X_{12}, \dots, X_{1n}$  is a random sample of size  $n$  from population 1;
- $X_{21}, X_{22}, \dots, X_{2n}$  is a random sample of size  $n$  from population 2;
- The differences between pairs,  $D_j = X_{1j} - X_{2j}$ , are **normal** (or central limit theorem applies);
- The variance of the differences between pairs is **unknown**

Under those assumptions, the quantity:

$$T = \frac{\bar{D} - \mu_D}{S_D / \sqrt{n}}$$

Follows a student-t distribution with  $n - 1$  degrees of freedom,

where:  $D_j = X_{1j} - X_{2j} \sim N(\mu_D, \sigma_D^2)$

(NOTICE: we are assuming differences to be random normal var.)

## The Paired $t$ -Test

Null hypothesis:  $H_0: \mu_D = \Delta_0$

Test statistic:  $T_0 = \frac{\bar{D} - \Delta_0}{S_D / \sqrt{n}} \quad (5-16)$

### Alternative Hypothesis

$$H_1: \mu_D \neq \Delta_0$$

$$H_1: \mu_D > \Delta_0$$

$$H_1: \mu_D < \Delta_0$$

### Rejection Region

$$t_0 > t_{\alpha/2, n-1} \text{ or } t_0 < -t_{\alpha/2, n-1}$$

$$t_0 > t_{\alpha, n-1}$$

$$t_0 < -t_{\alpha, n-1}$$

The  $100(1 - \alpha)\%$  confidence interval on the difference between the population means is given by:

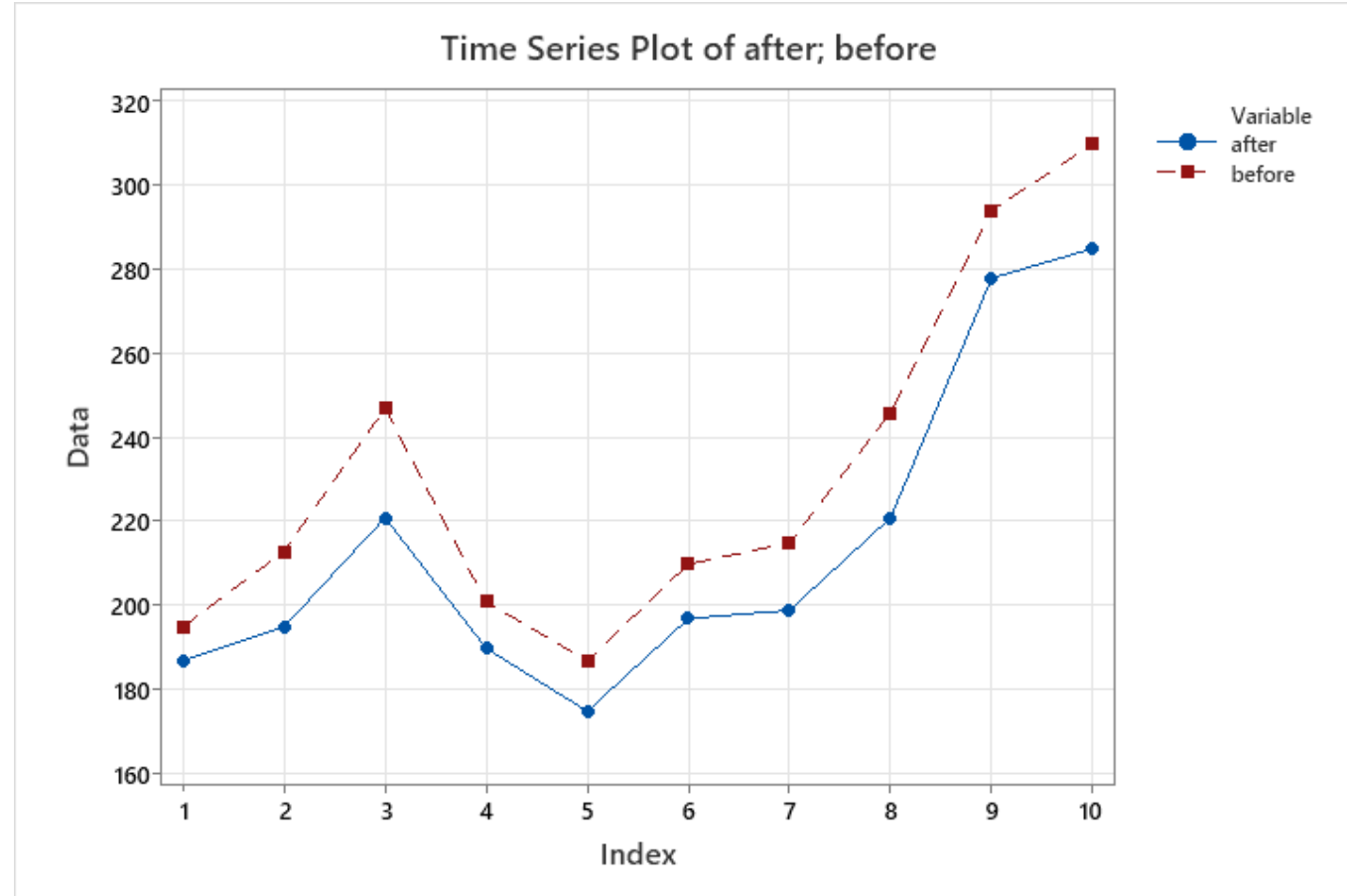
$$\bar{D} - t_{\frac{\alpha}{2}, n-1} S_D / \sqrt{n} \leq \mu_D \leq \bar{D} + t_{\frac{\alpha}{2}, n-1} S_D / \sqrt{n}$$

$S_D$  is lower than the pooled standard deviation: paired test is more precise (smaller c.i.)

# Exercise 1

Ten people are involved in a diet program. The weights before and after the program is reported in the table (expressed in pounds, 1 lb = 0.454). The data is stored in the file `weights.csv`.

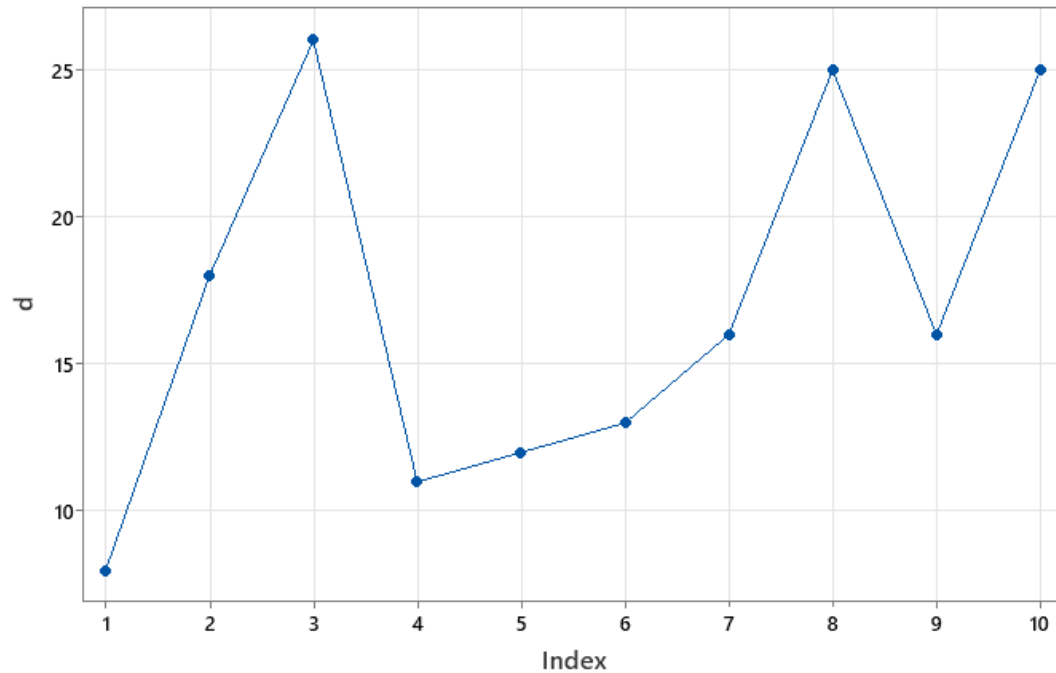
Is there statistical evidence (95%) to state that the diet program was effective? Design a two-sided confidence interval at 95% on the weight difference



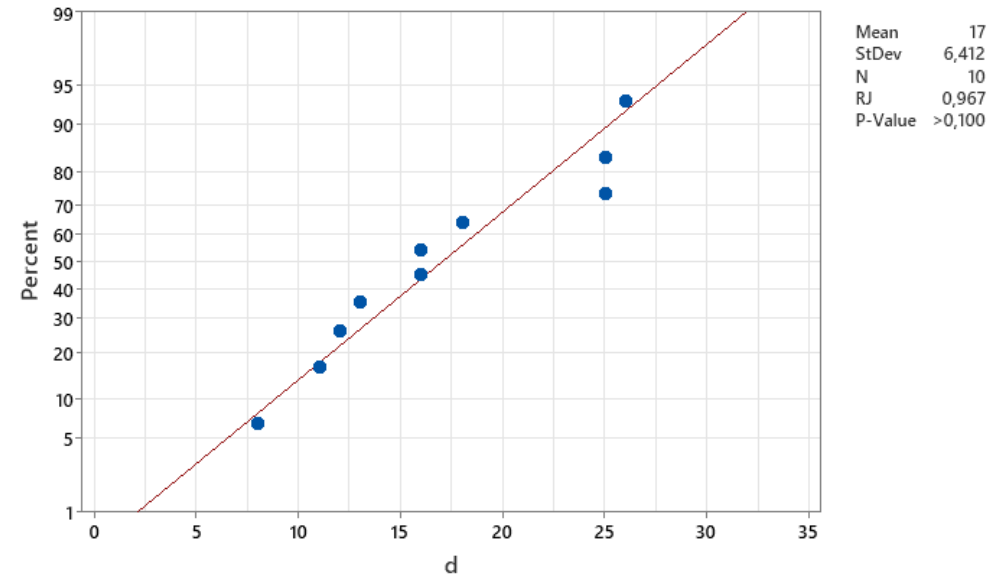
# Exercise 1

Preliminary activity: data visualization and **assumptions check** (on the difference “d”)

Time Series Plot of d



Probability Plot of d  
Normal



## Runs test

### Test

Null hypothesis  $H_0$ : The order of the data is random  
Alternative hypothesis  $H_1$ : The order of the data is not random

### Number of Runs

### Observed Expected P-Value

Observed	Expected	P-Value
6	5,80	0,888

*The p-value may not be accurate for samples with fewer than 11 observations above K or fewer than 11 below.*



# Exercise 1

Is there statistical evidence (95%) to state that the diet program was effective?  
Design a two-sided confidence interval at 95% on the weight difference

Knowing that the data are normally distributed, we can use the t-test to evaluate the following hypothesis:  $H_0 : \mu_d = 0$  vs  $H_1 : \mu_d > 0$

The t-test statistic is:

$$t_0 = \frac{\bar{d} - \Delta_0}{s_d / \sqrt{n}}$$

where  $\bar{d}$  is the sample mean of the differences,  $s_d$  is the sample standard deviation of the differences and  $n$  is the number of observations.

In this case,  $\Delta_0 = 0$

## Test

Null hypothesis  $H_0: \mu = 0$   
Alternative hypothesis  $H_1: \mu > 0$

T-Value	P-Value
8,38	0,000

Two-sided confidence interval at 95% on 'd':

## Descriptive Statistics

N	Mean	StDev	SE Mean	95% CI for $\mu$
10	17,00	6,41	2,03	(12,41; 21,59)

$\mu$ : population mean of d

# Not paired data and known variances

## Assumptions:

- $X_{11}, X_{12}, \dots, X_{1n_1}$  is a random sample of size  $n_1$  from population 1;
- $X_{21}, X_{22}, \dots, X_{2n_2}$  is a random sample of size  $n_2$  from population 2;
- The two populations are **independent**;
- Both populations are **normal** (or central limit theorem applies);
- The **variances** of the populations are **known**.

Under those assumptions, the quantity:

$$Z = \frac{\bar{X}_1 - \bar{X}_2 - (\mu_1 - \mu_2)}{\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}}$$

Follows a standard normal distribution,  $N(0,1)$

## Testing Hypotheses on the Difference in Means, Variances Known

Null hypothesis:  $H_0: \mu_1 - \mu_2 = \Delta_0$

Test statistic:  $Z_0 = \frac{\bar{X}_1 - \bar{X}_2 - \Delta_0}{\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}}$

### Alternative Hypotheses

$H_1: \mu_1 - \mu_2 \neq \Delta_0$

$H_1: \mu_1 - \mu_2 > \Delta_0$

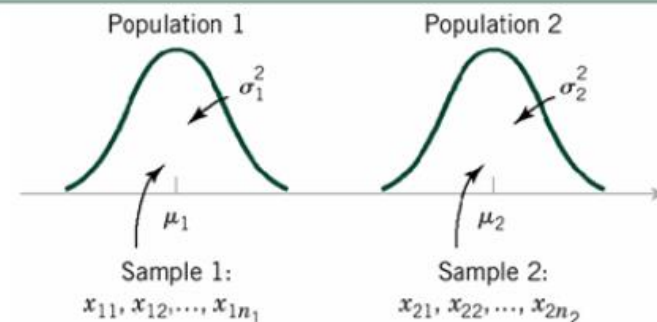
$H_1: \mu_1 - \mu_2 < \Delta_0$

### Rejection Criterion

$z_0 > z_{\alpha/2}$  or  $z_0 < -z_{\alpha/2}$

$z_0 > z_{\alpha}$

$z_0 < -z_{\alpha}$



Two independent populations.

# Not paired data and equal variances

**Assumptions** (case 1:  $\sigma_1^2 = \sigma_2^2 = \sigma^2$ ):

- $X_{11}, X_{12}, \dots, X_{1n_1}$  is a random sample of size  $n_1$  from population 1;
- $X_{21}, X_{22}, \dots, X_{2n_2}$  is a random sample of size  $n_2$  from population 2;
- The two populations are **independent**;
- Both populations are **normal** (or central limit theorem applies);
- The **variances** of the populations are **unknown and equal**.

Under those assumptions, the quantity:

$$T = \frac{\bar{X}_1 - \bar{X}_2 - (\mu_1 - \mu_2)}{S_p \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}}$$

Follows a student-t distribution with  $n_1 + n_2 - 2$  degrees of freedom.

**Pooled variance:**

$$S_p^2 = \frac{(n_1 - 1)S_1^2 + (n_2 - 1)S_2^2}{n_1 + n_2 - 2}$$

## Testing Hypotheses on the Difference in Means of Two Normal Distributions, Variances Unknown and Equal<sup>1</sup>

Null hypothesis:  $H_0: \mu_1 - \mu_2 = \Delta_0$

Test statistic:  $T_0 = \frac{\bar{X}_1 - \bar{X}_2 - \Delta_0}{S_p \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}} \quad (5-9)$

### Alternative Hypothesis

$$H_1: \mu_1 - \mu_2 \neq \Delta_0$$

$$H_1: \mu_1 - \mu_2 > \Delta_0$$

$$H_1: \mu_1 - \mu_2 < \Delta_0$$

### Rejection Criterion

$$t_0 > t_{\alpha/2, n_1 + n_2 - 2} \text{ or}$$

$$t_0 < -t_{\alpha/2, n_1 + n_2 - 2}$$

$$t_0 > t_{\alpha, n_1 + n_2 - 2}$$

$$t_0 < -t_{\alpha, n_1 + n_2 - 2}$$



# Not paired data and not equal variances

*Assumptions* (case 2:  $\sigma_1^2 \neq \sigma_2^2$ ):

- $X_{11}, X_{12}, \dots, X_{1n}$  is a random sample of size  $n_1$  from population 1;
- $X_{21}, X_{22}, \dots, X_{2n}$  is a random sample of size  $n_2$  from population 2;
- The two populations are **independent**;
- Both populations are **normal** (or central limit theorem applies);
- The **variances** of the populations are **unknown and not equal**.

Under those assumptions, the quantity:

$$T = \frac{\bar{X}_1 - \bar{X}_2 - (\mu_1 - \mu_2)}{\sqrt{\frac{S_1^2}{n_1} + \frac{S_2^2}{n_2}}} \quad v = \frac{\left(\frac{S_1^2}{n_1} + \frac{S_2^2}{n_2}\right)^2}{\frac{(S_1^2/n_1)^2}{n_1 - 1} + \frac{(S_2^2/n_2)^2}{n_2 - 1}}$$

Follows a student-t distribution with  $v$  degrees of freedom

# Exercise 2

We want to evaluate the resistance of resistors provided by two different suppliers.  
The data is stored in the file `resistance.csv`.

1. What can we **infer about the mean** resistance of the resistors provided by the two different suppliers?
2. Compute the **Type II error** expression in the variance equality test and compute the **test power** when the true variance of the first supplier is 1.5 times larger than the one of the second supplier

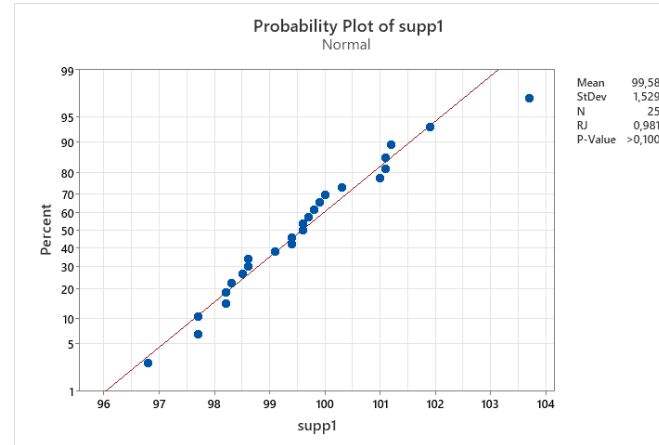
# Exercise 2

## Preliminary activity: data visualization and assumptions check

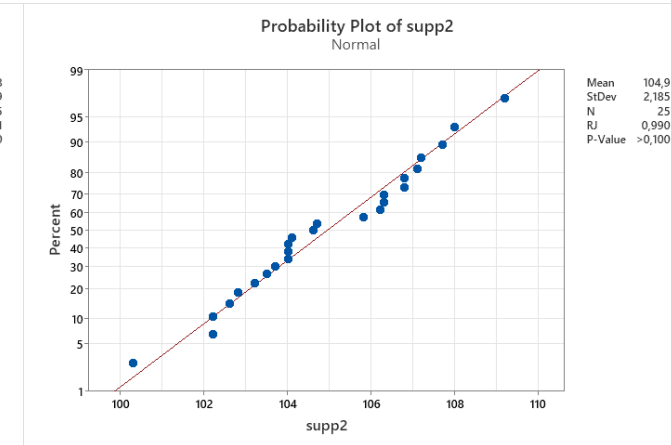
### Statistics

Variable	N	N*	Mean	SE Mean	StDev	Minimum	Q1	Median	Q3	Maximum
supp2	35	0	105,069	0,331733	1,96256	100,3	103,7	104,7	106,8	109,2
supp1	25	10	99,576	0,305793	1,52896	96,8	98,4	99,6	100,65	103,7

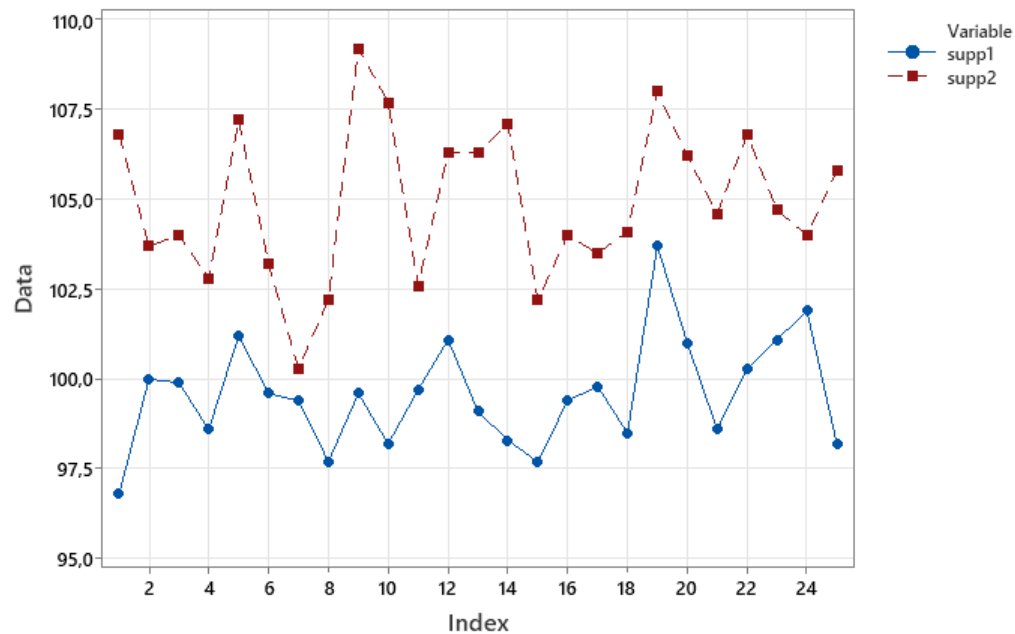
Supplier 1:



Supplier 2:



Time Series Plot of supp1; supp2



### Test

### Runs test

Null hypothesis

$H_0$ : The order of the data is random

Alternative hypothesis  $H_1$ : The order of the data is not random

#### Number of Runs

Variable	Observed	Expected	P-Value
supp1	15	13,48	0,534
supp2	13	13,32	0,894

# Point 1

What can we **infer about the mean** resistance of the resistors provided by the two different suppliers?

We want to compare the means of two populations. Variances are unknown thus there are two possible situations:

- Equal (unknown) variances
- Different (unknown) variances



We can use the F-test to test the equality of variances.

## First step: hypothesis test on the equality of variances

Null hypothesis: the two variances are equal

$$H_0 : \sigma_1^2 = \sigma_2^2$$

Alternative hypothesis: the two variances are different

$$H_1 : \sigma_1^2 \neq \sigma_2^2$$

This hypothesis test is equivalent to:

$$H_0 : \frac{\sigma_1^2}{\sigma_2^2} = 1$$

$$H_1 : \frac{\sigma_1^2}{\sigma_2^2} \neq 1$$

# Point 1

What can we **infer about the mean** resistance of the resistors provided by the two different suppliers?

## F-TEST

### Assumptions:

- $X_{11}, X_{12}, \dots, X_{1n_1}$  is a random sample of size  $n_1$  from population 1;
- $X_{21}, X_{22}, \dots, X_{2n_2}$  is a random sample of size  $n_2$  from population 2;
- The two populations are **normal** (or central limit theorem applies);
- The two populations are **independent**;
- The variances of the populations are **unknown** (obvious)

Under those assumptions, the quantity:

$$F = \frac{S_1^2 / \sigma_1^2}{S_2^2 / \sigma_2^2}$$

Follows an  $F$  distribution with  $n_1 - 1$  and  $n_2 - 1$  degrees of freedom.

### Testing Hypotheses on the Equality of Variances of Two Normal Distributions

Null hypothesis:  $H_0: \sigma_1^2 = \sigma_2^2$

Test statistic:  $F_0 = \frac{S_1^2}{S_2^2}$  (5-21)

#### Alternative Hypotheses

#### Rejection Criterion

$$H_1: \sigma_1^2 \neq \sigma_2^2$$

$$f_0 > f_{\alpha/2, n_1-1, n_2-1} \text{ or } f_0 < f_{1-\alpha/2, n_1-1, n_2-1}$$

$$H_1: \sigma_1^2 > \sigma_2^2$$

$$f_0 > f_{\alpha, n_1-1, n_2-1}$$

$$H_1: \sigma_1^2 < \sigma_2^2$$

$$f_0 < f_{1-\alpha, n_1-1, n_2-1}$$

The  $100(1 - \alpha)\%$  confidence interval on the difference between the population means is given by:

$$\frac{S_1^2}{S_2^2} f_{1-\alpha/2, n_2-1, n_1-1} \leq \frac{\sigma_1^2}{\sigma_2^2} \leq \frac{S_1^2}{S_2^2} f_{\alpha/2, n_2-1, n_1-1}$$



# Point 1

What can we **infer about the mean** resistance of the resistors provided by the two different suppliers?

## F-TEST

Two-Sample Variance

C1 supp1  
C2 supp2

Each sample is in its own column

Sample 1: supp1

Sample 2: supp2

Select Options... Graphs... Results...

Help OK Cancel

### Ratio of Standard Deviations

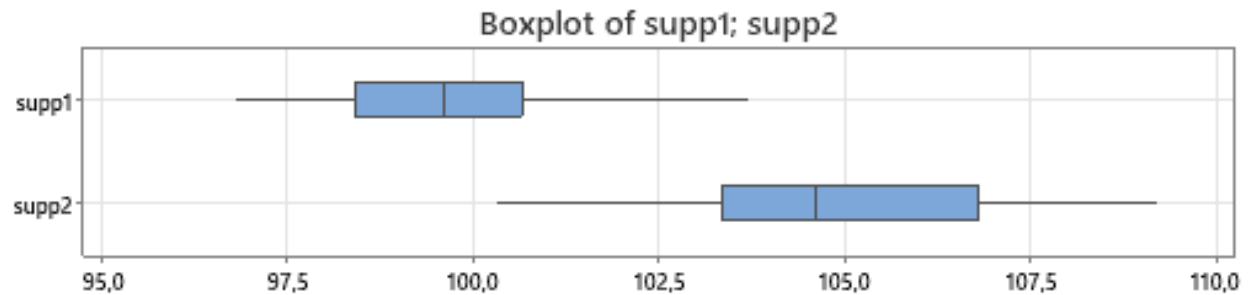
Estimated 95% CI for Ratio

Ratio	using F
0,699654	(0,464; 1,054)

### Test

Null hypothesis  $H_0: \sigma_1 / \sigma_2 = 1$   
Alternative hypothesis  $H_1: \sigma_1 / \sigma_2 \neq 1$   
Significance level  $\alpha = 0,05$

Test				
Method	Statistic	DF1	DF2	P-Value
F	0,49	24	24	0,087

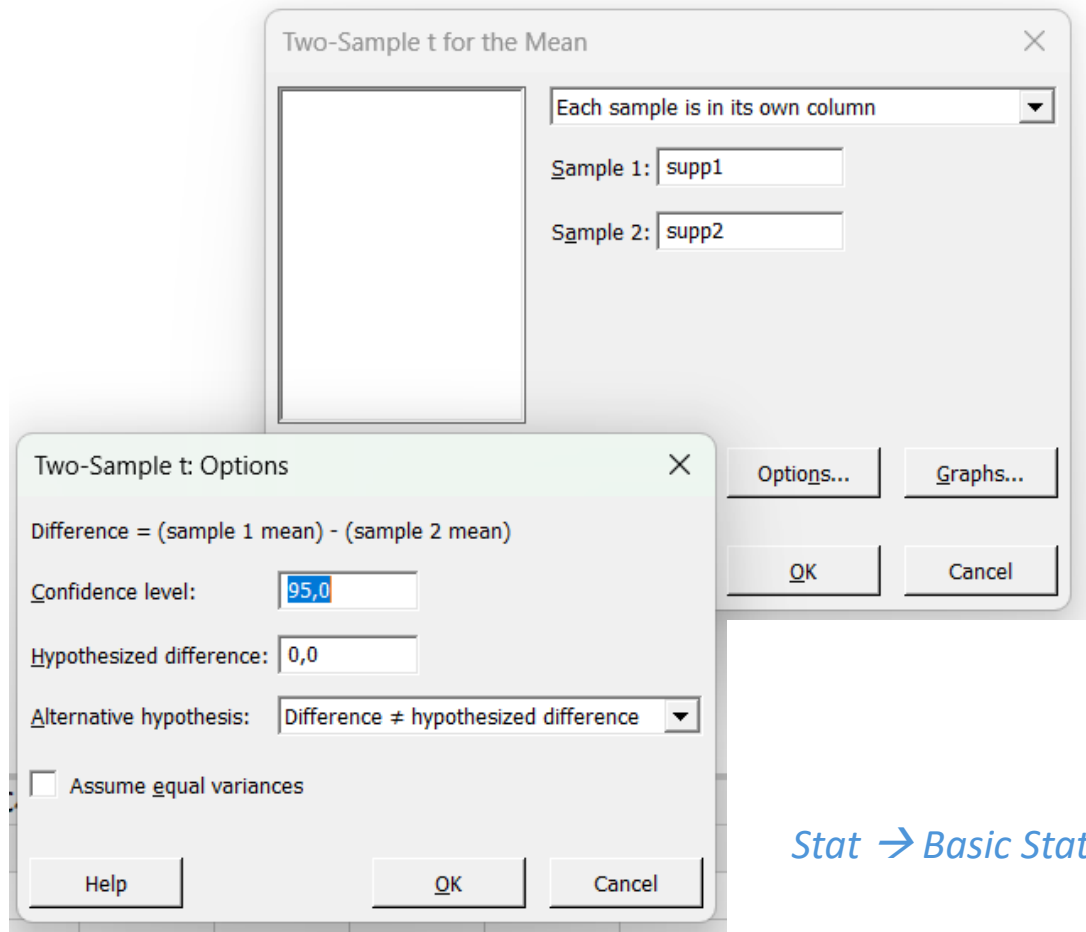


Stat → Basic Statistics → 2 Variances

# Point 1

What can we **infer about the mean** resistance of the resistors provided by the two different suppliers?

Now that we have verified the equality of variances, we can perform the **t-test (with equal variances)**:



## Estimation for Difference

95% CI for  
Difference    Difference  
-5,356 (-6,432; -4,280)

## Test

Null hypothesis  $H_0: \mu_1 - \mu_2 = 0$   
Alternative hypothesis  $H_1: \mu_1 - \mu_2 \neq 0$

T-Value	DF	P-Value
-10,04	42	0,000

*Stat → Basic Statistics → 2-sample t*

# Point 2

Compute the **Type II error** expression in the variance equality test and compute the **test power** when the true variance of the first supplier is **1.5 times larger** than the one of the second supplier

The Type II error is the probability of accepting the null hypothesis when it is false.

$$\beta = Pr(\text{accept } H_0 \text{ when } H_1 \text{ is true})$$

Let's expand the formula for the F-test:

$$\beta = Pr \left( F_{1-\alpha/2, n_1-1, n_2-1} \leq \frac{s_1^2}{s_2^2} \leq F_{\alpha/2, n_1-1, n_2-1} \mid \frac{\sigma_1^2}{\sigma_2^2} = \delta \neq 1 \right)$$

If we multiply all the terms by  $\sigma_2^2/\sigma_1^2$  we get:

$$\frac{S_1^2/\sigma_1^2}{S_2^2/\sigma_2^2} \sim F_{n_1-1, n_2-1}$$

If we substitute  $\sigma_2^2/\sigma_1^2$  with the ratio we want to test, we get:

$$\beta = Pr \left( \frac{F_{1-\alpha/2, n_1-1, n_2-1}}{1.5} \leq \frac{S_1^2/\sigma_1^2}{S_2^2/\sigma_2^2} \leq \frac{F_{\alpha/2, n_1-1, n_2-1}}{1.5} \right)$$

Rearranging the terms we get:

$$\beta = Pr \left( \frac{S_1^2/\sigma_1^2}{S_2^2/\sigma_2^2} \leq \frac{F_{\alpha/2, n_1-1, n_2-1}}{1.5} \right) - Pr \left( \frac{S_1^2/\sigma_1^2}{S_2^2/\sigma_2^2} \leq \frac{F_{1-\alpha/2, n_1-1, n_2-1}}{1.5} \right)$$

# Point 2

Compute the **Type II error** expression in the variance equality test and compute the **test power** when the true variance of the first supplier is **1.5 times larger** than the one of the second supplier

$$\beta = P\left(F_{1-\alpha/2, n_1-1, n_2-1} < \frac{S_1^2}{S_2^2} < F_{\alpha/2, n_1-1, n_2-1} \mid \frac{\sigma_1^2}{\sigma_2^2} = \delta \neq 1\right) = P\left(\frac{\sigma_2^2}{\sigma_1^2} F_{1-\alpha/2, n_1-1, n_2-1} < \frac{S_1^2 / \sigma_1^2}{S_2^2 / \sigma_2^2} < \frac{\sigma_2^2}{\sigma_1^2} F_{\alpha/2, n_1-1, n_2-1} \mid \frac{\sigma_1^2}{\sigma_2^2} = \delta\right)$$

$$\frac{S_1^2 / \sigma_1^2}{S_2^2 / \sigma_2^2} \sim F_{n_1-1, n_2-1} \quad \delta = 1.5 \quad n_1 = 25 \quad n_2 = 35 \quad \left\{ \begin{array}{l} F_{0.025, 24, 34} = 2.07 \\ F_{0.975, 24, 34} = \frac{1}{F_{0.025, 34, 24}} = \frac{1}{2.18} = 0.459 \end{array} \right.$$

$$\beta = P\left(\frac{0.459}{1.5} < \frac{S_1^2 / \sigma_1^2}{S_2^2 / \sigma_2^2} < \frac{2.07}{1.5}\right) = P\left(0.306 < \frac{S_1^2 / \sigma_1^2}{S_2^2 / \sigma_2^2} < 1.38\right) = P\left(\frac{S_1^2 / \sigma_1^2}{S_2^2 / \sigma_2^2} < 1.38\right) - P\left(\frac{S_1^2 / \sigma_1^2}{S_2^2 / \sigma_2^2} < 0.306\right) = 0.8073$$

## Cumulative Distribution Function

F distribution with 24 DF in numerator and 34 DF in denominator

x	P( X <= x )
0.3060	0.0018

## Cumulative Distribution Function

F distribution with 24 DF in numerator and 34 DF in denominator

x	P( X <= x )
1.3800	0.8091

**The test power is**  
 **$P = 1 - \beta = 19.27\%$**