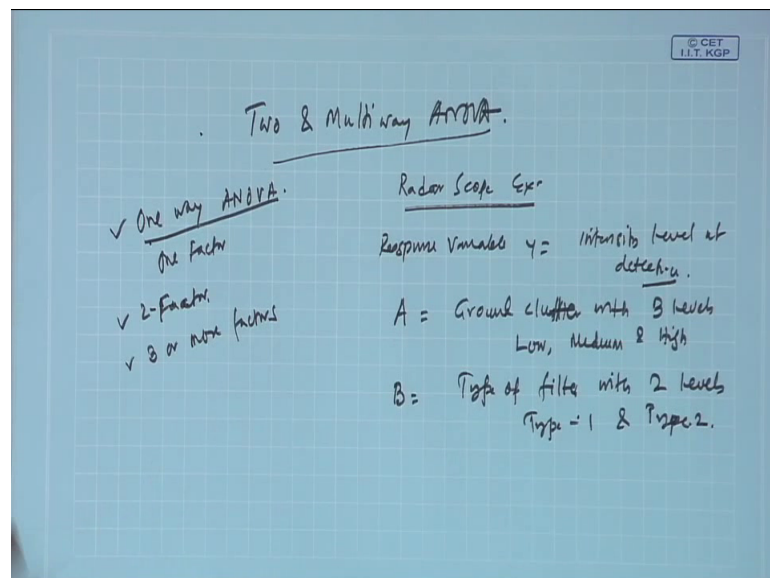


Design and Analysis of Experiments
Prof. Jhareswar Maiti
Department of Industrial and Systems Engineering
Indian Institute of Technology, Kharagpur

Lecture - 17
Two -Way ANOVA

Welcome. Today we will discuss Two and Multi-way ANOVA

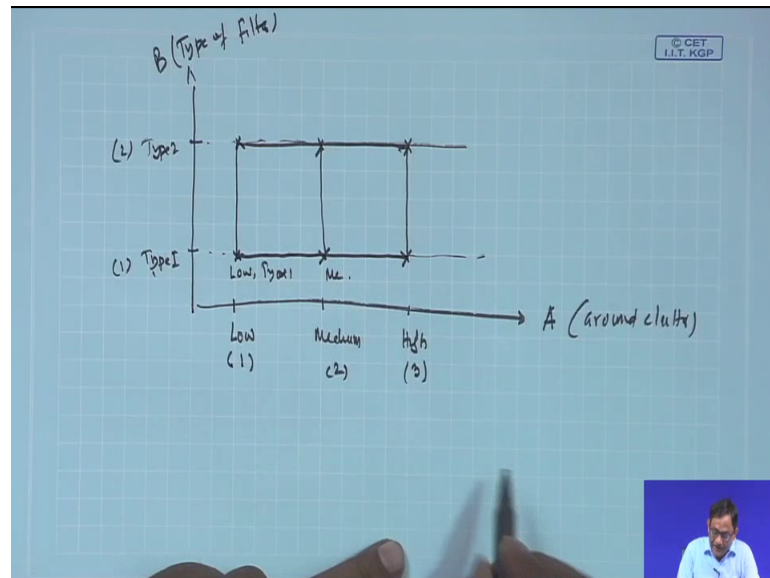
(Refer Slide Time: 00:23)



I hope that you have understood one way ANOVA. So, I also equally hope that you will be able to work with two and multi-way ANOVA. This way, one way means one factor two-way means 2 factor, multi-way means 3 or more factor. So, what is that two-way? If you consider radarscope example; so the response variable y that is the intensity level at detection, detection of targets. And if you recall we say that ground one factor A is ground clutter with three levels low, medium and high. And type of filter B; type of filter with two levels filter type 1 and filter type 2.

So, what I can say that if you do a random experiment actually you are having a situation like this.

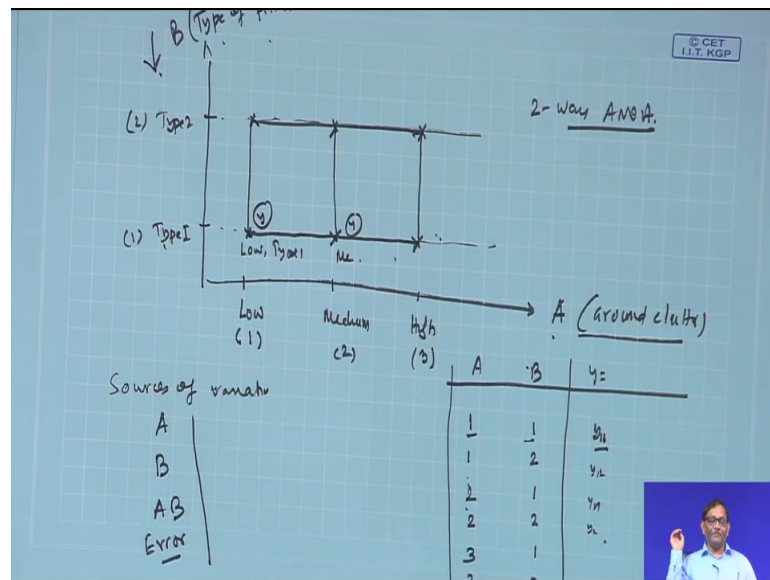
(Refer Slide Time: 02:22)



This side A ground clutter, this said B type of filter. So, you have three levels of ground clutter level low I can say 1; medium maybe 2; and high 3. And you have two levels of; so type 1 filter and another one is type 2 filter.

So, this is level 1 this is level 2. So, you have how many experimental settings. You have six number of experimental settings or treatment combinations. This is one, this one is another one, this one another one, another one, another one, another one. So, here low; low and type 1, here medium will type 1 medium the type 1 like this. What does it mean?

(Refer Slide Time: 03:58)



You are conducting an experiment in such a manner that suppose A factor level 1 then B factor level 1, then you are getting y value here the intensity level at, so intensity level at. So, this is maybe y_{11} if I go for.

So again this is 1, this may be 2, and this is 2 1 2 2 3 1 3 2; so you will be getting several. So, what happened here this means $y_{A1, B1, 11, 12, 21, 22}$ so like this. But when we I will show you the data I will we will not represent in this manner there will be replication one more things will be there one more direction will come.

So anyhow, then in this case this is a 2 factor experiment. Now if you want to develop a ANOVA model for this it will be a 2 factor ANOVA model. And when you analyze the data, you have to analyze considering that 2 factor and their interactions are there. So, what does it; does it mean that why here the response values y values here also y values will be there all those response values not only depends on the ground clutter it may depend on the type of filter. In addition it may depend on the interaction between ground clutter and type of filter that interaction between A and B.

So, effectively sources of variation will be not only A it will be B and their interaction AB and plus there will be a random source that is error. So, accordingly the observations will be partitioned reference to A factor, with reference to B factor, with reference to their interactions and obviously, error will be there. So, this kind of data will be analyzed using two-way ANOVA.

So, we will come to some other example, but for the time being you please understand two-way mean there are 2 factors and each of the factors has at least 2 or more levels. So that means, there will be A level for A factor B level for B factor and there will be a into B AB treatment combinations, and in each combination there will be n number of experiment; obviously, in the random order it will be done. So, in total you will be having AB into n number of experimental runs or the object data will be AB n.

So, let us see; what will be the model for two-way ANOVA. You see the basic 2 factor model.

(Refer Slide Time: 07:46)

Two-Factor ANOVA

- The basic two-factor ANOVA model is

$$y_{ijk} = \mu + \tau_i + \beta_j + (\tau\beta)_{ij} + \varepsilon_{ijk}, \begin{cases} i = 1, 2, \dots, a \\ j = 1, 2, \dots, b \\ k = 1, 2, \dots, n \end{cases}$$

μ = an overall mean, τ_i = i -th treatment effect, β_j = j -th level effect of column factor
 $(\tau\beta)_{ij}$ is the effect of interaction between τ_i and β_j , ε_{ij} = experimental error, $NID(0, \sigma^2)$

IIT KHARAGPUR | NPTEL ONLINE CERTIFICATION COURSES

So, here again what we will do; we will see the table first.

(Refer Slide Time: 07:57)

Data representation for two-factor experiment

		Factor B			
		1	2	...	b
Factor A	1	$y_{111}, y_{112}, \dots, y_{11n}$	$y_{121}, y_{122}, \dots, y_{12n}$		$y_{1b1}, y_{1b2}, \dots, y_{1bn}$
	2	$y_{211}, y_{212}, \dots, y_{21n}$	$y_{221}, y_{222}, \dots, y_{22n}$		$y_{2b1}, y_{2b2}, \dots, y_{2bn}$
	...				
	a	$y_{a11}, y_{a12}, \dots, y_{a1n}$	$y_{a21}, y_{a22}, \dots, y_{a2n}$		$y_{ab1}, y_{ab2}, \dots, y_{abn}$

$$y_{i.} = \sum_{j=1}^b \sum_{k=1}^n y_{ijk} \quad \bar{y}_{i.} = \frac{y_{i.}}{bn} \quad i = 1, 2, \dots, a$$

$$y_{.j} = \sum_{i=1}^a \sum_{k=1}^n y_{ijk} \quad \bar{y}_{.j} = \frac{y_{.j}}{an} \quad j = 1, 2, \dots, b$$

$$y_{i.} = \sum_{k=1}^n y_{ijk} \quad \bar{y}_{i.} = \frac{y_{i.}}{n} \quad i = 1, 2, \dots, a$$

$$y_{.j} = \sum_{i=1}^a \sum_{k=1}^n y_{ijk} \quad \bar{y}_{.j} = \frac{y_{.j}}{an}$$

- In general, there is two factors, row-wise there will be **a levels** of the factor, or **a row treatments**, and **b column treatments**, and **n replicates** of the experiment, run in **random order** - a completely randomized design (**CRD**)
- $N = abn$ total runs
- We consider the **fixed effects** case. The **random effects** case will be discussed later
- Objective is to test hypotheses concerning the **row, and column** treatment effects and interactions

IIT KHARAGPUR

NPTEL ONLINE
CERTIFICATION COURSES

Now see this table. Factor A, factor B: A has A number of levels, B has been number of levels, and the each cell here in this table is A treatment combination; so A at level 1 B at level 1 this treatment combination.

Similarly A at 1 B 2, so there will be AB treatment combination unique independent treatment combinations. And another one interesting one is that in each of the combination you find out that there are n number of experimental runs; n number of data available. So, as a result we index this data like this.

(Refer Slide Time: 08:50)

$i = 1, 2, \dots, a$
Number levels for factor A

$j = 1, 2, \dots, b$
Number levels for factor B

Replicates
 $k = 1, 2, \dots, n$

	1	2	...	b
1	$y_{111}, y_{112}, \dots, y_{11n}$			
2				
...				
a				

$$y_{ijk} = \mu + \alpha_i + \beta_j + (\alpha\beta)_{ij} + \epsilon_{ijk}$$

μ → Grand mean α_i → Row effect β_j → Column treatment effect $(\alpha\beta)_{ij}$ → Interaction effect ϵ_{ijk} → Error

$$= \mu + \mu_{ij} + \epsilon_{ijk}$$

Now see this table. Factor A, factor B: A has A number of levels, B has been number of levels, and the each cell here in this table is A treatment combination; so A at level 1 B at level 1 this treatment combination.

Y_{ijk} : so i stands for that 1, 2, a; that means, this is the treatment level for factor A, j stands for 1, 2, b treatment levels for factor B, k is replications. So, it will be 1, 2, n ; n number of applications will be there.

So, this is the general data and our table is like this: 1, 2 like dot dot dot a and here 1, 2 dot dot dot b. And here y_{11} as I shown you earlier also that y_{11} , but and that 11 level there will be n number of experimental run; so y_{111} , y_{112} , like this y_{11n} . So, in general y_{ij1} , y_{ij2} like this, y_{ijk} , y_{ijn} .

So, there will be general observation y_{ijk} . You can partition this y_{ijk} equal to grand mean μ plus that is A factor affect B factor affect also. So, τ_i is the factor A i treatment A i effect, β_j for B treatment effect plus $\tau\beta_{ij}$ that is the interaction effect plus ϵ_{ijk} . So, this is my ANOVA model.

So, you can understand now i from 1, 2, a; j from 1, 2, B; k from 1, 2, n and this is your model. So, I can say other way also this is row effect from considering this kind of table this is column effect. So, other way row treatment effect, column treatment effect that is interaction effect this is error and this is grand mean, ok. So, now you may be interested to add this together, so you can write this one equal to $\mu + \mu_{ij} + \epsilon_{ijk}$; $\mu + \mu_{ij}$. So, this μ_{ij} is row column and treatment this all those things. So μ_{ij} here, μ_{ij} every cell this image every cell there will be a mean value. So, y_{ijk} , so the general value will be grand mean plus mean of this plus error of this; so where μ_{ij} equal to $\tau_i + \beta_j + \tau\beta_{ij}$; i a τ β j .

So, let us see in the table again. Now, there will be if you take sum total across rows that will be $y_{i \dots}$ this one. If you take sum total across columns that will be $y_{\dots j \dots}$. If you take sum total of observation in a cell that is $y_{ij \dots}$; if you tell all the grand total considering AB and this is the case. And you can calculate average also. If you compute average or across rows then the row basically you see there are B number of columns and under every cell there is n number of data points so that why it will be divided by B n . For column average, the column total will be divided by n . For cell average, the cell total will be divided by cell num number n . And for overall average, the overall total will be divided by total number of observations, ok.

In this case also there will be random fixed effect models and random effect models, we are considering fixed effect model. And please keep in mind that we are basically

interested to know whether the row treatment and column treatment effects present and their interaction if it is there or not; So that means, we want to know whether the factory affecting the response variable factor B affecting the response variable, and their interaction also playing a role in the observed response values. So, that will be tested.

So, as a result here there will be three kinds of three different null hypothesis and we will be testing it.

(Refer Slide Time: 15:32)


Models for the Data

There are several ways to write a model for the data:

Means model

$$y_{ijk} = \mu_{ij} + \varepsilon_{ijk} \quad \begin{cases} i = 1, 2, \dots, a \\ j = 1, 2, \dots, b \\ k = 1, 2, \dots, n \end{cases}$$

where

$$\mu_{ij} = \mu + \tau_i + \beta_j + (\tau\beta)_{ijk}$$


(Refer Slide Time: 15:38)

Decomposition of total sum of square


- Total variability** is measured by the total sum of squares:

$$SS_T = \sum_{i=1}^a \sum_{j=1}^b \sum_{k=1}^n (y_{ijk} - \bar{y}_{...})^2$$

- The basic **ANOVA partitioning** is:

$$\begin{aligned} \sum_{i=1}^a \sum_{j=1}^b \sum_{k=1}^n (y_{ijk} - \bar{y}_{...})^2 &= \sum_{i=1}^a \sum_{j=1}^b \sum_{k=1}^n [(\bar{y}_{i..} - \bar{y}_{...}) + (\bar{y}_{.j.} - \bar{y}_{...}) + (\bar{y}_{ij.} - \bar{y}_{i..} - \bar{y}_{.j.} + \bar{y}_{...}) + (y_{ijk} - \bar{y}_{ij.})]^2 \\ &= bn \sum_{i=1}^a (\bar{y}_{i..} - \bar{y}_{...})^2 + an \sum_{j=1}^b (\bar{y}_{.j.} - \bar{y}_{...})^2 + n \sum_{i=1}^a \sum_{j=1}^b (\bar{y}_{ij.} - \bar{y}_{i..} - \bar{y}_{.j.} + \bar{y}_{...})^2 + \sum_{i=1}^a \sum_{j=1}^b \sum_{k=1}^n (y_{ijk} - \bar{y}_{ij.})^2 \end{aligned}$$

Therefore,

$$SS_T = SS_A + SS_B + SS_{AB} + SS_E$$


Now that test will be done using analysis of variance. So, the similar type of calculation we will be doing. First partitioning the general observations into overall average, into row average, into column average, and into cell average plus errors and then making their sum square total as the treatment color row treatment sum square, column treatment sum square, then interaction sum square and error sum square.

So, this is what we will be doing. In one way ANOVA the sources of variability a variability are the factor itself, and the error in two-way ANOVA sources of variability are factor A, factor B, their interaction and error; four sources will be there.

Now see the partitioning. As I told you let me; what is the general y_{ijk} . That is the general observation.

(Refer Slide Time: 16:47)

$$y_{ijk} = \underbrace{\bar{y}_{...}}_{\text{Grand average}} + \underbrace{(\bar{y}_{i..} - \bar{y}_{...})}_{\text{Row average}} + \underbrace{(\bar{y}_{.j.} - \bar{y}_{...})}_{\text{Column average}} + \underbrace{(-\bar{y}_{i..} + \bar{y}_{.j.} + \bar{y}_{...} + \bar{y}_{ij.})}_{\text{Cell average}} + (y_{ijk} - \bar{y}_{ij.})$$

$$\sum \sum (y_{ijk} - \bar{y}_{...})^2 = \sum \sum [(\bar{y}_{i..} - \bar{y}_{...}) + (\bar{y}_{.j.} - \bar{y}_{...}) + (\bar{y}_{ij.} - \bar{y}_{i..} - \bar{y}_{.j.} + \bar{y}_{...}) + (y_{ijk} - \bar{y}_{ij.})]^2$$

So, this can be written like this $y_{...}$, this is the grand average. So, you can write now $y_{i..}$, what is this? This is row average. I can subtract this using this one grand average also. Plus I will go to $y_{.j.}$ minus $y_{...}$, so this is my column average. Otherwise, you can say row means factor A average factor B level average, so plus if you see that here what happened this is one time plus minus this but these two time you have already taken. So, if I write like this $y_{i..}$ plus $y_{.j.}$ then, I write this one minus this; minus this because this is got plus consider plus consider plus $y_{...}$ this. So this, this; and I add one more term here $y_{ij.}$ this is my cell average.

Then, what is left here? This, this one will be canceled out by this one sorry; this will be cancelled by this one, $y_{i \cdot \cdot \cdot}$ this will be canceled out by this one, y_{\cdot} this will be canceled out by this one, then this will be canceled out by this one, so remaining this, this also to be canceled out. So, if I write $y_{ijk} - y_{ij \cdot}$ then this will be canceled by this then $y_{ijk} = y_{ijk}$. So, essentially what you have done then? You have written a general observation in this way $y_{ijk} - y_{i \cdot \cdot \cdot}$, this is the deviation; individual observation deviation from the grand mean. I can write this equal to $y_{i \cdot \cdot \cdot} - y_{i \cdot \cdot \cdot}$ plus $y_{\cdot j \cdot}$ minus $y_{\cdot \cdot \cdot}$ plus $y_{ij \cdot \cdot}$ minus $y_{i \cdot \cdot \cdot}$ minus $y_{\cdot j \cdot}$ plus $y_{i \cdot \cdot \cdot}$, then plus I can write $y_{ijk} - y_{ij \cdot}$.

So, this is the general observed deviation for individual observation. This is the deviation of the row average from grand mean, grand average, this is the deviation from column average from the grand average, and this is the basically the deviation in the cell from cell average to the grand average and other things because you are combiningly; and this is what is the deviation from every observation to itself average. Now, this is the partition.

Now what you can do, you can square it then take sum triple sum will be there here also triple sum and you will ultimately get this kind of results. So, let us see the slide here. In the slide itself, that means the left hand side you see the left hand side that is the individual observation is subtracted by the grand average and when we take the square and this is known as sum square total. And this quantity when you base expand it and then take the appropriate sums you will find out the inter sum of the cross term products will be vanished.

So what will remain here would generally, it will remain only this B_n into these, A_n into these, n into this, and this. So, this if you carefully look the first one you see that it is basically talking about the difference row level row difference my average deep from the grand average, here column average from the grand average, here cell lever is from the rests, and here basically the individual observation from the cell average.

So, this term is known as the row effect; row sum square row or sum square A this is sum square B, this one is sum square AB, and this one is sum square S E.

(Refer Slide Time: 22:34)

$N = abn$
 $SS_T = SS_A + SS_B + SS_{AB} + SS_E$
 $Dof \frac{N-1}{abn-1} = (a-1) + (b-1) + (a-1)(b-1) + ab(n-1)$
 A: $\mu_1, \mu_2, \dots, \mu_a$ $H_0: \mu_i = 0$
 $H_1: \mu_i \neq 0$ —
 B: $H_0: \beta_j = 0$ $AB: \mu: (\mu\beta)_{ij} = 0$
 $H_1: \beta_j \neq 0$ — $H_1: (\mu\beta)_{ij} \neq 0$ —

So, the variation part which is basically sum square total is now partition into sum square factor A sum square B sum square AB and sum square E. Like in 1 way ANOVA you have seen SS T equal to SS A plus SS E, here two more terms coming because the B source of interaction and; your B source and interaction source are coming.

(Refer Slide Time: 23:18)

Two-way ANOVA Table

- Two-way ANOVA compares the mean differences between groups that have been split on two independent variables (called factors).
- The primary purpose of a two-way ANOVA is to understand if there is an interaction between the two independent variables on the dependent variable.

Two-factor ANOVA table

Source of Variation	Sum of Squares	Degrees of Freedom	Mean Square	F_0
A treatments	SS_A	$a - 1$	$MS_A = \frac{SS_A}{a - 1}$	$F_0 = \frac{MS_A}{MS_E}$
B treatments	SS_B	$b - 1$	$MS_B = \frac{SS_B}{b - 1}$	$F_0 = \frac{MS_B}{MS_E}$
Interaction	SS_{AB}	$(a - 1)(b - 1)$	$MS_{AB} = \frac{SS_{AB}}{(a - 1)(b - 1)}$	$F_0 = \frac{MS_{AB}}{MS_E}$
Error	SS_E	$ab(n - 1)$	$MS_E = \frac{SS_E}{ab(n - 1)}$	
Total	SS_T	$abn - 1$		

$SS_T = SS_A + SS_B + SS_{AB} + SS_E$

So, here what happened ultimately you will find out that the in earlier case like this that when we talk about the one way ANOVA we say that this if; what is N here total number of observation here will be ab into n. So, the degree of freedom here will be N minus 1,

and this will be a minus 1, this will be B minus 1 plus this will be a minus 1 into B minus 1, and this last one will be your ab into n minus 1. So, the degree of freedom is also partitioned and if you add all those things this is nothing but ab N minus 1 because n equal to ab n. So, other way this can be written like abn minus 1.

From here you can develop the ANOVA table, see the table here. So, if you see the table sources of variation A treatment B treatment interaction error SS A, SS B, SS AB, SS E, SS T; these are the degrees of freedom. Compute MS A, MS B, MS AB, MS E. And then what is this? The SS A by their respective degree of freedom; so, what do you want to test whether A treatment effect is there or not, you create a statistics value for A that is MS A by MS E. Similarly we want to test whether B treatment effects are there or not, so compute F 0 for B MS B by MS E. Similarly you want to know the interaction effects are there or not, so compute F 0 for MS AB by MS E.

So then, all those things will be tested using appropriate F statistics. So, what will be the degree of freedom for this first F 0? That F statistics there will be numerator a minus 1 and denominator ab into N minus 1. The second one: b minus 1 ab into N minus 1. Third one: a of minus 1 B minus 1 into ab into N minus 1, that way.

(Refer Slide Time: 25:26)

Hypothesis Testing

Hypothesis testing for the equality of row treatment effects	$H_0 : \tau_1 = \tau_2 = \dots = \tau_a$ $H_1 : \text{At least one } \tau_i \neq 0$
Hypothesis testing for the equality of column treatment effects	$H_0 : \beta_1 = \beta_2 = \dots = \beta_b$ $H_1 : \text{At least one } \beta_j \neq 0$
Hypothesis testing for the equality of row & column treatment interaction effects	$H_0 : (\tau\beta)_{ij} = 0$ $H_1 : \text{At least one } (\tau\beta)_{ij} \neq 0$

IIT KHARAGPUR | NPTEL ONLINE CERTIFICATION COURSES

What are the hypothesis? You see the hypothesis, hypothesis testing for equality of row treatment effects; that mean all treatment are equal here, I mean at least one treatment. So, you just write down this one that tau i equal to 0 and this is basically mu 1 equal to;

μ related to μ_1 to μ_a . This is similarly β_j equal to 0, at least one of the β_j not equal to 0 and this one is like this. So, I am writing this one carefully.

What happened here? When I am talking about A treatment, so I have μ_1 to μ_2 to μ_a levels or H_0 I am creating like this τ_i equal to 0; H_1 τ_i not equal to 0 or other way μ_1 equal to μ_2 like the other one. Similarly for beta B you are creating that β_j equal to 0 β_j not equal to 0. Similarly for AB what your H_0 , this is H_0 and H_1 and this H_0 is τ_{β_j} equal to 0 and H_1 τ_{β_j} not equal 0. Obviously for these, these, and this case at least one of the treatment or interaction is not 0. That is what is the test.

So, now you see the computation part.

(Refer Slide Time: 27:14)

Two-Factor ANOVA-Formulae for sum square calculations

$$SS_T = \sum_{i=1}^a \sum_{j=1}^b \sum_{k=1}^n y_{ijk}^2 - \frac{y_{..}^2}{abn}$$

$$SS_A = \frac{1}{bn} \sum_{i=1}^a y_{i..}^2 - \frac{y_{..}^2}{abn}$$

$$SS_B = \frac{1}{an} \sum_{j=1}^b y_{.j}^2 - \frac{y_{..}^2}{abn}$$

$$SS_{\text{Subtotals}} = \frac{1}{n} \sum_{i=1}^a \sum_{j=1}^b y_{ij}^2 - \frac{y_{..}^2}{abn}$$

$$SS_{AB} = SS_{\text{Subtotals}} - SS_A - SS_B$$

$$SS_E = SS_T - SS_{AB} - SS_A - SS_B$$

You calculate how do compute manually or at a eg computation for equal samples are SS T here sum total of all the observations squares minus grand total square by abn. For SS A sum total of row total squares divided by bn, and in my end of the correction factor we subtracted, similarly SS B. Now another concept here is SS B SS subtotals; SS subtotal means the cell k for every cell we are interested in.

In the cell means the every observation is y_{ij} and k is changing. So, k is 1 to n not k ij. Now if you make for every cell the total then this is basically $\sum y_{ij}$; that square when

you sum up across all the cells you will be getting this quantity and it will be divided by n and subtracted by the substitution factor you will get subtotal.

Then what is SS AB? SS AB is subtotal minus SS A SS B something like this. So, let me repeat the calculation.

(Refer Slide Time: 28:42)

$$SS_T = \sum_{i=1}^a \sum_{j=1}^b \sum_{k=1}^n y_{ijk}^2 - \frac{y_{...}^2}{N = abn}$$

$$SS_A = \frac{1}{bn} \sum_{i=1}^a y_{i..}^2 - \frac{y_{...}^2}{abn}$$

$$SS_B = \frac{1}{an} \sum_{j=1}^b y_{.j.}^2 - \frac{y_{...}^2}{abn}$$

$$SS_{Subtotal} = \frac{1}{n} \sum_{k=1}^n y_{.k.}^2 - \frac{y_{...}^2}{abn}$$

$$SS_{AB} = SS_{Subtotal} - SS_A - SS_B$$

$$SS_E = SS_T - SS_A - SS_B - SS_{AB}$$

The calculation is SS total equal to three sums i equal to suppose you write i equal to 1 to a, j equal to 1 to b, k equal to 1 to n, then y_{ijk} square you have seen this one. Minus this is the grand total square by N; N equal to abn. Suppose you want to know SS A, what do you require? You require I equal to 1 to a, then y_{i..} square minus y_{...} dot dot dot square by abn, but here you have to divide it by 1 by B n.

Similarly, SS B you will find out 1 by a n j equal to 1 to b y_{.j.} dot square minus y_{...} dot dot dot square by abn. Everywhere you see y_{...} grand total square by abn is subtracted. Then you are creating SS subtotal which is basically 1 by n sum total y_{.k.} dot square k equal to 1 to n minus y_{...} dot dot dot square by abn. Then you are creating find out interaction which is SS subtotal minus SS A minus SS B.

Then you find out SS E which is SS T minus SS A minus SS B minus SS AB. This is the computation part for sum squares. And this you will be using ANOVA table. Now see the table SS A, SS B, SS AB, SS E, SS T all those the computation I have shown and now you will be able to find out this.

So, let us see one example.

(Refer Slide Time: 31:20)

Two-way ANOVA – Example

- An engineer is studying methods for improving the ability to detect targets on a radar scope. Two factors she considers to be important are the amount of background noise, or “ground clutter,” (A) on the scope and the **type of filter** (B) placed over the screen.
- The Response variable is intensity level.
- It is experienced that the ground clutter can be categorized into three levels, i.e., Low, Medium and High and two filter types are available in the market.

Factor	Filter types							
	Type-1				Type-2			
Ground clutter								
Low (1)	90	96	100	92	86	84	92	81
Medium (2)	102	106	105	96	97	90	97	80
High (3)	114	112	108	98	93	91	95	83

IIT KHARAGPUR | NPTEL ONLINE CERTIFICATION COURSES

Now example is this. Now, what happened we introduced that filter type; please remember I told in the beginning that these data; data this is experimental data is taken from one of the example of montgomery's book. And here what happened not only this ground clutter and filter type, but there is another one operated different kind of operator. So, it is a 2 factor with a blocking that experiment was done and the same data he is taken in. I, first what happened we ignored the filter types an operator in one way ANOVA case and we found there is no effect.

Now, when two-way effect is coming that the operator effect is ignored. And we are assuming that thus this type of data we will get even if operator effect we will not consider. So, as a result this is the type 1 and type 2 filter type; this we have data points and here we have data points for different on clutter level.

(Refer Slide Time: 32:27)

Two-way ANOVA – Example (Contd.)

$$SS_T = 220108 - \left(\frac{2288^2}{24}\right) = 1985.333$$

$$SS_{clutter} = \frac{1}{(2 \times 4)} \{721^2 + 773^2 + 794^2\} - \frac{2288^2}{24}$$

$$= 218475.75 - 218122.667 = 353.083$$

$$SS_{filter_type} = \frac{1}{(3 \times 4)} \{1219^2 + 1069^2\} - \frac{2288^2}{24}$$

$$= 219060.167 - 218122.667 = 937.50$$

$$SS_{interaction} = \frac{1}{(2 \times 2)} \{378^2 + 343^2 + 409^2 + 364^2 + 432^2 + 362^2\} - \frac{2288^2}{24} - SS_{clutter} - SS_{filter}$$

$$= 219494.5 - 218122.667 - 353.083 - 937.5$$

$$= 81.25$$

$$SS_E = SS_T - SS_{clutter} - SS_{filter_type} - SS_{interaction}$$

$$= 1985.333 - 353.083 - 937.50 - 81.25$$


$$= 613.5$$

ANOVA table

Sources of variation	Sum of squares	DOF	Mean squares	F0	Decision
Ground Clutter types	353.083	2	176.541	5.18	Significant
Filter type	937.5	1	937.5	27.5	Significant
Interaction	81.25	2	40.625	1.192	Insignificant
Error	613.5	18	34.083		
Total	1985.33	23	86.318		

Summary
Ground clutter and filter types are significant; but interaction is insignificant

IIT KHARAGPUR | NPTEL ONLINE CERTIFICATION COURSES



And what we have done basically, we have computed all SS T is a squatter, SS filter type, SS interactions using the formula I have already given to you. What are the formula? Let us see this formula; these are the formulas; so what I have done. So these are the SS T, SS A, SS B, SS subtotal these are the formula. Using this formula you have found out a SS T for that example and similarly up to a SS E. Come to the slides now, what happened SS T value is 1985.3, this is clutter value is this, this is filter type is this like this. And then, then using ANOVA we found out the F 0 values 5.18 to determine these.

Now, you know that this 5.18 value is little higher value. Now when we compare with the tabulated value considering the respective degrees of freedom 2 and 18 and that value is less than this for alpha equal to 0.05. And obviously then, they 1 and 18 also less than this 27.5; so fill the clutter type and filter type becomes significant.

That means, there is difference in the response value variable value which is here in the; that is the intensity level at the time of detection they differ if the ground clutter level changes from the low to medium to high. We do not know whether it is low medium high (Refer Time: 34:12), but there is a at least if you change from low to medium or low to high or medium to high in one of the case there will be difference. That is also true for filter type; that means, if you choose use filter type 1 the intensity at level at target detection will be different if you use filter type 2. But in one factor case when we have

considered the same data with only ground clutter and as if there is no other factors affecting and the data what we have got actually same if I do that one factor level experiment. There then we have analyzed and we found that there is no significant difference for different ground clutter level when he introduced the type 1 and type 2 that filtered type then there is a difference. So, this is the beauty of introducing more factors.

And another interesting thing is that the interaction between the ground clutter and filter type is insignificant. There is no significant, means there is no dependency between these 2 factors. So, fantastic; so what we have discussed then in this lecture? In this lecture we have I conclude them all two ANOVA part.

(Refer Slide Time: 35:47)

Two - fact

A → 1, 2, ... a.
 B → 1, 2, ... b.

$\tau_i, \tau_j, (\tau\beta)_{ij}, \epsilon_{ijk}$

Sources of Variance	SS	DOF	MS	F ₀
→ A	SS _A	a-1	SS _A /(a-1) = M _A	M _A /M _{SE}
→ B	SS _B	b-1	SS _B /(b-1) = M _B	M _B /M _{SE}
→ AB	SS _{AB}	(a-1)(b-1)	M _{S AB}	M _{S AB} /M _{SE}
Error	SS _E	ab(n-1)	M _{SE}	
TOTAL	SS _T	abn-1		

$H_0: \tau_1 = \tau_2 = \dots = \tau_a$
 $H_1: \tau_i \neq \tau_j$
 $H_0: \tau_{11} = \tau_{12} = \dots = \tau_{1b}$
 $H_1: \tau_{ij} \neq \tau_{kl}$

$F_{(a-1, ab(n-1))}$
 $\alpha = 0.05$

Now that in two-way ANOVA there will be two factors: there will be factor A with 1, 2, a level; factor B 1, 2, b level; and the model will have tau i that is for factor A there will be the treatment tau beta j that will be for factor B and there will be their interaction tau beta ij and obviously there will be error epsilon ij k.

So, there are how many sources of variability? Sources of variability AB interaction and error and finally total. This is basically sources of variability. Then we have computed SS, so that is for total it is SS T, for error it is SS E, for interaction it is SS AB, for factor B and for factor A like this. Then you compute a degree of freedom a minus 1 for A, b minus 1 for B, a minus 1 into b minus 1 for AB, then your ab into n minus 1 for error, and abn minus 1 for total.

Then what you have done? You have computed MS. MS is nothing but SS A by a minus 1 this is for MS A, similarly MS B will be SS B by b minus 1; so MS B. That means, SS by its degree of freedom. And similarly you will calculate MS AB, you will come calculate MS E, you do not require to compute as nothing not required for other things.

So, then you create an F statistics or F_0 ; that means, when hypothesis is true then this is nothing but MS A for treatment A it is MS A by MS E; for treatment B it is MS B by MS E; for treatment AB it is MS AB by MS E. And what you are doing you are testing hypothesis like this: H_0 in this case that τ_i equal to 0 versus H_1 τ_i not equal to 0 for at least one i.

Similarly, here H_0 β_j equal to 0 versus β_j not equal to 0 for 1 and like this, and similarly your AB also interaction. Then what happened? There will be your threshold value which is F alpha numerator degree of freedom and denominator degree of freedom. For the first case factor A case this one is a minus nu 1 is a minus 1 and nu 2 will be always same because this is the arrow degrees of freedom. So, this is ab n minus 1.

Now if any of the computed statistics say F_0 value is greater than this value F alpha this; usually alpha we will considered at 0.05, then if this is the case then that corresponding treatment effect is there or interaction if it is there. If this is less than this the effect is not there. And with an example we have shown; the radarscope example we have shown that the ground clutter and it will filter type FFT is there, but interesting effect is not there. So, thank you very much for your patient hearing.

Thanks a lot.