

# **MEM-264 Applied Statistics**

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## Two-way ANOVA

- ▶ In the one-way ANOVA, we classified populations according to a single factor (categorical variable).
- ▶ In the two-way ANOVA, we will extend the model to take into account two factors.

### Example : Student's Grades

		Columns	
Department		Male (j=1)	Female (j=2)
Rows	Mathematics (i=1)	<u>310</u> (mean: 6.38) <sup>(i,j)</sup>	400 (mean: 6.52)
	Physics (i=2)	$\eta_{2,1} = \underline{510}$ (mean: 6.41) <sub>i=2, j=1, k=1, \dots, 510</sub>	330 (mean: 6.56)
	Computer Science (i=3)	190 (mean: 6.44)	210 (mean: 6.41)

mean:  $\frac{310 \cdot 6.38 + 400 \cdot 6.52}{710}$   
 $310 + 400 = 710$

- ▶ **With this analysis we can investigate interactions between factors!**

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Overall sample mean  $\bar{Y} = \frac{310 \cdot 6.38 + 400 \cdot 6.52 + \dots + 210 \cdot 6.41}{310 + 510 + \dots + 210}$

## Formulation of the Two-way ANOVA model

One-way:

$$Y_{ik} = \mu + \alpha_i + \varepsilon_{ik}, \quad \varepsilon_{ik} \sim \mathcal{N}(0, \sigma^2) \quad \sum \alpha_i = 0$$

individual      1st cat. var.      2nd cat. var.      interactions      Random error

$$Y_{ijk} = \mu + \alpha_i + \beta_j + \gamma_{ij} + \varepsilon_{ijk}, \quad \varepsilon_{ijk} \sim \mathcal{N}(0, \sigma^2)$$

$$\sum_{i=1}^I \alpha_i = 0, \quad \sum_{j=1}^J \beta_j = 0, \quad \sum_{i=1}^I \gamma_{ij} = 0, \quad \sum_{j=1}^J \gamma_{ij} = 0$$

for all  $j=1, \dots, J$       for all  $i=1, \dots, I$

$$\begin{bmatrix} \alpha_1 \\ \vdots \\ \alpha_I \end{bmatrix} \quad \begin{bmatrix} \beta_1 \\ \vdots \\ \beta_J \end{bmatrix} \quad \Gamma = \begin{bmatrix} \gamma_{11} & \dots & \gamma_{1J} \\ \vdots & & \vdots \\ \gamma_{I1} & \dots & \gamma_{IJ} \end{bmatrix}$$

- ▶ PAIR ( $i=1, j=1$ ):  $\{Y_{1,1,1}, \dots, Y_{1,1,n_{1,1}}\}$  i.i.d sample
  - ▶ PAIR ( $i=1, j=2$ ):  $\{Y_{1,2,1}, \dots, Y_{1,2,n_{1,2}}\}$  i.i.d sample
  - ⋮
  - ▶ ~~GROUP~~ PAIR ( $i=I, j=J$ ):  $\{Y_{I,J,1}, \dots, Y_{I,J,n_{I,J}}\}$  i.i.d sample
- } IJ-i.i.d samples

- ▶ There are three null hypotheses (Rows, Columns, interactions)

### The three null hypotheses of the model

$$H_{I0} : \alpha_1 = \alpha_2 = \cdots = \alpha_I$$

$$H_{J0} : \beta_1 = \beta_2 = \cdots = \beta_J$$

$$H_{IJ0} : \gamma_{11} = \cdots = \gamma_{IJ}$$

# Two-way ANOVA

## Sample mean for the estimation of the mean $\mu_{ij}$

$$\bar{Y}_{ij, n_{i,j}} = \frac{1}{n_{i,j}} \sum_{k=1}^{n_{i,j}} Y_{ijk}$$

In our example

$$\bar{Y}_{21,510} = 6.41 \text{ (see the table)}$$

## Sample variance for the estimation of the variance $\sigma^2$

$$S_{ij, n_{i,j}}^2 = \frac{1}{n_{i,j} - 1} \sum_{k=1}^{n_{i,j}} e_{ijk}^2, \quad e_{ijk} = Y_{ijk} - \bar{Y}_{ij, n_{i,j}}$$

## Pooled sample variance

$$S_p^2 = \frac{\sum_{i,j} (n_{i,j} - 1) S_{ij, n_{i,j}}^2}{\sum_{i,j} (n_{i,j} - 1)} = N - IJ$$

← this is a better approximation of  $\sigma^2$

## Two-way ANOVA

$$N = \sum_{i=1}^I \sum_{j=1}^J n_{i,j}$$

### Overall sample mean

$$\bar{Y} = \frac{1}{\sum_{i,j} n_{i,j}} \sum_{i,j,k} Y_{ijk}$$

### Sample variance between groups

$$S_b^2 = \frac{\text{SSM}}{\text{DF}} = \frac{1}{IJ - 1} \sum_{i,j} (\bar{Y}_{ij, n_{ij}} - \bar{Y})^2$$

## Two-way ANOVA

In two-way ANOVA, the terms SSM and DF can be decomposed as follows:

$$SSM = SSI + SSJ + SSIJ$$

$$DF = DFI + DFJ + DFIJ$$

$\rightarrow SS_{IJ} = SSM - (SS_I + SS_J)$

### Two-way ANOVA table

		DF	F (numerator)	F (denominator)		
$F_I$	Rows	I-1	SSI/(I-1)	$S_p^2$	$DF = N - IJ$	$F_I = \frac{SS_I}{S_p^2}$
$F_J$	Columns	J-1	SSJ/(J-1)	$S_p^2$	$DF = N - IJ$	
$F_{IJ}$	Interaction	(I-1)(J-1)	SSIJ/[(I-1)(J-1)]	$S_p^2$	$DF = N - IJ$	

$P\text{-value}_I$

$P\text{-value}_J$

$P\text{-value}_{IJ}$

$\alpha = 0.05$

After the analysis we will obtain three p-values

### p-value for I (or/and J)

- ▶ If this p-value is smaller than the considered  $\alpha$ , this means that the first (or/and second) factor have a statistically significant effect on the values.

### p-value for IJ

- ▶ If this p-value is smaller than the considered  $\alpha$ , this means that there is a statistically significant interaction effect between the factors.



## Back to the student's grades

$P\text{-value-}II < \alpha$

Department	Male (j=1)	Female (j=2)	<u>Mean</u>
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Physics (i=2)	510 (mean: 6.41)	330 (mean: 6.56)	
Computer Science (i=3)	190 (mean: 6.44)	210 (mean: 6.41)	
<u>Mean</u>			

