

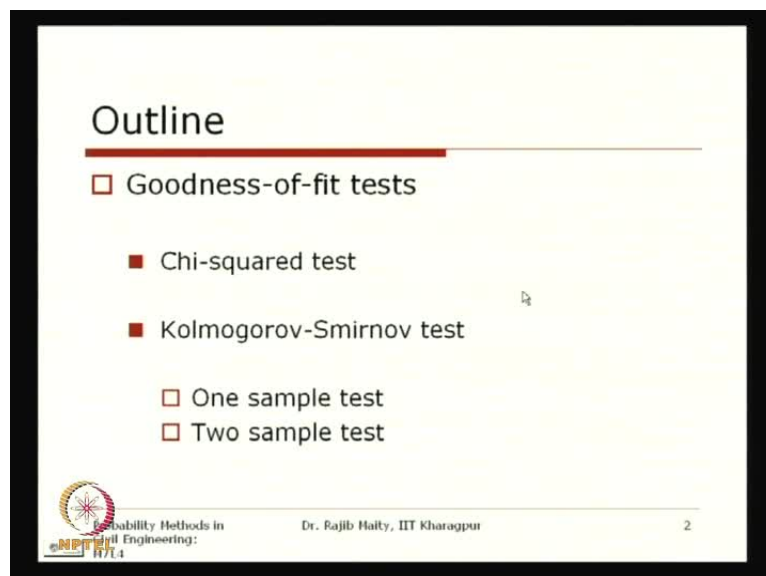
Probability Methods in Civil Engineering
Prof.Dr.RajibMaity
Department of Civil Engineering
Indian Institute of Technology, Kharagpur

Lecture No.# 38
Goodness - of – fit tests

Hello and welcome to this lecture; in this lecture, we will learn some of the statistical methods to test the observed data, whether they are fitting to some particular distribution or not. You know that in earlier lectures, as well as in earlier modules also, we have referred several times that while handling some problem, we generally assume that the dataset that is following a particular distribution and we have seen now, we will see that how we can test these things that whether the dataset is really following this distribution or not.


There are some of the statistical test that I am just going to tell in this in this lecture is that, so to use that one and use the knowledge of this hypothesis testing will be testing it statistically that how the data is fitting to a particular module. So, there are some tests and so we will just pick up some of this popular test to see that how we can test this goodness of fit test, these tests are known as the goodness of fit test, so, this is, so that is our today's lecture title is the Goodness of fit test.

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Outline

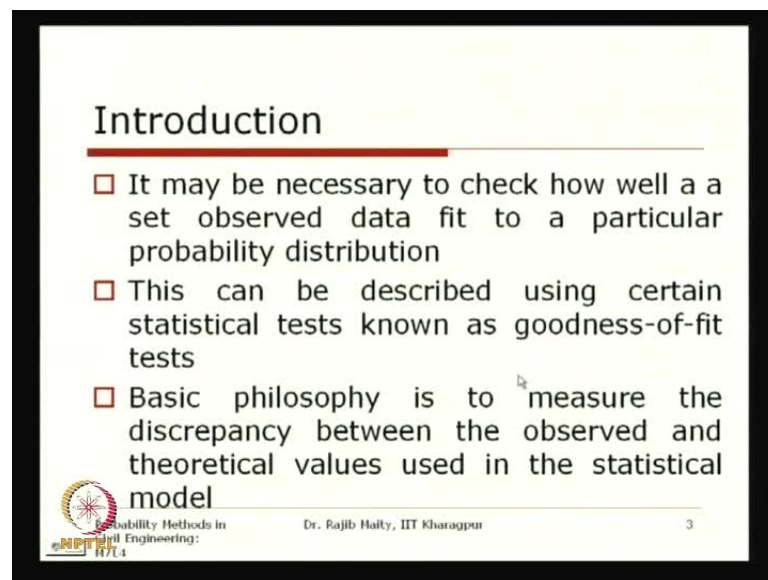
- Goodness-of-fit tests
 - Chi-squared test
 - Kolmogorov-Smirnov test
- One sample test
- Two sample test

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So, first we will just see that how this can work, and after that we will take two tests for this lecture. One is the chi square test, this will be chi square, not d; d is a mistake, that is a chi square test and the other one is the Kolmogorov-Smirnov test, it is also known as the K-S test as an abbreviated form. And this K-S test again can be that one sample test and two sample test. Basically, when we take that this one sample test, then we generally test it for one sample that we take and we generally take that from whether that particular dataset follows a particular distribution or not.

So, in this case, one is our sample and the other one is the standard distribution and when you go for the two sample test, it is basically both of our sample data is there and we look for the answer whether both the samples are from the same population or not, same population means, they are following the same distribution or not. So, this is how we go for one sample test and this two sample test. As you know that for this kind of statistical test, we need some kind of significance level and some statistical significance level. So, whenever we generally conclude or we draw some decision from whatever the statistical test we do, we have generally that decision is associated with statistical significance level, so that is important.

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Introduction

- It may be necessary to check how well a set of observed data fits to a particular probability distribution
- This can be described using certain statistical tests known as goodness-of-fit tests
- Basic philosophy is to measure the discrepancy between the observed and theoretical values used in the statistical model

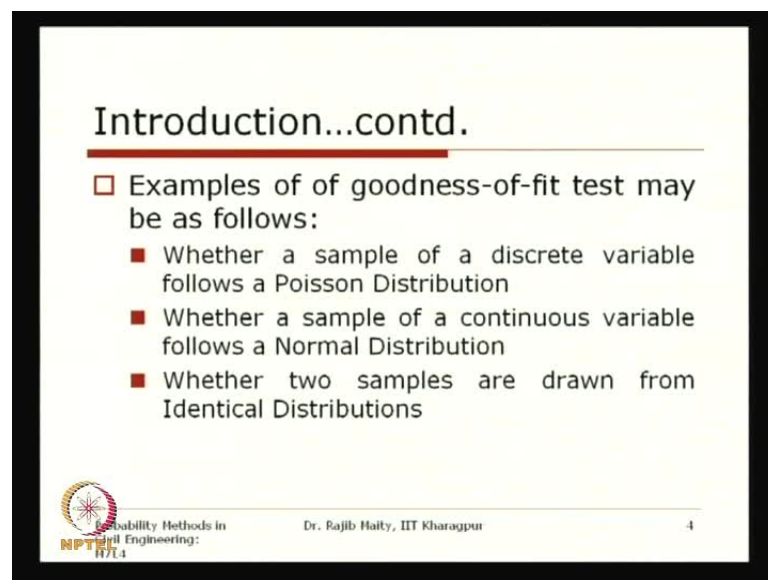
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So, what significance level that we are considering either beforehand that **ok**, at this significance level, whatever we are going to test is satisfied or not in a statistical sense that is what we will do. Well, so now, first, we will go to this, whatever I just mentioned so

far is that, it may be a necessary to check how well a set of observed data fit to a particular probability distribution, and this one, this can be described using the certain statistical test known as the goodness-of-fit test. And as also, I just try to indicate that the basic philosophy is to measure the discrepancy between the observed and theoretical values used in the statistical model. So, this basically are the general thing for all the test that we going to describe now, **that** and this one what we are describing is particularly for the one sample test.

Suppose, when we are saying that I have a dataset and that follows a particular distribution, so what is our interest is that, whether there is any discrepancy between that whatever the observed distributions that we can see from this data. And whatever the distribution that we are assuming to follow, whether it is normal or log normal or even in the discrete side Poisson or so, whether those distribution that theoretical values and this whatever the observed from this data are matching or not. So, that discrepancy we have to see, and that discrepancy we have to decide **which** through some distribution, we have to draw some inference in a statistical way.


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Introduction...contd.

- Examples of of goodness-of-fit test may be as follows:
 - Whether a sample of a discrete variable follows a Poisson Distribution
 - Whether a sample of a continuous variable follows a Normal Distribution
 - Whether two samples are drawn from Identical Distributions

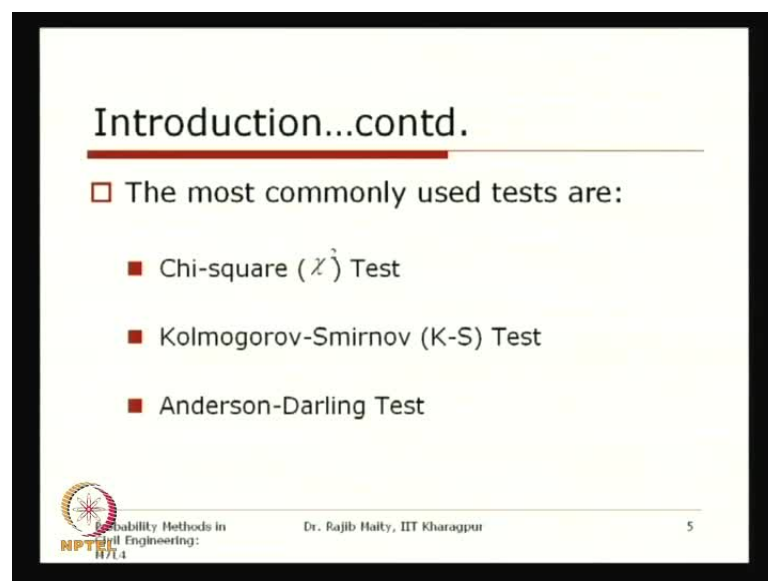
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So, the example of this goodness-of-fit test may be as follows, say, **thus** that what type of question that we are looking for their answer, say that, whether a sample of a discrete variable follows a Poisson distribution or not. So, these are just the example, it is not that always I am looking for this Poisson distribution or so. So, any distribution, so I know that

this random variable is a discrete variable, so I have some hypothesis that whether that sample can follow a particular discrete distribution that we know already, and those distribution we have discussed in earlier lectures, in earlier modules. So, like that question, I have a sample whether that sample follow a Poisson distribution or not.

Similarly, whether a sample of a continuous variable follows a normal distribution or not, or gamma distribution or not, or log normal distribution or not like that, or in case of the two samples, whether both the samples are drawn from the identical distribution or not. So, when we are testing it, we will just take try to take that all this several types of this problems, where we will try to cover almost all these different cases discrete variable case, continuous variable case, as well as the two samples case taking from two different sample we will take and whether we will test that whether they are from the identical distribution or not.

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The slide is titled "Introduction...contd." and lists the most commonly used tests for goodness-of-fit. It includes a red square icon next to the title and a list of three tests, each preceded by a red square bullet point. The footer contains the NPTEL logo, the text "Probability Methods in Civil Engineering: IIT Kharagpur", the name "Dr. Rajib Maiti, IIT Kharagpur", and the slide number "5".

Introduction...contd.

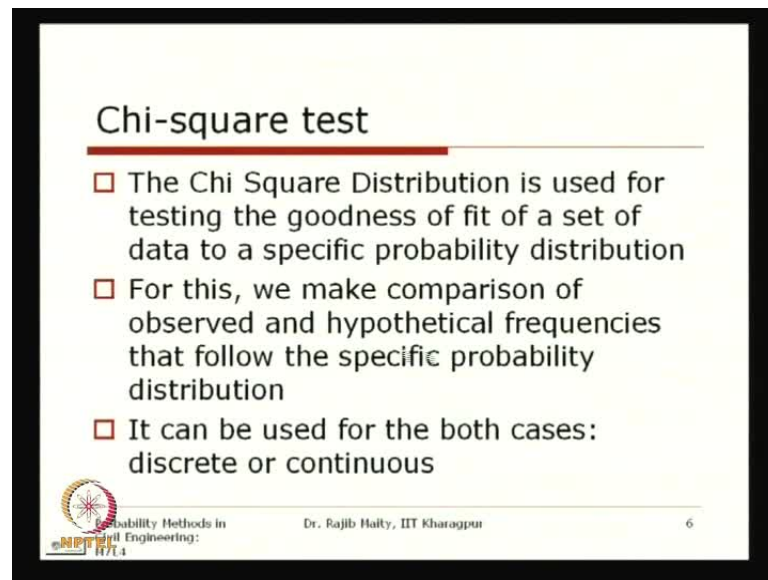
□ The most commonly used tests are:

- Chi-square (χ^2) Test
- Kolmogorov-Smirnov (K-S) Test
- Anderson-Darling Test

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So, as we mentioned, there are some of these commonly used tests, are there these are standard tests that we generally use for this goodness-of-fit test. The first one is this, chi square test and then, the Kolmogorov-Smirnov test, there is also another test which is an Anderson-Darling test which is generally known to be a little improvement over this K-S test. And we will take this one also, but maybe for this lecture we will just consider this chi square test and K-S test.

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Chi-square test

- ❑ The Chi Square Distribution is used for testing the goodness of fit of a set of data to a specific probability distribution
- ❑ For this, we make comparison of observed and hypothetical frequencies that follow the specific probability distribution
- ❑ It can be used for the both cases: discrete or continuous

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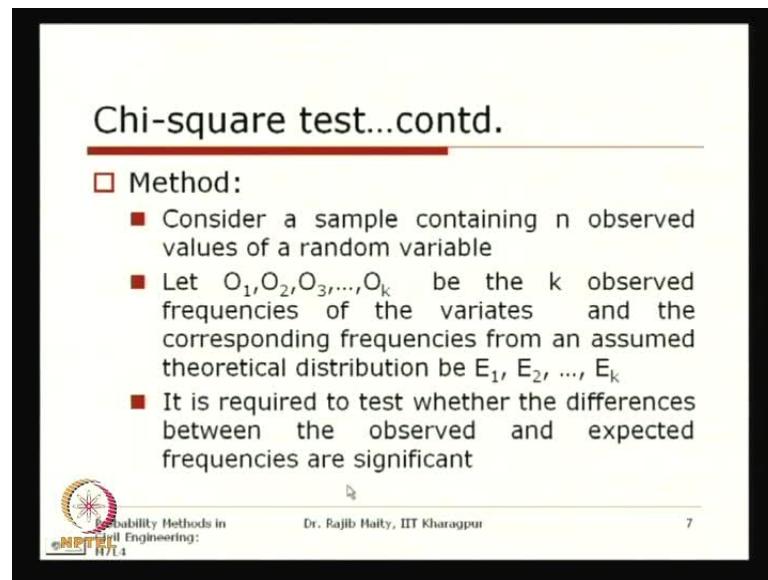
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Now, this chi square test, the chi square distribution is used for testing the goodness of fit of a set of data to a specific probability distribution. For this, we make the comparison of the observed and hypothetical frequencies that follow the specific probability distribution, so this is **that** what we have discussed for this basic philosophy. And this is basically, if you see that overall approach is same for all this test, so the comparison of this whatever you have observed from this data and whatever we are suppose to get from that hypothesize distribution, whether they are matching or not.

So, here basically this chi square test is based on their frequencies, so we will find out the observed frequency and also see, what is the hypothetical frequency and obviously, as we are talking about these frequencies, so we have to categorize the data into different bins and each bin what is the frequency, what is the observed frequency? This observed frequency means, whatever the data that we are having, based on that, what is the frequency that we can see, and we are hypothesize one distribution. So, based on that distribution obviously, that distribution should have some parameters, with that parameters, what are the frequency that we can observed. And this one can be used for the both the cases; means, both for this discrete random variable as well as for the continuous random variable, we can use this test.

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Chi-square test...contd.

- Method:
 - Consider a sample containing n observed values of a random variable
 - Let $O_1, O_2, O_3, \dots, O_k$ be the k observed frequencies of the variates and the corresponding frequencies from an assumed theoretical distribution be E_1, E_2, \dots, E_k
 - It is required to test whether the differences between the observed and expected frequencies are significant

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So, what is done that may be we will discuss now in a **step in** different steps, first of all let us consider that a sample contains n observed values of the random variable, so whatever the data that we are having that is **that** n numbers of data that we are having. And now this notation, that is, O_1, O_2, O_3 up to O_k , be the K observed frequencies of the variates and the corresponding frequencies from the assumed theoretical distribution be E_1, E_2, E_3 up to E_k . So, this is that observed, that is, this is from whatever that data that we are having is the $O_1, O_2, O_3, \dots, O_k$ and these $E_1, E_2, E_3, \dots, E_k$, these are from **that** whatever the distribution, **that** we are hypothesize that this dataset may follow that particular distribution. So, from that distribution whatever the frequency that we are getting is, **E_i 's** and whatever from the data that is O_i 's.

So, this it is required to test whether the difference between the observed and the expected frequencies are significant or not. So, this are the $O_1, O_2, O_3, \dots, O_k$ and these are from the hypothesize distribution, now they are difference, whether this difference from the observed and to the theoretical is significant, then we can say that whatever the data we are having is not following whatever the distribution that we have **(())**. And on the other hand, if they are almost same, then we can say that, yes, that it is following that particular distribution.


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Chi-square test...contd.

- Thus the goodness is checked as:

$$\chi^2 = \sum_{i=1}^k \frac{(O_i - E_i)^2}{E_i}$$

- As n approaches infinity, the sampling distribution of χ^2 tends to a χ^2 distribution with $\nu (=k-1)$ is the degrees of freedom

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Now, so, how to do that one, how to find out that discrepancy is through this parameter, we generally call it as a statistics and that statistics is obtained as that O minus E whole square divided by E and sum of are all K S means, all this groups that is from 1 to K. Now, this quantity **this quantity** is, as I told that this is the statistics and denoted by here χ^2 square that we have denoted this one, and it has been found that this χ^2 square or this statistics, it follows a chi square distribution. If n tends to infinity, that is, **that** that mathematical limit for the large number of this sample data, that is available, this statistics follow a chi square distribution and this chi square distribution is having some degrees of freedom, here the degrees of freedom is k minus 1.


So, this one shows, as we know that this particular statistics follows the chi square distribution with k minus 1 degrees of freedom; that means, we know that what are its properties, and based on that at what significance level I want to test whether, **this** **is means significantly**, this is significant or not that we have to test. One thing is, I should mention here that sometimes it is required, for example that when we are calculating this E_i , it may require to estimate some of the parameter of that hypothesized distribution from the data itself, if that parameter is not known. And in that case, as many parameters as we are required to estimate from this data then, those many degrees of freedom will be lost. So, if you are not estimating any parameter from that data available, that time this the degree of freedom is k minus 1, but if we estimate one parameter then, this degrees of

freedom will be k minus 2, if you are estimating 2 parameters then degrees of freedom will be k minus 3.

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Chi-square test...contd.

- **Property :**
 - The χ^2 test for goodness of fit should generally have $k \geq 5$ and $E_i \geq 5$
- **Special case:**
 - Most of the cases we may not have the parameters of theoretical distribution
 - Then the parameters should be estimated from the data itself and the statistic remain valid if the degree of freedom is reduced by one for every unknown parameter


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So, similarly, so as many parameters, you are estimating from the data those many more degrees of freedom will be lost. So, now this one some more things, some more properties, these are not that, means, directly I cannot say that why these are required, but for the better results that is within quote, that for the better result, this we should take care before we go ahead with the test. The first thing is that, the K should be greater than equal to 5, so that number of a groups that we have divided that observed data it should be greater than equal to 5 and that E_i is that is the frequencies that we got in each beam should be at least 5, so this E_i should be greater than equal to 5.

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Chi-square test...contd.

- Assuming that the distribution follows as:
$$\sum_{i=1}^k \frac{(O_i - E_i)^2}{E_i} < C_{1-\alpha, \nu}$$
where $C_{1-\alpha, \nu}$ is value of approximate χ^2_ν distribution at cumulative probability $1-\alpha$
- The assumed theoretical distribution is an acceptable model at the significance level α

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Now, the special case, the most of the cases we may not have the parameters of the theoretical distribution and then, the parameters should be estimated from the data itself, and the statistics remain valid if the degree of freedom is reduced by one for every unknown parameter. So, that is what I just explained, so if you need to estimate the parameter of the hypothesized distribution from the data then, you need to **then you need to** allow that much degrees of freedom will be lost. Now, assuming that the distribution follows as this one that I told, there will be square here, as you have seen earlier this statistics, this observed minus, this hypothesized square divided by E_i and their summation. So, if this one is less than this value, what is shown is, $C_{1-\alpha, \nu}$ as it is already explained that, this is the degrees of freedom, this $C_{1-\alpha, \nu}$, that is $C_{1-\alpha}$ is a value of approximate that this chi square distribution with ν degrees of freedom at the cumulative probability $1-\alpha$.

Now, this α is that significance level that we are talking about, the assumed theoretical distribution is an acceptable model at the significant level of this α . So, this is what that I **was** mention initially, that all these test should be associated with some significant level. Generally, this significance level are kept around, say, 0.01 or, sorry, 0.01 or 0.05. So, at this cumulative probability one minus this α , so if it is that 5 percent significance level if we say; that means that cumulative probability at which we are testing it is at the 95 percent. So, if the observed statistics is less than **is less than** that cumulative probability of that distribution, here it is the chi square distribution of this 95

percent if the alpha is 0.05 significance level and then, we can say, yes, that whatever we have hypothesized may not be rejected.

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Chi-square test...contd.

- Thus the goodness is checked as:

$$\chi^2 = \sum_{i=1}^k \frac{(O_i - E_i)^2}{E_i}$$

- As n approaches infinity, the sampling distribution of χ^2 tends to a χ^2 distribution with $\nu (=k-1)$ is the degrees of freedom

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Example

Q. Consider a given station in a watershed where the severe rainstorms are recorded over a period of 70 years. Out of this 70 years, 22 years were without severe rainstorms and 25, 14, 6, 3 years with 1, 2, 3 and 4 rainstorms annually. Test whether the data can be assumed to follow Poisson distribution at 5% significance level.

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So, yes, as I told that this should be the square, as we have seen it here also, this statistics, this y minus E power square. Now, we will take up one example and this example, we have taken for the discrete case and because, this chi square distribution can be used for the discrete parameters. So, it can be used for the continuous as well as discrete, but the other than next one what we are going to cover is,

the K S test that is for the continuous distribution, so here we are taking a discrete example.

Consider a given station in a watershed, where the severe rainstorms are recorded over a period of 70 years. In the last 70 years, we have recorded the how many rainstorms are there in a particular year. And out of these 70 years, 22 years were without severe rainstorms, so there are in 22 years there are no severe rainstorms, so number rainstorms is 0. And 25 years, 14 years, 6 years and 3 years are with 1 rainstorm, 2 rainstorms, 3 rainstorms and 4 rainstorms respectively. So, 25 years we have observed there is 1 rainstorm, 14 years we have seen there are 3 rainstorms, 16 years 3 rainstorms, and 3 years, there are 4 rainstorms are there. Test whether the data can be assumed to follow a Poisson distribution at 5 percent significance level. So, here as you can see that we are hypothesizing that whether the data that is given that whatever we have recorded, whether it is following the Poisson distribution or not, and the significance level is given as 5 percent.

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Example...Contd.

Sol.:


Average occurrence rate of rain storms:

$$\lambda = \frac{1}{70} (25 \times 1 + 14 \times 2 + 6 \times 3 + 3 \times 4)$$

$$= 1.1857 \text{ rainstorms / year}$$

Now to check the goodness of fit, we use the chi-square distribution at $\alpha = 5\%$ significance level

As the data is so small, the data of four storms/year is combined to data of three storms/year



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So, here now you see, so the Poisson distribution you know there is some lambda, there is one parameter that we have discussed earlier modulus, that parameter that lambda is the mean rate of occurrence. So, rate so average occurrence rate of this rainstorm, so if you want to calculate then we have seen that there are 22 years where there is no rainstorms. So, 22 multiplied by 0, basically, the first thing, then 25 years 1 rainstorm, 14 years 2 rainstorm, 6 years 3 rainstorm and 3 years 4 rainstorms are there are total. So, in this way

out of these 70 years what are how many total rainstorms that we have seen that divided by 70 gives you that average rainstorm that can occur in a year, so 1.1857 rainstorms you can see in a year. So, λ here is that 1.1857 and the λt - that is the total quantity that t is here the per year, so one year how much rainstorms has occurred, so this is the quantity 1.1857.

Now remember that, we have estimated this one from the data itself, so this was not supplied. Sometimes what happens? these data could have been supplied directly, whether the test, whether the data is following the Poisson distribution with the λ is equal to, say, 1.3 or 1.1 like that. If that is given to us that means, we are not estimating that one from the data, so there the degrees of freedom whatever I told that, it is $k - 1$ will be there; but here, we have already estimated one parameter, so now the degrees of freedom will be $k - 2$.

So, now to the check, the goodness-of-fit, we use the chi-square distribution at α equals to 5 percent significance level, as the data is so small, the data for the 4 storms per year is combined with the 3 storms per year. So, this one is generally done, but so there is some discussion is required. We can Now, we have to think that what distribution you have to hypothesize, so it is the Poisson distribution that you have hypothesize and what is the support for that distribution. Now, the Poisson distribution that support that we are looking for is basically starting from the 0 to 0, 1, 2 the discrete values and goes up to infinity.

Now, the data that is obviously is given to us is that 1 rainstorm, 2 rainstorm, 3 rainstorm up to the 4 rainstorm per year, but when we are hypothesizing that this is the Poisson distribution of obviously, I cannot scuttle at any particular point. So, here what is the general practice is that for the higher side where data is becoming very small, we can combine those things to take care two things, one is that so I can declare that, yes, more than equal to this value is having this frequency and the second thing is that we will also be testing whether that each group is having that minimum requirement of this frequency is available or not.

Because this kind of asymptotic distribution, that is, it is going towards plus infinity, so this type of it should be open bounded, but the way the problem is given it is just as a close boundary at 4 rainstorm per year. So, here if we just consider that 2 rainstorms are


combined together, that is, the 3rainstorm per year and 4rainstorm per year case, then we can say that will greater than equal to 3rainstorm per year. So, that remains that, positive side remains unbounded and also, this will help to check whether that at least greater than 5, that is, for the better result that we have mentioned earlier.

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Example...Contd.

- Null hypothesis H_0 : The random variable has a Poisson distribution with $\lambda=1.1897$
- Alternate hypothesis H_1 : The random variable does not follow the distribution specified in null hypothesis.
- Level of significance: $\alpha = 0.05$
- We have $k=4$; degree of freedom, $v=k-2=2$
- Critical region:

$$\chi_v^2 \geq \chi_{2,0.05}^2$$



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So, here also what we are doing is that the 4storms per year and the 3rainstorms per year are combined together. So, here the null hypothesis is the random variable has a poisson distribution with lambda equals to 1.1897; alternate hypothesis is the random variable does not follow the distribution specified in null hypothesis, what is specified here? Significance level is 0.05 at 5 percent significance level mentioned. And we have k equals to 4, so the degrees of freedom here are the k minus 2, so k minus 1 is from there and one parameter we have estimated, so it is k minus 2, the degree of the freedom is 2. So, that critical region here is this chi square distribution, that is, chi square distribution with 2 degrees of freedom at α equals to 0.05.

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Example...Contd.

Cumulative Chi-square Distribution with v degrees of freedom: $F(x^2) = \frac{1}{2} \int_0^{x^2} t^{-v/2} e^{-t/2} dt$ entry in table is $F(x^2)$

$F =$	0.001	0.005	0.010	0.025	0.050	0.100	0.200	0.500	1.000	2.000	5.000	10.000	20.000	50.000	100.000
1	0.000002	0.000044	0.000090	0.000175	0.000353	0.000675	0.001348	0.002707	0.005413	0.010826	0.021651	0.043302	0.086604	0.173208	0.346416
2	0.000020	0.000090	0.000183	0.000359	0.000718	0.001436	0.002872	0.005744	0.011488	0.022976	0.045952	0.091904	0.183808	0.367616	0.735232
3	0.000071	0.000297	0.000594	0.001187	0.002374	0.004748	0.009496	0.018992	0.037984	0.075968	0.151936	0.303872	0.607744	1.215488	2.430976
4	0.000154	0.000615	0.001230	0.002460	0.004920	0.009840	0.019680	0.039360	0.078720	0.157440	0.314880	0.629760	1.259520	2.519040	5.038080
5	0.000261	0.001044	0.002088	0.004176	0.008352	0.016704	0.033408	0.066816	0.133632	0.267264	0.534528	1.069056	2.138112	4.276224	8.552448
6	0.000393	0.001572	0.003144	0.006288	0.012576	0.025152	0.050304	0.100608	0.201216	0.402432	0.804864	1.609728	3.219456	6.438912	12.877824
7	0.000540	0.002160	0.004320	0.008640	0.017280	0.034560	0.069120	0.138240	0.276480	0.552960	1.105920	2.211840	4.423680	8.847360	17.694720
8	0.000701	0.002804	0.005608	0.011216	0.022432	0.044864	0.089728	0.179456	0.358912	0.717824	1.435648	2.871296	5.742592	11.485184	22.970368
9	0.000877	0.003444	0.006888	0.013776	0.027552	0.055104	0.110208	0.220416	0.440832	0.881664	1.763328	3.526656	7.053312	14.106624	28.213248
10	0.001067	0.004180	0.008360	0.016720	0.033440	0.066880	0.133760	0.267520	0.535040	1.070080	2.140160	4.280320	8.560640	17.121280	34.242560
11	0.001271	0.004968	0.009936	0.019872	0.039744	0.079488	0.158976	0.317952	0.635904	1.271808	2.543616	5.087232	10.174464	20.348928	40.697856
12	0.001490	0.005808	0.011616	0.023232	0.046464	0.092928	0.185856	0.371712	0.743424	1.486848	2.973696	5.947392	11.894784	23.789568	47.579136
13	0.001724	0.006704	0.013408	0.026816	0.053632	0.107264	0.214528	0.429056	0.858112	1.716224	3.432448	6.864896	13.729792	27.459584	54.919168
14	0.001973	0.007664	0.015328	0.030656	0.061312	0.122624	0.245248	0.490496	0.980992	1.961984	3.923968	7.847936	15.695872	31.391744	62.783488
15	0.002237	0.008688	0.017376	0.034752	0.069504	0.139008	0.278016	0.556032	1.112064	2.224128	4.448256	8.896512	17.793024	35.586048	71.172096
16	0.002516	0.009776	0.019552	0.039104	0.078208	0.156416	0.312832	0.625664	1.251328	2.502656	5.005312	10.010624	20.021248	40.042496	80.084992
17	0.002809	0.010928	0.021856	0.043696	0.087392	0.174784	0.349568	0.699136	1.398272	2.796544	5.593088	11.186176	22.372352	44.744704	89.489408
18	0.003117	0.012144	0.024288	0.048576	0.097152	0.194304	0.388608	0.777216	1.554432	3.108864	6.217728	12.435456	24.870912	49.741824	99.483648
19	0.003441	0.013424	0.026848	0.053632	0.107264	0.214528	0.429056	0.858112	1.716224	3.432448	6.864896	13.729792	27.459584	54.919168	109.838336
20	0.003781	0.014768	0.029536	0.058976	0.117952	0.235904	0.471808	0.943616	1.887232	3.774464	7.548928	15.097856	30.195712	60.391424	120.782848

$\chi^2_{2,0.05} = 5.99$

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Now, if you see, so this is that chi square distribution table, basically, you know that we have discussed this distribution earlier and these standard values are listed in any standard text book. So, here if you see that **this point** at this 0.95 these that cumulative probability are given here and for these degrees of freedom 2, this value is 0.5, 0.99, so these value is our critical value, that we have to test against.

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Example...Contd.

Thus from the table we get

$$\chi^2_{2,0.05} = 5.99$$

No of storms/year	Observed frequency O_i	Theoretical frequency E_i	$(O_i - E_i)^2 / E_i$
0	22	21.3019	0.0229
1	25	25.3428	0.0046
2	14	15.0752	0.0767
3	9	8.2801	0.0626
Total ->	70	70	0.1668

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So, this is that 5.99 that we got from this table and **these are the** these are the beans here now, so 0 rainstorm per year, 1 rainstorm per year, 2 rainstorm per year and this is the


greater than equal to 3 rainstorm per year. So, now, as it is given in this data, that there are 22 years where this 0 rainstorms are there and the 25 years are there is that 1 rainstorm, 14 years 2 rainstorm and 6 years 3 rainstorm and that 3 years 4 rainstorms are there, so we have combined it here to get the 9 years data, where it is greater than equal to 4 rainstorm per year.

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Example...Contd.

- Null hypothesis H_0 : The random variable has a Poisson distribution with $\lambda = 1.1897$
- Alternate hypothesis H_1 : The random variable does not follow the distribution specified in null hypothesis.
- Level of significance: $\alpha = 0.05$
- We have $k=4$; degree of freedom, $v=k-2=2$
- Critical region:

$$\chi^2_v \geq \chi^2_{2,0.05}$$

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Example...Contd.

□ Thus from the table we get

$$\chi^2_{2,0.05} = 5.99$$

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Total ->	70	70	0.1668

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So, we have avoided that one, in one occurrence it is becoming 3 which is less than 5, so that is also avoided and that right side is kept open. And this theoretical frequencies, that


is, these values what we are getting is that this from this Poisson distribution that lambda t here equals to this lambda equals to that 1.1897 and t equals to 1, so 1 year, so this lambda t value we will get, and from this Poisson distribution if you just put that x equals to 0, x equals to 1, x equals to 2 and x equals to 3 with that value of lambda t equals to that value, then we will get these values, which are the, so are the theoretical frequencies for this Poisson distribution.

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Example...Contd.

- From the table we get

$$\sum_{i=1}^k \frac{(O_i - E_i)^2}{E_i} = 0.1668 < 5.99 (= \chi^2_{2,0.05})$$
- Hence the Poisson distribution is a valid model at 5% significance level
- Decision: The null hypothesis can not be rejected at 5% significance level



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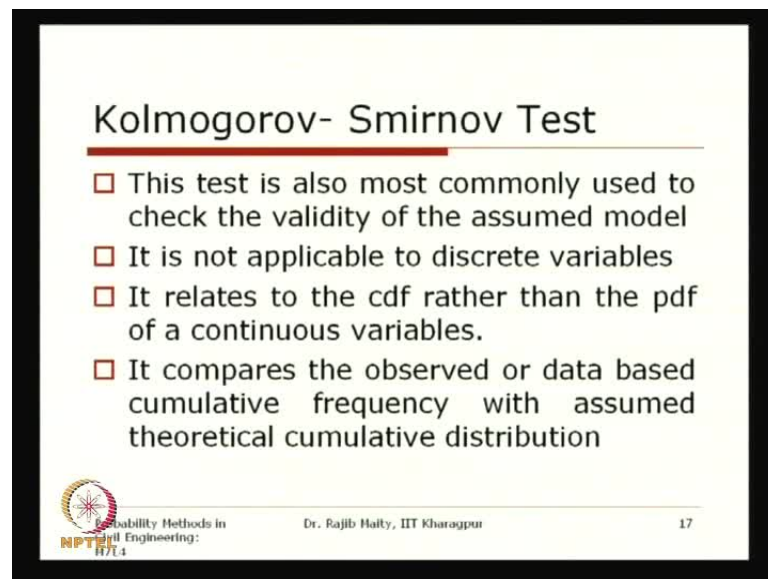
Now, so these are so these values we are getting when we are putting this x equals to 0, and this value you are putting when you are putting x equals to 1 from this Poisson distribution, so this is the theoretical frequencies and these are the observed frequencies. Now, what we have to do? We have to get this statistics, that is, y minus Ei whole square divided by Ei, so if we do this one then we get these values and we have to sum it up this one, so this is the summation from is that 0.1668.

So, these one this is our statistics which is equals to 0.1668 that we have seen from the table. So, that so, which is now is the less than 5.99 that we have seen from the table, so this is less than this this critical value, so hence the Poisson distribution is a valid model at 5 percent significance level. So, the decision is that for this test is that the null hypothesis cannot be rejected at 5 percent significance level.

So, there could be some other words that we can express, ok, the null hypothesis is not rejected and all, so what I feel is that this should be the proper decision,


proper probabilistic or the proper statistical inference should be written as this one that null hypothesis is accepted is not the right thing to declare. So, rather the complete thing that we can declare is that the null hypothesis cannot be rejected at 5 percent significance level because, the result that we have got it depends on what significance level that we have opted for. So, that is what we have to mention that at this significance level the null hypothesis cannot be rejected.

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Kolmogorov- Smirnov Test

- This test is also most commonly used to check the validity of the assumed model
- It is not applicable to discrete variables
- It relates to the cdf rather than the pdf of a continuous variables.
- It compares the observed or data based cumulative frequency with assumed theoretical cumulative distribution

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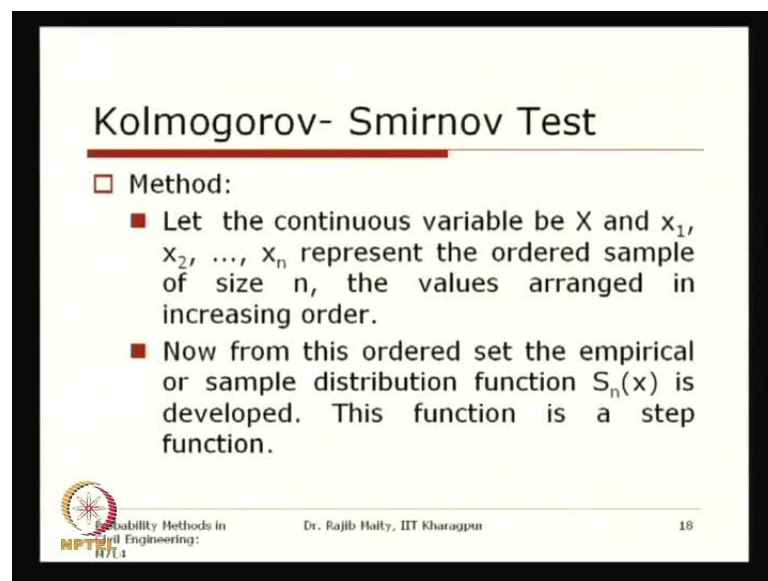
So, next we will go to that our second test, which is the kolmogorov-Smirnov test, in brief we generally mention here K S test. This test is also most commonly used to check the validity of the assumed model, and it is not applicable for the discrete variables, so mostly that continuous random variable if the dataset is available with that we generally get this one. It relates to the cdf rather than the pdf to a continuous variable, so here earlier what we have seen **that** the frequency that when we are talking about in this chi square test that is basically a pdf that we are talking about. Here, this K S test is generally based on this what is the cdf, that is cumulative distribution function, and it compares the observed or data based cumulative frequency with the assumed theoretical cumulative distribution, so data based cumulative distribution, sorry, this should be distribution with the assumed theoretical cumulative distribution.

So, there are some kinds of discrepancies or shortcomings I should say, in this chi square test is that, **we need to define that**, we need to first of all get that **that** means that we have to

first categorize the dataset, the full dataset that is, ok, this is my range. And in the discrete it is fine that whatever the example that we have seen, now **for the** as I told this chi square distribution is also applicable for the continuous distribution.

Now, if we take some continuous distribution in case of this chi square test, what we have to do, we have to first **we have to first** categorize the data into different bins and each bin we have to see the frequency and also we have to check that whether each bin is having. So, the number of bins should not be less than 5, as well as each bin should have **minimum** that minimum frequency should be 5, for the better results that we are mentioning. So, these two things are not there in this K-S test, so it is directly getting **it is directly the** access it **is** results from the cdf itself directly, so there is no need to categorize the data into bins. And so, it is avoiding those requirements of these minimum 5 bins and that each bin should have that minimum frequency of 5, so which is not there in this K-S test.


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Kolmogorov- Smirnov Test

□ Method:

- Let the continuous variable be X and x_1, x_2, \dots, x_n represent the ordered sample of size n , the values arranged in increasing order.
- Now from this ordered set the empirical or sample distribution function $S_n(x)$ is developed. This function is a step function.

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Now, so, if suppose, that there is a random variable x and we have some dataset of this X_1, X_2, X_3 up to X_n , so represent the ordered sample of size n . So, whatever the data that we are having from this actual observation, **we can** first of all we can make it **in an** arranged in an increasing order, and **that increasing order** if that increasing order is that X_1, X_2, \dots, X_n if that is available. Now, from this ordered set the empirical or sample distribution function $S_n(x)$ is developed and this function is basically a step function.

So, this is from this data, so from this sample how to get that this cumulative distribution function is as follows.


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Kolmogorov- Smirnov Test

- Thus the cumulative frequency step function is defined as:

$$S_n(x) = \begin{cases} 0 & x < x_1 \\ \frac{k}{n} & x_k \leq x < x_{k+1}; \quad k = 1, 2, \dots, n-1 \\ 1 & x \geq x_n \end{cases}$$

- $S_n(x)$ is the step function and $F(x)$ is the proposed theoretical distribution



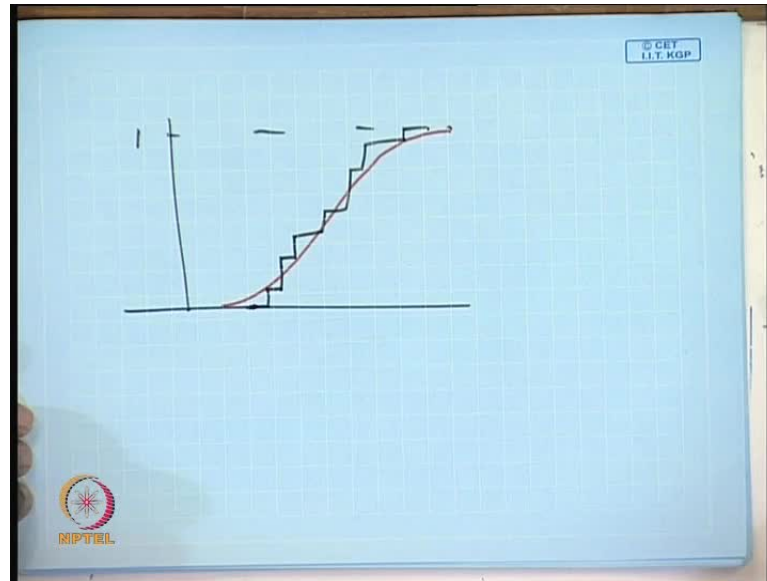
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That is for, when this $S_n(x)$ is equal to 0, in case, when x is less than x_1 , $S_n(x)$ is equal to k/n , when this x is in between k to $k+1$ and k can vary from 1, 2, 3 up to $n-1$. So, basically for all this x_1 to x_n for this thing we are defining these values, the value of this cumulative distribution is k/n . And for x greater than x_n is equal to 1. So, this is basically what we are getting is, whatever the representative, cumulative distribution directly from the data and each point it is changing, this value is changing, so it will look like a step function.

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So, obviously as you can see this essence **should** will start from a value 0, it will start from a value from 0 and at some point it will go, it will increase, and again from the next one, it should go and like this, it will go somewhere, it will go flat, and in this way what will happen? It will ultimately attain the value where it is 1.

Now, this is the distribution basically we got it from this data. Now, the hypothesized distribution, suppose that I want to match that **that** normal distribution, so that normal distribution with the parameters of course, whether it is supplied or obtained from this sample data **is that** with that data, I can also plot what should be the shape of this **that** particular distribution, say, that normal distribution if I say.


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Kolmogorov- Smirnov Test

- Thus the cumulative frequency step function is defined as:

$$S_n(x) = \begin{cases} 0 & x < x_1 \\ \frac{k}{n} & x_k \leq x < x_{k+1}; \quad k = 1, 2, \dots, n-1 \\ 1 & x \geq x_n \end{cases}$$

- $S_n(x)$ is the step function and $F(x)$ is the proposed theoretical distribution



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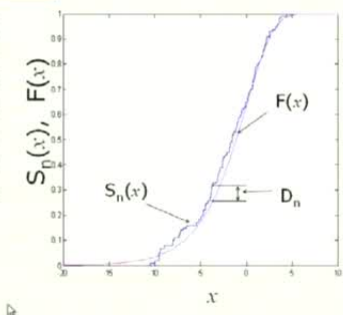
Now, if whatever we have observed the black line here, if this is very close to this red one which is the theoretical distribution obviously, then the data is from that particular distribution. So, this discrepancy, now again, keeping the same approach **same** for all that test that I mention at the beginning of this class, is that here also we have seen that what is the maximum difference between this two distribution that cdf, so one is the theoretical and the other one is the **other one is the** database. So, that discrepancy we have to assess through some statistical test, **this is what**, that is why we have got this $S_n(x)$.


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Kolmogorov- Smirnov Test

- The discrepancy between the theoretical model and the observed data is computed and the maximum difference (D_n) between $S_n(x)$ and $F(x)$ over the entire range of x is obtained

$$D_n = \max_x |F(x) - S_n(x)|$$





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So, this $S_n(X)$ is the step function and this F_X is the proposed theoretical distribution that what I have drawn in the reading, just now. Now, here if we see the discrepancies between the theoretical model and the observed data is computed, and the maximum difference between the D_n between the $S_n(X)$ and the F_X is over the entire range of X is obtained, which is which is denoted as D_n , which is the maximum for all X , F_X minus $S_n(X)$ obvious the absolute value.


So, now you can see here, for a particular point here as you can see, so from the blue one that I have drawn, this is also a step function here - the blue line, and this pink one is that your is that theoretical distribution. Now, the difference between any point to that theoretical distribution is your that value is the difference between two things. Now, what we have to pick up from this two information is that what is the maximum difference? So, at each point there will be some difference between this blue line and this and this pink line, so that difference at each and every point have to find out and we have to select, we have to pick up the maximum one, so what is the maximum difference. This blue line obviously, as we are coming leaps towards this, this is 0, and from where it is ending towards the right to that one it is equal to 1. And so over this entire range, we have to pick up what is the maximum difference.

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Kolmogorov- Smirnov Test

- Thus for a specified significance level α , the K-S test compares the maximum difference with the critical value D_n^α
- D_n^α is defined as:

$$P(D_n \leq D_n^\alpha) = 1 - \alpha$$
- If the observed value is less than the critical value, then the proposed distribution is valid at the significance level α .



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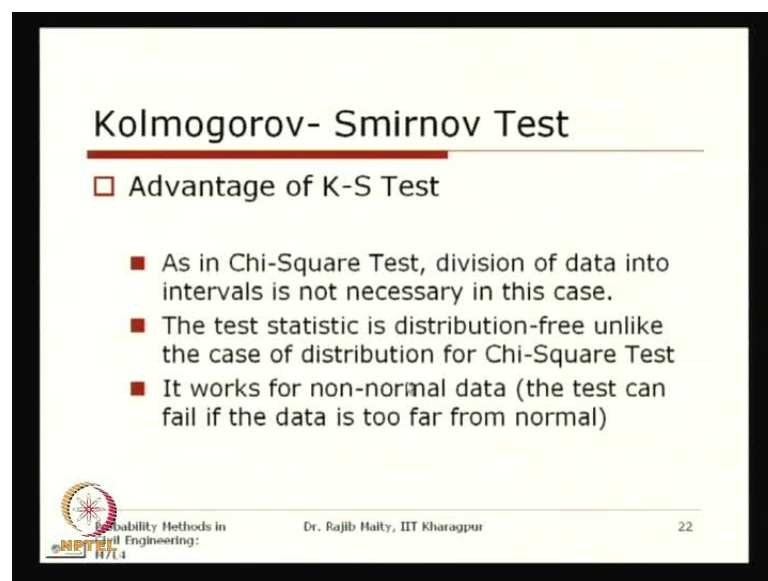
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So, thus for a specified significance level at α that K S test compares the maximum difference with the critical value D_n^α . Now, what is this D_n^α ? This D_n^α is

defined as the probability that $D_n \leq D_{n, \alpha}$ is equal to $1 - \alpha$, again this α here is that significance level that we have mentioned at the beginning of this class. So, if the observed value is less than the critical value, then the proposed distribution is valid at the significance level α . So, we have to check that whether whatever the maximum difference that we get, and whatever the critical value. So, this probability, that is, that observed $D_n \leq D_{n, \alpha}$ is equal to that critical value whether, it is equal to this $1 - \alpha$ or not.

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Kolmogorov- Smirnov Test

□ Advantage of K-S Test

- As in Chi-Square Test, division of data into intervals is not necessary in this case.
- The test statistic is distribution-free unlike the case of distribution for Chi-Square Test
- It works for non-normal data (the test can fail if the data is too far from normal)

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Now, the advantage of this K S test, as in the chi square test, division of the data into interval is not necessary in this case, so I think this thing I was just mentioning while at this starting of this K S test. So, **these** intervals are not necessary here, because we are just observing this at each and every data point. The test statistics is distribution free unlike in the case of the distribution of the chi square test, it works for this log normal data, however, the test can fail if the data is too far from this normality.

So, there is no such restriction that with the data should be approximate normal or so, but it is better to get the better result again that data should be somewhere near normality that is why the popularity of this K S test is, we have seen, it is very frequently, it is applied to test whether the data follow the normal distribution or not, that is, where the maximum application of this K S test has been found.


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Kolmogorov- Smirnov Test

- If the sample distribution n is large, Smirnov has given the limiting distribution of $\sqrt{n}D_n$ as:

$$\lim_{n \rightarrow \infty} [P(\sqrt{n}D_n \leq z)] = \left(\frac{\sqrt{2\pi}}{z} \right) \sum_{k=1}^{\infty} \exp \left[-\frac{(2k-1)^2 \pi^2}{8z^2} \right]$$
- For $n > 50$ for $\alpha = 0.05$ and 0.10

$$D_n^\alpha = 1.36/\sqrt{n} \text{ and } D_n^\alpha = 1.22/\sqrt{n} \text{ respectively}$$



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Now, the sample distribution, if the sample distribution that n is large, not the distribution sample size, if the sample size n is large, Smirnov has given the limiting distribution of square root n multiplied by this D_n , so this is that D_n that we have defined the maximum difference that multiplied by square root n , these quantity follow a distribution like this that limit n tends to infinity, probability of this quantity square root n multiplied by that D_n , the statistic less than equals to z is equals to square root 2π by z multiplied by summation of k equals to 1 to infinity, exponential minus $2k$ minus 1 square multiplied by π square by $8z$ square.

So, this is what it is giving is that for this n tends to infinity means, when this n is very large that time what we can get is that, how this D_n is varying is that through this, that we have to find out. **suppose that,** Suppose now, so it depends on this what is the significance level that we have fixed, suppose that this significance level if it is 0.05; that means, this probability is equals to 1 minus α that means this 0.95. Now, if we solve this right hand part with equal to that, what is that, 0.95, then we will get that is z becomes 1.36, if you see this quantity here inside this exponential term, that is, $2k$ minus 1 whole square multiplied by π square by $8z$ square.

So, this term basically is changing that is from this k equals to 1 to infinity, now k if k is 1, you can see that this is just minus of this quantity plus exponential of, if k becomes 2 then, it becomes 9, so minus 9, so exponential minus 9 times of this one and if k


becomes 3 then it is 25 times of this one, so basically if you just do just a hand calculation also you will see that, if you just consider the first term itself that is k equals to one only and remaining if you just ignore it then, also you will see for this, this is almost very closely matching with this one may be, it is just varying after third decimal or so.

So, for this n greater than 50, sometimes in some textbook can refer to that if n is greater than 35 itself, and for this α equals to 0.05; that means, this right hand side is equated with this 0.95 and if we just calculate what should be the value of this z , if we just consider only one value of k equals to 1 then, you will see that this z becomes very close to this 1.36. So, for this significance level α equals to 0.05 that critical value is 1.36 divided by square root of n , so that means this Z becomes 1.36 and this is that 1.22 square root of n . So, this critical value this should be, so our observed statistics that is D_n should be less than this particular value to declare that, at that significance level, we can accept that particular hypothesis or that null hypothesis cannot be rejected.

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Example

Q. The data of fracture toughness of plain concrete specimens made with burnt brick aggregates is shown in the table in next slide. The data appears to fall approximately a straight line on a Normal probability paper with $N(0.540, 0.051)$. Perform the Kolmogorov-Smirnov test at 5% significance level to statistically justify the assumption for the given data.



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And similarly, if you put that say α equals to 0.1; that means, this right hand side if we equate to... if with that 0.9 then, we will see that this quantity, this z is becoming 1.22, so the critical value is 1.22 divided by square root n and for these lower values of n , this is also available in the standard table from that distribution and we can refer to those tables to get these critical values.


So, we will take up one example just to discuss all these things, and here we have taken one example of this continuous random variable. And as I mentioned that this is mostly used when we are considering that when, whether the dataset is following normal distribution or not. So, that data of the fracture toughness of the plain concrete specimen made with the burnt brick aggregate is shown in the table in the next slide. That data appears to fall approximately a straight line on a normal probability paper, that if it falls approximately normal on a normal probability paper, there is a possibility of it may follow a normal distribution, and that the parameters are $\mu = 0.54$ and $\sigma = 0.051$.

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Example...Contd.

Fracture toughness (MPa√m) of plain concrete specimens (in increasing order)

m	K _{IC}	m	K _{IC}	m	K _{IC}
1	0.451	10	0.508	19	0.557
2	0.481	11	0.531	20	0.59
3	0.484	12	0.532	21	0.591
4	0.484	13	0.538	22	0.602
5	0.489	14	0.538	23	0.605
6	0.494	15	0.544	24	0.611
7	0.494	16	0.548	25	0.658
8	0.494	17	0.548		
9	0.502	18	0.551		



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To perform the Kolmogorov-Smirnov test at 5 percent significance level to statistically justify the assumption of the assumption for the given data, so data is supplied here that is the fracture toughness, which is having an unit of mega pascals square root of meter of the plain concrete specimen, and this is already arranged in an increasing order. So, you can see that, so that there are total 25 samples are there 1 to up to this 25 and this one that K_{IC} which is the notation for this fracture toughness is arranged in an increasing order, so from 0.451 to 0.658.

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Example

Q. The data of fracture toughness of plain concrete specimens made with burnt brick aggregates is shown in the table in next slide. The data appears to fall approximately a straight line on a Normal probability paper with $N(0.540, 0.051)$. Perform the Kolmogorov-Smirnov test at 5% significance level to statistically justify the assumption for the given data.



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Example...Contd.

Sol.:

- Null hypothesis H_0 : The random variable has a Normal Distribution
- Alternate hypothesis H_1 : The random variable does not have the specified distribution.
- Level of significance: $\alpha = 0.05$
- Critical region (from table in next slide) :

$$D_n^{\alpha} = D_{25}^{0.05} = 0.264$$


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So, we have to test that whether this data set is following the normal distribution or not, and normal distribution having the parameters 0.54 and 0.051. So, over null hypothesis here is the random variable, has a normal distribution with those parameters of course. Alternative hypothesis, the random variable does not have the specified distribution in this null hypothesis, level of significance is 0.05. And the critical region from the table just show, that is, here what the number of data is 25 that is available and significance level is 0.05.

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Example...Contd.

□ From table:

$$D_n^\alpha = D_{25}^{0.05} = 0.264$$

Values of $D_{n,\alpha}$ for the Kolmogorov-Smirnov goodness-of-fit test

n	$D_{n,0.10}$	$D_{n,0.05}$	$D_{n,0.02}$
10	0.369	0.409	0.457
11	0.352	0.391	0.437
12	0.338	0.375	0.419
13	0.325	0.361	0.404
14	0.314	0.349	0.390
15	0.304	0.338	0.377
16	0.295	0.327	0.366
17	0.286	0.318	0.355
18	0.279	0.309	0.346
19	0.271	0.301	0.337
20	0.265	0.294	0.329
21	0.259	0.287	0.321
22	0.253	0.281	0.314
23	0.247	0.275	0.307
24	0.242	0.269	0.301
25	0.238	0.264	0.295
26	0.233	0.259	0.290
27	0.229	0.254	0.284

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Now, if you see this table here, that is, these are the values of this K S test goodness-of-fit test. This is that for different n is listed in the first column 10, 11, 12 like this, and the second one is that, α is your alpha 0.05. So, this kind of table is available to any standard text book and here, if you just see this value is highlighted for this n equals to 25 and the alpha equals to 0.05, the value is 0.264, so from this table we have just picked up these the critical value of that test.

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Example...Contd.

□ The cumulative frequency of the given data is plotted in figure with respect to the equation of K-S Test

□ The theoretical distribution function of normal model is also shown

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Now, if we just do this one, do these same methods that we have explained; now whatever the data that we are having, we have plotted it for its distribution with this step function as shown in the blue line here. And the theoretical distribution for this normal distribution that cdf of this normal distribution with parameter 0.54 and 0.051 is shown in the line magenta, now whatever the maximum difference between these two that we have to pick up.

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Example...Contd.

- From the figure, the maximum discrepancy of two functions, $D_{\max} = 0.1348$ occurring at $K_{IC} = 0.5080 \text{ (MPa}\cdot\sqrt{\text{m}})$
- i.e. the maximum discrepancy 0.1348 is less than the critical value 0.264
- Therefore model $N(0.540, 0.051)$ is a valid model at 5% significance level, in other words, the null hypothesis can not be rejected at 5% significance level



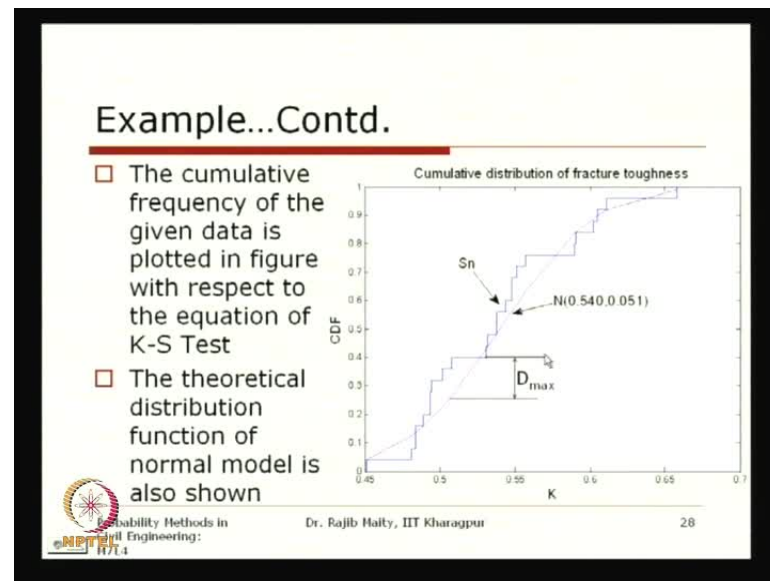
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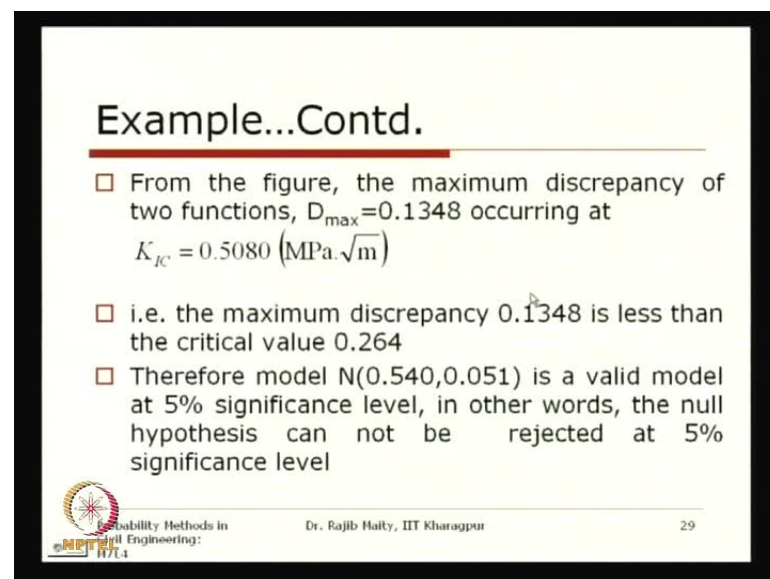
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So, the cumulative frequency of the given data is plotted in this figure with respect to the equation of this K S test and the theoretical distribution function of the normal model is also shown, what is shown here. From the figure and of course, you can **you can** check it in this calculation also, the maximum discrepancy of the two functions is d_{\max} equals to 0.1348 which is occurring at K_{IC} equals to 0.508, so at this value the D_{\max} is 0.1348.

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So, this is the only value **that we can** that we have to pick up from this comparison, this is what that K-S test and sometimes from this point onwards, may be that further improvement, we will look for that, but that is the later part. But here from this graph, only one information that we are picking up is, what is the maximum difference, just one particular value we have to pick up and that value is 0.1348. The maximum discrepancy is 0.1348, now we can test that it is less than that critical value that we have seen that also **which is** the critical value is 0.264 that we have seen from the table.

So, what we can see here, so this model is a normal distribution with parameter 0.54 and 0.051 is a valid model at 5 percent significance level, in other words or I should say that this should be used to declare that the null hypothesis cannot be rejected at 5 percent significance level.

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Kolmogorov-Smirnov two-sample test

- The same test used in the case of one sample test can be used to evaluate whether two samples come from the same distribution.
- Let the maximum absolute difference between two empirical distribution functions be $D_{m,n}$

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Well, now, we will go to that two sample test, keeping the basic philosophy, again here is the same thing, one is that, **so in the two sample**; in one sample test what we have done is that one the data that we have obtained and other one is that sum of some standard distribution that we already know. So, in the example we have seen that one normal distribution that we have used. Now, the two sample test means, that we are not using any standard distribution, the two samples are there, two samples both will have their own that **(())** event that observed cumulative distribution function and we have to pick up the difference between two.

So, in the one sample test basically, that is one observed data and other one is some standard known theoretical distribution. And in two sample test, both are the observed data, and we are plotting, and what value we **have** picking up are exactly the same thing. So, **the same test used** in the case of the one sample test can be used to evaluate whether the two samples come from the same distribution or not. Let the maximum absolute difference between two empirical distribution functions be $D_{m,n}$.

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Kolmogorov-Smirnov two-sample test

- Let the two functions be represented as step functions $G_m(x)$ and $S_n(x)$ based on two samples of sizes m and n , respectively.
- Thus the difference becomes as:

$$D_{m,n} = \max_x |G_m(x) - S_n(x)|$$


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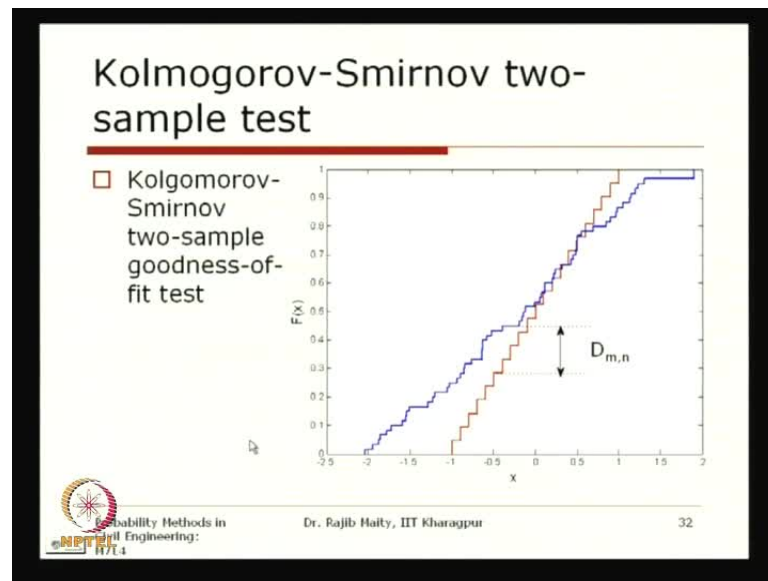
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Now, let the two functions be represented as the step function $G_m(x)$ and $S_n(x)$ based on the two samples of size m and n respectively, so that two samples are there, one sample is having the m data other one is having n data, so this should be flexible. It is not that both the samples may not have the same size of this data. And what we have to go what we have to get is that from here, is that we have to find out what is the $G_m(x)$ and what is the $S_n(x)$, and following the same equation that we have shown in this one sample test using K-S test.

So, thus here, the difference will be that maximum difference that we have to get, which is the absolute value of this difference between this $G_m(x)$ and that $S_n(x)$ and that one we have just pick up only single value again similar to the one sample test, which is denoted as that $D_{m,n}$ this m is the size of this one sample, other one is the size of the other one.

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Now, this is one such typical example, how it will look like, the blue one is for the one sample and the red one is for another one. So, now we have to find out, so basically the red is again approaching towards this all are 0 value here, and here also red is going all are one- values are one. So, now each and every point we can find out what is the difference there, so at this point may be around point minus 0.4 or so, the difference is shown here between this point and this one, this is $D_{m,n}$ for this particular value. So, for all such differences we have to pick up the maximum one, so this K S test for the two sample goodness-of-fits as looks like this and we have to pick up the maximum one.

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Kolmogorov- Smirnov Test

- If the sample distribution have large values of m and n, Smirnov has given the limiting distribution as:

$$\lim_{m,n \rightarrow \infty} P\left(\sqrt{\frac{mn}{m+n}} D_{m,n} \leq z\right) = \left(\frac{\sqrt{2\pi}}{z}\right) \sum_{k=1}^{\infty} \exp\left[-(2k-1)^2 \frac{\pi^2}{8z^2}\right]$$

replacing from one sample test:

$$\sqrt{n} \text{ to } \sqrt{\frac{mn}{m+n}}$$

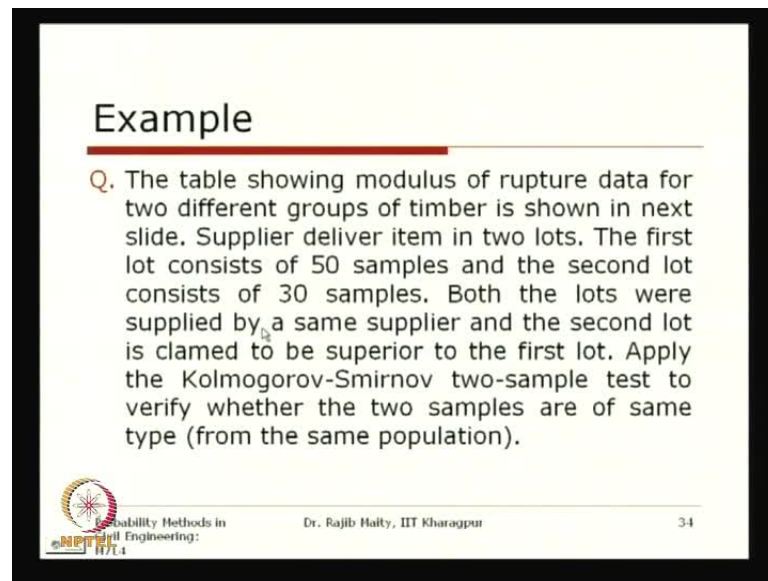
and continuing the test procedure.



Now again, the sample distribution have the large values of the m and n, if this **if** **this** values are large enough that is m and n, the Smirnov has given the limiting distribution as this. The square root of m n by m plus n, which is the data size for the two samples, multiplied by this D m n less than equals to z is equals to square root 2 pi by z summation of k equals to 1 to infinity, exponential of minus 2k minus 1 square multiplied by pi square by 8z square. So, basically what happens from the single sample, **it was here** it was square root n and for the two samples test, it is replaced by this square root of m n by m plus n.


So, basically the sample size in the one sample test we have used it for this n and for the representative sample size for the two sample test is this quantity m n by m plus n. So, if we just change this one, so **if we** once, we are having this two samples, so what is the representative data length we have to calculate first. And the remaining thing is same, whatever we have discussed for the one sample test, that is, if we take that the significance level is 0.05 say, then this quantity be equate with this 0.95 then, it will again come that same value which is 1.36. So now, the critical value that is that D m n should be less than equals to 1.36 divided by the square root of this full quantity, earlier it was square root of n, now it is square root of this m n by m plus n that is the difference.

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Example

Q. The table showing modulus of rupture data for two different groups of timber is shown in next slide. Supplier deliver item in two lots. The first lot consists of 50 samples and the second lot consists of 30 samples. Both the lots were supplied by a same supplier and the second lot is claimed to be superior to the first lot. Apply the Kolmogorov-Smirnov two-sample test to verify whether the two samples are of same type (from the same population).

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So, we will take one example here, the table showing the modulus of rupture data for two different groups of timber is shown in the next slide. Supplier delivered items in two lots. The first lot consists of the 50 samples and the second lot consists of 30 samples. Both the lots were supplied by the same supplier and the second lot is claimed to be superior to the first lot. Apply the K S test, the Kolmogorov-Smirnov test, two sample test, to verify whether the two samples are **from the** of the same type or not, that is, whether in the statistical sense, we should say that whether both the samples are from the same population or not.

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Example...Contd.

35.3	33.18	30.05	32.68	26.63
36.85	36.81	36.38	34.44	23.25
27.9	38.81	37.78	35.88	28.46
24.55	29.9	35.03	37.51	30.33
28.71	17.83	34.63	33.47	38.05
31.33	23.15	33.06	32.48	34.56
23.37	27.93	36.47	34.12	36
23.56	30.02	38.64	35.58	37.65
28	33.71	28.98	36.92	28.83
25.39	28.76	32.02	33.61	32.4


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So, this is what we have to decide and this 50 samples and 30 samples for all these timbers what is the modulus of rupture is shown in this table. This is the first lot that is lot A, this modulus of rupture is given in Newton per millimeter square. So, this is 35.3 and like this, you can see that this is 5 by 10 columns, so all these data refers to the modulus of rupture for the 50 samples supplied in lot A.

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Example...Contd.

33.19	34.4	28.97	35.89
28.69	36.53	35.17	39.33
37.69	31.6	38.71	29.11
25.88	22.87	32.76	34.49
27.11	36.88	25.19	38
29.93	32.03	25.84	35.67
33.92	38.16	28.13	30.53
	33.14	39.2	

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Similarly, in the lot B, these are 30 data points of this modulus of rupture in newton per millimeter square which is supplied in the lot B. Now, we have to test whether both this dataset is the same is from the same population or not.

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Example...Contd.

□ Sol.:

- Null hypothesis H_0 : The random variables sampled by the first 50 values and the random variables sampled by the next 30 values have the same distribution.
- Alternate hypothesis H_1 : The random variables have different distributions.
- Level of significance: $\alpha = 0.05$.
- Calculations: The data from each sample are ranked separately with values of the step functions $G_m(x)$ and $S_n(x)$.



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So, the null hypothesis is that the random variables sampled by the first 50 values and the random variables sampled by the next 30 values, have the same distribution or not. And so and the alternative hypothesis are the random variables have the different distributions, so whatever we have hypothesized in the null hypothesis is not valid. Level of significance here is 0.05 and the calculation that we have to do is that the data from each sample are ranked separately, so we have to make it, we have to sort it. In the last example, it was already sorted and that data was supplied, but here we have to sort it, first we have to give the rank and from there we have to calculate their respective cumulative distribution, so both are that step functions.

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Example...Contd.

- The samples are sorted in increasing order and ranked accordingly for both samples and $G_m(x)$ and $S_n(x)$ are determined as shown in table in next slide
- Then the step function of both samples are plotted
- The maximum absolute difference between the empirical distribution is then determined



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Thesamples are sorted in an increasing order and the ranked accordingly for both the samples $G_m(x)$ and $S_n(x)$ are determined as shown in this table in the next slide. And then, the step functions of the both samples are plotted, the maximum absolute difference between the empirical distribution is then determined.

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Example...Contd.

Rank	x at lot A	MR at A	Rank	x at lot A	MR at A
1	0.02	17.83	26	0.52	33.06
2	0.04	23.15	27	0.54	33.18
3	0.06	23.25	28	0.56	33.47
4	0.08	23.37	29	0.58	33.61
...
22	0.44	32.02	47	0.94	37.78
23	0.46	32.4	48	0.96	38.05
24	0.48	32.48	49	0.98	38.64
25	0.5	32.68	50	1	38.81



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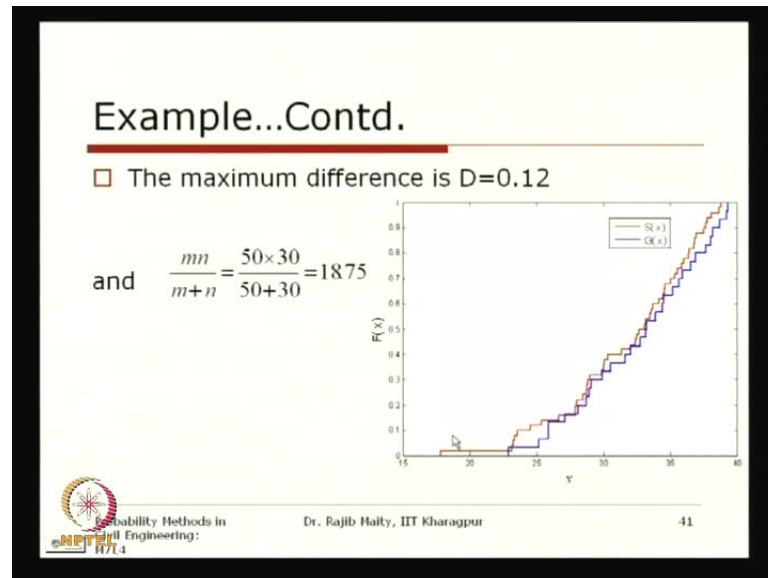
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So, the basic steps what we have seen in the single sample and this two samples are same. So, this is for the first lot, and you can see this rank 1, 2, this is a shortened this is a

table is shortened just to accommodate in a single slide that the rank is 1,2,3,4 and this continuing up to 22,23,24,25,26,27 and going up to 50.

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Now, this is that value of that k by n , that is, that rank m by n , so it is 0.02, 0.04, and 0.006 like that from this 0 it will go on, and it will come to that 1. So, this is for this lot A, and similarly, this is for the lot B, and if we plot this one, it looks like these two plots are there, one is that shown by red and the other one is that blue. Now, the maximum difference we can observe that the difference and we can get it and this maximum difference is found to be 0.12 and as I told that now, there are two samples, one is that m is the 50 data samples and n is the 30, so if we just get it becomes at 18.75, so we have to see what is the critical value against this sample size of 18.75.

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Example...Contd.

□ 0.12 is less than the critical values 0.301

□ Thus, null hypothesis can not be rejected at 5% significance level

Values of $D_{n,\alpha}$ for the Kolmogorov-Smirnov goodness-of-fit test

n	$D_{n,0.10}$	$D_{n,0.05}$	$D_{n,0.025}$	$D_{n,0.01}$
10	0.369	0.409	0.437	0.489
11	0.352	0.391	0.437	0.468
12	0.338	0.375	0.419	0.449
13	0.325	0.361	0.404	0.432
14	0.314	0.349	0.390	0.418
15	0.304	0.338	0.377	0.404
16	0.295	0.327	0.366	0.392
17	0.286	0.318	0.355	0.381
18	0.279	0.309	0.346	0.371
19	0.271	0.301	0.337	0.361
20	0.265	0.294	0.329	0.352
21	0.259	0.287	0.321	0.344
22	0.253	0.281	0.314	0.337
23	0.247	0.275	0.307	0.330
24	0.242	0.269	0.301	0.323
25	0.238	0.264	0.295	0.317

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And here if you see, that this sees approximately we have taken is 19 and obviously that for the proper value. We can go for this linear interpolation between 18 and 19, but here we have just taken this 0.301 as against this n equals to 19 and this $D_{n,0.05}$, which is this significance level - at 5 percent significance level. So, this **so** 0.12 that is what the maximum difference that we got here is now less than the critical values of this point - critical value not s - so critical value of 0.301 which I have seen from this table, so thus the null hypothesis cannot be rejected at 5 percent significance level.

So, that means, both the samples are basically from the same population, so what the supplier has claimed that the second sample is superior than the first one is not valid at least at this 5 percent significance level. So, in this lecture, we have discussed two statistical tests to test whether now, what type of particular distribution if particular sample is following through two statistical tests, one is the chi square test and second one is the K S test. Generally, the chi square test can be applied for both that discrete and continuous random variable and this K S test is for the continuous random variable, but most of this application, whether the dataset is following the normal distribution or not that we test, and also we test that whether the two samples that we are having whether, they are following the same distribution or not, that is what we can test using this K S test.

So, we will take up some other test in the next lecture; thank you.