

Design and Analysis of Experiments
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Lecture – 13
Analysis of Variance (ANOVA)

We will discuss certain parts of Analysis of Variance.

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Analysis of Variance
(ANOVA)

One factor expt.

$A_i \Rightarrow a \text{ levels, } i=1, 2, \dots, a$
 $j=1, 2, \dots, n$

$y_{ij} =$ Response for the j th obs when expt done for i th treatment level.

Expt data	A	Replicate (j)					Total	Average
		1	2	3	...	n		
1		y_{11}	y_{12}	y_{13}	y_{1j}	y_{1n}	$y_{1.}$	$\bar{y}_{1.}$
...	2	y_{21}	y_{22}	y_{23}	y_{2j}	y_{2n}	$y_{2.}$	$\bar{y}_{2.}$
...
...	i	y_{i1}	y_{i2}	y_{i3}	y_{ij}	y_{in}	$y_{i.}$	$\bar{y}_{i.}$
...
...	a	y_{a1}	y_{a2}	y_{a3}	y_{aj}	y_{an}	$y_{a.}$	$\bar{y}_{a.}$
							$y_{..}$	$\bar{y}_{..}$

No. of total obs = $N = a \cdot n$

Today's contents are first we will introduce with an example and then we will see that how one factor experimental data can be represented in tabular form, then we will discuss the partitioning of data, observations for single factor ANOVA and also the models for the data and then how to decompose the sum of squares into treatment effects, error effects and that same thing with an example with later we will explain.

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Source: This lecture is prepared based on "Design and Analysis of Experiments" by D C Montgomery, Wiley, 8th Edition

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So, let us start with the same example what we have seen in earlier lectures.

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Introduction – An 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	Observations (Replications)							
Ground clutter								
Low (1)	90	86	96	84	100	92	92	81
Medium (2)	102	97	106	90	105	97	96	80
High (3)	114	93	112	91	108	95	98	83

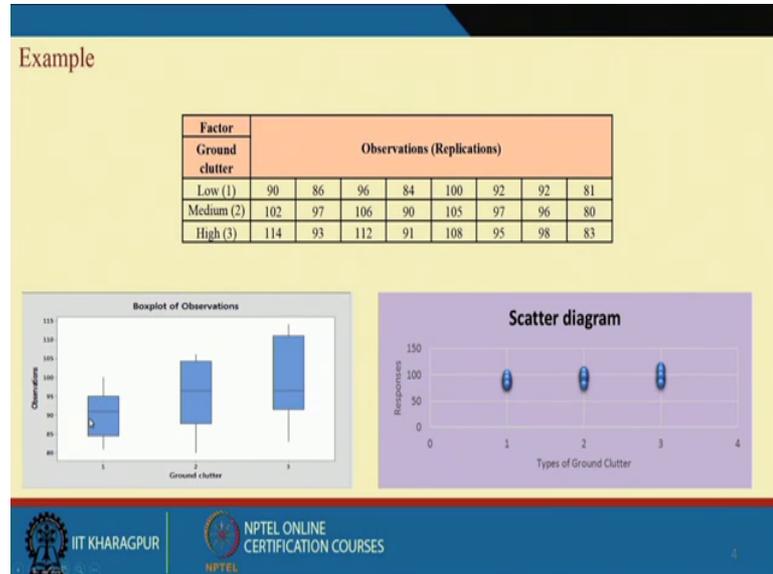
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Suppose the engineer is interested to study the methods of improving the ability to detect targets on a radar scope two factors considered to be important are ground clutter and type of filter response variable is the intensity level at the time of detection that will be that is what is measured. And in this example we are considering only one factor that is ground clutter which is having 3 levels low, medium and high.

For the time being like earlier in earlier lectures we assumed that this data comes from an experiment that is one factor experiment and ground clutter level at low medium and

high and there are 8 observations or 8 replicates again against each of the 3 factorial settings or 3 treatment levels.

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So, I also explained in earlier lectures that a use of boxplot. So, from the boxplot we wanted to know whether there is difference in means as and medians when the factor level changes from level 1 to level 2 to level 3. In fact, in this boxplot the median lines are given this horizontal lines middle lines are median. So, there can be somewhere mean here mean here and mean here.

So, from this plot it is visible that there is difference in mean response from when you observe with ground clutter level low to medium to high. But this is a visual one only just and it all it all depends on the scale you are using. So, it is, it can give you go for some kind of other analysis, but it will not clearly give you that whether the difference is statistically significant or not, and in the right hand side scatter diagram is shown. This is just to see that how the data behaves in terms of the relation between this two variables scatter plot region is full of one or to talk about the relation, but from the scatter plot again it all again depends on the response measurement that the scale of the figure. So, here also here from here we are not able to find any significant we are not able to find out much difference between the mean levels.

So, essentially what happened then that if there is one factor experiment let the factor is a denoted by a then this factor has a levels, let a mean level 1, level 2 like a levels. So, we

say i from 1 to a and then against each level you have n replications j equal to 1 to n and each of the response responses, each of the responses now can be denoted by y_{ij} . What does it mean? This is the response for the j th observations when experiment done for i th treatment level, i th treatment level what I mean to say your factor a is having level 1, level 2 like level i similarly suppose level a .

And there is replica replicate that denoted by j a j equal to 1 2 3 like j like n . So, then the general observation is y_{ij} . So, i th level j th observation that is what I have written here and this is what is the representation of data when you do experiment you do experiment for one factorial experiment again which is having a levels each having n n replications then your data set will be y_{11}, y_{12}, y_{13} .

So, like y_{1j} like y_{1n} this is y_{21} this is y_{22} y_{23} y_{2j} , y_{2n} then $y_{i1}, y_{i2}, y_{i3}, y_{ij}$ y_{in} , in the same manner $y_{a1}, y_{a2}, y_{a3}, y_{aj}, y_{an}$. So, that means, the number of observations total observations number of total observations if I say equal to a n then this is nothing, but a level times n a n , a level time times n , now one factor experimental data.

So, let us see the slides, you see this, this is data representation for one factor experiment the what I have given in this, this slide this is nothing, but data representation for one factor experiment you may be interested in total here you may be interested in average across each level here. So, by total what I mean to say suppose the i th total you can write down $y_{i\cdot}$. So, then first row total is $y_{1\cdot}$ $y_{2\cdot}$ dot like here $y_{a\cdot}$, if you want to get average then you write $y_{1\cdot}$ average, $y_{2\cdot}$ average $y_{i\cdot}$ average $y_{a\cdot}$ average and then someone may be interested to know the total of all the observations $y_{\cdot\cdot}$ and grand average $y_{\cdot\cdot}$ bar. So, you know how do compute average how to make total all those things. So, this is what is there, but we will proceed further. So, I want to go to the slide.

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Data representation for one factor experiment

Treatment (level)	Observations				Totals	Averages
1	y_{11}	y_{12}	...	y_{1n}	$y_{1.}$	$\bar{y}_{1.}$
2	y_{21}	y_{22}	...	y_{2n}	$y_{2.}$	$\bar{y}_{2.}$
...
a	y_{a1}	y_{a2}	...	y_{an}	$y_{a.}$	$\bar{y}_{a.}$
					$y_{..}$	$\bar{y}_{..}$

- In general, there will be a levels of the factor, or a treatments, and n replicates of the experiment, run in random order...a completely randomized design (CRD)
- $N = an$ total runs
- We consider the fixed effects case...the random effects case will be discussed later

Objective is to test hypotheses about the equality of the a treatment means

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So, you see n replicates and you all know this is random, random experiment. So, how to make randomization that also we have discussed in first few lectures, this is a complete randomized design case.

I told you n total a n total runs, and here we can consider two kinds of models here a fixed effect it means the treatment fixed treatment otherwise random treatment random effects. So, we will discuss later. So, why we are we have done this experiment, we have done this or we will be doing this experiment to know that whether the means across different level levels differ or not, that is what we want to test it will be a hypothesis testing case where you have more than two levels or two treatments and you want to test whether the mean response across the different treatments are equal or not. So, equality of treatment means across various treatment levels.

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Single Factor ANOVA

- The name “analysis of variance” stems from a **partitioning** of the total variability in the response variable into components that are consistent with a **model** for the experiment
- The basic single-factor ANOVA model is

$$y_{ij} = \mu + \tau_i + \varepsilon_{ij}, \begin{cases} i = 1, 2, \dots, a \\ j = 1, 2, \dots, n \end{cases}$$

μ = an overall mean, τ_i = *i*th treatment effect,
 ε_{ij} = experimental error, $NID(0, \sigma^2)$

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So, with this we will now formally define the single factor model. So, single factor model will be defined like this y_{ij} equal to $\mu + \tau_i + \varepsilon_{ij}$. This is more general.

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Fixed effect model.

$$y_{ij} = \mu + \tau_i + \varepsilon_{ij}$$

$i = 1, 2, \dots, a$
 $j = 1, 2, \dots, n$

General obs. (Grand mean), Factor effect (treatment), Random error.

$\varepsilon_{ij} \sim N(0, \sigma_i^2)$
 $\sim N(0, \sigma^2)$
 Homoskedasticity

$\tau_i = \mu_i - \mu$

$\mu_i = \mu + \tau_i$

So, $\mu + \tau_i$ is basically μ_i that also we can write, but we will start with this $\mu + \tau_i$. So, what is the y_{ij} ? This is a general observation. What does it mean by general observation? It means the response value for the *j*th observation and when you expand *i*th experiment level 1 or on the experiment of *i*th level of factor *a*. Then what is

this μ ? This is known as grand mean. What is this τ_i ? τ_i is because you have one factor the factor effect other way we can see the treatment effects what is this portion this is the this is the random error path. So, if τ_i is fixed it has a particular effect that is constant then this model will be known as fixed effect model fixed effect model.

Now, this in this diagram show what is i , i equal to 1 to a , what is j , j equal to 1 to n this is our, this is our model that is one factor ANOVA, single factor ANOVA model. So, here very important concept is this, this is the random error this random error means it has certain distribution. So, ϵ_{ij} will be normally distributed with 0 mean and σ_i^2 square variability, if the variance across the different levels i we have a with 1 2 like a levels if the variance of y when experiment done is done at the factor level a at level 1 then the variance is suppose σ_1^2 , then this will be σ_2^2 and this will be σ_a^2 or i th case this will be σ_i^2 then we will write like this.

Suppose we do one kind of assumptions here that that σ_1^2 equal to σ_2^2 square equal to σ_i^2 square equal to σ^2 then we can write that this ϵ_{ij} is 0 σ^2 ; that means, the response values y values across different levels. So, level 1 n observation level 2 another n observation it will if you have substantially large number of observe experimental data you will find that the from grater the variability across this values across this values, across this values, across this values will remains same or similar all most equal then we can assume that variability across different level of y across different level is equal that is σ^2 .

Then this error term this error term follows normal distribution with mean 0 and σ^2 variance this is a very very important assumption. Later on I when I talk about assumption we will come back to this, this is known as Homoskedercity as assumption skadersty city. Those things we will discuss later on.

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Models for the Data

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

Means model

$$y_{ij} = \mu_i + \varepsilon_{ij}$$

Effects model

$$\mu_i = \mu + \tau_i \quad i = 1, 2, \dots, a$$
$$y_{ij} = \mu + \tau_i + \varepsilon_{ij} \quad \text{where } j = 1, 2, \dots, n$$

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So, as I told you that $\mu + \mu_i$. So, I you can also write that y_{ij} equal to $\mu + \mu_i + \varepsilon_{ij}$. So, then what is μ_{ij} , μ_{ij} equal to $\mu + \tau_i$. So, this one is known as partitioning also you see that the total observation a particular experimental run y of values what you got you partition into grand mean into fixed factor effect plus random error.

So, it is partition into 3 components overall mean, factor effect and error. So, overall mean plus factor effect is giving you the mean at each level. So, mean at this level is μ_1 , mean at second level is μ_2 mean at i th level is μ_i . So, then what we are saying that these τ_i equal to $\mu_i - \mu$ you see from here τ_i equal to $\mu_i - \mu$. So, μ_i is mean at i th level i th level μ_i , if i equal to 1 this is a first level. So, this is what is our model data representation and model.

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ANOVA model

❑ **Fixed effect model**

- In the effects model, we break the i -th treatment mean μ_i into two components such that $\mu_i = \mu + \tau_i$. We usually think of μ as an overall mean so that

$$\frac{\sum_{i=1}^a \mu_i}{a} = \mu$$
 where a is the number of treatments.
- This definition implies that $\sum_{i=1}^a \tau_i = 0$.

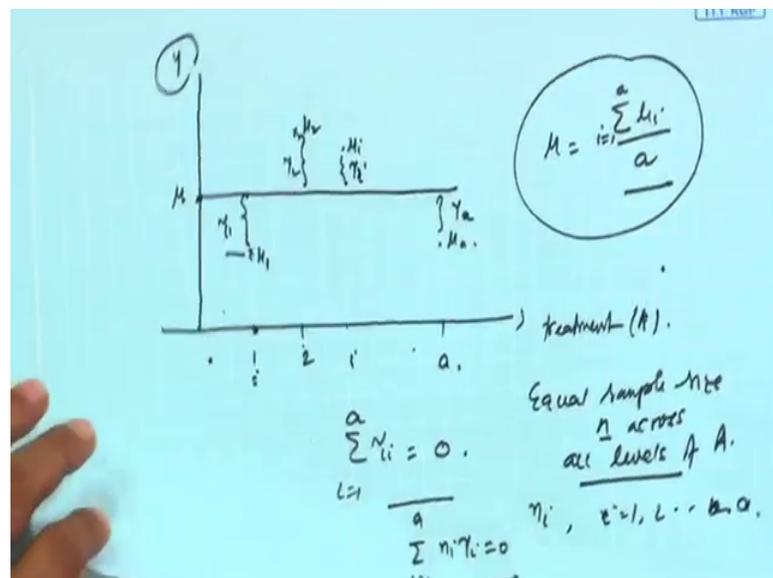
❑ **Random effect model**

Here, τ_i are random variables and knowledge about the particular ones investigated is relatively useless. Instead, hypothesis is tested about the variability of the τ_i

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And here interestingly what is happening. Suppose I plot the data at this side y , suppose this is my mean, suppose you are talking about these are the treatments treatment our treatment is a .

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So, at different level first level second level like you it is i th level this may be a th level. So, everywhere there will be a μ . So, I first level mean maybe here would you which is μ_1 , second level mean maybe here which is μ_2 , i th level meal may be here which is

μ_i , i th level mean may be here which is μ_a . So, these are the mean at of y of y at different level and then average or mean of the means is the grand mean.

Then what is the effect? Effect is the difference between those two. This is τ_1 , this one is τ_2 , this is τ_3 , τ_i and this one will be τ_a . So, it is it is nothing, but difference between individual level mean and the grand mean, again grand mean can be computed from the individual level means like grand mean is some total of μ_i by i equal to 1 to a . So, these a restriction or constraint this leads to some of τ_i equal to 0, i equal to 1 to a means this plus this plus this plus this if you do you will get it 0.

Obviously we are considering equal sample size equal sample size that is n across all levels of, levels of a when you are doing experiment (Refer Time: 19:20) otherwise what will happen if the sample size varies sample size is n_i i equal to 1 to 1 to a then it will be i equal to 1 to a $n_i \tau_i$ this will become 0. Now, this is the, this is the restriction and this help us in doing many things. Now come back to the slide.

So, in fixed effect model we consider that τ_i is fixed you just see that within μ_a s and then in the random effect model you see here τ_i are random variables and knowledge about particular once investigated is relatively used less instead hypothesis tested over the variability of τ_i . So, you do not have fixed value for τ_i it is a random one, it is a random variable. So, instead of mean you will we will be interested in the variability of τ_i , but ANOVA is analysis of variance even though it is analysis it is named as analysis of variance, but it is actually tests the equality of population means.

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Decomposition of total sum of square

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

$$SS_T = \sum_{i=1}^a \sum_{j=1}^n (y_{ij} - \bar{y}_{..})^2$$
- The basic **ANOVA partitioning** is:

$$\begin{aligned} \sum_{i=1}^a \sum_{j=1}^n (y_{ij} - \bar{y}_{..})^2 &= \sum_{i=1}^a \sum_{j=1}^n [(\bar{y}_{i.} - \bar{y}_{..}) + (y_{ij} - \bar{y}_{i.})]^2 \\ &= n \sum_{i=1}^a (\bar{y}_{i.} - \bar{y}_{..})^2 + \sum_{i=1}^a \sum_{j=1}^n (y_{ij} - \bar{y}_{i.})^2 + 2 \sum_{i=1}^a \sum_{j=1}^n (y_{ij} - \bar{y}_{i.})(\bar{y}_{i.} - \bar{y}_{..})^2 \\ &= n \sum_{i=1}^a (\bar{y}_{i.} - \bar{y}_{..})^2 + \sum_{i=1}^a \sum_{j=1}^n (y_{ij} - \bar{y}_{i.})^2 \\ SS_T &= SS_{Treatments} + SS_E \end{aligned}$$

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So, that mean when you say y ANOVA model you have to say y_{ij} equal to μ_i plus τ_i plus ϵ_{ij} and then definitely you will be saying ϵ_{ij} normally distributed 0 and σ^2 . This you can also write like this y_{ij} equal to μ_i plus τ_i plus ϵ_{ij} and ϵ_{ij} is normally distributed 0 and σ^2 that is that is what is what we have learned so far.

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Handwritten notes on a whiteboard:

$\rightarrow y_{ij} = \mu + \tau_i + \epsilon_{ij} ; y_{ij} = \mu_i + \epsilon_{ij}$

$\sum_{i=1}^a \tau_i = 0, \sum_{i=1}^a n_i \tau_i = 0$

$\epsilon_{ij} \sim N(0, \sigma^2)$

$\mu_i = \mu + \tau_i$

Sum of squares

$y_{ij} = \bar{y}_{..} + (\bar{y}_{i.} - \bar{y}_{..}) + (y_{ij} - \bar{y}_{i.})$

$y_{ij} = \bar{y}_{..} + (\bar{y}_{i.} - \bar{y}_{..}) + (y_{ij} - \bar{y}_{i.})$

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So, now, come to the point that you have done the experiment once you have done the experiment you will get this kind of data. So, well the row of interest for us now is this

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$$y_{ij} = \mu + \alpha_i + \beta_j$$

$$\sum_{i=1}^a \alpha_i = 0, \quad \sum_{j=1}^n \beta_j = 0$$

$$\mu_i = \mu + \alpha_i$$

$$y_{ij} \sim N(0, \sigma^2)$$

$$y_{ij} = \bar{y}_{..} + (\bar{y}_{i.} - \bar{y}_{..}) + (y_{ij} - \bar{y}_{i.})$$

Sum Squared.

$$\sum_{i=1}^a \sum_{j=1}^n (y_{ij} - \bar{y}_{..})^2 = \sum_{i=1}^a \left[(\bar{y}_{i.} - \bar{y}_{..}) + (y_{ij} - \bar{y}_{i.}) \right]^2$$

So, what you do you just write down $y_{ij} - \bar{y}_{..}$ left hand side then right hand side will be $\bar{y}_{i.} - \bar{y}_{..} + y_{ij} - \bar{y}_{i.}$. If you take square then you have to square it and then take summation then you sum here. So, i and j both are there. So, i stands from 1 to a j stands from 1 to n . Suppose I will write j equal to 1 to n here you write j equal to 1 to n here.

Similarly another summing will be there not only across not suppose if I put j equal to 1 to n that mean you are summing up $y_{i1} + y_{i2} + y_{i3}$ like this. So, this you are subtracting by $\bar{y}_{i.}$ and then making square and summing up. So, if you make then there is another portion is 1 to a . So, that also you sum up. So, i equal to 1 to a . Then another sum i equal to 1 to a .

What will happen if you expand this? Now, see the slides see what I have written here i equal to 1 to a j equal to 1 to n this square now this one when you take double sum then you see you are doing this and finally, these portion left hand side is becoming $n \times a$ times i equal to 1 to a $y_{i.} - \bar{y}_{..}$ square plus this here double sum is there. So, this portion when each of the observations are subtracted by their grand mean and taken square and makes sum up this is the sum square total, this is sum square total it is total. Now, here what is happening each of the treatment means are subtracted from the grand means or grand average each of the treatment average is subtracted from the grand

average and then you have taken their sum S square and sum across all the levels and then this one is nothing, but some square treatment and rest is some square error.

So, from here to here is a very simple one this quantity remains valid this quantity remains valid and this quantity becomes 0 because j equal to 1 to n if you sum up here you will find out that this quantity will become 0. So, one quantity this y i this y ij, one of these two will be this one will be I think this one, this one will become 0 and ultimately this is simple one it is not a big one big issue because if you write j equal to 1 to n then this will be n y bar square will be there and this also will be n y bar square then this n y bar minus n y bar that will become vanish, (Refer Time: 28:02) vanish and you will be having total sum square equal to treatments sum square equal to individual sum square. So, what you are getting then?

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$$\begin{aligned} \underline{SS_T} &= \sum \sum (y_{ij} - \bar{y}_{..})^2 \\ \underline{SS_{\text{treatment}}} &= n \sum_{i=1}^a (\bar{y}_{i.} - \bar{y}_{..})^2 \\ \underline{SS_E} &= \sum \sum (y_{ij} - \bar{y}_{i.})^2 \\ \underline{SS_T} &= \underline{SS_{\text{treatment}}} + \underline{SS_E} \end{aligned}$$

→ Decomposition of sum squares.

Hypothesis | $H_0: \mu_1 = \mu_2 = \dots = \mu_a$ ✓
 $H_1: \mu_i \neq \mu_k$ for at least one pair ✓

You will get SS T equal to double sum y ij minus y dot dot bar square. You will get treatment equal to n, 1 sum i equal to 1 to a y i dot bar minus y dot dot bar square and there will be SS error again double sum then you this is nothing, but y ij minus y dot bar this square. So, interested in what you have achieved SS T equal to SS treatment plus SS error this is known as sum square decomposition of sum squares, decomposition of sum squares.

So, we will see the one example here and ultimately in ANOVA as I told you at ANOVA we test null hypothesis that mu 1 equal to mu 2 equal to mu a and alternate hypothesis is

μ_i not equal to μ_k for at least one pair μ_1, μ_2, μ_3 like this for at least one pair there is a difference this. So, this one, we take this is the hypothesis what hypothesis we test what are the hypothesis null hypothesis is all means are equal alternate hypothesis at least one pair is not equal. So, that is H_1, H_0 and H_1 here.

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Example 1 - ANOVA

Factor	Observations (Replications)								Total	Average
Ground clutter										
Low (1)	90	86	96	84	100	92	92	81	721	90.125
Medium (2)	102	97	106	90	105	97	96	80	773	96.625
High (3)	114	93	112	91	108	95	98	83	794	99.25
									2288	95.33333

SS_E

0.015625 17.01563 34.51563 37.51563 97.51563 3.515625 3.515625 83.26563

28.89063 0.140625 87.89063 43.89063 70.14063 0.140625 0.390625 276.3906

217.5625 39.0625 162.5625 68.0625 76.5625 18.0625 1.5625 264.0625

$SS_{Treatment}$

27.12674

1.668403

15.34028

$$SS_E = \sum_{i=1}^a \sum_{j=1}^n (y_{ij} - y_i)^2 = 1632.25$$

$$SS_{Treatment} = n \sum_{i=1}^a (\bar{y}_i - \bar{y})^2 = 8 \times 44.13542 = 353.0833$$

$$SS_T = SS_E + SS_{Treatment} = 1632.25 + 353.0833 = 1985.333$$

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So, we will see how the sum square decomposition is taken place you just see this slide here we have H number of observations and you find out the total and their average second average, like this average this is grand average. We use this formula and we got that SS_E , SS_E equal to 1632.25, $SS_{Treatment}$ is coming 353.08 this is nothing, but using this formula only.

So, what is y_{ij} ? Suppose 90 minus grand average 95.33, this square is this, this square 90 minus 95 point like this then 86 minus this. So, make it and square it and this one this square. So, every observations will be subtracted by their respective by the grand min grand average and then square it and then when you make the total you will get this value, you will get this value.

Now, for the treatment how many treatments are there, there are 1 2 3 treatments. So, what is these 3 treatment method treatment average minus grand average, its square then second treatment average and grand average square and third treatment average square then take the sum you will get $SS_{Treatment}$ level like this. So, essentially what we

basically calculate, we calculate that SS treatment and SS total and then SS error will be the SS treatment minus SS total and accordingly you will get the values.

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Example 2 - ANOVA

An engineer is interested in investigating the relationship between the RF power setting and the etch rate for this tool. The objective of an experiment like this is to model the relationship between etch rate and RF power, and to specify the power setting that will give a desired target etch rate. She is interested in a particular gas (C2F6) and gap (0.80 cm) and wants to test four levels of RF power: 160, 180, 200, and 220 W. She decided to test five wafers at each level of RF power. She is interested in determining if the RF power setting affects the etch rate, and she has run a completely randomized experiment with four levels of RF power and five replicates.

Power (W)	Observations					Total	Averages
	1	2	3	4	5		
160	575	542	530	539	570	2756	551.2
180	565	593	590	579	610	2937	587.4
200	600	651	610	637	629	3127	625.4
220	725	700	715	685	710	3535	707

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Second one is a like this, another object here what happened this is the first example what I have given you in, the first from the first class onwards we have given you. This is the second example, second example is as engineer is interested in investigating the relationship between the RF power setting and etch rate for this tool and here basically response variable is the etch rate and you are using different power, power 160 watt, 180 watt, 200 watt and 220; that means, 4 levels. So, there are 5 replications and this is what are the experimental data coming and then these are the total and average. Now, find out SS error, SS total and treatment and SS total.

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Example 2 – ANOVA (Contd.)

RF Power (W)	Observed Etch Rate (Å/min)					Totals T_i	Averages \bar{y}_i
	1	2	3	4	5		
160	575	542	530	539	570	2756	551.2
180	565	593	590	579	610	2937	587.4
200	600	651	610	637	629	3127	625.4
220	725	700	715	685	710	3535	707.0
						$y_{..} = 12,355$	$\bar{y}_{..} = 617.75$

$$SS_E = \sum_{i=1}^a \sum_{j=1}^n (y_{ij} - y_i)^2 = 5339.20$$

$$SS_{Treatment} = n \sum_{i=1}^a (\bar{y}_i - \bar{y}_{..})^2 = 66,870.55$$

$$SS_T = SS_E + SS_{Treatment} = 5339.20 + 66,870.55 = 72,209.75$$





So, here let me give you tell you that there is some typographical mistake. So, please follow this when you are talking about SS T that mean each of the observation is subtracted by the grand average and then that difference is taking some taken squared and then overall sum. When you are talking about SS treatment? Every level average is subtracted from the grand average then there that difference is squared then using this you are getting this and when you are talking about SS error that mean your every average value y_{ij} sorry y_{ij} , not \bar{y}_{ij} every observation is subtracted from their respective from their, respective level average and then this one is summed up.

What does it mean? Here 575 minus 551.2 then 542 minus 551.2 so then the level one all these observations are subtracted by the respective average and then that one is squared and sum and summed up. So, there is similarly they are this one that this 4 different levels 4 into 5, 20 observations will be summed up using this formula will be summed up and then you will be getting SS error.

So, what did the mistake here typographical mistake here is, there is a bar here y_i dot bar. So, if you here it is y_i dot bar minus this the 3 levels, here one bar is missing please put that bar y_i dot bar. So, I hope the calculations are correct it is done in excel. If there is a mistake please mention in the in the discussion for to check.

So, this is what is decomposition of sum squares. So, I hope that you understand. So, what we have discussed?

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One factor CRD.

② Data Rep.

③ $y_{ij} = \mu + \tau_i + \epsilon_{ij}$ → Effect model
 $= \mu_i + \epsilon_{ij}$ → Mean "

$\epsilon_{ij} \sim N(0, \sigma^2)$

④ $y_{ij} = \bar{y}_{..} + (\bar{y}_{i.} - \bar{y}_{..}) + (y_{ij} - \bar{y}_{i.})$.

⑤ $SST = SS_{Treatment} + SS_{Error}$

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

① Radar Scope
A: 1, 2, 3.

② Etch rate.
160, 180, 200, 220
A: 1, 2, 3, 4

We have discussed one factor one factors CRD complete randomized design. Then we have we have seen that data representation in tabular form, then what we have seen that the ANOVA model, then we have seen that is the effect model this is what is effect model or you can write this one as mu i plus epsilon ij this is mean model and then here also we have said that the error is normally distributed with sigma square 0 sigma square. And then we have seen that a general observation how it will be partitioned to how it will be partitioned to different component, different component. From there we have seen that how the sum square total is partitioned into sum square treatment plus sum square error and we have given a general formula for calculation of these, these equal to double sum y ij minus y dot dot bar square equal to n i equal to 1 to a y i dot bar minus i double dot square plus here it will be double sum y ij minus y i dot bar square and here it is j equal to 1 to n, a equal, i equal to 1 to a, i equal to 1 to a, j equal to 1 to n.

And then I have given you two example, one is the that that is radarscope example where ground clutter level A has 3 levels and another one is the each etch rate e t c h etch rate example where basically some power is used 160, 180, 200 and 220 so that means, the factor a has 4 levels 1 2 3 and 4, 4 levels are there. And then how to compute the different things SS T, SS (Refer Time: 37:43) is given to you. Please practice it.

(Refer Slide Time: 37:52)

References

- Design and Analysis of Experiments by Douglas Montgomery, Wiley, 8th Edition, 2014

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Thank you very much and I have taken all these lecture, those things a from the book Design Analysis of Experiment by DC Montgomery, Wiley, 8th Edition.

Thank you very much. We will go to next, we will see next lecture now.