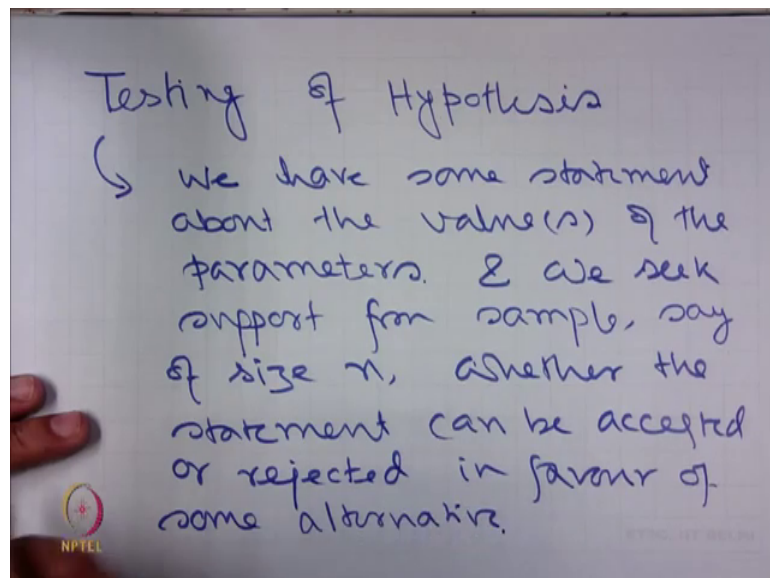


Statistical Inference
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Department of Mathematics
Indian Institute of Technology, Delhi

Lecture – 20
Statistical Inference

Welcome, students to the MOOCS lecture series on Statistical Inference, this is lecture number 20.

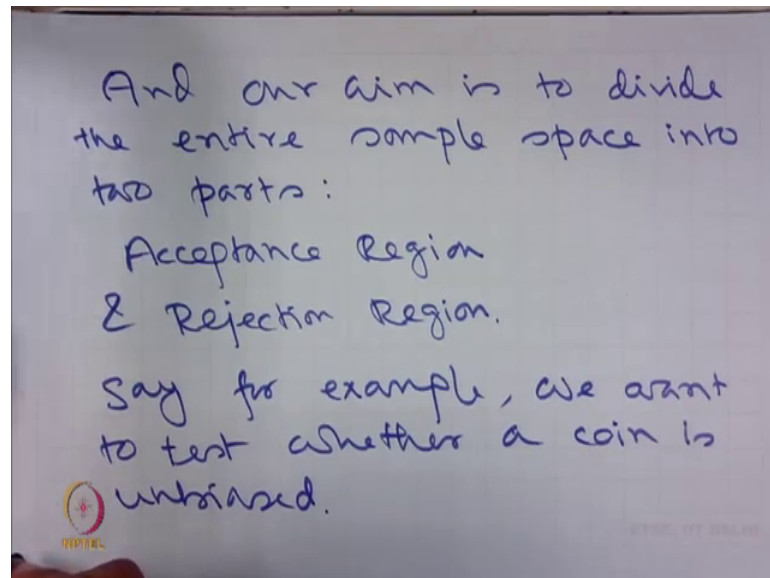
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If you remember we are doing testing of hypothesis as the last topic of the series of lectures. So, testing of hypothesis means we have some statement about the value or values of the parameters. And we seek support from sample say of size n , whether the statement can be accepted or rejected in favor of some alternative.

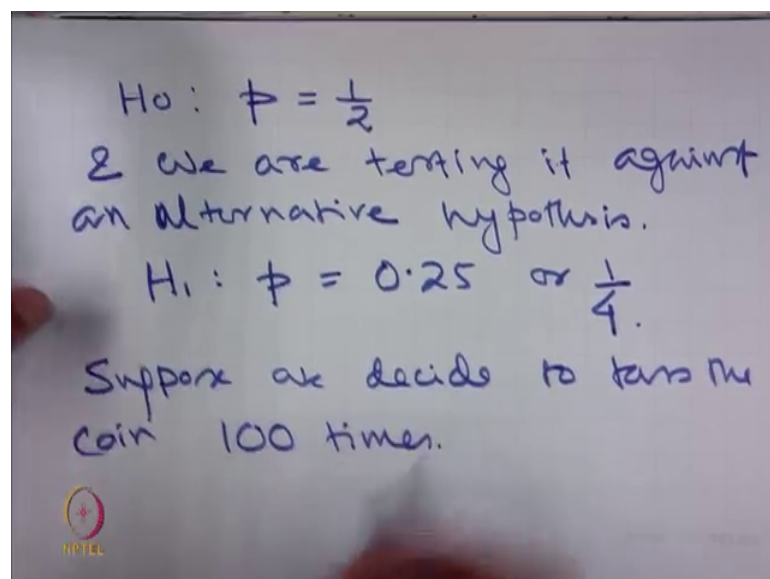
So, that is the whole purpose of testing of hypothesis.

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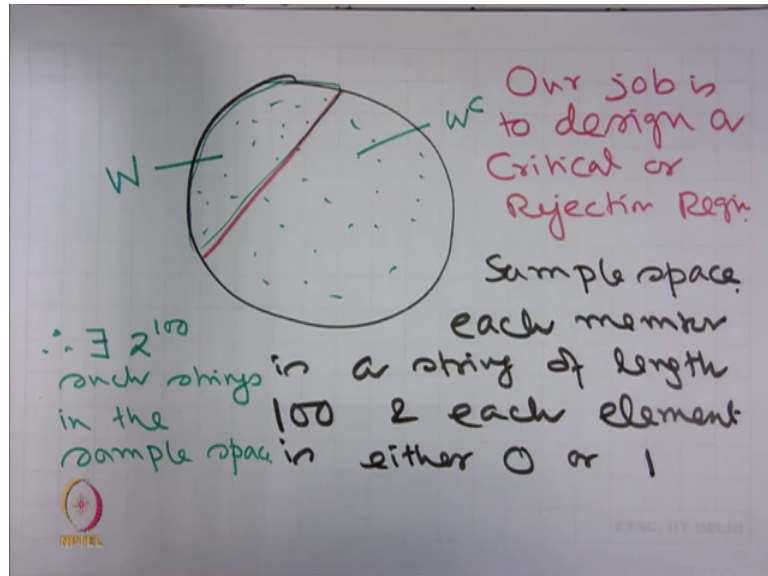
Our aim is to divide the entire sample space into two parts: Acceptance region and Rejection region. Say for example, we want to test whether a coin is unbiased. So, if we look at a simple null hypothesis.

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Then our H_0 is p is equal to half and we are testing it against an alternative hypothesis H_1 p is equal to say 0.25 or 1 by 4. Suppose, we decide to toss the coin 100 times then what is going to be the sample space.

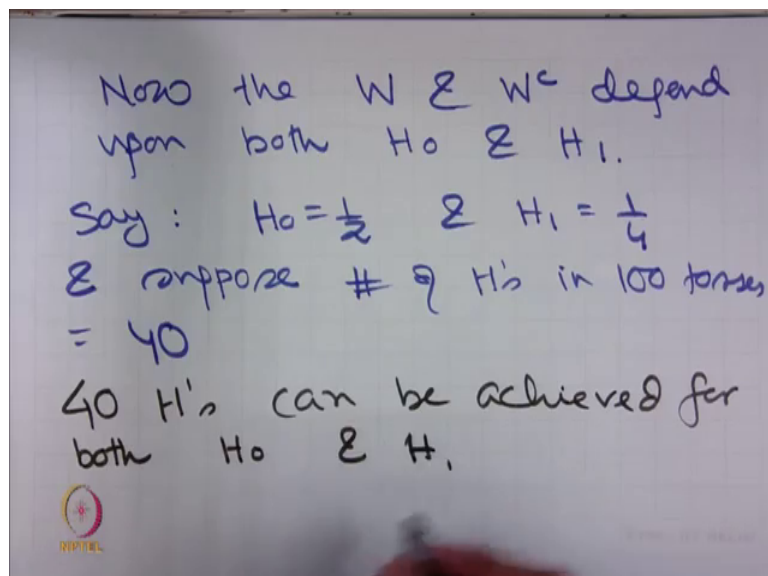
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So, the sample space is and each member of the sample space is a string of length 100 and each element is either 0 or 1. Therefore, there are 2^{100} such strings in the sample space. Suppose these are the points in the sample space. Now our job is to design a critical region or rejection region say we decide this is to be our critical region.

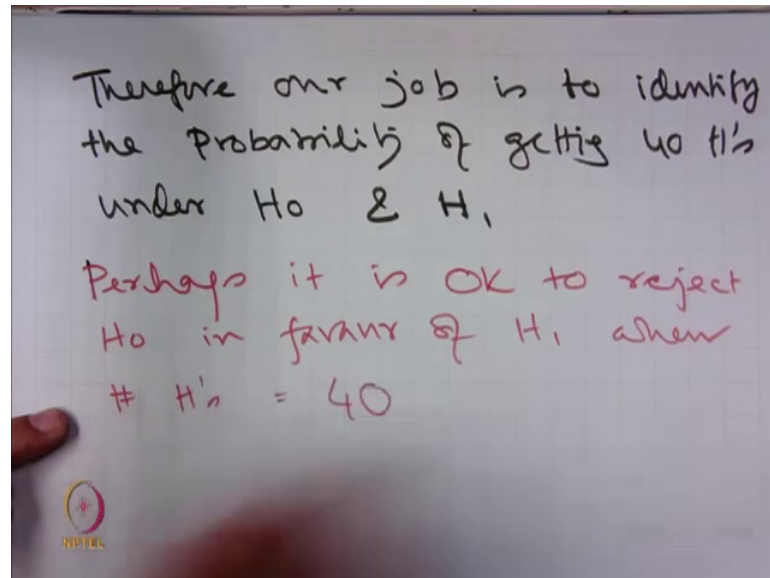
So, that if the outcome is falling in this part of the sample space then we are going to reject the null hypothesis. So, this is called w and this side is called w complement which gives us the acceptance region.

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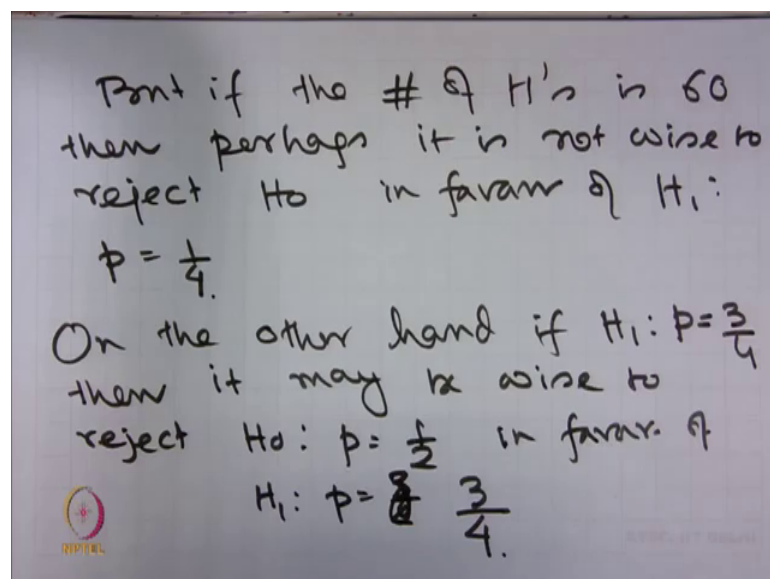
Now, the W and W complement depend upon both H_0 and H_1 say H_0 is half as I have already mentioned and H_1 is 1 by 4 and suppose number of heads in 100 tosses is equal to 40. Now, 40 heads can be achieved for both H_0 and H_1 .

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Therefore our job is to identify the probability of getting 40 heads under H_0 and H_1 . Perhaps, it is to reject H_0 in favor of H_1 when number of head is 40.

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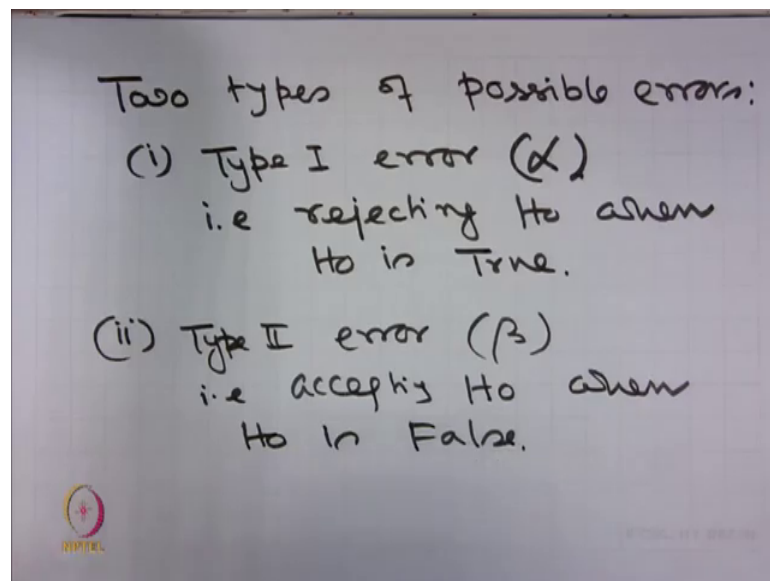


But if the number of heads is 60 then perhaps it is not wise to reject the null hypothesis H_0 in favor of H_1 that p is equal to 1 by 4. On the other hand, if H_1 is p is equal to

3 by 4 then it may be wise to reject H_0 if p is equal to half in favor of H_1 that p is equal to 3 by 4.

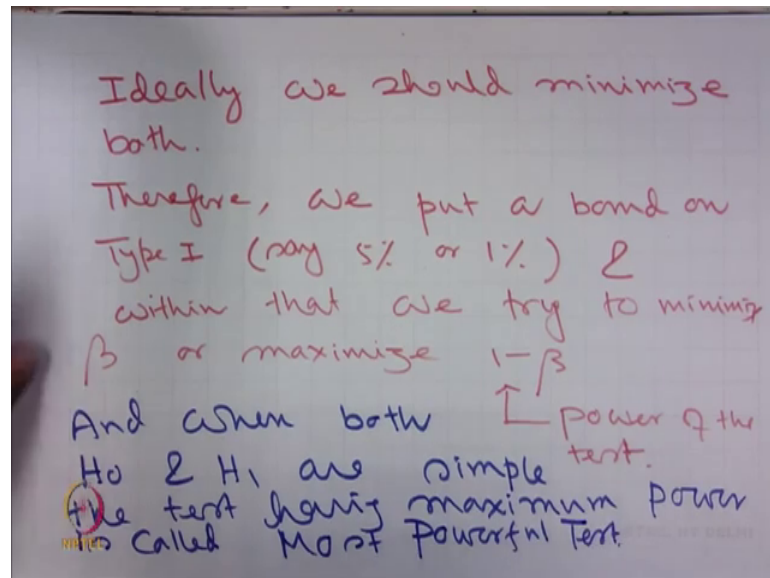
So, the job of a statistician for testing of hypothesis is that we have to decide that test. So, that we know the sample space and we need to decide the testing criteria. So, that we can accept or reject the null hypothesis against an alternative.

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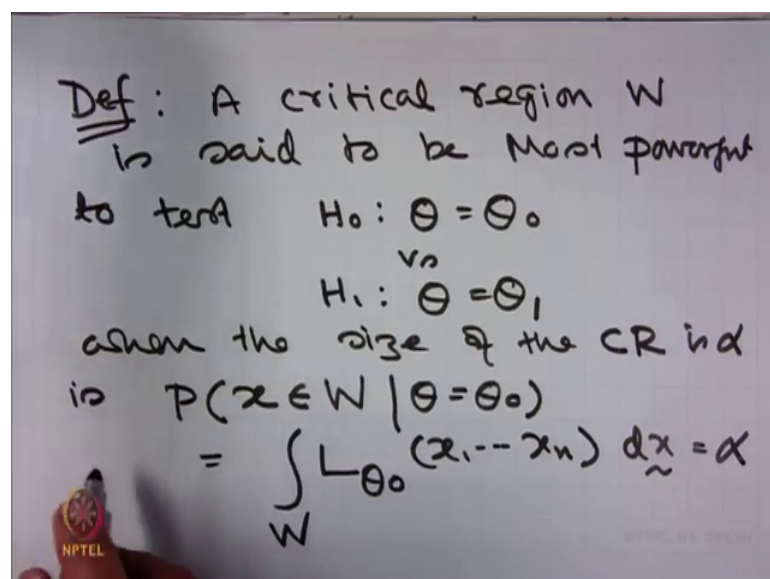
So, as I discussed two types of possible errors, one is type I error which you call alpha that is rejecting the H_0 when H_0 is true. That means, actually the coin is unbiased what we are rejecting that hypothesis based on the sample obtained. And, if I reject when the H_0 is true then we are committing an error which you call type I error and type II error is which you call beta is accepting H_0 when H_1 H_0 is false.

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Ideally, we should minimize both, but as I have explained that that cannot be done. Therefore, we put a bound on type I say 5 percent or 1 percent; that means, we can make a mistake only 5 out of 100 cases or 1 out of 100 cases. And within that we try to minimize beta or maximize 1 minus beta, this quantity is called the power of the test. And when both H_0 and H_1 are simple, the test having maximum power is called the most powerful test.

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So, definition are critical region W , we said to be most powerful to test $H_0: \theta = \theta_0$ versus $H_1: \theta = \theta_1$. When the size of the critical region is α such that probability x belonging to W given $\theta = \theta_0$ is equal to $\int_W L_{\theta_0}(x_1, x_2, \dots, x_n) dx = \alpha$.

So, we are saying that the making a mistake of rejecting or correct null hypothesis is this probability that the obtained sample x will belong to the rejection region or critical region W , these probabilities integrating on the space W of $L_{\theta_0}(x_1, x_2, \dots, x_n)$ this is the joint pdf of obtaining the sample x_1, x_2, \dots, x_n under the null hypothesis that is the parameter $\theta = \theta_0$ and that probability should be α .

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and among all CR regions W_1 of size α

$$\int_W L_{\theta_0}(x_1, \dots, x_n) dx \geq \int_{W_1} L_{\theta_0}(x_1, \dots, x_n) dx$$

Among all critical regions W_1 of size α integration over W of $L_{\theta_0}(x_1, x_2, \dots, x_n) dx$ is greater than equal to integration of W_1 of $L_{\theta_0}(x_1, x_2, \dots, x_n) dx$.

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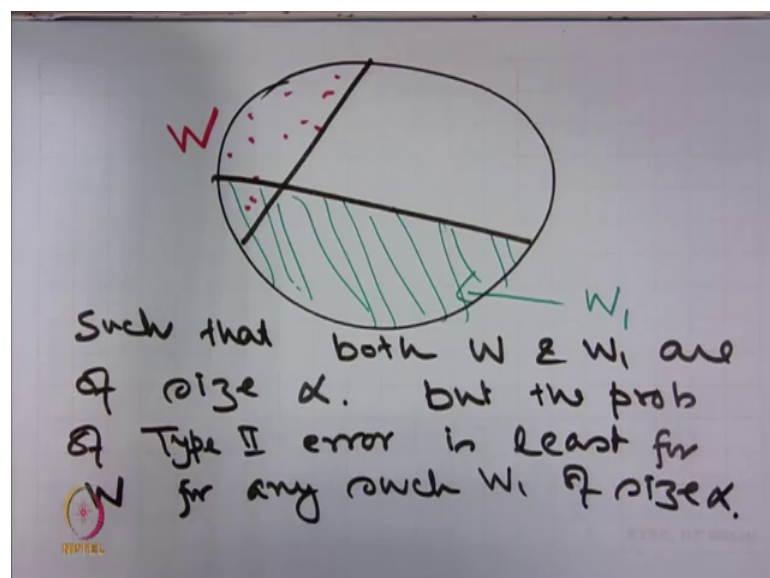
We want to maximize $1 - \beta$.

$$\beta = \text{Prob}(\text{Type II error})$$
$$= \text{Prob}(\text{Accepting } H_0 \text{ when } H_1 \text{ is True})$$
$$\therefore 1 - \beta = \text{Prob}(\text{Rejecting } H_0 \text{ when } H_1 \text{ is True})$$
$$= P(\text{The sample } x_1, \dots, x_n \in W \text{ when } H_1 \text{ is True})$$
$$= \int_W L_{\theta_1}(x_1, \dots, x_n) dx$$

This is because we want to maximize 1 minus beta, what is beta? Beta is equal to probability of type II error is equal to probability of accepting H_0 when H_1 is true. Therefore, 1 minus beta is equal to probability of rejecting H_0 when H_1 is true is equal to probability that the sample x_1, x_2, x_n belongs to the critical region when H_1 is true.

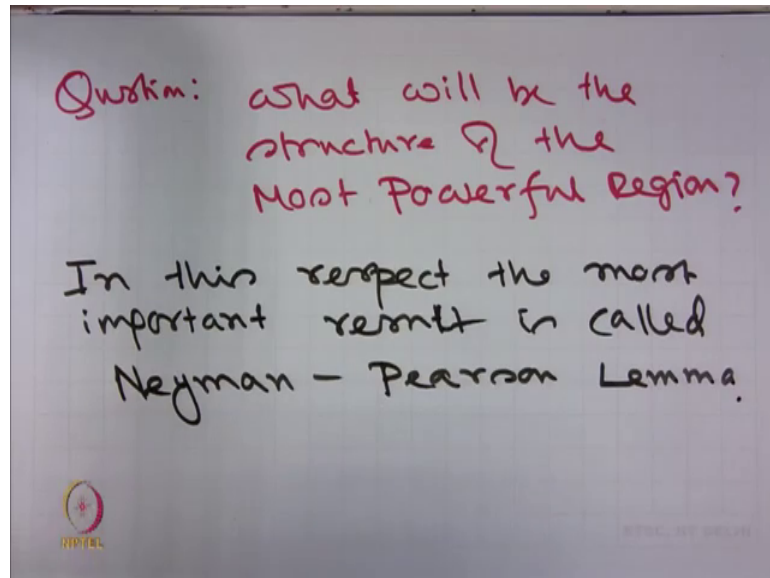
And therefore, this is, is equal to integration over W L_{θ_1} , because H_1 is true the parameter is θ_1 and this is $x_1, x_2, x_n dx$.

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Therefore, if this is the sample space and suppose this is one critical region W . And this is another critical region, say W_1 such that both of them are of size α , but the probability of type II error is least for W for any such W_1 of size α .

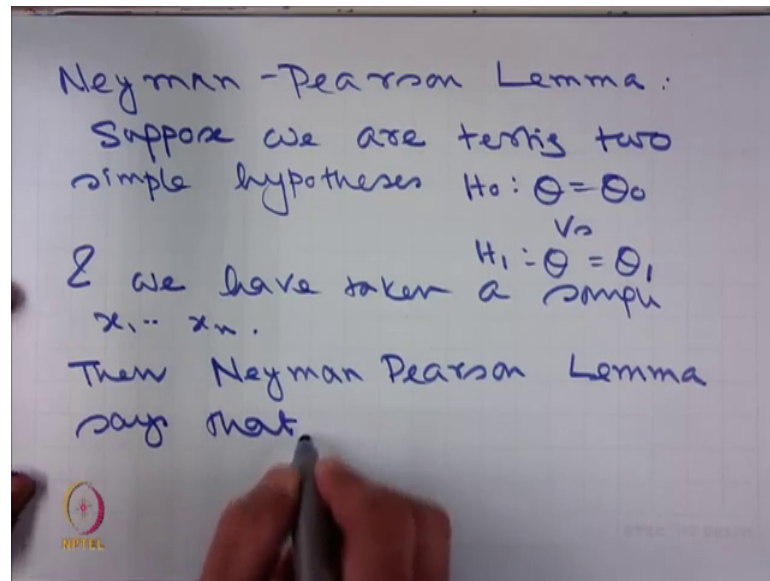
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In this respect questions, what will be the structure of the most powerful region or in other words which function of the sample X_1, X_2, X_n we should take?

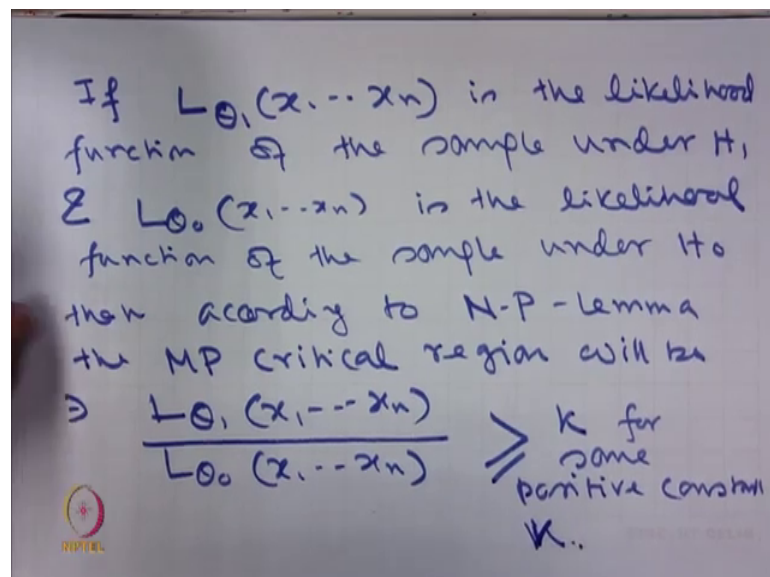
So, that based on that, we can make the decision of rejecting or accepting the null hypothesis and the corresponding test is most powerful. In this respect the most important result is called Neyman-Pearson Lemma. Today I will give you the Lemma and solve a few problems based on the Neyman-Pearson Lemma. In the next class I shall give you the proof of the Lemma so that you can understand how such a result has come.

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Neyman-Pearson Lemma: suppose we are testing two simple hypothesis H_0 that θ is equal to θ_0 versus H_1 that θ is equal to θ_1 and we have taken a sample x_1, x_2, \dots, x_n .

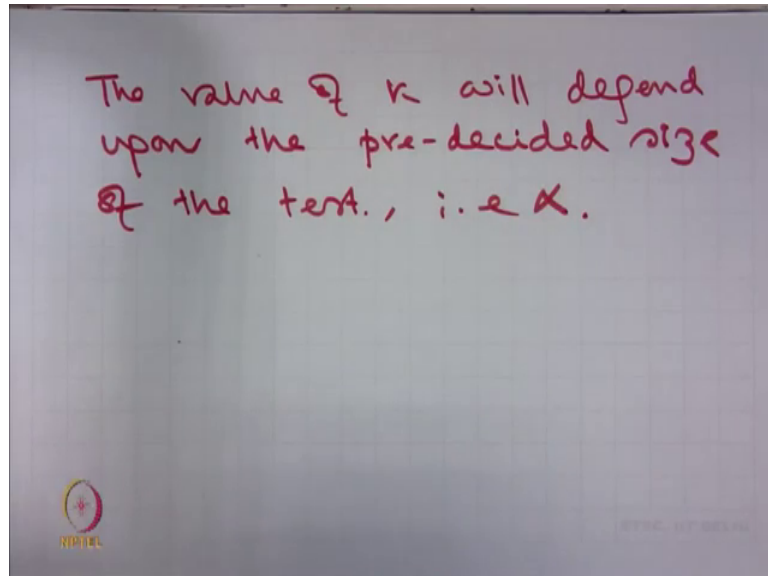
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If $L_{\theta_1}(x_1, x_2, \dots, x_n)$ is the likelihood function of the sample under H_1 and $L_{\theta_0}(x_1, x_2, \dots, x_n)$ is the likelihood function of the sample under H_0 . Then according to Neyman-Pearson Lemma the most powerful critical region will be such that

$L(\theta_1, x_1, x_2, \dots, x_n)$ upon $L(\theta_0, x_1, x_2, \dots, x_n)$ is greater than K for some positive K for some positive constant K .

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The value of K will depend upon the pre decided size of the test that is α . So, at the beginning as a statistician one will decide upon the size α whether it is 5 percent or 1 percent or 10 percent. And based on that we should get a constant such that $L(\theta_1, x_1, x_2, \dots, x_n)$ upon $L(\theta_0, x_1, x_2, \dots, x_n)$ has to be greater than that constant. Before proving that as I said in today's class I will solve a few problems using the above Neyman-Pearson Lemma.

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Ex-1 Suppose Bernoulli (p) distribution & we are testing $H_0: p = 0.5$ or $\frac{1}{2}$ vs $H_1: p = 0.75$ or $\frac{3}{4}$

\therefore By NP-Lemma, the most powerful CR will be

$$\frac{L_{3/4}(x_1, \dots, x_n)}{L_{1/2}(x_1, \dots, x_n)} > K.$$

where x_1, \dots, x_n is the sample.

Let us consider the first example suppose we are tossing a coin. So, we are looking at Bernoulli p distribution and we are testing say $H_0: p$ is equal to 0.5 or half versus $H_1: p$ is equal to 0.75 or 3 by 4. Therefore, by Neyman-Pearson Lemma, the most powerful region critical region will be $L_{3/4}$ of x_1, x_2, \dots, x_n upon $L_{1/2}$ of x_1, x_2, \dots, x_n is greater than K where x_1, x_2, \dots, x_n is the sample.

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Now $L_{3/4}(x_1, \dots, x_n)$
 $= \left(\frac{3}{4}\right)^{\sum x_i} \left(\frac{1}{4}\right)^{n - \sum x_i}$

Similarly
 $L_{1/2}(x_1, \dots, x_n)$
 $= \left(\frac{1}{2}\right)^{\sum x_i} \left(\frac{1}{2}\right)^{n - \sum x_i}$

This we have seen as
 $P(X=1) = P^x (1-P)^{1-x}$
where $x = 0$ or 1
which comes from Bernoulli trial.

Now, $L_{3/4}$ of x_1, x_2, \dots, x_n is equal to $\left(\frac{3}{4}\right)^{\sum x_i} \left(\frac{1}{4}\right)^{n - \sum x_i}$. This we have seen as probability X is equal to 1 is equal to P to

the power x_1 minus P to the power $1 - x_1$ where x_1 is equal to 0 or 1 which comes from Bernoulli trial. Similarly, $L_{1/2}$ of x_1, x_2, \dots, x_n is equal to half to the power $\sum x_i$ into half to the power $n - \sum x_i$.

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$$\begin{aligned} &\therefore \frac{L_{3/4}(x_1, \dots, x_n)}{L_{1/2}(x_1, \dots, x_n)} \\ &= \frac{\left(\frac{3}{4}\right)^{\sum x_i} \left(\frac{1}{4}\right)^{n - \sum x_i}}{\left(\frac{1}{2}\right)^{\sum x_i} \left(\frac{1}{2}\right)^{n - \sum x_i}} \geq K \\ \text{Or } &\left(\frac{3}{4}\right)^{\sum x_i} \left(\frac{1}{4}\right)^{n - \sum x_i} \geq K \cdot \left(\frac{1}{2}\right)^n \end{aligned}$$

Therefore, $L_{3/4}$ of x_1, x_2, \dots, x_n upon $L_{1/2}$ of x_1, x_2, \dots, x_n is equal to $3/4$ to the power $\sum x_i$ into $1/4$ to the power $n - \sum x_i$ into half to the power $\sum x_i$ into half to the power $n - \sum x_i$. And that has to be greater than equal to K . Or $3/4$ to the power $\sum x_i$ $1/4$ to the power $n - \sum x_i$ is greater than equal to K times half to the power n .

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$$\text{Or } \frac{\left(\frac{3}{4}\right)^{\sum x_i}}{\left(\frac{1}{4}\right)^{\sum x_i}} \times \left(\frac{1}{4}\right)^n \geq K \left(\frac{1}{2}\right)^n$$
$$\text{Or } 3^{\sum x_i} \geq K \cdot 2^n$$

By taking \log_2 :

$$\sum x_i \log_2 3 \geq \log_2 K + n$$
$$\text{Or } \sum x_i \geq \frac{\log_2 K + n}{\log_2 3} \leftarrow C.$$

Or $3^{\sum x_i} \geq K \cdot 2^n$ is greater than equal to $K \cdot 2^n$. Or $3^{\sum x_i}$ is greater than equal to $K \cdot 2^n$. By taking log base 2, we get $\sum x_i \log_2 3 \geq \log_2 K + n$ or $\sum x_i$ is greater than equal to $\frac{\log_2 K + n}{\log_2 3}$. This is a constant, let me call it C.

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∴ The CR will be of the form $\sum x_i \geq c$

So in general the CR for Bernoulli (θ) will be from NP-Lemma as follows:

Therefore, the critical region will be of the form $\sum x_i \geq c$. Or in other words, we will reject the null hypothesis that the probability of a

head for this coin is half, we will reject that in favor of that the probability of getting a head is 3 by 4, if $\sum x_i$ or the number of heads is greater than some constant; which is very intuitive because the alternative here is 3 by 4 which is bigger than the null hypothesis value that is half.

So, in general the critical region for Bernoulli p will be from NP Lemma as follows.

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$$\frac{p_1^{\sum x_i} (1-p_1)^{n-\sum x_i}}{p_0^{\sum x_i} (1-p_0)^{n-\sum x_i}} \geq K$$

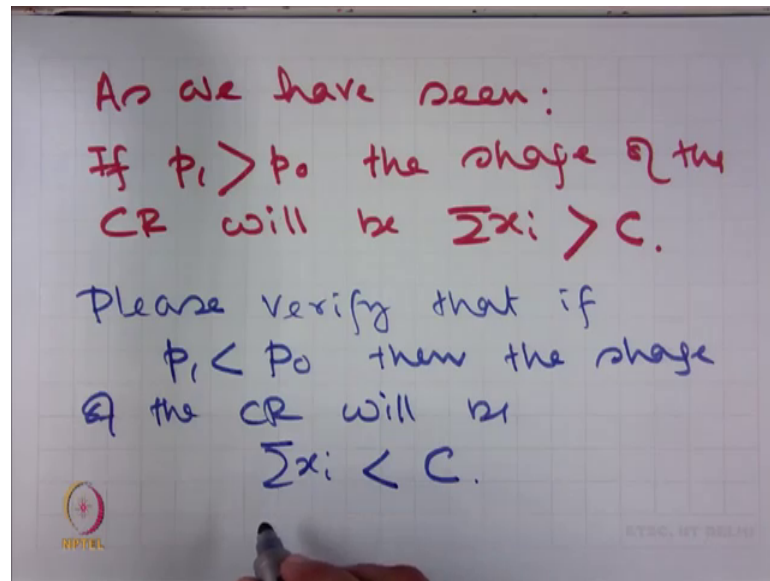
$$\text{or } \left(\frac{p_1}{p_0}\right)^{\sum x_i} \times \frac{(1-p_0)^{\sum x_i}}{(1-p_1)^{\sum x_i}} \geq \frac{(1-p_0)^n}{(1-p_1)^n} K$$

Depending upon whether $p_1 > p_0$ or $p_1 < p_0$ we shall get two different types of CR.

p_1 to the power $\sum x_i$ into $1 - p_1$ to the power $n - \sum x_i$ upon p_0 to the power $\sum x_i$ into $1 - p_0$ to the power $n - \sum x_i$ is greater than equal to K . Or, p_1 upon p_0 $\sum x_i$ into $1 - p_0$ $\sum x_i$ upon $1 - p_1$ $\sum x_i$ is greater than equal to $1 - p_0$ to the power n upon $1 - p_1$ to the power n times K .

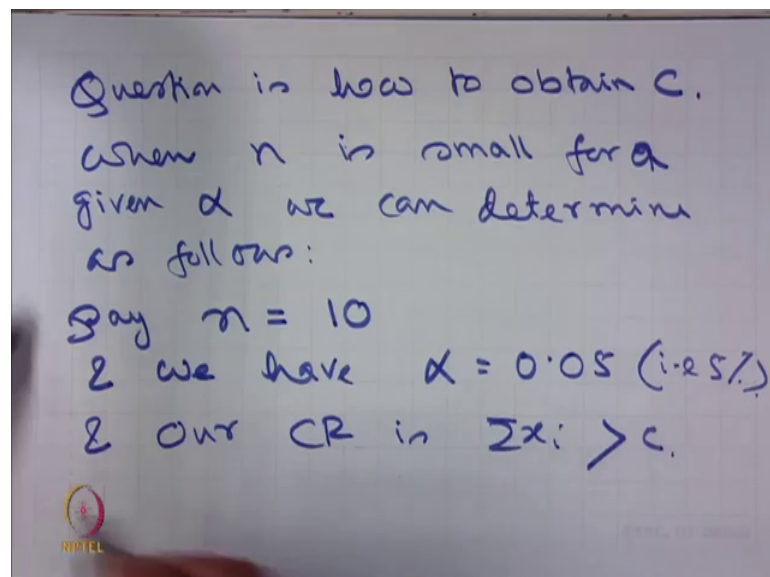
This is the general structure. So, depending upon whether p_1 is greater than p_0 or p_1 is less than p_0 we shall get two different types of Critical Region as we have seen.

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If p_1 is greater than p_0 the shape of the critical region will be $\sum x_i$ is greater than some constant. I will leave it on you, you verify that if p_1 is less than p_0 , then the shape of the critical region will be $\sum x_i$ is less than some constant C .

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The question is how to obtain C ? When n is small for a given α , we can determine this easily as follows say n is equal to 10. And we have α is equal to say 0.05 that is 5 percent and with respect to the earlier problem our CR is $\sum x_i$ is greater than sum C .

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Under H_0 : i.e. $p = \frac{1}{2}$

$$P(\sum x_i = 10) = \frac{1}{2^{10}}$$
$$P(\sum x_i = 9) = \binom{10}{1} \frac{1}{2^{10}} = \frac{10}{2^{10}}$$
$$P(\sum x_i = 8) = \binom{10}{2} \frac{1}{2^{10}} = \frac{10!}{8! \cdot 2!} \frac{1}{2^{10}}$$
$$= \frac{45}{2^{10}}$$
$$\therefore P(\sum x_i \geq 8) = (1 + 10 + 45) \frac{1}{2^{10}}$$

Now, under H_0 that is p is equal to half probabilities $\sum x_i$ is equal to 10 is equal to half to the power 10 probability $\sum x_i$ is equal to 9 is equal to $10 C 1$ half to the power 10 is equal to 10 divided by 2 to the power 10 probabilities $\sum x_i$ is equal to 8 is equal to $10 C 2$ half to the power 10 is equal to factorial 10 upon factorial 8 factorial 2 into half to the power 10 is equal to 45 upon 2 to the power 10.

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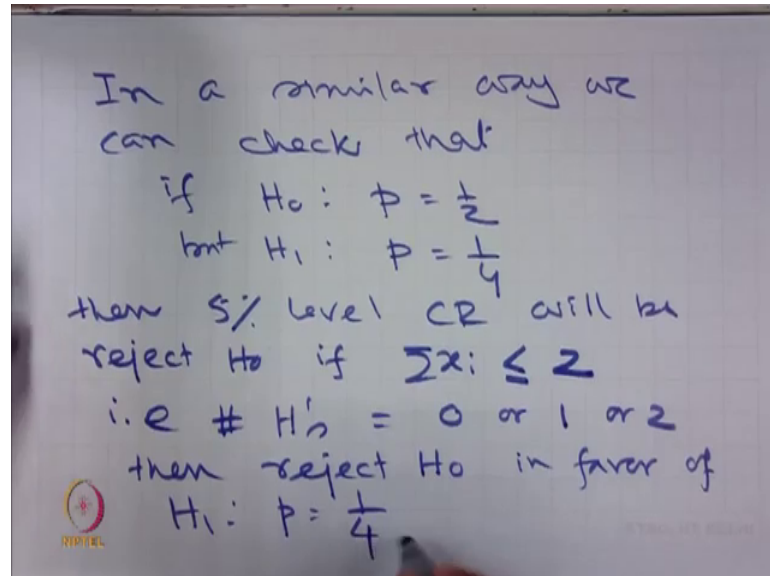
$= \frac{56}{1024} \approx 0.05$

\therefore wrt the problem
Our CR with $\alpha = 0.05$
to test $p = \frac{1}{2}$ against $p = \frac{3}{4}$
i.e. $\sum x_i \geq 8$

Therefore, probability $\sum x_i$ greater than equal to 8 is equal to 1 plus 10 plus 45 up to the power 10 is equal to 56 upon 1024 which is roughly equal to 0.05.

Therefore, with respect to the problem our critical region with alpha is equal to 0.05 to test p is equal to half against p is equal to $\frac{3}{4}$ is $\sum x_i$ is greater than equal to 8. Or if number of it is obtained is 8 or more, then we are going to reject the null hypothesis that the coin is unbiased otherwise we are going to accept that this coin is unbiased when testing against p is equal to $\frac{3}{4}$.

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In a similar way, we can check that if H_0 is p is equal to half. But H_1 is p is equal to $\frac{1}{4}$, then the 5 percent level critical region will be reject H_0 , if $\sum x_i$ is less than equal to 2 that is number of heads is equal to 0 or 1 or 2 then reject H_0 in favor of H_1 p is equal to $\frac{1}{4}$.

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If the no. of samples is large then we can make Normal approximation.

Say $n = 100$

\therefore under H_0 : $E(\# \text{ Heads}) = 50$

$\sum \text{Var}(\sum X_i) = 100 \times \frac{1}{2} \times \frac{1}{2} = 25$

$\therefore \frac{\sum X_i - 50}{\sqrt{25}} \sim N(0, 1)$

If the number of samples is large, then we can make normal approximation say n is equal to 100. Therefore, under H_0 expected number of heads is equal to 50 and variance of $\sum X_i$ is equal to 100 into half into half is equal to 25. Therefore, $\sum X_i$ minus 50 upon root over 25 may be approximated as normal with 0 1.

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Therefore the 5% CR is

$$\left| \frac{\sum X_i - 50}{5} \right| \geq 1.65$$

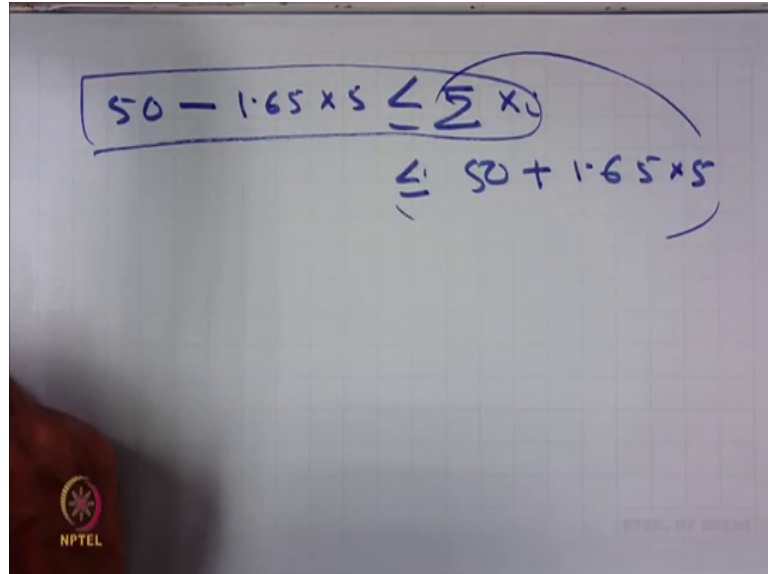
\therefore Depending upon $p_i > p_0$ or $p_i < p_0$ we shall decide the acceptance or rejection of H_0 is

This can be obtained from the Normal Table.

Therefore, the 5 percent critical region is modulus of $\sum X_i$ minus 50 upon 5 is greater than equal to 1.65. This is, this can be obtained from the normal table. Therefore,

depending upon p_1 is greater than p_0 or p_1 is less than p_0 we shall decide the acceptance or rejection of H_0 .

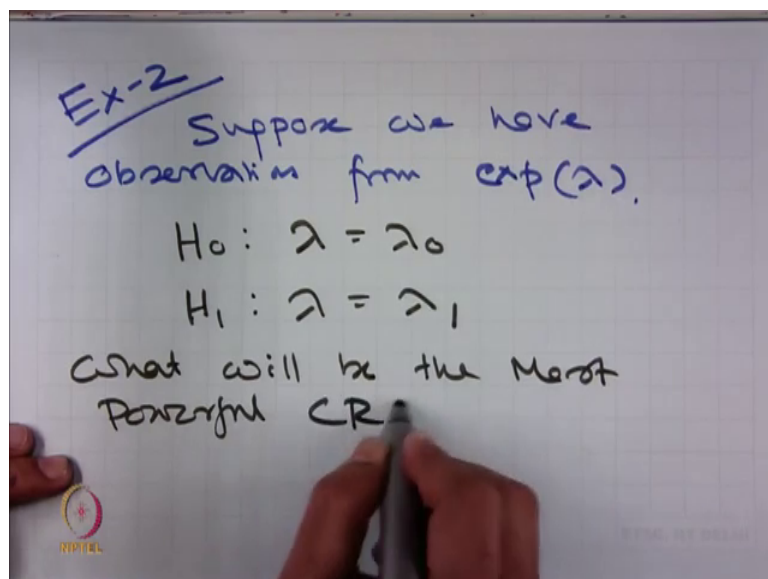
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A whiteboard with a grid pattern. The text is handwritten in blue ink. The top line is $50 - 1.65 \times 5 \leq \sum x_i$, with a large blue circle around the entire expression. The bottom line is $\leq 50 + 1.65 \times 5$, also with a large blue circle around the entire expression. In the bottom left corner, there is a small circular logo with the text 'NPTEL' below it.

If $50 - 1.65 \times 5$ is less than or equal to $\sum x_i$ is less than or equal to $50 + 1.65 \times 5$ and we know that if p_1 is greater than p_0 , then this should be the condition. If p_1 is less than p_0 then this should be the condition. Therefore the shape of the critical region will be obtained accordingly. Let us now consider a second example.

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A whiteboard with a grid pattern. The text is handwritten in blue ink. The top line is 'Ex-2' with a horizontal line underneath. The second line is 'Suppose we have observations from $\exp(\lambda)$.' The third line is $H_0: \lambda = \lambda_0$. The fourth line is $H_1: \lambda = \lambda_1$. The fifth line is 'What will be the Most Powerful CR'. In the bottom left corner, there is a small circular logo with the text 'NPTEL' below it. A hand holding a marker is visible at the bottom center.

Suppose, we have observations from exponential lambda and our H naught is lambda is equal to lambda naught and H 1 is lambda is equal to lambda 1, what will be the most powerful critical region?

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Using NP Lemma:

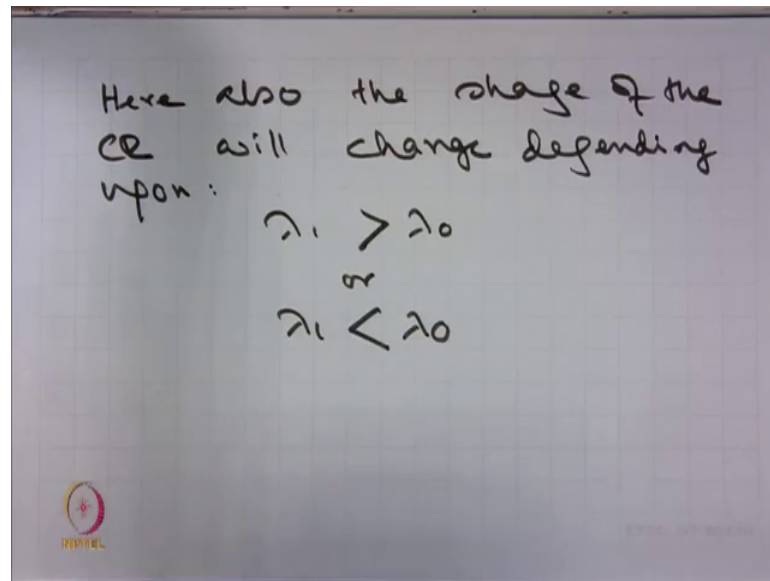
$$\frac{L_{\lambda_1}(x_1, \dots, x_n)}{L_{\lambda_0}(x_1, \dots, x_n)} \geq K$$

$$\text{or } \frac{\lambda_1^n e^{-\lambda_1 \sum x_i}}{\lambda_0^n e^{-\lambda_0 \sum x_i}} \geq K$$

$$\text{or } \left(\frac{\lambda_1}{\lambda_0}\right)^n e^{-(\lambda_1 - \lambda_0) \sum x_i} \geq K$$

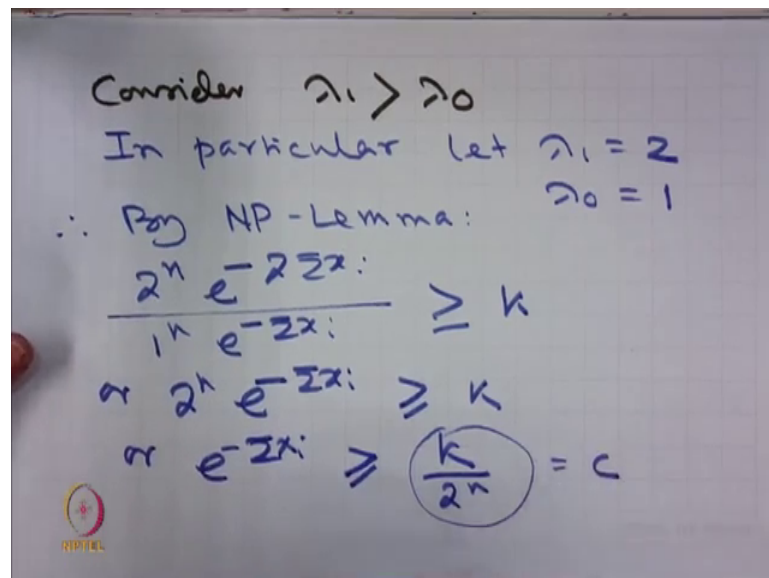
Using, Neyman-Pearson Lemma L_{λ_1} of x_1, x_2, \dots, x_n L_{λ_0} of x_1, x_2, \dots, x_n is greater than equal to some positive constant K or $\lambda_1^n e^{-\lambda_1 \sum x_i}$ upon $\lambda_0^n e^{-\lambda_0 \sum x_i}$ is greater than equal to K or λ_1^n upon λ_0^n to the power n $e^{-\lambda_1 \sum x_i}$ upon $e^{-\lambda_0 \sum x_i}$ is greater than equal to K or $\left(\frac{\lambda_1}{\lambda_0}\right)^n e^{-(\lambda_1 - \lambda_0) \sum x_i}$ is greater than equal to K .

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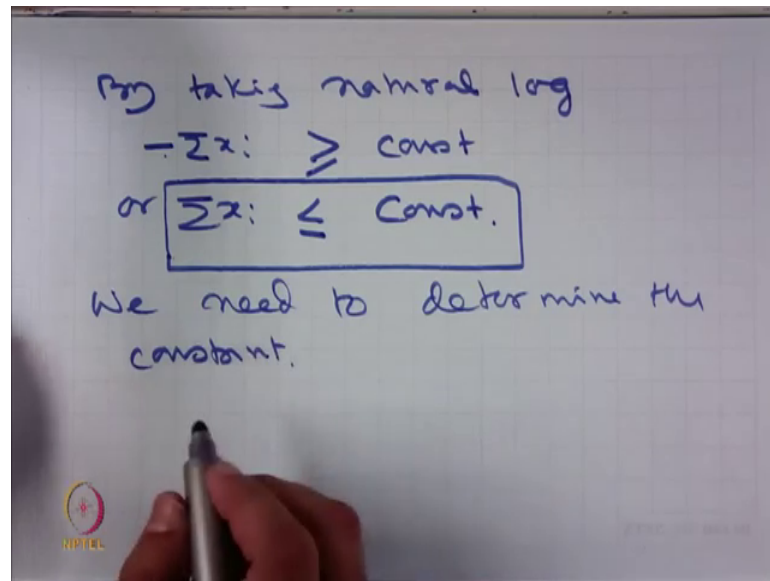
Here, also shape of the critical region will change depending upon lambda 1 is greater than lambda 0 or lambda 1 is less than lambda 0.

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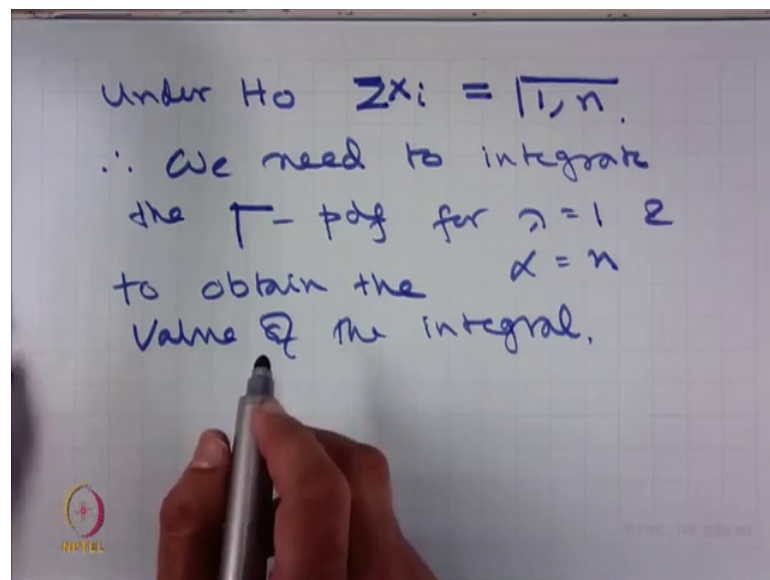
Consider lambda 1 is greater than lambda 0 in particular let lambda 1 is equal to 2 lambda 0 is equal to 1. Therefore, by NP Lemma $2^n e^{-2 \sum x_i} / 1^n e^{-\sum x_i} \geq K$ or $2^n e^{-\sum x_i} \geq K$ or $e^{-\sum x_i} \geq \frac{K}{2^n} = c$

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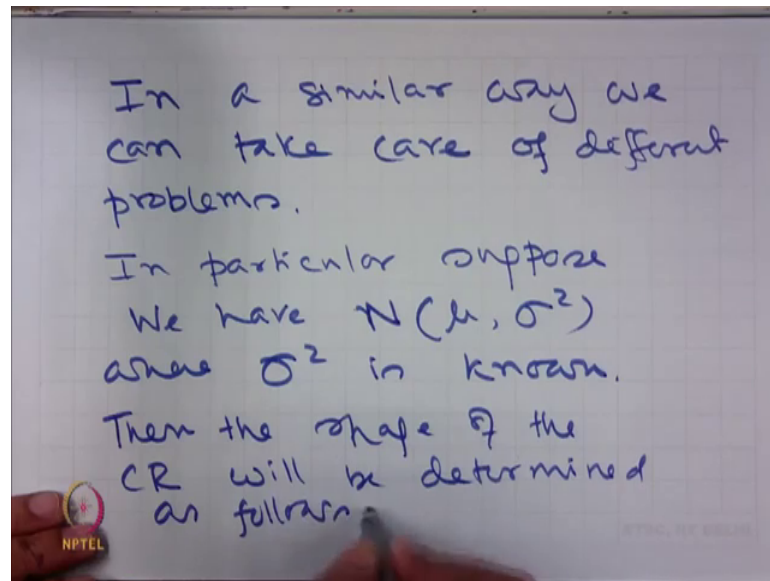
Let us call it a constant equal to C by taking natural log minus sigma x_i is greater than equal to some constant or sigma x_i is less than equal to some constant because this is negative there is a change in the inequality. Now, we need to determine the constant.

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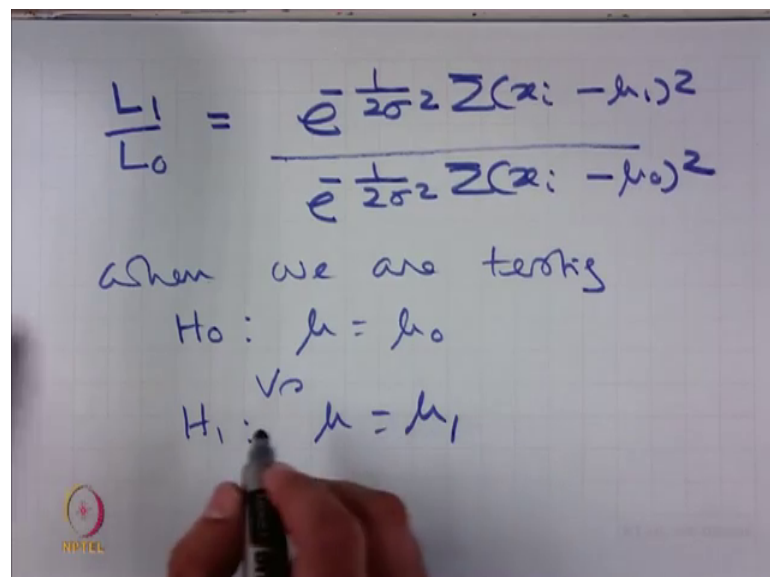
Under H_0 sigma x_i is equal to gamma 1 comma n. Therefore, we need to integrate or the gamma pdf for lambda is equal to 1 and alpha is equal to n to obtain the value of the integral.

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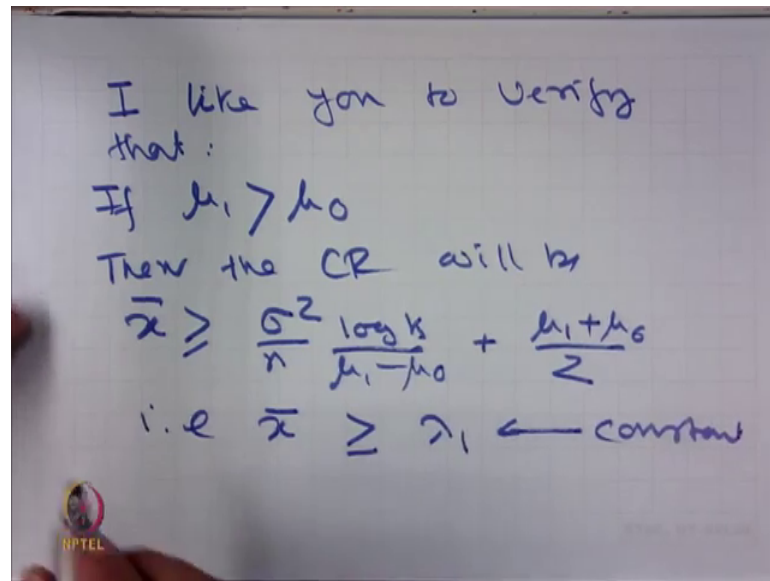
In a similar way we can take care of different problems. In particular, suppose we have normal μ comma σ^2 where σ^2 is known, then the shape of the critical region will be determined as follows.

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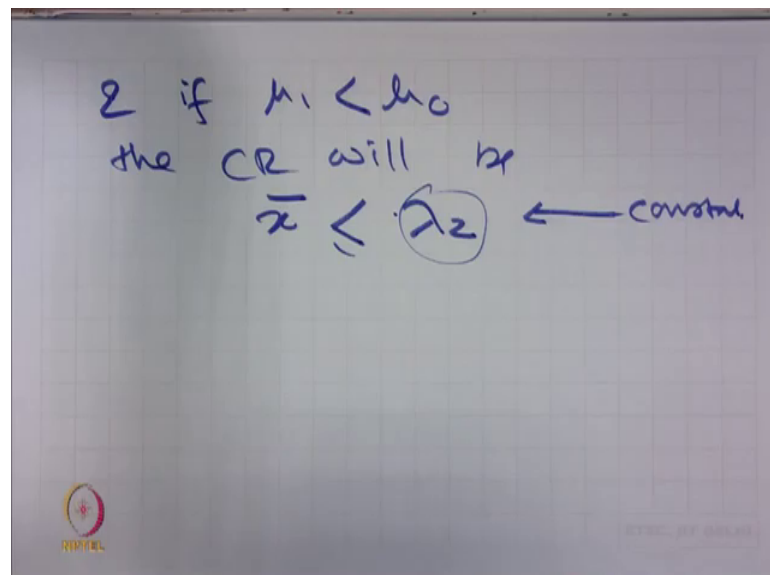
So, L_1 upon L_0 is equal to e to the power minus $\frac{1}{2\sigma^2}$ into $\sum (x_i - \mu_1)^2$ upon e to the power minus $\frac{1}{2\sigma^2}$ into $\sum (x_i - \mu_0)^2$. When, we are testing $H_0: \mu = \mu_0$ versus $H_1: \mu = \mu_1$.

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I like you to verify that, If μ_1 is greater than μ_0 , then the critical region will be \bar{x} is greater than equal to $\frac{\sigma^2}{n} \log$ of $K \mu_1$ minus μ_0 plus μ_1 plus μ_0 by 2. That is \bar{x} has to be greater than equal to λ_1 which is a constant.

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If μ_1 is less than μ_0 , the CR will be \bar{x} less than equal to λ_2 constant. And we can obtain the value of λ_1 or λ_2 by using the normal table.

So, this is how we decide whether to accept a simple null hypothesis against a simple alternative hypothesis where the distributions are known. If the statistic is such that we know it is pdf and we get its table then we can obtain the value from there otherwise we will have to integrate the pdf or you have to numerically obtain the value of the threshold. So, that depending upon whether the statistics is greater than that or less than that we can take a decision of accepting or rejecting the null hypothesis.

Friends, I stop here today. In the next class I shall prove the Neyman-Pearson Lemma and also solve a few problems to understand the method of testing of hypothesis.

Thank you.