

One Sample mean test: $H_0: \mu = \text{constant}$

$$Z = \frac{\bar{x} - \mu}{\sigma/\sqrt{n}}$$

$$t_{n-1} = \frac{\bar{x} - \mu}{s/\sqrt{n}}$$

Two Sample mean test: $H_0: \mu_1 = \mu_2$

σ_1, σ_2 known

σ_1, σ_2 unknown assuming $\sigma_1^2 = \sigma_2^2$

σ_1, σ_2 unknown

More sample mean test: $H_0: \mu_1 = \mu_2 = \mu_3 = \dots = \mu_a$

H_{a1} : at least one of the mean is equal to others.

13.2 Analysis of Variance (F-test)

Example:

	Level ①	level ②	level ③	level ④
		5.8	5.9	6.2
	7.5	7.3	$y_{32} = 6.2$	6.8
	6.2	8.2	<u>6.2</u>	5.7
$y_{1.}$	6.9	7.1	5.8	4.9
$\bar{y}_{1.}$	<u>7.4</u>	7.8	4.7	6.2
	9.2	<u>7.8</u>	8.3	7.1
	8.3	$n_2 = 5$	7.2	5.8
	7.6		6.2	5.4
	<u>7.6</u>		<u>6.2</u>	<u>5.4</u>
	$n_1 = 7$		$n_3 = 7$	$n_4 = 8$

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 H_a : at least of ^{one} mean is different.

② Test Stat:

Notation: $a = \#$ of treatment

$y_{i.} =$ total of i th treatment.

$\bar{y}_{i.} =$ mean of i th treatment.

$y_{..} =$ grand total

$\bar{y}_{..} =$ grand mean

$N = \#$ of total observations.

$n_i = \#$ of i th group's observations.

$y_{ij} = j$ -th observation in the i th group.

$S_i^2 =$ sample variance of i th group.

Total sum of squares (SST)

$$SST = \sum_{i=1}^a \sum_{j=1}^{n_i} (y_{ij} - \bar{y}_{..})^2 = \sum_{i=1}^a \sum_{j=1}^{n_i} y_{ij}^2 - \frac{y_{..}^2}{N}$$

Within groups or error sum of squares (SSE)

$$SSE = \sum_{i=1}^a \sum_{j=1}^{n_i} (y_{ij} - \bar{y}_{i.})^2 = \sum_{i=1}^a (n_i - 1) S_i^2$$

Variation among groups or treatments sum of squares (SSTr)

$$SSTr = \sum_{i=1}^a n_i (\bar{y}_{i.} - \bar{y}_{..})^2 = \sum_{i=1}^a \frac{y_{i.}^2}{n_i} - \frac{y_{..}^2}{N}$$

$$SST = SSE + SSTr$$

ANOVA Table - (F-test)

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F

ANOVA Table - (F-test)

ANOVA	SS	DF	MS	F
Treatment	SS_{Tr}	$a-1$	$MSTr = \frac{SS_{Tr}}{a-1}$	$F = \frac{MSTr}{MSE}$
Error	SSE	$N-a$	$MSE = \frac{SSE}{N-a}$	
Total	SST	$N-1$		

Test Stat = $F_{a-1, N-a}$

Example: (Cont.)

$$\begin{array}{cccc}
 y_{1.} = 53.1 & y_{2.} = 36.2 & y_{3.} = 44.3 & y_{4.} = 48.1 \\
 \bar{y}_{1.} = 7.59 & \bar{y}_{2.} = 7.24 & \bar{y}_{3.} = 6.33 & \bar{y}_{4.} = 6.01 \\
 n_1 = 7 & n_2 = 5 & n_3 = 7 & n_4 = 8
 \end{array}$$

$$y_{..} = 181.7 \quad \bar{y}_{..} = 6.73 \quad N = 27 \quad a = 3$$

① Hypothesis: $H_0: \mu_1 = \mu_2 = \mu_3 = \mu_4$
 $H_a: \text{at least one mean is different.}$

② Test Stat:

$$SSTr = \sum_{i=1}^a \frac{y_{i.}^2}{n_i} - \frac{y_{..}^2}{N} = 11.673$$

$$SSE = \sum_{i=1}^a (n_i - 1) s_i^2 = 20.3043$$

ANOVA Table

	SS	DF	MS	F
Treatment	11.673	$a-1 = 3$	3.891	$\frac{3.891}{0.8828} = 4.4$
Error	20.3043	$N-a = 27-4 = 23$	0.8828	
Total	Test Stat		Critical Value	

Total	Test Stat	Critical Value	
③	F	F_{α}^*	reject H_0 . (at $\alpha=5\%$)
	$ F=4.4 $	$ F_{3,23,0.05}^* = 3.03 $	

④ Conclusion: reject H_0 . There is at least one of the mean is different.

13.2 Test for individual pairs of mean.

$$4C_2 = 6.$$

- 1-2
- 1-3
- 1-4
- 2-3
- 2-4
- 3-4

Fisher's LSD Test:

① Hypothesis: $H_0: \mu_i = \mu_j$ for $i \neq j$ from 1 to a.
 $H_{a1}: \mu_i \neq \mu_j$

② Step 1: $\bar{y}_{i.} - \bar{y}_{j.}$

Step 2: $LSD = t_{n-a, \frac{\alpha}{2}} \cdot \sqrt{MSE(\frac{1}{n_i} + \frac{1}{n_j})}$
 \uparrow
 from ANOVA

Compare $\bar{y}_{i.} - \bar{y}_{j.}$ and LSD.

Reject H_0 if $|\bar{y}_{i.} - \bar{y}_{j.}| > \underline{LSD}$

Reject H_0 if $\underbrace{|y_i - y_j|} > \underbrace{LSD}$

Example (Cont.):

$$t_{n-a, \frac{\alpha}{2}} = t_{23, \text{two tail}, \alpha=5\%}$$

$$= 2.069$$

Hypothesis	LSD	Difference	Conclusion
$H_0: \mu_1 = \mu_2$	1.138	0.35	X
$H_0: \mu_1 = \mu_3$	1.039	1.26	✓ reject
$H_0: \mu_1 = \mu_4$	1.006	1.58	✓ reject
$H_0: \mu_2 = \mu_3$	1.138	0.91	X
$H_0: \mu_2 = \mu_4$	1.108	1.23	✓ reject
$H_0: \mu_3 = \mu_4$	1.006	0.32	X

Confidence Interval, Fish's LSD:

100(1- α)% C.I for $\mu_i - \mu_j$:

$$\bar{y}_i - \bar{y}_j \pm LSD$$

Note: If zero is in this interval, do not reject H_0
 is not ..., reject H_0

Example: $\mu_1 - \mu_2$ 95% CI:

$$(7.59 - 7.24) \pm 1.138 = (-0.788, 1.488)$$

ANOVA'S Assumption

1. Each population must be normal.
2. The population have equal Variances.
3. The Samples must be independent.
4. All the other factors must be consistent,

5. $\hat{e}_{ij} = y_{ij} - \bar{y}_i$. (residual for ANOVA)

\hat{e}_{ij} follows a normal distribution.

6. The variance of residuals should be similar for groups.