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

Lecture No. #13 Probability Distribution of Continuous RVs(Contd...)

Welcome to this lecture 8 of this module three. This is the last lecture for this module, wherewe are discussing about random variable their properties, and some standard distribution for continuous random variables as well as discrete random variables. So, up to the last class we discussed some of the continuous random variable, and we discuss that there are few more is pending, but still whatever is covered, and whatever will be covered in this lecture may not be the full list of the standard probability distribution, there could be some other as well, but these are the distribution which are mostly used for different problems in civil engineering.

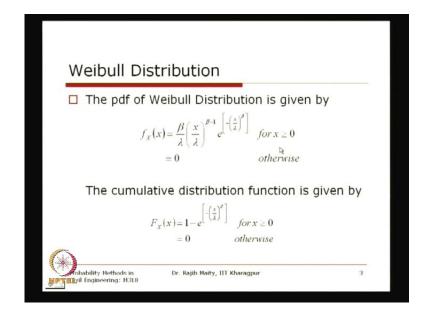
So, in the last classwethere is the Weibull distribution was pending. So, we will start with the distribution this class, and we will also discuss some of the distribution after that, that we mentioned in the last class that chi square distribution, t distribution, and F distribution which are mainly used for the different statistical test. So, our this lecture on that probability distributions of continuous random variable, we will continue in this lecture as well.

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Outline		
□ Weibull Distribution		
□ Chi-square Distribution	A	
☐ Student's t Distribution		
☐ F Distribution		
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And we will discussWeibull distribution first, and then some of the distributions which are specifically important for the different statistical test; and the this statistical test will be carried out in last module- module7, where the properties of this distribution particularly this three distribution that chi square distribution, student's t distribution, and F distribution will be used for different statistical test, that we will discuss. In this lecture, we will just discuss about their basic properties, and which distribution is suitable for what type of test that we will discuss in this class.

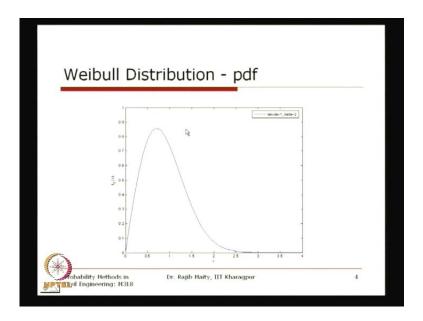
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So, we will start with this weibull distributionthis general form of the p d f of weibull distribution looks like this there is two parameters one is beta and another one is lambda. So, this p d f probability density function beta by lambdax by lambda power beta minus 1 exponential minus x by lambda power beta for x greater than equal to 0.

So, for this support of this distribution isbounded at 0. So, lower bound is 0and upper bound isinfinity and for thenegative values of x this is not defined this is 0and we know that thuscumulative distribution for anyp d f is theintegration from this lower support that is here is 0 up tohere is 0 up to that particular valuex, if you do that integration we get the cumulative distribution function and this is this can be shown that, this is 1 minus e power minus x by lambda power betafor x greater than equal to 0. So, if we compare this distribution from the whatever the distribution we have discussearlier then we will see that, we discussearlier the exponential distribution we discuss earlier that gamma distribution and this particular distribution is having some kind of sense of the generalization of those distributions and we will show that how this distribution is related to those distributions as well that is exponential distribution and gamma distribution.

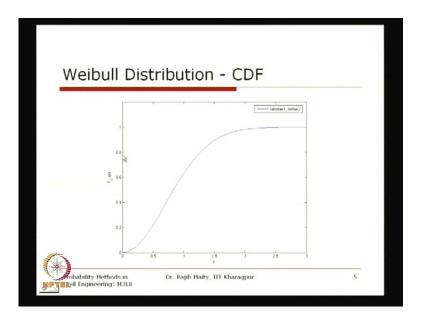
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So, the otherproperties of this distribution is means first this isone example of that weibull distribution p d f, how the p d f looklike here the parameters that is taken as this lambda equals to 1 and beta equals to 2. So, taking this this set of parameter this is the general shape of this one which we can see that this a positively skewed distribution and lower

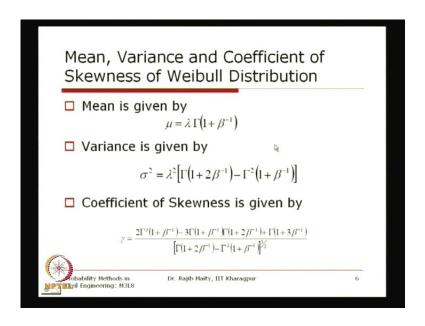
bound is 0 and upper bound this is coming andasymptotic to 0. So, this is your x and this is your f x that isdensity of probability.

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And for the same combination of the parameters that is lambda equals to 1 and beta equals to 2that is same p d fthat is shown in last slide if we take that cumulative distribution that is f x is capital f xx,then we will get this type of distribution which is starting from 0 and asymptotic to 1likethistype of distribution we can see.

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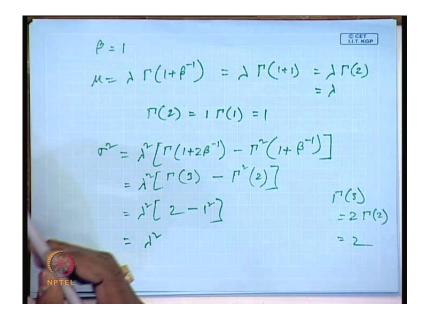


So, now, this first few moments if we can discuss about thisone that this mean is given bymu equals to lambda this isone parameter and this gamma function of 1 plus beta inversethat is 1 plus 1 bybeta. We discuss about this gamma function, some previous lectures while discussing that beta distribution and gamma distribution. So, if we know the value of this betawe will be able to know what is the value of this gamma function and we will be able to calculate this mu. So, this mu is dependent on both this parameter we have seen and similarly the variance also is given by this expression which is lambda square multiplied by gamma of 1 plus 2 beta inverse and minusgamma square 1 plus beta inverse.

Basically thisone if you just thislambda square if tou take inside this bracket this becomes thatmean. So, that mean square. So, we can say this quantitythat is the lambda square multiplied bygamma 1 plus 2 beta inverse minus mu square will be the variance of the distribution and similarly, if we take that skewnesswe will gets along expressions likes this which is the measure of the skewness of the weibull distribution.

One thing is important here that, if we discuss that if we simply prove that beta equals to 1 here then, you know that this will become that gamma 1 plusbeta inverse.

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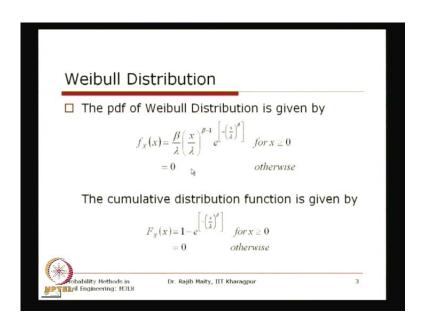


So, what we are trying to say here if we just put that beta equals to 1 beta equals to 1then that mu that lambda gamma 1 plus beta inverse which is equals to lambda gamma 1 plus 1 lambda gamma 2. So, and from theearlier lectures we know this gamma if this1is 1

integerthen, weknow that this if gamman equals to n minus 1 gamman minus 1.So, this 1 gamma 1 and this gamma 1 value is equals to 1. So, this is basically 1. So, this expression that is mu becomes lambda. Similarly, if I take that sigma square which is the expression islambda square multiplied bygamma 1 plus 2 beta inverse minus gamma square 1 plus beta inverse. So, putting this beta equalsto 1 it will eventually come that lambda, this lambda square gamma this is three minus gamma square 2.

So, this is 1 and this will becomeyour 2. So, lambda square this is 2. So, gamma 3 you know gamma 3 equals to 2 gamma 2 and gamma 2 equals to 1. So, it is 2 minus 1 square. So, this is equals to lambda squareif we see that for the exponential distribution we got exactly the same value of this moments at least the first moments that is for a exponential distribution; that means, is equals to lambda and variance is equals to lambda square. So, if we put this beta equals to 1 it seems that it is a, it is becoming an exponential distribution.

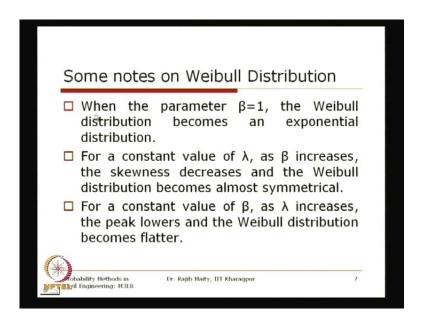
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So, if we just go back to theto this distribution of this weibullp d f and here if you put that beta equals to 1 then we can see that that is the 1 by lambda and here it is beta equals to 1. So, this is 1 and this epower minus x by lambda for x greater than 0 and otherwise it is 0. So, which is nothingbut exactly same to this exponential distribution, exponential distribution. So, here sowith respect to the exponential distribution we can say that one more parameter is introducedwhich is beta.

If this beta is equals to 1 this weibull distribution becomes exponential distribution and we will also showthat how this weibull distribution is also that gamma distribution is a special case for this weibull distribution that we will discuss in a minute. So, now, this we discuss and this first few moments mean variance and coefficient of skewness we discuss.

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Some note that what just now, we discuss is the first note that when parameter equals to 1the weibull distribution becomes an exponential distribution which just now we discuss.

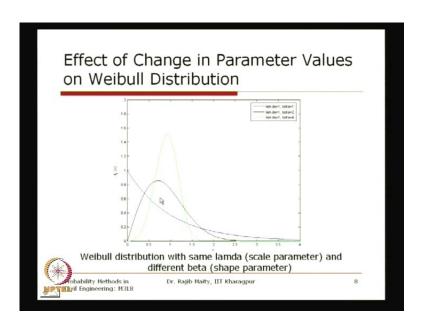
Now,if we just want to see the effect of the parameters that is both for thislambda and beta. So, if we just keep one parameterconstant and if we just the increase the other parameter then, what will be the effect on the shape of this p d f of weibull distribution?

So, the firstitsays that for a constant value of lambda. So, if you just take constant value of lambda and if we increase thisbeta. So, as beta increases theskewness of, the skewness of, the skewness decreases and the weibull distribution becomes almost symmetrical. So, this isimportant in the sense that when we need to model some kind of, some random variable which issymmetrical aswell as this is lower bounded by 0. Then, we can say, we can say that this weibull distribution maybe a suitable choice for that case because we can

estimate this beta is high and then this weibull distribution becomessymmetrical. That we will see that,how it ischanging with best on the different values of this beta.

On the other hand, if we take this beta as constant if we take some constant value of beta and if weincrease this lambda, then the peak lowers and the weibull distribution becomes more flatter. So, weibull distribution becomes more flatter means that, that its peakedness reduces, its skewness also reduces, it becomes more flatter. So, if we see that its, its shape depending on this different combination of this parameter that is alpha and beta then this will be more clear to us.

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So, this is the firstthing when we are keeping thislambda constant and adding this beta and beta equals to 1 2 and 4 and keeping the all main all three cases almbda is equals to 1. What you can see here that this for this blue line this pdf shown by blue linethis is for the combination that lambda equals to 1 and beta equals to 1 for this black one, the combination is the lambda equals to 1 and beta is equals to 2 and for this greenone this lambda equals to 1 means all the lambdas are same for all three and beta equals to 4.

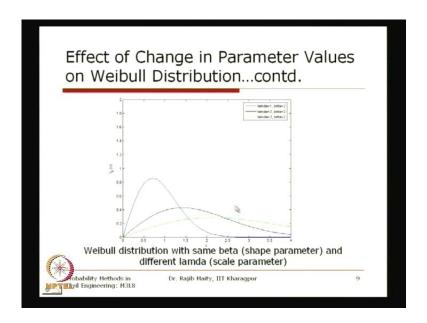
So, from this blue then black, then green the beta is increasing from 1 2 and then 4. So, here 2 things we can see, the first is that just now we discussed that if beta is equals to 1 when we say that beta is equals to 1; that means, we are talking that irrespective of any value of lambda. So, when beta is equals to 1 it approaches to the exponential

distribution. So, that we can see herethat, this is anthis an exponential distribution and we know that this at 0 at x is equals to 0 the value of fx is 1 by lambda.

So, as we are taking the lambda is equals to 1that is whythat x isequal to 0it iscoming to be 1. Similarly, now when we are increasing this beta from 1 2 and 4, we cansee that this is becoming some, this is these are positively skewed we know exponential distribution is positively skewed this can be seen graphically as well as mathematically also we've discussed inprevious lectures. Now, when we increase beta from beta equals to 1 to 2 the skewnessthat asymmetry decreases.

So, it approaching to the symmetrical and it is more clear when we are going to this four that it is becoming some more to moresymmetrical compared to the othertwo graphs where beta equals to 1 andbeta equals to 2. So, if we further increase beta we can see that it is, it is becoming more symmetrical with respect to its mean. Mathematically also this can be this can be shown that we willsee in a minute.

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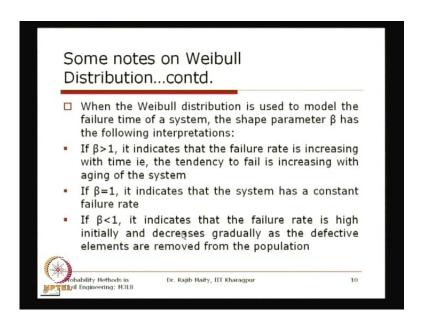


Second case if we just keep this betaconstantand if we,if we change the value of lambdahere, we have shown this three differentp d fs withdifferent combination of lambda and beta. So, this first one this blue one is for this lambda equals to 1, beta equals to 2; this black one is the lambda equals to 2, beta equals to 2 and this green one is for lambdaequals to 3 and beta equals to 3. Thus for all three p d fs here the beta is same which is equals to 2 and for the, and the lambda is increasing from 1 then for the black it is

2 and for green from itis 3. So, what we can see is that when this lambda islow, then itspeak is more and with the increase of thislambda this peak becomes flatter.

Basically, when we are having some set of data there are some test tounderstand that which distribution it is following and after we select a particular distribution, we generally discuss that we generally want to know, what is its parameter?, then this parameters are estimated from the sample data that is available to us that is basically the first step for any probability model, whatever the problem is faced. So, the sample data is collected from the sample data first we test that which distribution this is following and then we estimate the parameter. So, that while estimating the parameter there are different methods are available. So, here what we are discussing that the effect of this parameter the graphical significance of this parameter. If we selectone parameter, if we change one parameter how the shape and location of particular p d fchanges, that is what we are discussing for all the, all the models, all the probability distributions.

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Now, one important discussion with respect to this weibull discussion is this theparameter beta, this parameter beta is playing a vital role and it is having different meaning for the different ranges of this parameter first. So, when the weibull distribution is used to model the failure time in asystem the shape parameter beta has the following interpretations.

So, how this beta we can interpreted for the different values or the different ranges of this parameter beta. So, first if we start with this beta equals to 1 that is the, when we say

that beta equals to 1.We know thatthis is the approaching to the exponential distribution. Now the failure of a system; that means, that we know that value is constant that is that failure rate, if we while discussing that exponential distribution we discuss that this the failure rate or the interarrival rate, averagetime on which a particular aphenomenon is occurring that is constant overtime based on that assumption, we develop that exponential distribution this is exactly what is stated here that is e beta equals to 1 it indicates that the system has a constant failure rate; that means, which is the basic assumption of the exponential distribution and we have also discuss that when beta equals to 1 the distribution is approaching to the exponential distribution. Now there are two other region if beta is less than 1 or if beta is greater than 1, what does it implies?

So, when we say that this beta is less than 1, it indicates that the failure rate is increasing with time. That is the tendency to fail is increasing with the aging of the system, this is important for different civilengineering problem for the, for any structure that we take due to its aging problem orsome. Some apparatus for its fatigue problem when the time increases sometimes the inter arrival rate of the phenomenon may increase.

So, here we are deviating for from the basic assumption of this exponential distribution that this rate of this failure need not be constant over the time starting from 0 to the the anypositive infinite towards the plus infinity direction. So, when we say that this rate of failure is, rate of failure is increasing with time; that means, that is captured when the distribution is having the parameter that having the particular beta is greater than 1.

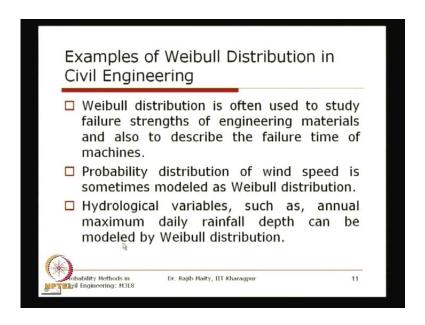
Similarly, So, this is you know that when someparticular, some particular structural memberor the structure itself due to itsaging problem orwhen it is passing more time the chances offailure isincreases. So, that kind of casesthisweibull distribution with beta greater than 1 will be the, will be a suitable case. On the other hand when the beta is less than 1 it indicates that the failurerate is high initially and decreases gradually as the defective elements are removed from the population. Now, rather to come to this civil engineering problem, if we just take a social problem then it will be a more clear that, if we say that due to the health means health condition due to the this is the health system. If the mortality of the babies is increase at birth then this is a one of this, one of this very suitable application to model that kind of situation with this beta less than equals to 1. So, at the birth is shown the mortality rate is more compared to when the age increases of

this of the child. Similarly, for if we come to this some application of civil engineering

that time when we are producingsome kind of product and we are justmarketing it. Now, if there is any defective element in the population then what happens and if there is any the initial check of this whatever the product is developthen there is a chance that, if we can separate out those defectivematerials then, that can eitherfail at the initial stage itself or if we pass from this initial stage then it can go for a long time. So, what we what I am trying to communicate is that this rate of failureat to the initial time is more compared to that when tincreases. So, at that case that beta is, beta is less than 1. So, that that the failure rate is high initially and decreases gradually as time pass.

So, these arethe interpretation of this parameter beta for a weibull distribution and suitably we canthink of that forwhat kind of situation which range of beta is suitable fordifferent problems in hand.

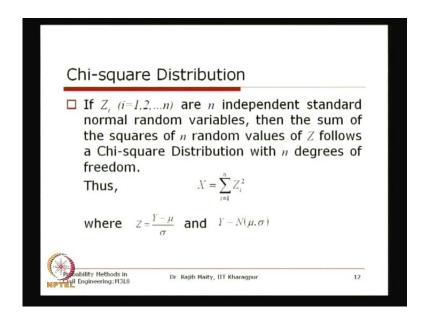
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This examples of the weibull distribution in civil engineering as we discussed that this weibull distribution is often used, to study the failure strength of engineering materials and also to describe the failure time of themachine. So, it can. So, the failure time of a machine or the failure strength of engineering material. If I assumed to be its to be constant then it can be as well an exponential distribution; that means, weibull distribution with parameter beta equals to 1 or it depends on that, what kind of situation? what kind of problem? in hand based on that the beta parameters may change.

Secondly thesometimes theprobability distribution of the wind speed particularly the hourly wind speed is also has seen to follow weibull distribution. So, that is also designed using theweibull distribution, in the application of water resource and hydrology the annual maximum daily rainfall depth, if I take un to that data also has found to follow a weibull distribution with suitable combination of its parameter lambda and beta.

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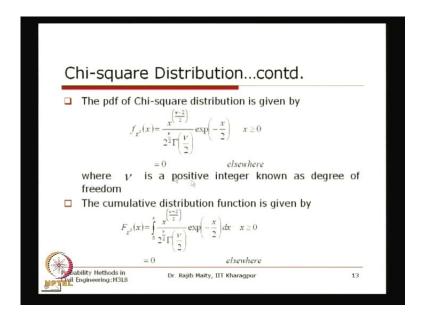
Next,we will discuss three distribution one after another,that we will start with the chi square distribution and this distributions are mostly as I mentioned earlier these distributions are mainly important for the statistical test and different types of test is carried out using different types of distributions. So, first were discussing this chi-square distribution which is generally used for the purpose, of for the purpose of goodness of fit. So, many times I have mentioned that ones the sample data is available to us. We generally have to first we have to test that distribution it is following and. So, they this is also can be tested statistically and that test is generally done through this chi square testit is called and based on this chi square distribution. So, we will understand now the what are this distribution and what are the different properties.

So, first we are starting this chi square distribution, if Z i equals to 1 to n that is Z 1 Z 2 Z three up to z n are independent and identically distributed of course, we use the word this

3,3 were together independent and identically distributed sometime it isabbreviated as iid herethese are independent of course and following the standard normaldistribution; that means, these are identically distributed. So, independent standard normal random variables then their sum of the squares. So, this is important the square the sum of the squares of this n random variables, that follow a chi-square distribution with n degrees of, degrees of freedom. So, here that x that therandom variable that I am defining is equals to the summation of those Z 1 square plus Z 2 square plus up to Z n square. So, where this Z are the standard normal distribution.

So, if I say thatthis Y is following a normal distribution with mu and sigma their parameter. Then we can ake it, we can get their reduced variate Z which is equals to Y minus mu by sigma and this Z for this n numbers of independent random variables are their. If we square them of add them then these random variable that summation x that is follow a chi square distribution.

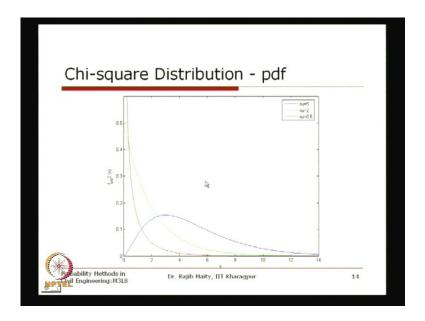
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Now the p d f of thischi square distribution is like this that is,this is f chi square x is x power nu minus 2 by 2divided by 2 power nu by 2 gamma nu by 2 exponential minus x by 2for x greater than equals to 0. Here nu is the parameter known as thedegrees of freedom this isan integer it is an positive integer and this distribution isagain lower bound by 0 and the fornegative values it is 0. So, this nu here mentioned that this nu is a

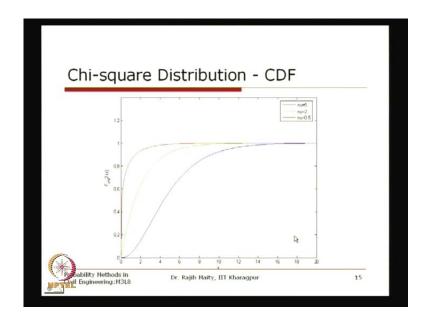
positive integer known as degree offreedom. Again, we know that from this distribution this cumulative distribution is nothing, but theintegration from the lower support to a specific value x of the p d f and this is also defined for this x greater than equal to 0 and 0 0elsewhere.

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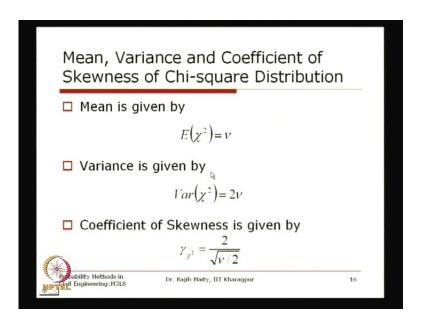
So, we will know that its differentmoments and different initial moments. Now, before that we will show the how this for this different values of this nuwhich is that the degrees of freedom for this chi square distribution is this. So, here we have used different combination of this nu and we are. So, showing that as this nuincreases, then this is becoming this. So, this blue is for the maximum values of new compared to this green and this redhere. So, this is if its if the degrees of freedom increases is generally become more symmetrical.

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And for the and these are the for those three chi square distribution, these are the cumulative distribution for those of f that is shown in this last slide.

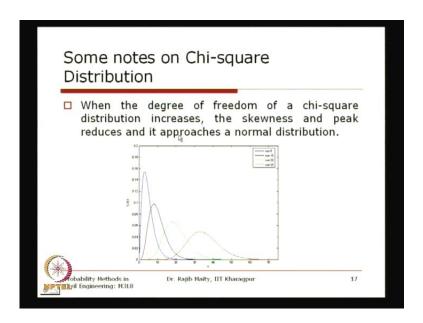
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Now, theinitial fewmoments if we see then there are some interpretation particularly itsskewnessas just we have seenfrom this graph the first is this its mean. The mean of this chi square distribution is equals to its degree offreedom which is nu thevariance of the

chi-square distribution is 2 nu that is double of themeanand the coefficient of skewness if we see then, this is that 2 by square root nu by 2. So, this square root nu by 2. So, this is always positive. So, the chi square distribution always positivelyskeweddistribution and you can see here what we have seen graphically in a minute back. Also, thatif we increase this nu that is if the degree of freedom is increased. Thenum then this becomes lower andlower and for every very high degreeof freedom, this becomes almost symmetrical as this skewness is reducing with the increase of this nu.

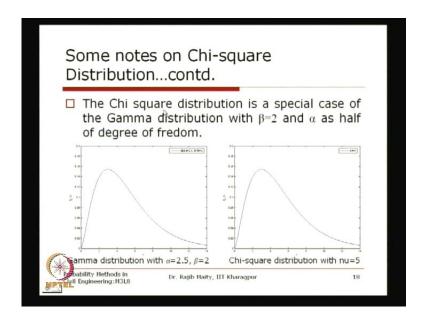
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So, this isagain explain here graphically that is when thedegree of freedom of achi square distributionincreases the skewness andpeak reduces, and is approachesto anormal distribution in the sense that it is it is approaching to a symmetrical distribution and going away from itsorigin. Here we have takenthefour examples with different value of degree of freedom. So, the firstone the blue one that we see that is for the degrees of freedom that is nu equals to 5, the black one that nu equals to 10 then this green one is 20 and this nu is 35.

So, here you can see that as this degrees of freedom is increasing this distribution becomes almost symmetrical of course, all this distributions are lower bounded by 0 that is the basic property of this chi square distribution but as this one increases this approaches to a approximate normal distribution with high value of this degrees of freedom.

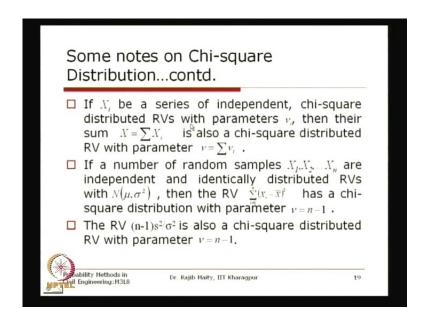
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Now, thischi squaredistributionis a special case of the gamma distribution withbeta equals to 2 and this alpha asthe half of the degree of freedom. So, this is there are two graphs are shown is that, this is the, this is this if you see this is chi square distribution with nu that that is the nu the parameter of this chi square distribution is equals to 5.

So, if we take thisone that nuequals to 5 then the shapelook like this the p d f shape look like this now, if we this is a gamma distribution withtwo parameters we knowthat this gamma distribution is having two parameters. The first that beta is 2 and alpha is equals to half of the these degrees offreedom that is 2.5. So, if we plot a gamma distribution with alpha equals to 2.5 then beta equals to 2, then we get the exactly the same p d f with for which is for the chi square distribution with degrees of freedom 5. So, this is therelation between this chi square distribution and the gamma distribution.

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Now, if that x i be aseries of independent chi square distribution, chi square distributed random variables withparameternu ithat is for first that X 1 this degrees of freedom that the parameter is nu 1 X 2 parameter is nu 2 then their some, if I take the someof those all.

All those X i s' thatis from 1 to n then the a newrandom, will get a new random variable X which is the summation of themsimple arithmetic summation then this new random variable is also a chi square distribution with the parameter is equals to summation of all those degrees of freedom. So, this is the additive rule you can say for this chi square distribution that is if we addmore than one chi square distribution and if we know they are what is the degrees of freedom. Then we can get a new random variable which is also chi square distribution and degree of freedom for this new random variable is equals to the summation of the degrees of freedom for all those random of, all those random variables we added up. So, that is explained here for this first note of this chi square distribution.

The second is the, if a number of random variable that is X 1X 2 X n are independent and identically distributed random variables with normal distribution, with the parameters mu and variance sigma square mean mu and variance sigma square then, therandom variable if we take here that is X i minus X barsquare then, sum them up these variable this is

also arandom variable which will follow achi square distribution withparameter that nu is equalsto n minus 1.

Now, there are two things here the first one is that we have n random variable all are, all are randomall are normal distribution that is identical distribution which isnormal distribution with their parameters. Now, if we get the another parameter we knowthat if we just make them square this is what we started the discussion of our chi squaredistribution earlier, but here what we are doing is that we are instead of callingthisrandom variable to the standard normaldistribution, we are just making it generalthat is it these are the normal distribution with some parameter of mean and variance. Then, if we add themup after deducting their mean this X bar is the sample mean. So, this X bar is equals to the summation of all this their values divided by the n. So, that is that average of, that is the mean that is the sample estimate of the mean. So, if we take out thatmean and then square them up and then add them of, then we will also onenew random variablewhich is; obviously, follow the chi-square distribution, but earlier we told that this follow a chi-square distribution withparameter that nu equals to n because we are adding n degrees of freedom, but here the degrees of freedom isthat n minus 1. So, thisone degrees of freedom is lost and this loss of this degrees of freedom is due to thecalculation oftheir mean. Because, I have to calculate their mean from the sample that I have to deduct from they from them to and then square it up. So, while calculating this mean the one degrees offreedom is lost and that is why this, degrees of freedom for this randomvariable which ischi squaredistribution thisdegrees of freedom is n minus 1.

Anotherrandom variable if I take that this s, sis now the sample estimate of, sample estimate of the standard deviation. So, this s square divided by this sigmasquare, sigma square is this population variance of this population. So, this ratio multiplied by the n minus 1so this is also this total quantity is also a random variable, which is also a chi square distribution with parameternu equals to nminus 1 with this degrees of freedom n minus 1.

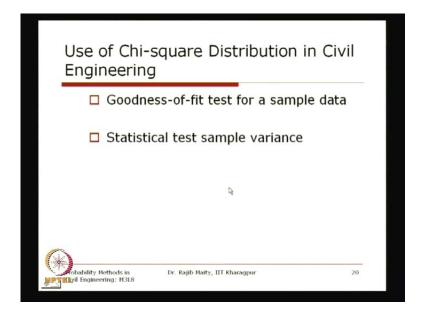
So, this one is basically thesetwo properties why I am stressing is that thesetwo are the basicproperties which is generally used for the different problems. The first one is that when we just take that difference from a particular value take the difference square

them and add them up. So, this property isgenerally used to test the goodness of fits. So, when we take a sample data and wetake that how, what is the distance? from that, that target distribution. Suppose that I have the sample distribution and I can calculate their empirical cumulative distribution, and for some known standard probability distributions say for example, gamma distribution for that gamma distribution with some set of parameter I knowhow the CDF looks like.

So, what I can do with the fromthesample data the calculated CDF and for a standarddistribution there is ain this case what I am discussingis gamma. So, that if we take the gamma distribution then what we can do for eachpoint of thesample I can calculate what is the distance? of the two cumulative distribution function. And that distance, if I just square them of add them up and; obviously, then we can see from this discussion that, that particular random variable will follow a chi square distribution.

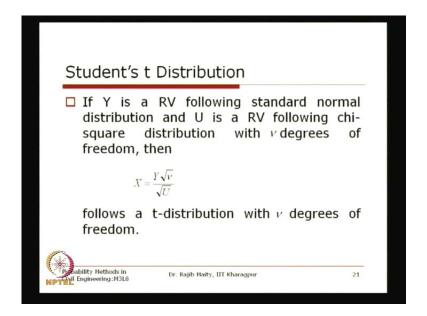
So, generally for the goodness offittest that is why we use the chi square test. Secondly the anotherapplication is that, is that for this a square. So, when this asquare; that means, the sample estimate of the variances of that, that is also follow a chi square distribution. So, if we need to know some properties of the sample estimate of the variance then also you can follow this distribution to have some statistical test. This will be discussed in more detail for this module seven while we are using make use of this chi square distribution and of course, the other distribution as well which we which I am going to discuss now.

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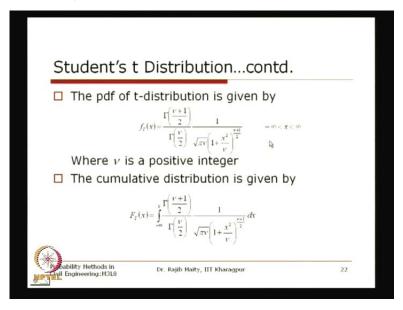
So, first of all I want to do this one the use ofthe title of the slide that use of chi square distribution in civilengineering. So, so far mostly forthis all this distribution we have use this word that use ofthat particular distribution in civilengineering. But here I just draw your attention becausethatthis use of this chi squaredistribution is need not be for the civil engineering only this is more general and this canbe used for anyparticular application. Because this applications are more generally, more general in natureandthis is the civil engineering is also not violated. So, even thoughI mention it here that in civil engineering, but this is in general forall the problems in hand. So, this two are we are just now we discuss goodness-of-fit test fora sampledatathat did not be only the civil engineering problem data that can be any data that is why, I just mention thatone that is the general concept of thisgoodness-of-fit test andagain the statistical test for the sample variancewhich also we discussin this last slide.

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The next another important distribution is the student tdistribution. If Y is a random variable following the standard normal distribution and U is a random variable following the chi square distribution with nu degrees of freedom. Then if we get another random variable like this that is Y by the square root of the chi square distribution normalized by its degrees of freedom, that is U by nu whole square root then, we will get this quantity which is another random variable and this is as then found to follow a student, is student t distribution this is known ast distribution or student t distribution with the degrees of freedom equals to nu.

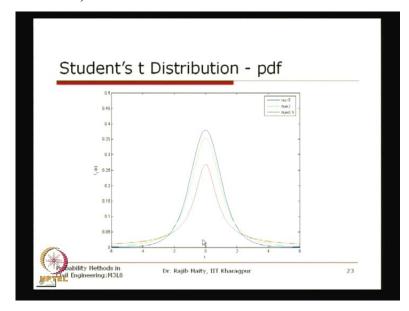
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So, its distribution form looks like this the p d f of this t distribution is given bythisform thatis gamma of nu plus 1 divided by 2 divided by gamma nu by 2, this is the normalizing constant to make that mixture the total area under this p d f is equals to 1 multiplied by 1 bysquare root pi nu multiplied by 1 plusx square by nu power nu plus 1 by 2 and itsupport here is minus infinity to plus infinity which is here that. So, there is no bound of this support this is aentire realaxis, over the entire real axis this random variable can take. So, this nuagain is the positive integer which is the degree of freedom for this distribution. And to get this cumulative distribution again it is known that from this left hand support that is minus infinity in this case to specific value x this integration will give you the cumulative distribution of this p d f.

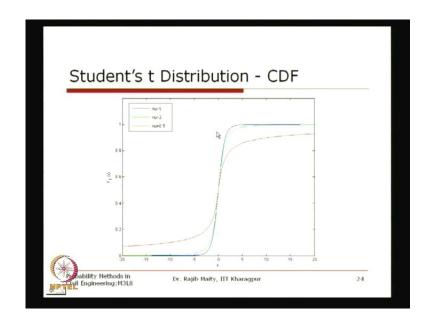
These are generally donenumerically, because of close form of this integration is not available.

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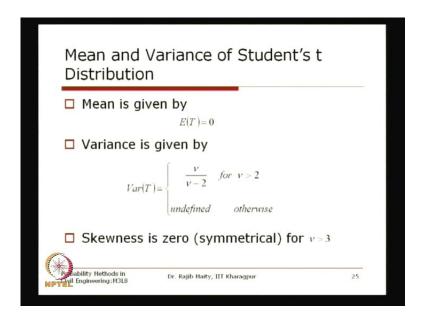
Now, again here we have shown different examples of this different values of this nu and you can see that here what we can see isthat, if we increase the value of this nu this isthis increase is peak isincreases with theincrease of this degrees of freedom and it can be shown that for the highvalue of this nu, high value of this degrees of freedom. It is becomes exactly similar to the normal distribution again. So, it will go and approach to this normal distribution with the increase of degrees of freedom that is what is shown here.

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And for, the for those combinations of this degrees of freedom here we have shown that their distribution of this CDF that cumulative distribution of the student t distribution for different combination of that degrees of freedom. Here the red is the lowest then green and then highest is the bluewith. So, far as this degrees of freedom is concerned.

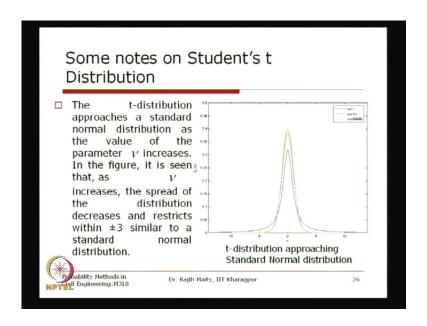
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Now these are again the first fewmoments of this distribution the first of all the mean the mean of the students T distributionis equals to 0 when we are talking about this standard T distribution because this is the standard T distribution when we say. So, that is basically is beingused when we do this statistical test and for this standard student t distribution, this you know that mean is 0 similar to the fact that for the standard normal distribution this mean is equals to 0 for this variance of this distributionis equals to is related to its degrees of freedom. That is nu bynu minus 2 and this is for when this nu is greater than 2 the degrees of freedom greater than 2 otherwise it is undefined. So you can see that this is a this is again. So, this as this newincreases. So, this is also become this is this is, this variance when this v is greater than 0 then you can say that we are having a positive variance; obviously, the variance is always positives that is why when we're taking this for v less than 2 you know that. So, that is why this variance for the student t distribution for nu less than 2 is undefined and skewness is zero.

We know that we have seen that nature of this p d f. So, this skewnessis zero as this distribution is distribution is symmetric the skewness coefficient if you just then this is generally defined for this nu greater than three, that is the degree of freedom greater than 3 we can say that this distribution is the skewness is zero.

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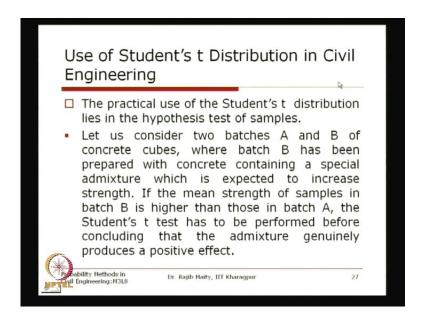


So, hereif we see that this t distribution, this t distributionapproaches to a standard normal distribution as the value of theparameter nu increases. In the figure it is seen that as nu increases the spread of the distribution decreases and restrict within this plus minusthree to a similar to a standard normal distribution. This is what we are just discussing while giving this example of this the effect of this different values of this nu. Herealso there are threeplots are shown: the first is blue one is for the lowest nuhere in this case this nuequals to 1. In this case for this blue line and then, and then if we increase that nu that is for this red one you can see that it is nu equals to 10 for this green it is n is a very high value taken and shownthat. So, it will mostly. So, it is not that it will go on in increasingit is, it is asymptotically with respect to this degrees of freedom it is asymptotically approaching to a standard normal distribution.

This is basically used in the case of this statistical test when you knowthat anysample data that is having some finite values of sample data. Now when wegenerally see that this dataset is the number of data points are less than,hence we even thoughwe know

that particular statistics follow a normal distribution we generally go for this t distribution because for this t distribution that degrees of freedom is defined here and, we know that for the small number of this samplesize instead of using that standard normal distribution we use the student t distribution which is more logical in this case because, for this standard normal distribution effectively the sample size should be infinity.

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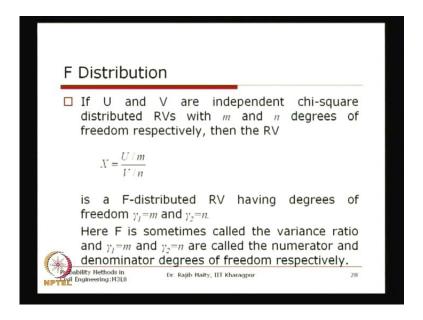
So, use of student t distribution againI am discussing with respect to the civilengineering, but this applications application of all this three distributions student t distribution, chisquare and F distribution are moregeneral in nature. The practical use of student t distribution lies in the hypothesis test of this sample particularly, for the samples mean. So, two samples are available to us and if we have the two different mean then thethen the target is the whether, we have to estimate that weather the which mean is more, which mean is less. So, if that kind of thing that I cannot simply calculate the sample mean and say that the more and less by, we have to test this statistically in the sense that, that whether that with the some statistical significance level whether theone mean is really larger than others are not.

So, oneexample from the civil engineering application we can take that take in the in case of this admixture in the concrete sometimes we mix some admixtures to a concrete mix and claim the strength as increased. So, in that kind of test what we have to do is that we

have to do the two different batches of this concrete cube inone batch that admixture will be added and otherbatch without that particular admixtures. Then we have to take that sample of the calculate that sample mean and then, if you want to test that which mean is more whether that is that, that is an effect of this mixing this admixture are not then that, we have to test it that hypothesis we have to test while testing this kind of hypothesis where the mean is in involved then we have to follow the student t distribution.

This is what is explained here that let us consider two batches A and B of the concrete cube where the batch B has been prepared with the concrete containing a special admixture which is expected to increase the its strength. If the mean strength of the sample in the batch B is higher than those in the batch A the student t test has to be performedbefore concluding that the admixture generallyproduce a positive effect or not. So, it is not only forthis one. So, wherever we have we can say that there is a that mean is in further sample mean is ininvolved we use this student t distribution.

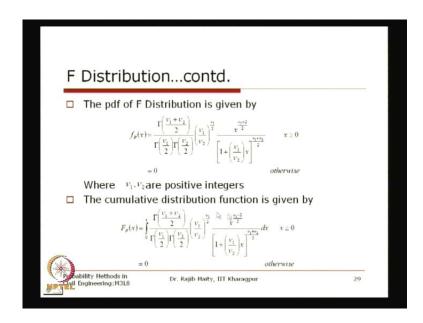
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The last one in this series is the F distribution if the u and vare independent chi squaredistributed with m and n degrees of freedom. Then thistheir ratio that is U by m and v by nwe will get a new random variable which is F distributed and for the F distribution they are two degrees of freedom and here. So, that is gamma 1 and gamma 2

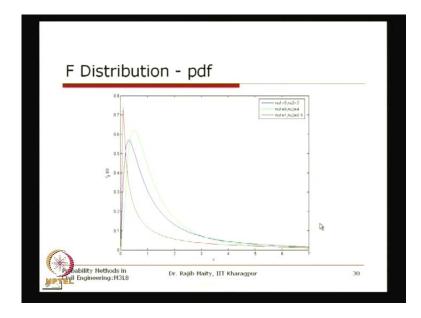
asdenoted is here the gamma 1 equals to m and gamma 2 equals to n. Here the f is sometime called as variance ratio and this gamma 1 equals to m and gamma 2 equals to n are called the numerator and denominator degrees of freedom respectively.

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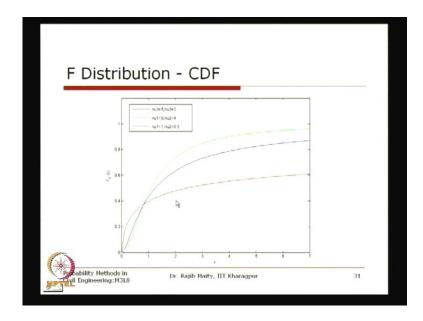
So, this is the general form of thisp d f of the F distribution that is gamma nu 1 plus nu 2 by 2 divided bygamma nu 1 by 2 gamma nu 2 by 2 nu 1 by nu 2 powernu 1 by 2 this is the full the normalizing constant, multiplied by x power nu 1 minus 2 by 2 1 plus nu 1 by nu 2 into x power nu 1 plus nu 2 by 2 and this is for this x greater than equals to 0 and otherwise it is 0. And its cumulative distribution again at the close form is not available to go for some numerical integration.

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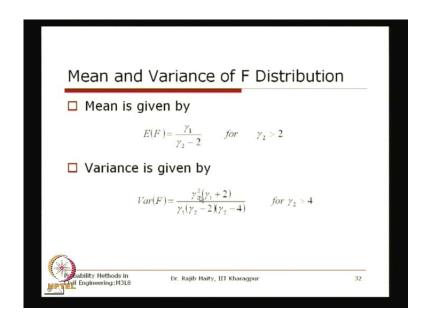
Now, this is the effect of this different values of this combination of two degrees of freedom one is the nu 1 and nu 2. We can see that thus red, blue and green different combinations are shown here for different set of combination of degrees of freedom.

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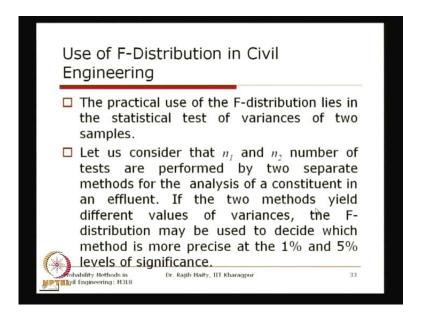
And their respective cumulative distribution functions are shownherewhich is starting; obviously, from 1 and it is asymptotic to 1 starting from 0 going and asymptotic to 1.

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The mean of this F distribution is thethis nu 1 by nu 1nu 2 minus 2 for nu 2 greater than 2 and this variance is defined by this nu 1 square multiplied by a nu 1 plus 2 divided by nu 1 in to nu 2 minus 2 multiplied by nu 2 minus 4 for nu 2 greater than 4.

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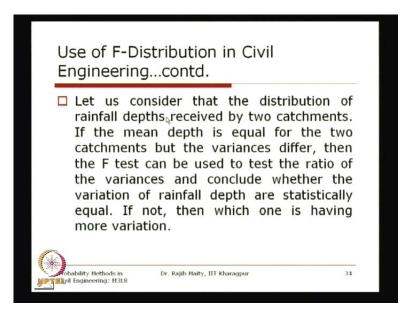


This use of F distribution here arein civil engineering and mostly it is for thus general casewhen now for this t distribution we told that when the mean of the sample is

involved that time weuse the t distribution similarly and when the sample variance is involved we here use this chisquare distribution now suppose that there are two samples are available and two the ratio of the sample variance from the two samples are concerned then that ratio to test that ratio we have to follow that F distribution. Suppose that one sample is having one standard deviation is therewhich is estimated from the sample data. Similarly, for the other one now, if I takeratio and if I want to test which 1 is more than; that means, that ratio should be either greater than 1 or less than 1 are equal to 1 say if equal to 1 then both the samples, both the samples' standard deviation is same and if it is more depending on which one is more and less that can be decided.

So, for the ratio of the variances fromtwo population follow a F distribution. Keeping that in mind we can just see where this will be effective the practical use of F distribution lies in the statistical test of the variance oftwo samples. Let us consider that n 1 and n 2 number of test are performed by two separate methods of this analysis of a constituent in an in an effluent if the two methods yield different values of this variance the F distribution may be used to decide which method is more precise at 1 percent and 5 percent level of significance.

So, this level of significance and whichone isstandard we willbe discuss in more detail in the substituent module andthat last module sevenwhen we discuss about the statistical test. (Refer Slide Time: 55:45)



Andin civil engineering specially, we can also say that let usconsider that the distribution of the rainfall depth received by two catchments are there, if the meandepth is equal for this two catchment, but the variances differs, thenthe f test can be used to testthe ratio of the variances and concludewhether the variation of the rainfall depth is statistically equal are not. If not then whichone is havingthe more variation.

So, with this I mention once again that this may not be the complete listof this all the standard probability distributions, but these are the mostly used into address the different problems in civil engineering. Thank you.