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فاكولتي ڦرتانين

# **AGR5201**

## **Advanced Statistical Methods**

### **Factorial experiments**

# Topic outline

## **1.0 Factorial experiment**

### **1.1 Two factors experiment**

### **1.2 Interaction**

### **1.3 Result presentation**

### **1.4 Three factors experiment**

### **1.5 Two-factorial example**

### **1.6 Review question**

#### **Reference:**

Gomez and Gomez (page 84)



# 1.0 Factorial experiment

## So far...

- We have seen experiments with single factor only (rate of nitrogen, variety, etc.)
- The result can only be recommended in condition used in those experiment.
- If a different rate of fertilizer were tested using a different variety, the result will be different.
- Thus, a factorial experiment is conducted to study two (or more) factors at once.

# 1.0 Factorial experiment

## Single factor experiment

In a single factor ANOVA, the effect of nitrogen and variety is determined using separate ANOVA table as shown below:

### Nitrogen experiment

- Blocks = 4
- Nitrogen rates: 5
- Variety = 1

Source	DF
Blocks	3
Nitrogen	4
Error	12
Total	19

### Variety experiment

- Blocks = 7
- Variety : 3
- Nitrogen rate: 1

Source	DF
Blocks	6
Variety	2
Error	12
Total	20

# 1.0 Factorial experiment

## Definition

- A method for designing treatments, which allows testing and estimation of **main effects** and **interaction effects**.
- An experiment which looks at the effects of more than one factor (independent variable=IV)
- The interaction effects is measured from experiment that involves **at least two factors with at least two levels** each factor.
- Example:
  - i. Factor A (a1, a2)
  - ii. Factor B (b1, b2)



# 1.0 Factorial experiment

## FACTOR

**A specific type of treatment.**

- Alfalfa cultivar (A)
- Rates of N fertilizer (N)
- Methods of weed control (W)
- Diet feds to animal (D)

## LEVEL

**A specific state of a factor.**

- Alfalfa cultivars (A) → Saranac, Arc, Oneida (a=3)
- Methods of weed control (W) → herbicide, tillage, none, hand (w=4)
- Rates of N fertilizer (N) → 0, 50, 100, 150, 200 kg/ha (n=5)
- Diets fed to animals (D) → 100% hay, 70% hay, 30% hay (d=3)



# 1.0 Factorial experiment

- **Single factor** – only one factor (independent variable)
  - effect of watering: 100% FC, 50% FC, 20% FC
- **Two factors** – two factors (independent variables)
  - Effect of watering: 100% FC, 50% FC, 20% FC
  - Effect of Nitrogen: 0 kg/ha, 50 kg/ha, 100 kg/ha
- **Three factors** – three factors (independent variables)
  - Effect of watering: 100% FC, 50% FC, 20% FC
  - Effect of Nitrogen: 0 kg/ha, 50 kg/ha, 100 kg/ha
  - Effect of variety: A, B, C.



**Factorial  
experiment**

# 1.0 Factorial experiment

## Factorial treatment

- A treatment defined by the combination of one level of each factor.
- Factorial treatment designs have one main purpose:

→ **TO INCREASE THE SCOPE OF INFERENCE.**

- Identify important factors.
- Identify optimum or suboptimum levels of a factor.
- Identify and study interactions between factors.

to increase the scope of inference



# 1.0 Factorial experiment

## Single factor ANOVA

1 factor: 3 levels 4 reps = 12 Experimental units (3 levels x 4 reps)

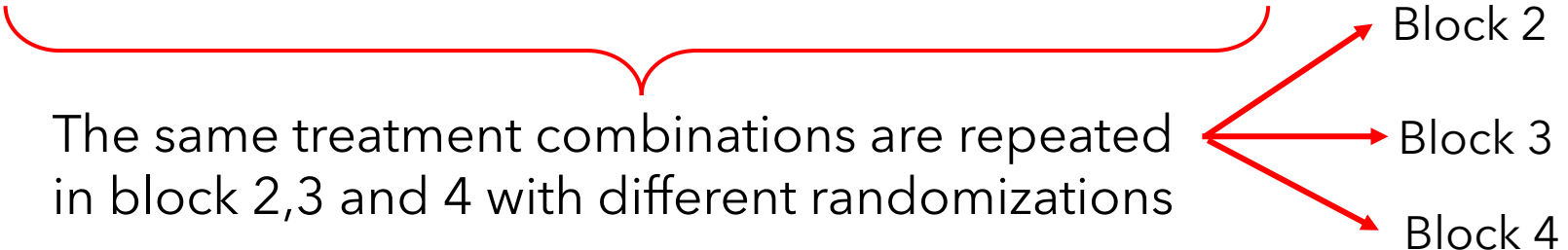
Source	df	SS	MS	F	Pr>F
Water	2				
Error	9				
Total	11				

# 1.1 Two factors experiment

- Water: 3 levels; Nitrogen: 3 levels; Rep = 4.
- Design RCBD =  $3 \times 3 \times 4 = 36$  experimental units

		Nitrogen			
		N 0	N 50	N 100	
Water	20% FC	N0W1	N50W1	N100W1	Block 1
	50% FC	N0W2	N50W2	N100W2	
	100% FC	N0W3	N50W3	N100W3	

The same treatment combinations are repeated in block 2,3 and 4 with different randomizations



# 1.1 Two factors experiment

## Linear model (RCBD)



$$Y_{ijk} = \mu + \rho_k + v_i + \pi_j + (v\pi)_{ij} + \varepsilon_{ijk}$$

Where,

$Y_{ijk}$  = observation

$\mu$  = overall mean

$\rho_k$  = the effect of the  $k^{\text{th}}$  block

$v_i$  = the effect of the  $i^{\text{th}}$  level of factor 1

$\pi_j$  = the effect of the  $j^{\text{th}}$  level of factor 2

$(v\pi)_{ij}$  = the interaction effect of the  $ij^{\text{th}}$  factorial treatment.

$\varepsilon_{ijk}$  = random error

$v_i + \pi_j + (v\pi)_{ij} = \tau_i$  if we were to ignore the factorial structure.

# 1.1 Two factors experiment

## ANOVA table

- Water (3 levels) Nitrogen (3 levels) Blocks (4)

	Source	df	SS	MS	F	Pr>F
	Block	3				
Main effect →	Water (W)	2				
Main effect →	Nitrogen (N)	2				
Interaction effect →	W x N	4				
	Error	24				
	Total	35				

# 1.1 Two factors experiment

## ANOVA table - CRD

Linear model:  
$$Y_{ij} = \mu + \nu_i + \pi_j + (\nu\pi)_{ij} + \varepsilon_{ij}$$

Source	df	SS	MS	F
Factor A	a-1	SSA	$MSA = SSA/(a-1)$	$F_A = MSA/MSE$
Factor B	b-1	SSB	$MSB = SSB/(b-1)$	$F_B = MSB/MSE$
A x B (interaction)	$(a-1)(b-1)$	SSAB	$MSAB = SSAB/(a-1)(b-1)$	$F_{AB} = MSAB/MSE$
Error	$ab(r-1)$	SSE	$MSE = SSE/(r-1)(ab-1)$	
Total	$rab-1$	SSTot		

# 1.1 Two factors experiment

## ANOVA table - RCBD

Linear model:

$$Y_{ijk} = \mu + \rho_k + v_i + \pi_j + (v\pi)_{ij} + \varepsilon_{ijk}$$

Source	df	SS	MS	F
Block	$r-1$	SSR	$MSR = SSR/(r-1)$	$F_R = MSR/MSE$
Factor A	$a-1$	SSA	$MSA = SSA/(a-1)$	$F_A = MSA/MSE$
Factor B	$b-1$	SSB	$MSB = SSB/(b-1)$	$F_B = MSB/MSE$
A x B (interaction)	$(a-1)(b-1)$	SSAB	$MSAB = SSAB/(a-1)(b-1)$	$F_{AB} = MSAB/MSE$
Error	$(r-1)(ab-1)$	SSE	$MSE = SSE/(r-1)(ab-1)$	
Total	$rab-1$	SSTot		

# 1.1 Two factors experiment

## Definition formulae

$$SSTot = \sum_i \sum_j \sum_k \left( Y_{ijk} - \bar{\bar{Y}} \right)^2$$

$$SSR = ab \sum_k \left( \bar{Y}_{..k} - \bar{\bar{Y}} \right)^2$$

$$SSA = rb \sum_i \left( \bar{Y}_{i..} - \bar{\bar{Y}} \right)^2$$

$$SSB = ra \sum_j \left( \bar{Y}_{.j.} - \bar{\bar{Y}} \right)^2$$

$$SSAB = r \sum_i \sum_j \left( \bar{Y}_{ij.} - \bar{Y}_{i..} - \bar{Y}_{.j.} + \bar{\bar{Y}} \right)^2$$

$$SSE = SSTot - SSR - SSA - SSB - SSAB$$

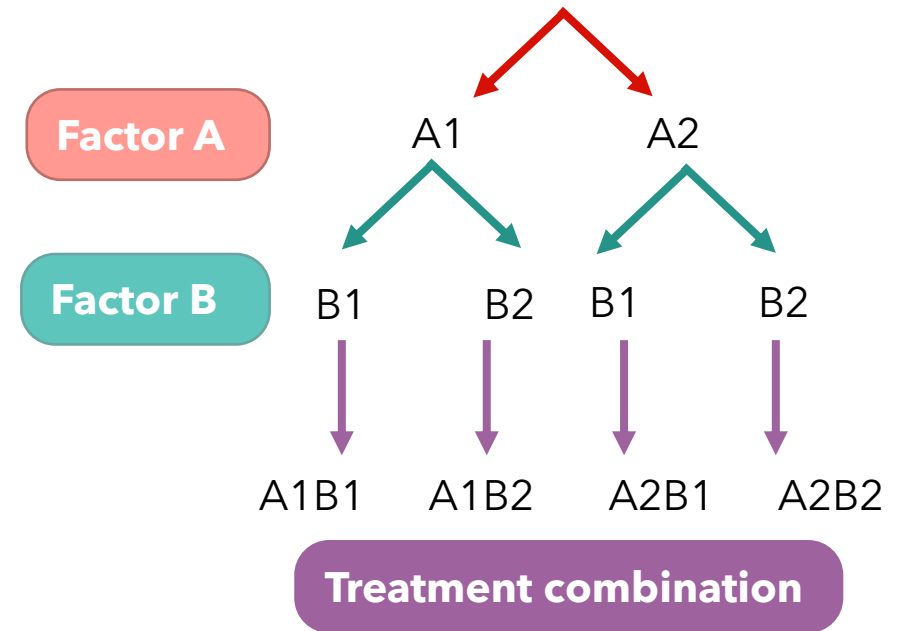
$$SS_{\text{treatment}} = SSA + SSB + SSAB$$

# 1.1 Two factors experiment

- **If there are interactions**, we should be able to measure and test them.
  - We **cannot** do this if we vary only one factor at a time
- We can combine two or more factors at two or more levels of each factor
  - Each level of every factor occurs together with each level of every other factor
  - Total number of treatments = the product of the levels of each factor

This has to do with the selection of treatments:

- Can be used in any design - CRD, RCBD, Latin Square - etc.
- “Designs” generally refer to the layout of replications or blocks in an experiment
- A “factorial” refers to the treatment combinations





# 1.2 Interaction

## Interaction between two factors:

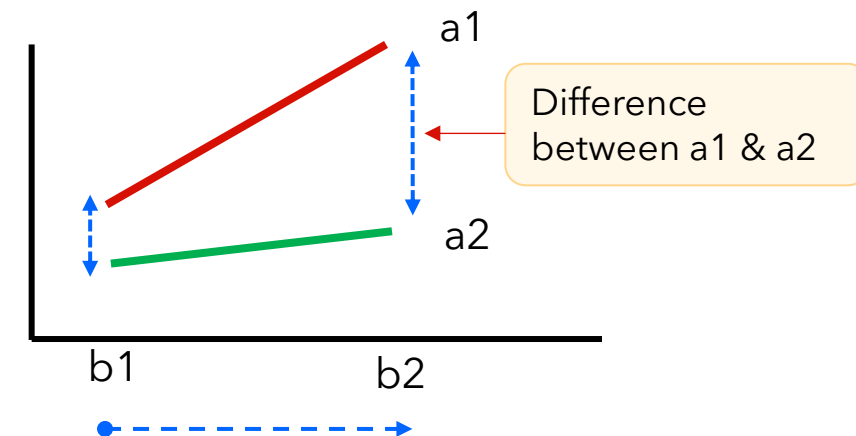
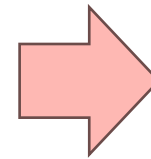
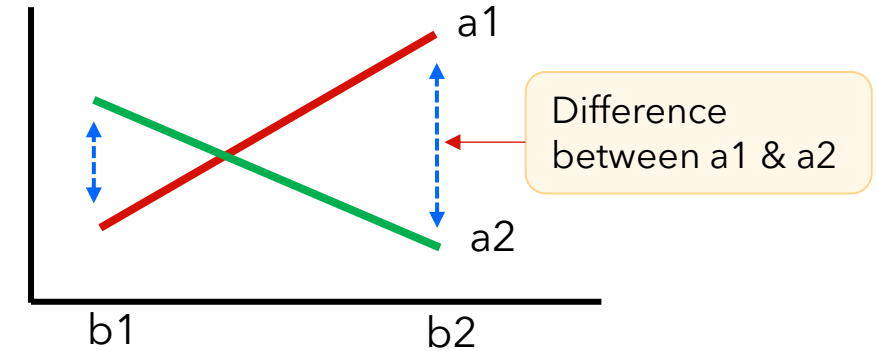
The effect of one factor changes as the level of the other factor changes

### Example:

As we move from b1 to b2, there is a change in the differences between a1 and a2 (blue arrow)

The interaction can be measured from a two factor experiment with at least 2 levels for each factor. Example:

- Factor A (a1, a2)
- Factor B (b1, b2)



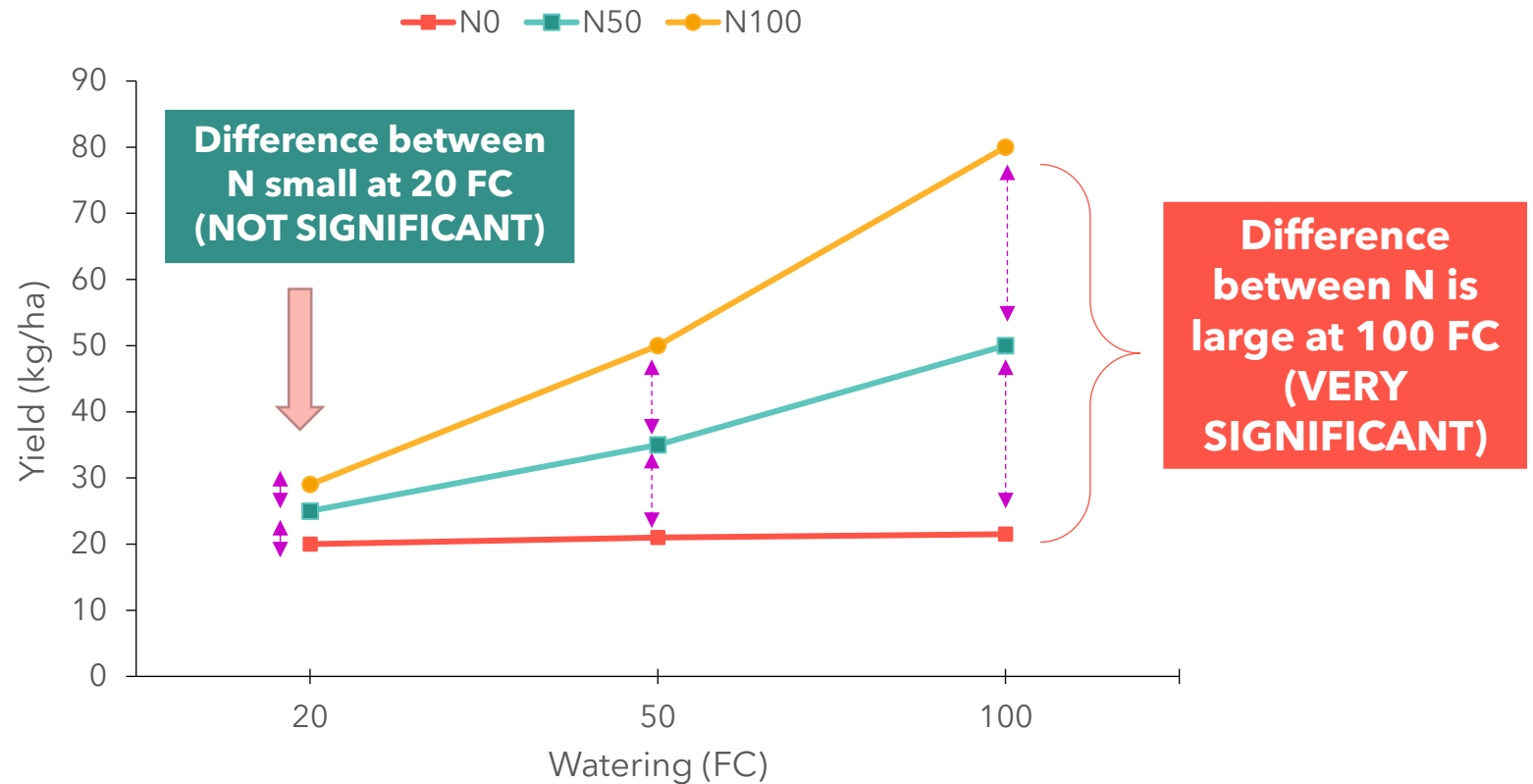
# 1.2 Interaction



## INTERACTION

- The effect of one factor **DOES** affect the other factor

Yield as affected by water and nitrogen



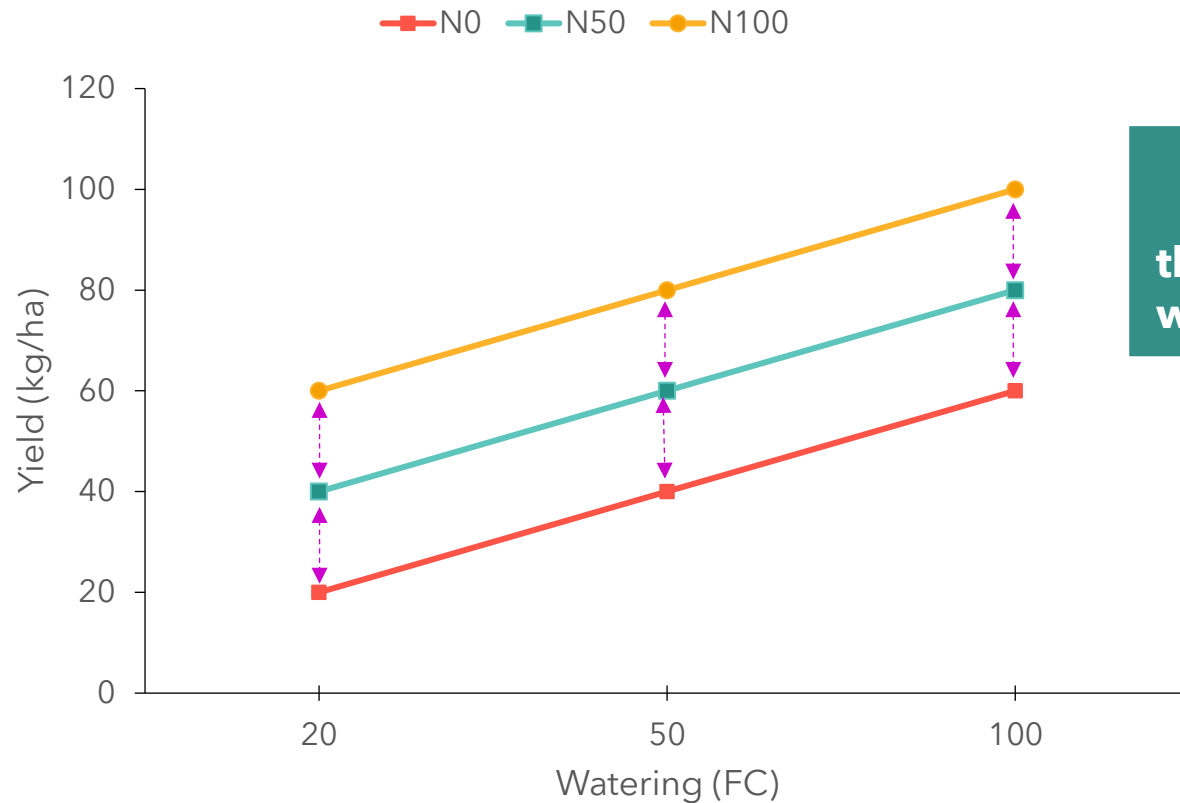
# 1.2 Interaction

## NO INTERACTION

- the effect of one factor DOES NOT affect of the other factor



Yield as affected by water and nitrogen

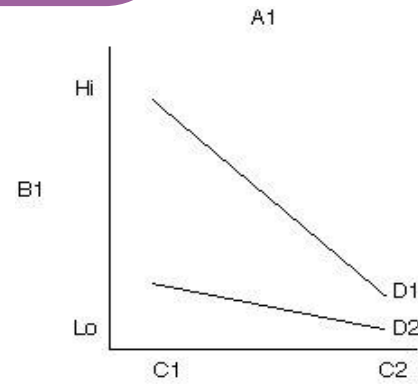


Differences between N is the same for all watering levels

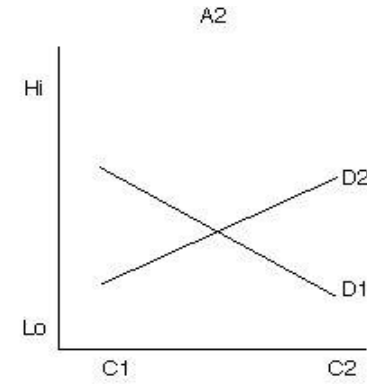
# 1.2 Interaction

## Types of interaction

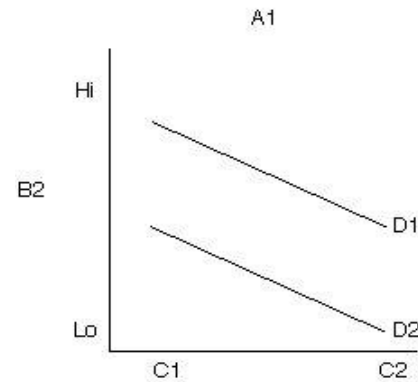
Non-crossover 1



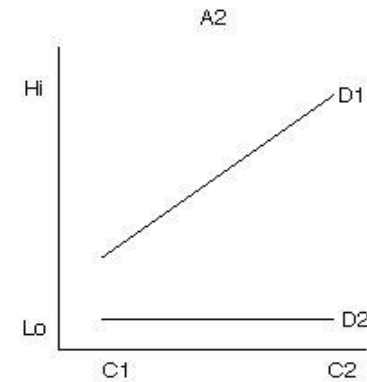
Crossover



No interaction



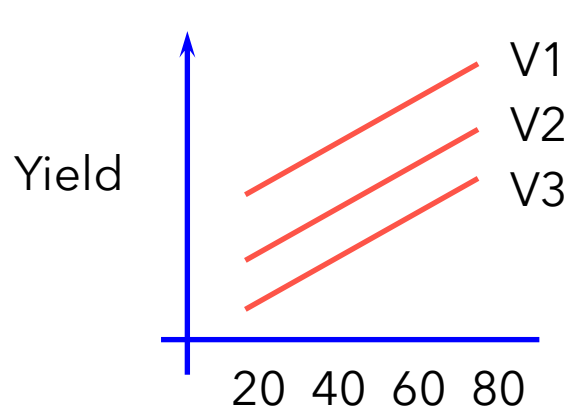
Non-crossover 2



# 1.2 Interaction

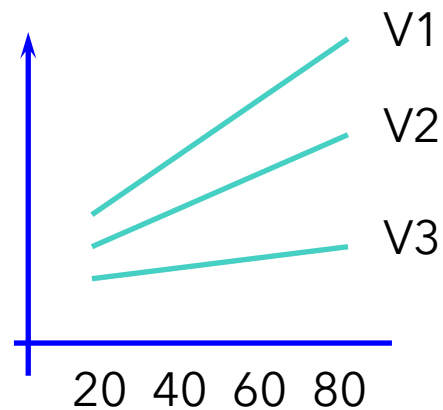
## Interaction graphs

Consider 3 varieties at four rates of nitrogen



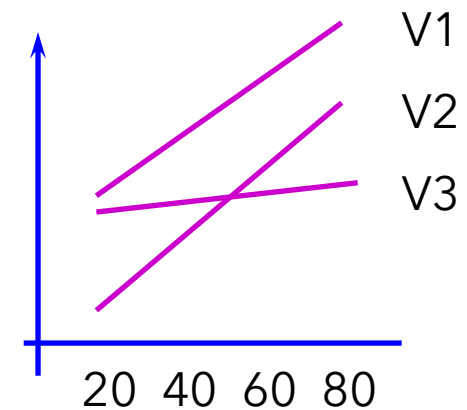
**No  
interaction**

Relative yield of varieties is the **same** at all fertilizer levels



**Non-crossover  
Interactions**

**Magnitude** of differences among varieties depends on fertilizer level



**Crossover  
Interactions**

**Ranks** of varieties depend on fertilizer level

# 1.2 Interaction

## Numerical examples of interaction

### Effect of two levels of phosphorus and potassium on crop yield

- ✓ **Main effects** are determined from the marginal means
- ✓ **Simple effects** refer to differences among treatment means at a single level of another factor

#### No interaction

	P <sub>0</sub>	P <sub>1</sub>	Mean
K <sub>0</sub>	10	18	14
K <sub>1</sub>	14	22	18
Mean	12	20	

$$(22-14)-(18-10) = 0$$

#### Positive interaction

	P <sub>0</sub>	P <sub>1</sub>	Mean
K <sub>0</sub>	10	18	14
K <sub>1</sub>	12	26	19
Mean	11	22	

$$(26-12)-(18-10) = +6$$

#### Negative interaction

	P <sub>0</sub>	P <sub>1</sub>	Mean
K <sub>0</sub>	10	18	14
K <sub>1</sub>	16	14	15
Mean	13	16	

$$(14-16)-(18-10) = -10$$

# 1.2 Interaction

## Interpretation

- If the **AB interaction IS significant**:
  - the main effects may have no meaning whether or not they test significant
  - Data summary:
    - Table: A two-way table of means for the various AB combinations
    - Figure: A bar chart of means for the various AB combination
- If the **AB interaction IS NOT significant**:
  - test the independent factors for significance
  - Data summary:
    - Table: A one-way table of means for the significant main effects
    - Figure: A bar chart of means for main effects only

# 1.2 Interaction

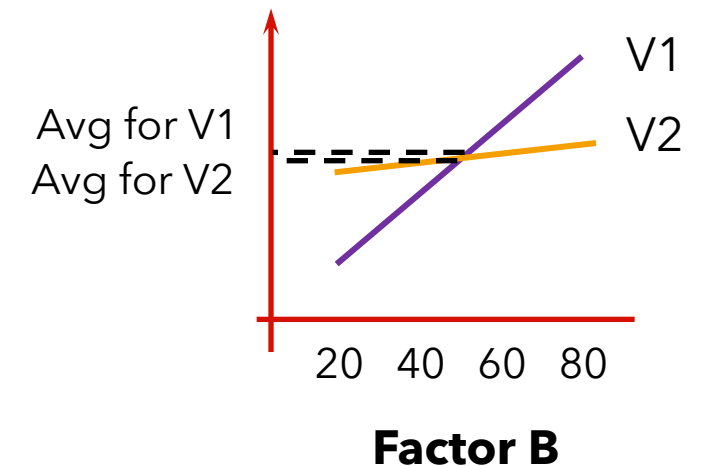
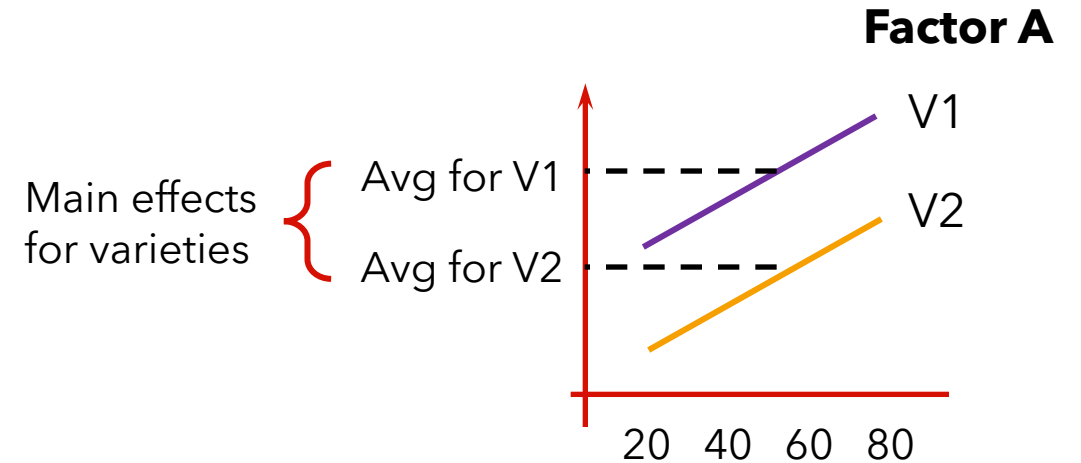
## Interaction and statistical test

### No Interaction

- Tests for **main effects are meaningful** because differences are constant across all levels of factor B

### Interaction

- Tests for **main effects may be misleading**
- In this case the test would show no differences between varieties, when in fact their response to factor B is very different





# 1.3 Result presentation

## Non-significant interaction

- Look at the **MAIN EFFECTS ONLY**

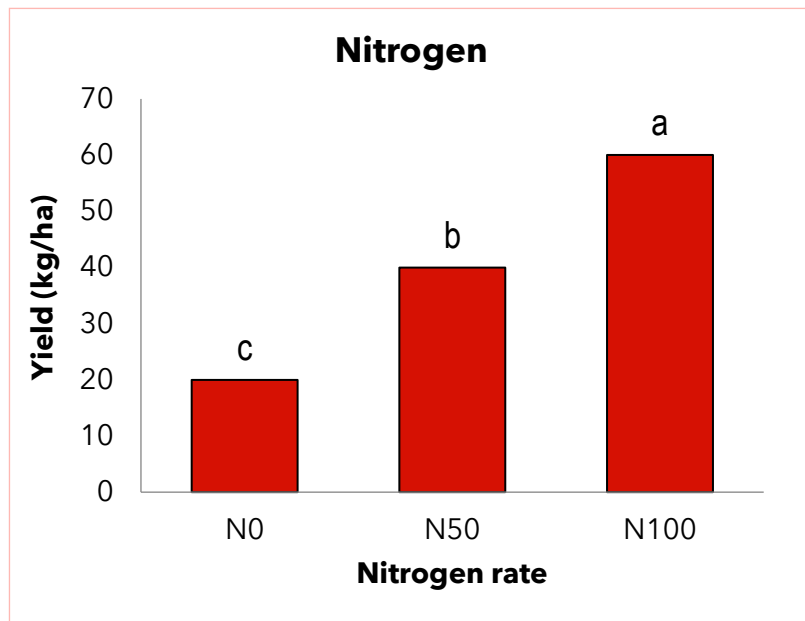


Figure 1.1. The effect of nitrogen rate on yield.

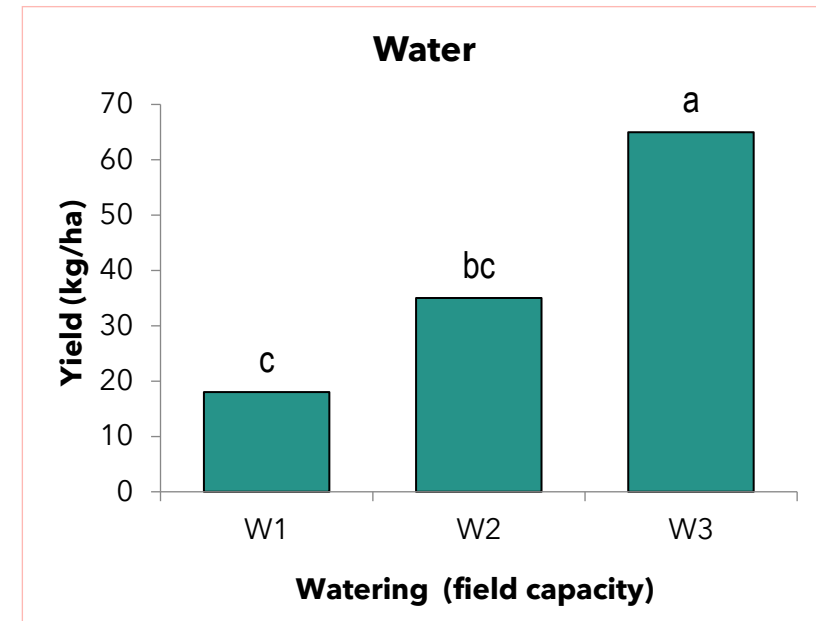
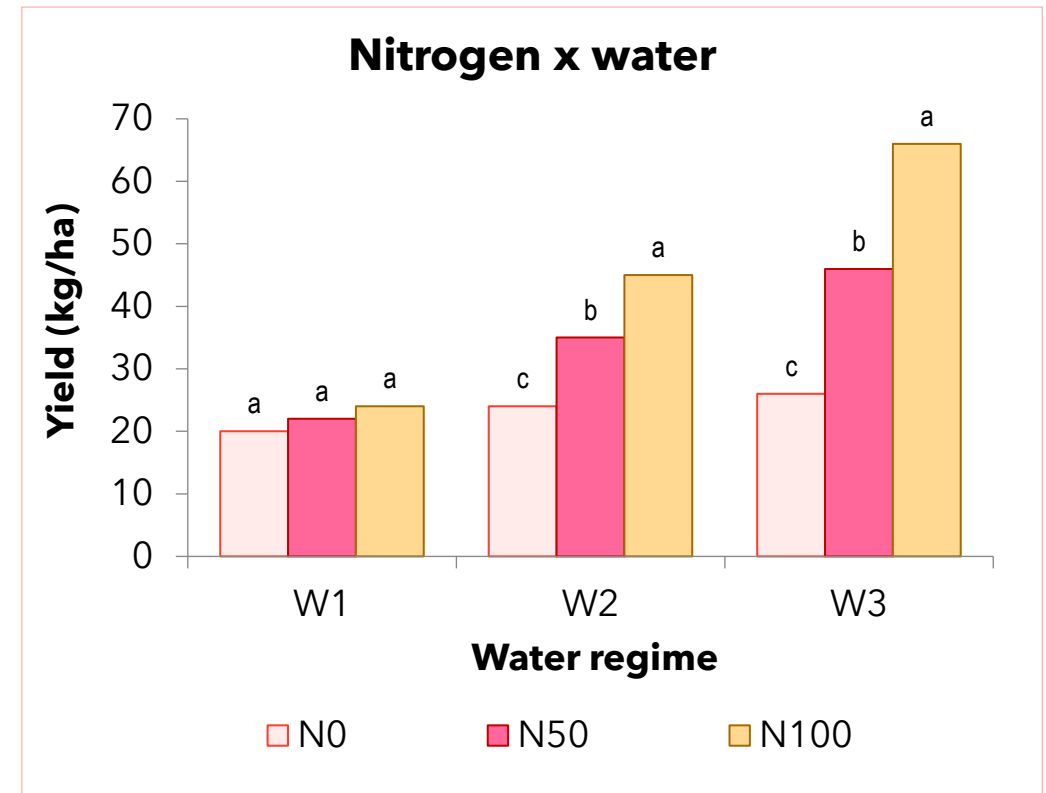


Figure 1.2. The effect of watering on yield.

# 1.3 Result presentation

## Significant interaction

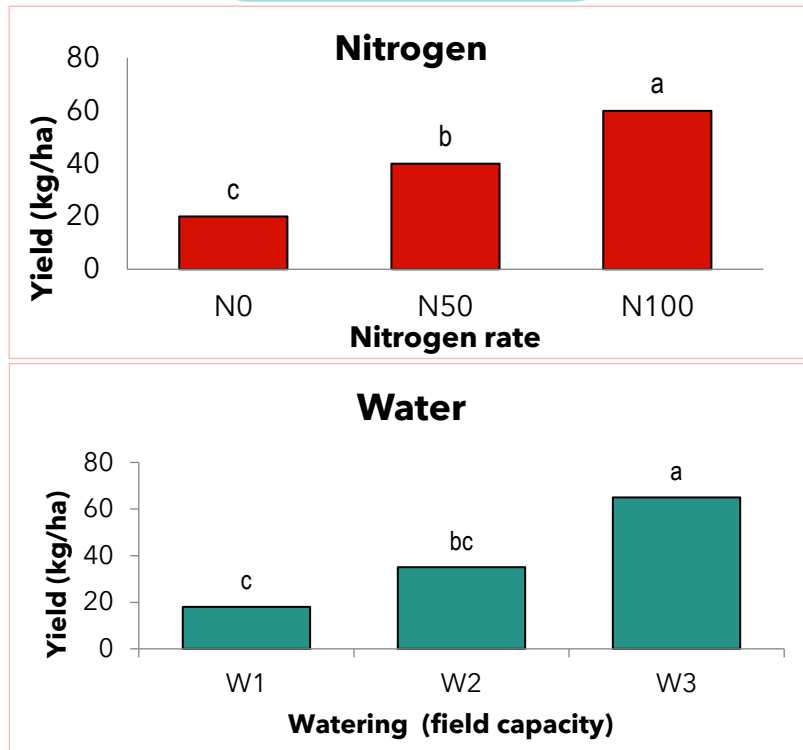
- Significant interaction means:
  - The effect of one factor influences the effect of the other factor.
- Data must be SORTED by one factor and comparisons must be made within a SORTED factor



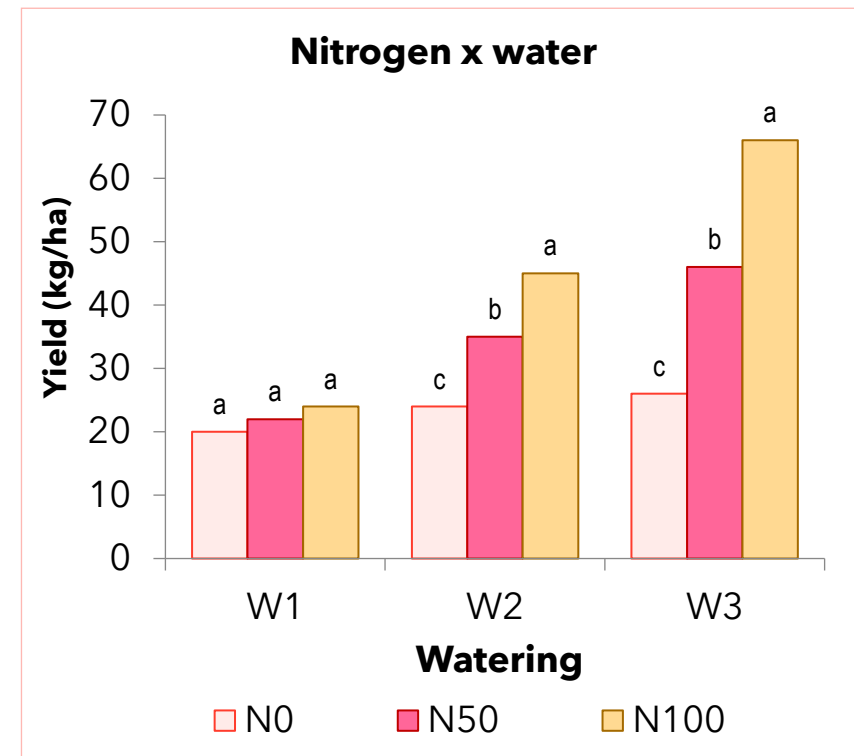
# 1.3 Result presentation

## Main vs. interaction effects

### Main effects



### Interaction effects



# 1.3 Result presentation

## Example

### Yield and quality of two kenaf varieties as affected by harvesting age.

**Masnira.M.Y<sup>1</sup>., Halim. R. A.<sup>2\*</sup>, Rafii M.Y.<sup>2</sup>, Mohd Jani, S.<sup>1</sup> and Martini, M. Y.<sup>2</sup>**

<sup>1</sup>Rice and Industrial Research Crop Centre, MARDI Headquarters, Serdang,  
P.O Box 12301, 50774 Kuala Lumpur Malaysia.

<sup>2</sup>Faculty of Agriculture, Universiti Putra Malaysia, 43400 Serdang Selangor Malaysia

\*Corresponding author: [ridzwan@upm.edu.my](mailto:ridzwan@upm.edu.my)

#### Factors:

1. Varieties: MHC 123 and V 36
2. Harvesting age: 8, 12, 16 and 20 weeks after planting (WAP)

# 1.3 Result presentation | Non-significant interaction

Table 1: Dry matter yield based on harvest age and variety

Treatment	Dry matter (kg plot <sup>-1</sup> )	Yield (t ha <sup>-1</sup> )
<b>Harvest age (H)</b>		
8 WAP	1.71b	8.5b
12 WAP	2.24a	11.2a
16 WAP	2.18a	10.9a
20 WAP	2.03a	10.2a
<b>Variety (V)</b>		
MHC 123	2.34a	11.7a
V 36	1.74b	8.7b
<b>Significance level</b>		
Harvest age	**	**
Variety	**	**
H x V	ns	ns
Mean	2.43	12.15
CV	13.66	13.66

**Main effects**

**ANOVA  
(simplified)**

**Note:**

The results can be presented using a table or a figure (graph).

Means with the same letter were not significantly different among harvest age and variety ( $p > 0.05$ ) using LSD

\*\*  $p < 0.01$ , ns: not significant.

# 1.3 Result presentation | Significant interaction (Crossover)

## Main effects and ANOVA

Table 2: Mean crude protein content

Treatment	CP (%)
<b>Harvest age (H)</b>	
8 WAP	20.3a
12 WAP	14.3b
<b>Variety (V)</b>	
MHC 123	18.1a
V 36	16.5a
<b>Significance level</b>	
Harvest age	**
Variety	ns
H x V	**
Mean	17.29
CV	9.62

Means with the same letter were not significantly different among harvest age and variety  $p > 0.05$  using LSD.

## Interaction effect: Graph

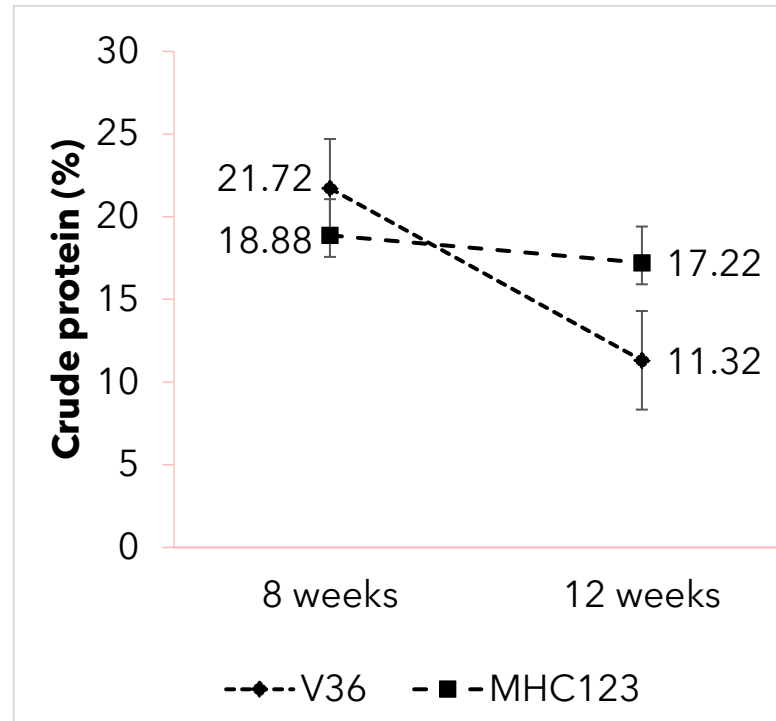


Figure 1. The effect of variety and harvest age on crude protein %.

## Interpretation:

When the interaction is significant, the interpretation should focus on the interaction effect as shown in the graph.

# 1.3 Result presentation | Significant interaction (non-crossover)

## Main effects and ANOVA

Table 3: Mean acid detergent fiber content

Treatment	ADF (%)
<b>Harvest age (H)</b>	
8 WAP	35.6b
12 WAP	46.2a
<b>Variety (V)</b>	
MHC 123	34.3b
V 36	47.6a
<b>Significance level</b>	
Harvest age	**
Variety	**
H x V	**
Mean	40.93
CV	7.61

Means followed by the different letters were significantly different among harvest age and variety using the LSD ( $P < 0.05$ )

\*\*  $P < 0.01$

## Interaction effect: Graph

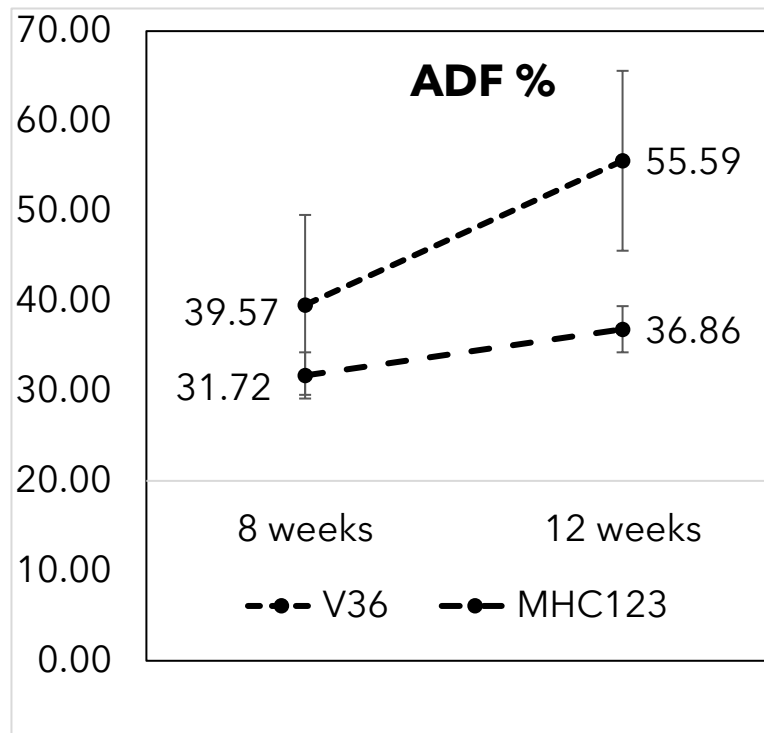


Figure 2. The effect of variety and harvest age on ADF %

## Interpretation:

When the interaction is significant, the interpretation should focus on the interaction effect as shown in the graph.

# 1.4 Three-factor experiment

- Three factors – three IV
  - Effect of watering: 100% FC, 50% FC, 20% FC
  - Effect of Nitrogen: 0 kg/ha, 50 kg/ha, 100 kg/ha
  - Effect of variety: A, B, C.



# 1.4 Three factor experiment

## ANOVA of 3 factor experiment - CRD

Source	df	SS	MS	F
Water (W)	2			
Nitrogen (N)	2			
Variety (V)	2			
W x N	4			
W x V	4			
N x V	4			
W x N x V	8			
Error	9			
Total	35			

3 main effects  
4 interaction  
terms

# 1.5 Two-factorial example

- To study the effect of row spacing and phosphate on the yield of bush beans
  - 3 spacings: 45 cm, 90 cm, 135 cm
  - 2 phosphate levels: 0 and 25 kg/ha

<b>S2P1</b> <b>60</b>	<b>S1P2</b> <b>45</b>	<b>S1P1</b> <b>55</b>
<b>S1P1</b> <b>65</b>	<b>S3P1</b> <b>55</b>	<b>S3P1</b> <b>51</b>
<b>S3P2</b> <b>66</b>	<b>S3P2</b> <b>57</b>	<b>S1P2</b> <b>43</b>
<b>S3P1</b> <b>59</b>	<b>S1P1</b> <b>58</b>	<b>S2P1</b> <b>54</b>
<b>S1P2</b> <b>56</b>	<b>S2P2</b> <b>50</b>	<b>S2P2</b> <b>45</b>
<b>S2P2</b> <b>62</b>	<b>S2P1</b> <b>59</b>	<b>S3P2</b> <b>50</b>

# 1.5 Two-factorial example

## Tables of Means

Treatment Means

Phosphate	Spacing			Mean
	S1	S2	S3	
P1	59.3	57.7	55.0	57.3
P2	48.0	52.3	57.7	52.7
Mean	53.7	55.0	56.3	55.0

Block Means

Block	I	II	III	Mean
Mean	61.3	54.0	49.7	55.0

# 1.5 Two-factorial example

## ANOVA table

Source	df	SS	MS	F
Block	2	417.33	208.67	31.00**
Spacing	2	21.33	10.67	1.58
Phosphate	1	98.00	98.00	14.56**
S X P	2	148.00	74.00	11.00**
Error	10	67.33	6.73	
Total	17	752.00		

\*\* Significant at the 1% level.  
CV = 4.7%  
StdErr Spacing Mean = 1.059  
StdErr Phosphate Mean = 0.865  
StdErr Treatment (SxP) Mean = 1.498

# 1.5 Two-factorial example

## Report of statistical analysis | Table ( A two-way table)

**Table 1. The effect of phosphate and spacing on yield of bush bean (kg/ha)**

Phosphate (kg/ha)	Spacing (cm)		
	45	90	135
0	59.33a	57.67ab	55.00b
25	48.00c	52.33b	57.67a

Within phosphate (row), the means with different letters are significantly different at  $P < 0.05$  using LSD

### Interpretation:

- Yield response depends on whether or not phosphate was supplied
- If no phosphate - yield decreases as spacing increases
- If phosphate is added - yield increases as spacing increases
- Blocking was effective

The explanation is based on "phosphate" level (separated by phosphate)

# 1.5 Two-factorial example

## Report of statistical analysis | Figure (Line graph)

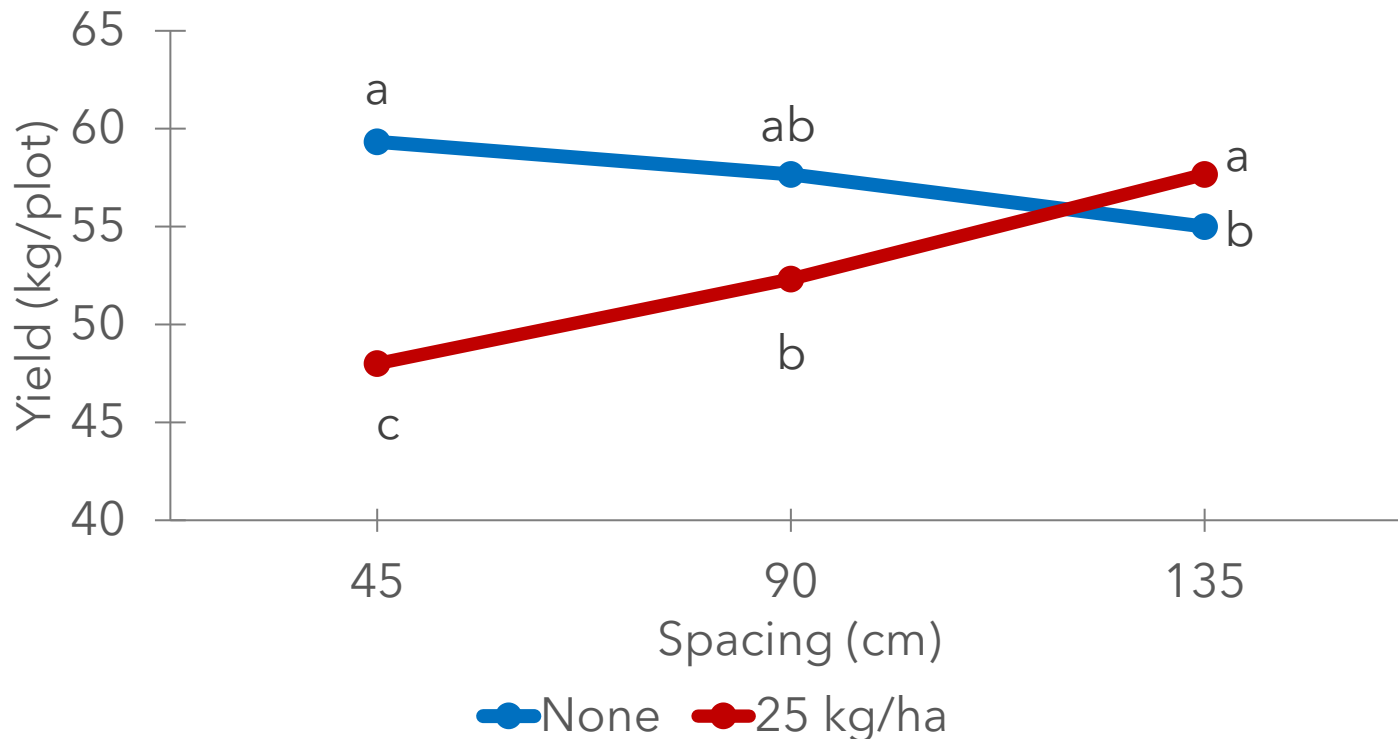


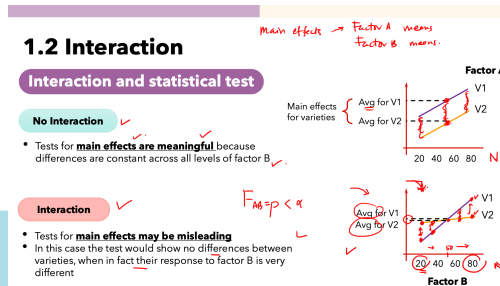


Figure 1. The interaction between spacing and rate of P on bush bean yield (kg/plot). Within phosphate, means with different letters are significantly different at  $P < 0.05$  using LSD.

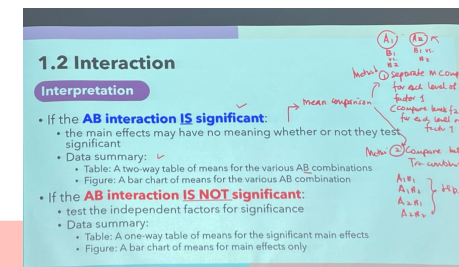
# 1.6 Review question



1. Two-factorial experiment is an experimental design. True or false? Explain your answer. 
2. Explain the requirement in an experiment to test the interaction effects between factors. 
3. List the interaction types and explain each of the interaction. see the figure below
4. How do you proceed with mean comparison if the interaction is significant in ANOVA ( $p\text{-value} < \alpha$  (e.g., 0.05))? see the figure below
5. Explain why the approach for mean comparison is different when the interaction term in ANOVA is significant compared to when the interaction is not significant. Q5



Q4



Q3

## 1.2 Interaction

### Interaction graphs

