

Probability Methods in Civil Engineering
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Lecture No. # 29
Introduction to Copulas

Hello and welcome to this lecture. We are in this multiple random variable module, and this is our eleventh lecture, and as we are mentioning for sometime in the some previous lecture that we will be give some brief introduction on copula and this copula is useful when we are looking for the joint distribution of multiple random variables.

So, today's lecture, we will be considering a brief introduction on these copulas. This copula theory is relatively newer to this statistical theory and just we will go through that brief introduction and some special class. We will just discuss which is known as the Archimedean copula, and throughout this lecture, we will try to communicate that how this theory or how this, **this**, function, this copulas or some functions, how this functions can be useful to derive the joint distribution of two random variable when their marginal's are known.

So, you know that we have discussed earlier that even though the marginal's are known, it may not be that easy to get the joint distribution. Only one meaningful conclusion that we have seen earlier is that if two random variables are jointly Gaussian or they are joint normal distribution they follow. Then we can conclude that their marginal's also will be, marginal's also will be normal distribution for each of the random variables involved, but the reverse is also not true, that is, when their marginal's are normal, we cannot conclude that their joint will be that joint normal distribution.

So, here we will be introducing some functions, some copula functions, and so, using those functions, we can derive that their joint distribution. This function follows some properties that has to be followed. So, those properties we will be discussing, and after that, we will be taking some of the specific example of these functions and which can be useful and what has recently found its application to the civil engineering problems.

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Probability Methods in Civil Engineering

Module 5: Multiple Random Variables

Lecture - 11: Introduction to Copulas

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Outline

- Definition and Explanation
- Properties
- Basic terminologies
- Sklar's theorem
- Measure of dependence
- Archimedean copulas
 - Example

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So, our today's lecture is on this introduction to copula and the outline that we have, that we will follow is that, first, we will go through some basic introduction, and after that, we will see its definition, its meaning and its explanation, how, what is this copula function and we will see their properties, the certain properties that it should follow.

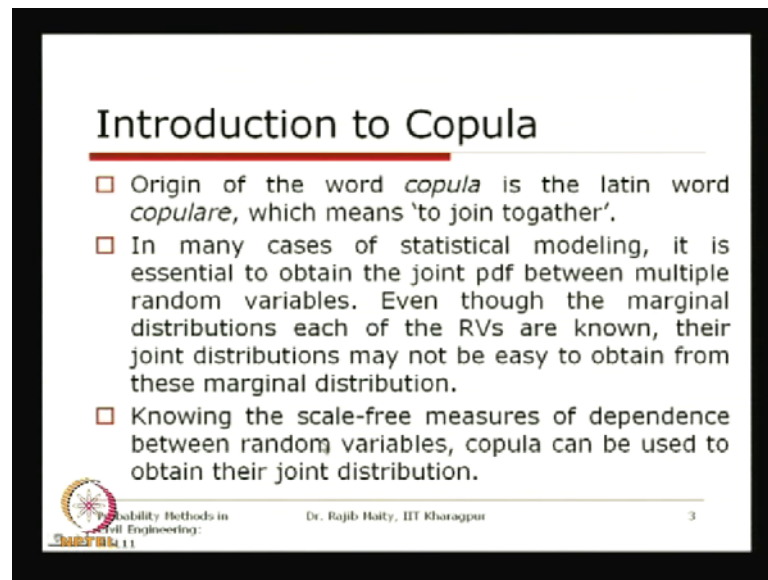
First of all to be a valid a copula and there are some basic terminologies as this is as I told that, this is a newer area. So, some of the basic terminologies we have to discuss, and while discussing this basic terminologies, we will try to relate it to their one-

dimensional counterpart because most of the theories we have discussed for the one-dimensional cases, and so, that as we have discussed it to the one-dimensional cases. Then, now basically this introduction that we are talking about like this similar to the earlier lectures. We will first discuss about terms of the two-dimensional cases, that is, two random variables are involved, and even though, later on we will just show that how this things can be extended to even more than two random variables involved.

So, so that while discussing the basic terminologies we will just try to relate that what is the counterpart on the, on the, single random variable, that is, the properties of the single random variable their p d f and c d f's. So, that we will see, and after that, there is very, one of the very important theorem which is known as the Sklar's theorem, and basically this is the backbone of this full theory from where I can use this function to, to, obtain their, their, joint distribution of the random variable involved, and that is generally done through some of the measure of dependence and that dependence is generally the scale free measure of dependence.


So, we will discuss soon the, on the, issues of this measure of dependence how we can measure the dependence from the population as well as from the sample estimate, that is, from the sample of this data and then we will discuss specifically about some on the Archimedean copulas. There are different classes of copulas. For example, the electrical copula, placket copula, but here, so far the many applications has been found in this different problems of the civil engineering is through the Archimedean copula. So, we will, we will specifically see that what does this Archimedean copula means and some examples of this Archimedean copula also we will see towards the end of this lecture.

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Introduction to Copula

- Origin of the word *copula* is the latin word *copulare*, which means 'to join together'.
- In many cases of statistical modeling, it is essential to obtain the joint pdf between multiple random variables. Even though the marginal distributions each of the RVs are known, their joint distributions may not be easy to obtain from these marginal distribution.
- Knowing the scale-free measures of dependence between random variables, copula can be used to obtain their joint distribution.

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Well, so, this to start with this word copula, this might be, even the word might be a little bit newer that, when we are, this copula word is basically obtained from the Latin word which is known as copulare, **copulare**. So, this meaning of this word, this is basically a verb and this meaning of this verb is that to join together. So, here, as you know that we are just joining two marginal distribution of two random variables to obtain their joint probability density functions or joint cdf.

So, basically we are taking two marginal's and we are joining them to obtain their joint distribution. So, that is the name has been taken from that Latin word and this is the background of this word copula. In many cases if we see of this statistical modeling, then we see that it is essential to obtain the joint pdf between multiple random variables else, even though sometimes in the marginal distributions of each of the random variable are known. Their joint distribution may not be easy to obtain from these marginal distributions. So, this we have discussed earlier also that we know that what is the, what is the marginal distribution, but for until and unless, there are very specific cases. We cannot conclude that what should be their joint behavior.

Say for example, if we just say that one distribution is say your one distribution is say that gamma distribution or otherwise the normal distribution, then what should be their joint. So, so far I think if I am not wrong, then without the help of this copula that kind of joint distribution cannot be obtained. So, that while, **while**, getting their joint distribution

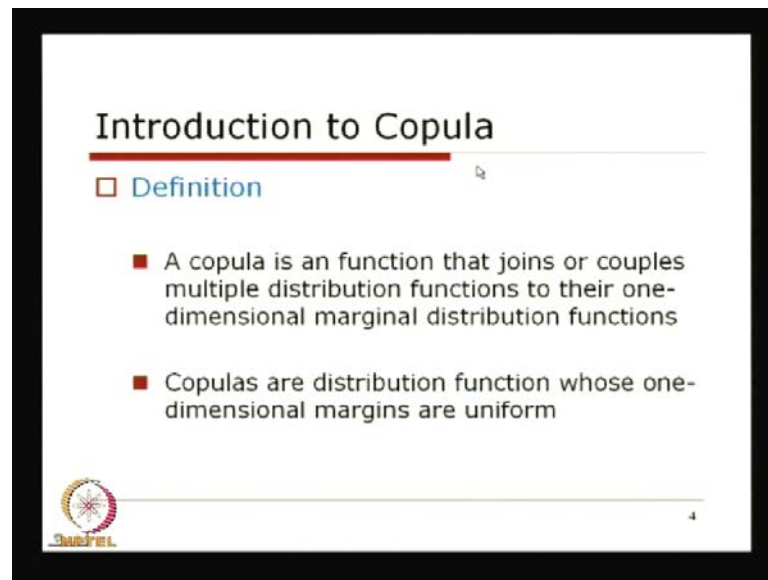
also, we have to follow some of the, some of the, measure of their dependence and one of the, **the**, scale free measure of dependence. Then, we will just discuss about what is this scale free measure, that, that dependence between the random variable. That is one of the important parameter that we have to consider before we can go through that copula.

Basically, so far as, now I take some time to discuss on this dependence. Basically, the whatever the measure of dependence in this, in the, in the statistical theory is there that can be broadly divided into two groups - one is that whether the dependence rather than measure of that dependence is **can**, **can**, change if I change arbitrarily the original data to some to some arbitrary scale, scale factor.

So, for example, that we are talking about the dependence between x and y , now, for one of this random variable either x or y , if I just make some change to some, **some**, scale changes, say that I make y I replace y as y square. So, now, the question is whether their dependence will change that the measure of the dependence that we are considering that whether that will change or that will not change.

Depending on these two situations, this, **this**, measure of dependence can be classified into two groups - so, one group is that it is, it is, irrespective of whatever the, whatever the, changes that we have made, the scale changes that we have made and other group may change if we do this type of changes. So, the, **the**, scale free measure of dependence means what we are talking? If we are going for that type of changes, that type, that type of a scale change, then this type of a measure of dependence will not change. So, two such measure of dependence that we will discuss in this lecture is, one is that that Spearman's rho and other one is that Kendall's tau that we will come after sometime. Before that, we will see the definition of, of, this copula.

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The slide is titled "Introduction to Copula" and features a red horizontal line. Below the title, the word "Definition" is written in blue. There are two bullet points, each preceded by a red square. The first bullet point states: "A copula is an function that joins or couples multiple distribution functions to their one-dimensional marginal distribution functions". The second bullet point states: "Copulas are distribution function whose one-dimensional margins are uniform". In the bottom left corner, there is a circular logo with a star-like pattern and the word "JWEL" below it. In the bottom right corner, the number "4" is displayed.

Introduction to Copula

Definition

- A copula is an function that joins or couples multiple distribution functions to their one-dimensional marginal distribution functions
- Copulas are distribution function whose one-dimensional margins are uniform

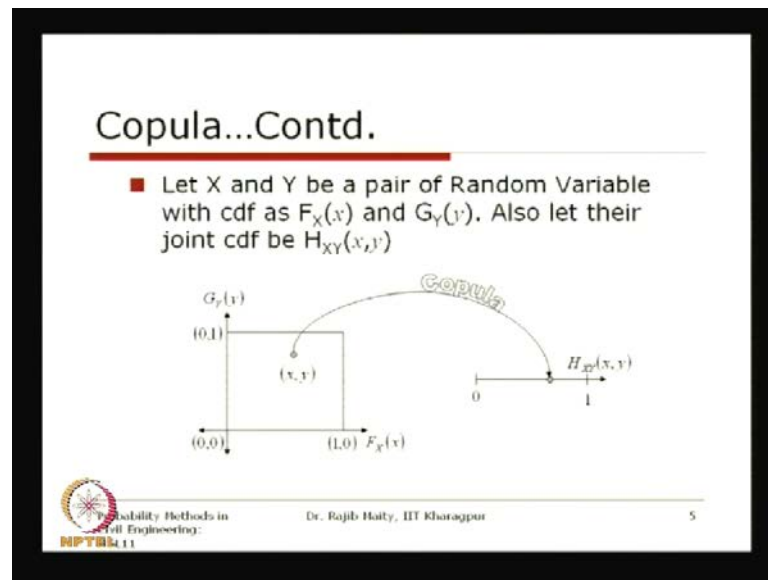
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So, the definition of copula - it says that this copula is a function that joins or couples multiple distribution function to their one-dimensional marginal distribution functions. So, so, one-dimensional marginal distribution for each of this random variable involved is available. So, these has to be joined, this has to be coupled together to obtain their joint distribution and this is done through a function and that function is indeed aa. So, that, that, correspondence is, is, 1 function and that function is nothing but what we are talking about is that copula. So, and the second is that copula are the distribution functions whose one-dimensional margins or the margins means the marginal distributions are uniform.

So, now, this one-dimensional marginal densities are uniform means when we are taking some random variable and if we know there, know there, c d f and, and, if so, from the c d f if we go for the reduced variate, you know that c d f changes from has a range from 0 to 1.

So, now, once we know what is their distribution, what is their cumulative distribution function, and if that cumulative distribution functions properly represents that data, then if we convert that original data set to their through their cumulative distribution function, then that will converted to the range between 0 and 1 and that will become, become, one uniform distribution and that thing can be joined together to get there, to get to obtain that copula that, that, function which is known as copula. So, that is why the copulas are the distribution function whose one-dimensional margins are uniform.

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Now, if we see it more intuitively, say here the examples have been taken for the two random variables only as I was just telling that, suppose that there are two random variables x and y and as I was telling there their cumulative distribution function that the, for x , the cumulative distribution function is $F_X(x)$ and this cumulative distribution function of y is $G_Y(y)$. Now, this $F_X(x)$, so, this, **this**, cumulative function can change can take the values from 0 to 1.

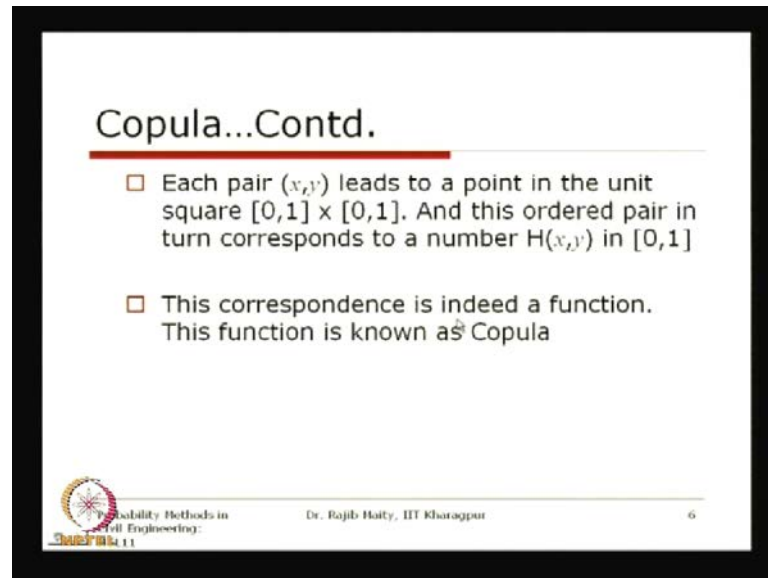
Similarly, that y also can take the value of this 0 to, to, 1. Now, when we are talking about their joint distribution, so any points on this unit square, that is, from 0, 0, 1, 1, this is 0, 1 and this is 1, 1. So, within these unit square, any point basically represents a pair between the x and y , pair of this x and y .

Now, suppose that this $h(x, y)$ represented through this lower case of this letter x, y . So, if this is their joint distribution and joint cumulative distribution, then following the properties of this cumulative distribution, this will also take the values between 0 and 1. So, so, the same, **same**, combination of this x and y will correspond to one point on this unit square over the plane constituted by their, by their, cumulative density function, cumulative distribution function $F_X(x)$ and $G_Y(y)$.

So, this is that point that we are talking about. Again, the same point with the same pair that x and y , the same pair will correspond to another point over this line from 0 to 1 which is their, **their**, joint distribution that is $h(x, y)$. So, so, so these two points - one is


on these units square and other one is on this line between 0 to 1 basically representing the same pair of this x and y . Now, these two points are basically, basically, these two pair basically they corresponds to each other through a function. So, these, these, correspondence the, the, function that makes the correspondence between these two function is what is known as the copula function.

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Copula...Contd.

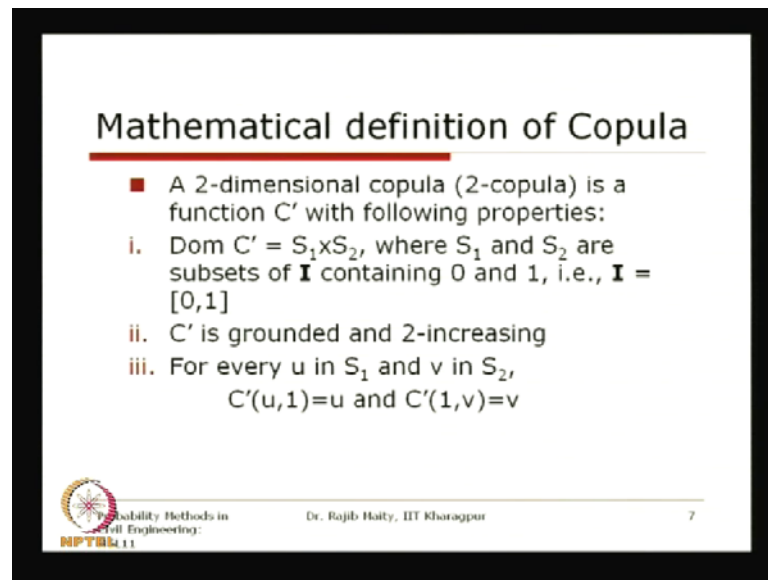
- Each pair (x, y) leads to a point in the unit square $[0,1] \times [0,1]$. And this ordered pair in turn corresponds to a number $H(x, y)$ in $[0,1]$
- This correspondence is indeed a function. This function is known as Copula


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So, that is why that each pair x, y leads to a point in the unit square $[0,1] \times [0,1]$ as we have shown that on this, on the plane made by this x and y . Here, that we have, we have mentioned. So, this point, so, each pair leads to that point, leads to a particular point on this unit square and this ordered pair in turn also corresponds to a, to a, number $h(x, y)$ in that $[0,1]$ real line. So, so, this is through that through their joint distribution. Now, this, both these things, both this point as correspondence, so this correspondence between these two points is indeed a function and this, this, function is, is, nothing but which is known as copula.

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Mathematical definition of Copula

- A 2-dimensional copula (2-copula) is a function C' with following properties:
 - i. $\text{Dom } C' = S_1 \times S_2$, where S_1 and S_2 are subsets of \mathbf{I} containing 0 and 1, i.e., $\mathbf{I} = [0,1]$
 - ii. C' is grounded and 2-increasing
 - iii. For every u in S_1 and v in S_2 ,
 $C'(u,1)=u$ and $C'(1,v)=v$

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Now, if we go to this mathematical definition of this copula, then this, first of all, we will discuss about this, this, two-dimensional cases as we have started. So, these mathematical definitions also here we are discussing in terms of this two-dimensional cases first.

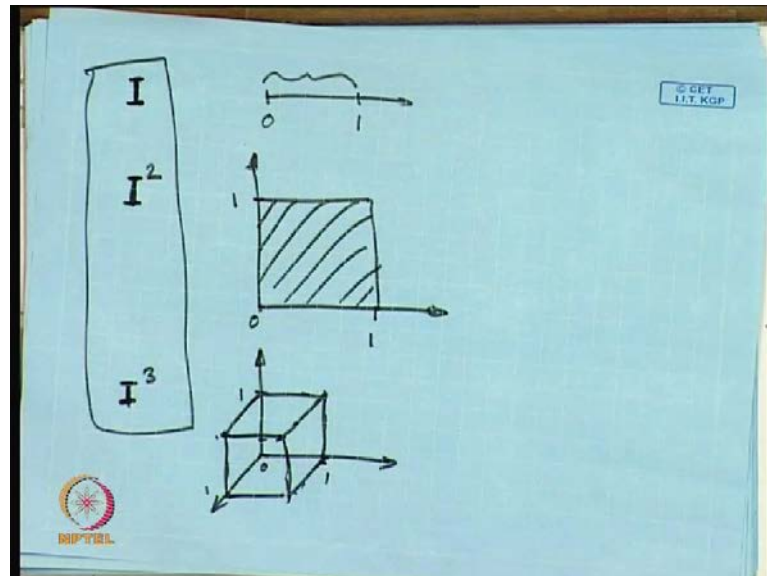
So, so, this two-dimensional copula which is also shortly written as this two copula is a function with the following properties. So, this copula should follow some property and those, there are three properties - the first one is this domain of this C' . So, domain means here you know that when we are talking about a single random variable that time, we have refer to the range of this random variable or the support of the random variable. Here, as it is the minimum is the two-dimensional cases and it can go up to means any dimensional cases.

So, for that function, we are defining is as a, for its entire range, we are just talking about its domain. So, on what domain that we have to specify some domain over which that function should be defined. So, the domain of that, domain of that function C' is $S_1 \times S_2$ - where this S_1 and S_2 are the subset of \mathbf{I} . This bold phase \mathbf{I} containing that 0,1, that is, \mathbf{I} is, \mathbf{I} is the line between 0 and 1.

This is basically what is meant through this mathematical notation is that. So, these domain of this C' for a two copula is the subset of the unit square which is made by these 0,1 in one direction and other direction also it is 0,1. So, this \mathbf{I} is generally a this bold phase \mathbf{I} will be, because we will be using this notation again. So, this bold phase \mathbf{I} is

generally retain and with some power and that power is basically indicating what is its dimension.

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So, here if you see, that is, if we write that this bold phase letter **I**, what is referred to is that is one line up to 0 to 1. So, these range we are referring to this one. So, now, if that bold phased **I** we, **we**, just put that square, that is, in the two-dimensional case, then what we are referring to is that in a two-dimensional space from 0 to 1 and from again other direction 0 to 1. So, we are referring to a square area over this two dimensional case.

Now, if we just, if we put that this bold phase **I** as cube, then you can see that this what we, what is referred is that is one three-dimensional space where each axis so 0 to 1, 0 to here 1 and 0 to here 1. So, you can just see that this will be a cube like this. So, this cube is that **I** see that we are referring to. So, this is, these are the notation what we mean by this by these symbols.

So, here, when we are talking about this domain of this c is that, this is a subset of this, of this, I containing that 01 , containing that 01 . In that, in this case, that is, I it is that 1 to 0 to 1. So, this is that domain of this function, function and means that function means that copula.

Now, the c prime is that function is grounded and two increasing. So, this grounded and two increasing when we are referring to, basically we are giving some, these two

properties that it should follow and it is having some correspondence to thus to the one-dimensional cases also. That we will take both these properties separately to discuss about this, what does this, **this**, terminologies basically means that, what is this grounded, what is this to increasing. That we will see in a minute.

And the third property that it should follow is that, for every u in s_1 and v in s_2 . Now, you can see that there are two that, values that we are referring to for the two-dimensional cases - one is in the direction of this s_1 which is also between that 0 and 1, and other one is in the direction of the s_2 which is between 0 and 1.

So, this c prime u_1 is equals to u ; that means, when the other variable reached to its maximum point, that is, its 1, then the values of this functions should be equal to the value of the other attribute. Similarly, when u reaches to the maximum value, that means it is 1, then the value of this function should be the value of the other, other, attribute, that is, v . So, these are the properties. So, this is the maximum.

Now, you can see that this the last one what we are talking about the maximum limit that we have, we have, just defined, that is, if one of the attribute is reaches to the maximum, what happens to this function? Now, if one of the function is the minimum one, then what is the value of that function. That is basically is, **is**, shown in this word that is grounded and another property is that two increasing. So, we will take this grounded a bit later.

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Basic Terminologies

□ 2-increasing function

- A two dimensional analog of non-decreasing functions in one dimension
- A 2-place real function is 2-increasing if H-Volume of rectangle B ,

$$V_H(B) \geq 0$$

for all rectangles B whose vertices lies in Dom H



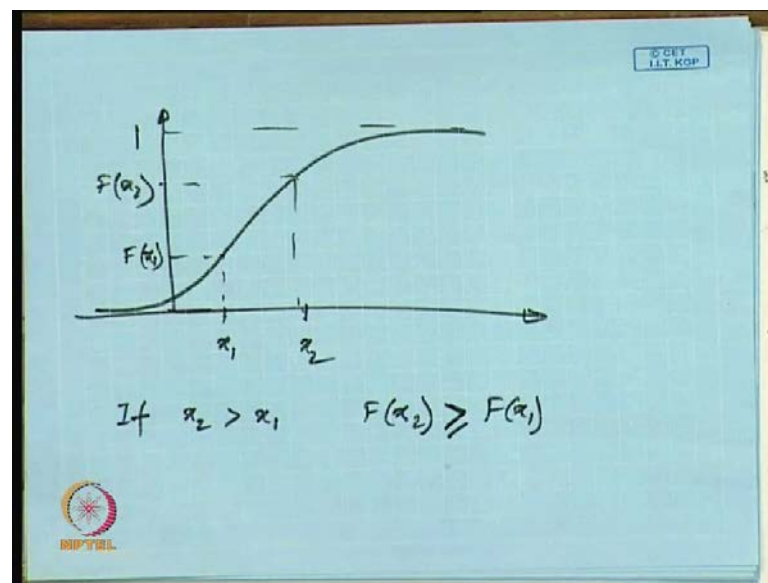
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First of all, we will see that what does it mean that two increasing. So, so this is the first that basic terminologies that we are taking if the first one is this 2-increasing, 2-increasing function. So, what does this 2-increasing, 2-increasing function means? So, a two-dimensional analog of the non decreasing functions is a one dimension. So, this is the first point that we, that we are talking about is that is, **is**, that how it is corresponds? What is its corresponding counterpart in the one-dimensional cases?

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Now, if you see here that for a, **for a**, one-dimensional function, we have seen for its c d f that it starts and reaches up to 1. It can be asymptotic or it may not be asymptotic. It can touch its entire support. It, **if**, should from the left extreme of the support, it should start from 0, and to the right extreme of the support, it should reach to 1. So, that is the range of this c d f.

But the other thing that we have, we have already seen that this function is, **is**, non decreasing function. So, non decreasing means it should, it should, that if that, if this is your say x_1 and this is your say x_2 and we can say that the x_2 is greater than x_1 , then these, then these values that is say $f(x_1)$ and this is $f(x_2)$. Then what we see that, if x_2 is greater than equals is, **is**, greater than x_1 , then that $f(x_2)$ is greater than $f(x_1)$; so, that means it is an increasing function; it is an strictly increasing function, but we should include one equality sign here also just to, **to**, state that this is a non decreasing function.

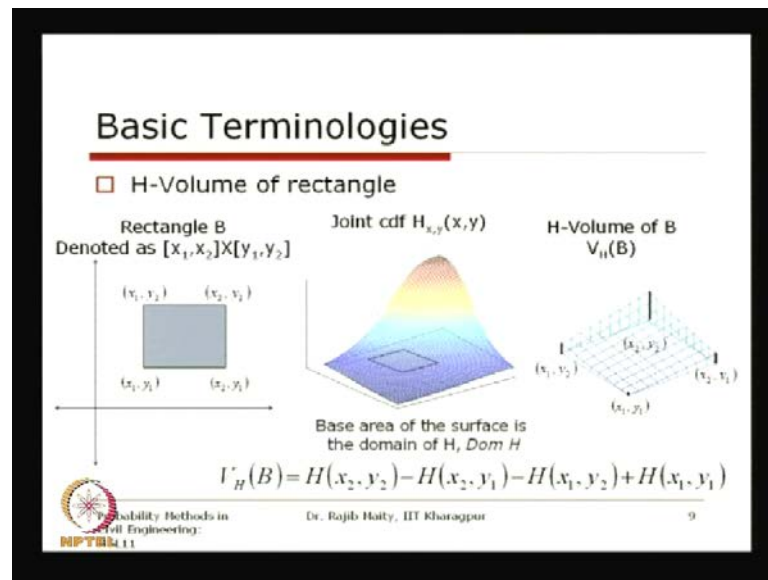
So, it does not mean that whether between these two point, it will increase or not; it can even remain flat for this reason and it can again go, but it will never come down. So, this will be either greater than or it should be, at least it should be equal between these two. The values of these functions at point x_2 and at point x_1 . So, so this is what we have seen in this, in this, one-dimensional cases.

Now, if I want to extend the same property to the two-dimensional cases, then what is known as is that, is the, is known as the two increasing functions. So, what does this? So, this is a, this is the correspondence between the one-dimensional cases what we have we have seen earlier, and the corresponding thing the same, same corresponding property in the two dimensional cases is known as that properties it is 2-increasing, 2-increasing functions.

Now, the a two place real function is 2-increasing if h volume of the rectangle b is greater than equal to 0. So, again, here we have got two more new terminologies - one is that h volume of the rectangle b . So, first of all you have to know the, which rectangle we are talking about, and if we know what is the rectangle, what does this h volume means?

So, that we will see and this is for this rectangles b and this should be valid for all rectangles b whose vertices lies in the domain of h . So, we have to know that. So, domain we have seen that it should be the domain of the c that we know that it is within this 0_1 and 0_1 is bound, and we have to know that in this domain if we can identify one rectangle and what is that h volume means and that h volumes should be greater than equal to 0 to satisfy the condition that the function is 2-increasing.

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Now, let us discuss about this what is this h volume means. Now, let us see this a, this area. First, that is the one rectangle, that is, whose vertices are the x_1, y_1 ; this is x_2, y_1 ; this is x_1, y_2 and this is x_2, y_2 , and if we say that this, **this**, rectangle is in the domain between these 0 and 1, that is the domain of that function, and if from this rectangle also you can say that or if we just specifically mention that x_2 is greater than x_1 and y_2 is greater than y_1 . Then this is this rectangle what we can say in this, in this, form. Now, so this is the plain by that two random variable. Suppose that this side it is x and this side it is y . Now, the same rectangle is again represented here on that plane. So, this side is your x and this side is your y and this is that rectangle that you can see here.

Now, to see that, what is the h volume we, **we**, mean that, for that one, we have to first see that how that how that joint cdf between that x, y looks like. So, this is as you know that this should start from 0 and it will go on and reach to a maximum value to that reach to a maximum value of one. So, so, this is your that $h \times y$ axis and this should be your here 0 and this one, this line it should be written as 1 which is missing in this figure, but you can, you can, easily make out that this is the starting point is your 0 and this is going up to a maximum value of this 1. So, this surface that you can see. Suppose that, that is, 1 that is the joint cdf function in a two-dimensional case. So, two-dimensional function that means this is a surface.

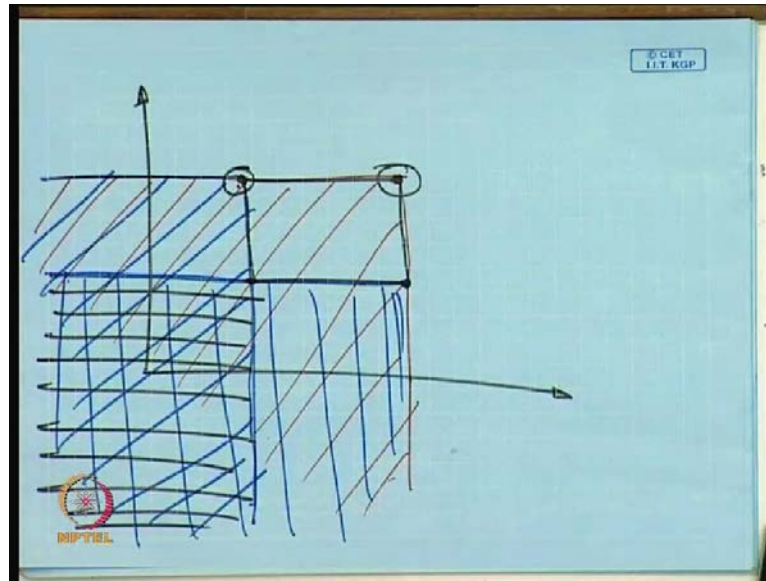
Now, that h volume when we are talking about, so it is just like that one surface is given there and you are cutting that surface through this, through the lines of that rectangle which is shown on the base. Base means on this surface considered by x and y . So, now, if you cut this, **this**, surface is just like a slicing the cake along this lines. So, what you will get is a volume like this. Now, if you get the volume like this, then what you are getting? So, you will get four ordinate at this four vertices of the, of the, rectangle shown on this, shown on this ground means, ground means here on the surface.

Now, that value that we are talking about is your this $h \times 2y^2$ is nothing but is the what is the ordinate at the point $x = 2y^2$; that means, at this, at this point, what is the, what is the value of that joint cdf function which is shown by a black vertical line here. So, now, that h volume of this rectangle is this one, this height minus this height minus this 1 plus this, **this**, height at $x = 1y^2$. So, if we just go back to this original rectangle, then the value of the function at this point minus value of the function at this point minus value of the function at this point plus value of the function at this point. So, this is giving you the h volume of that function h volume over the rectangle b which, for which the vertices are as shown in this figure that is $x = 1y^2, x = 2y^2, x = 1y^2$ and $x = 2y^2$.

Now, if just, if it is, it is not that difficult to even remember that this is what we are taking is that, first we are starting from this extreme right top corner minus other two diagonals plus this one, but if you say that why this one it is also, it can also be easily referred in terms of the cdf . Basically, what does it mean? The, what is this ordinate here at this $x = 2y^2$ means. If we see that, this is the joint cdf , that is, cumulative distribution function of this one. Then what, **what**, this ordinate means is nothing but what is the total volume covered by the pdf - probability density function - up to this point.

Now, instead of considering this cdf over this plain, if you just consider what is the c , what is the pdf , there is the probability density function over this one. So, what you will get is that, that, so that it can be you can see that as a , as a , surface again over this zone, over this area means depending on the support of this two random variable x and y and what is the total volume that is a covered up to this point which is represented by this ordinate when you are talking this, when we are taking this surface as its cdf .

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Now, if I just represent that same rectangle here and if we want to see that, if, for this rectangle if we want to know, what is the volume over this rectangle? When I am considering a, considering the, their joint p d f over this zone, then what will happen? For this one, for this point to get that, what is the volume above this rectangle? Then what I have to do that, I have to first take what is the total volume **support that**, suppose that this is your, the left extreme of the random variable that we are talking about. Then or, **or**, I can even just say that why to take that. It can be say that minus infinity. Then what I am talking that, I am first taking the full area up to this point. What is the total volume up to this, up to this point? **This should be...**

Now, to get only this rectangular area, then I have to deduct this area first. Let us use other colour. So, this total area minus this area minus this area; that means now you see that in, **in, in**, this process, this area has been deducted twice. So, this area should now be added which is nothing but the total volume up to this point. So, that is why the total volume under the p d f, remember, total volume under the p d f - probability density function - up to this point minus total volume under the p d f up to this point.

So, this area I have first deducted; then this area is again deducted. Then, I have to add this area because this area has been deducted twice while deducting this point and this point, and this is why this, when we are talking about this age volume; that means, that time we are taking that this area minus. So, this, **this**, ordinate, this ordinate is nothing but

the total volume under the p d f that just now what we discussed, so, this one minus this ordinate minus this ordinate plus this 1.

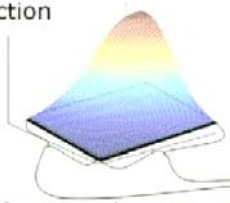
So, this gives you the h volume over this rectangle. Now, **now**, if I want to say that this function is always increasing in both the direction in, in, the direction in the, when this x_1 , when, what I say that that x_2 is greater than x_1 and y_2 is greater than y_1 . Then this volume should be the volume below that p d f should be greater than equal to 0. So, that is why so this volume should be greater than equal to 0, which is represented as, **as**, $v_{h,b}$ and that is why the requirement there is that $v_{h,b}$ should be greater than equal to 0. So, this is what actually is the meaning of this 2-increasing function.


But remember one thing that only this condition, that is, $v_{h,b}$, that is, h volume of this b if it is 2-increasing, that does not guarantee that each every direction, that is, x and y both the direction that function also will be a non decreasing that is not assured. So, to assured, but we know that the marginal's that the marginal distribution also is a non decreasing function. To assure that, that one what is the other condition is there that also should be satisfied. Then that other condition is that it should be grounded.

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Basic Terminologies

- ☐ Grounded function
- ☐ A 2-increasing function does not imply that H is nondecreasing in each argument.
- ☐ However, if it is 2-increasing and grounded H is nondecreasing in each argument also.





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So, what does that grounded function means? So, that is here; that is shown here, that is, this grounded function here means that if two increasing functions does not imply that h is non decreasing in each argument; however, if it is 2-increasing and grounded, the h is non decreasing in each argument also. So, these are the after we discuss this grounded;

these are the, the, the, conclusion, but before that, what is this grounded function means. So, you see here this, this, line that we are talking about and this line is the value of that function h is 0 at these line segments.

So, what does this line segment means, that is, when one of the argument is 0, either u or v means which are the reduced variate for the x and y . If either one of this u and v are 0, then the value of the function is 0 and that is what is known as the grounded function, that is, at both these line segment, the function is starting from the ground. So, now, if I go back to that one of this, this, this, property, what we discuss here that, if one of the values are reaching to its maximum value, then value of the function is equals to the value of the other attributes, that is, c'_{u1} is equals to u and c'_{v1} is equals to v .


Now, the, by the word grounded what we mean is that, if one of the argument is 0, then the value of that function equals to, will be equals to 0, that is, c'_{01} is equals to 0 and c'_{0v} is equals to 0. So, irrespective of what is the value of the other one; that means, here I should say that c'_{u0} equals to 0 and c'_{0v} equals to v , sorry, equals to 0. So, this is what is meant by this grounded function which is represented through this diagram here over this line segment. This, the value of this functions should be equals to 0.

Now, as I we are just telling that it is simply if we just say that a function is 2-increasing, that does not imply that the in both the argument, both the argument means in both the random, in the direction of the random variable is does not guarantee that whether they will also be non decreasing or not, but if we say both the properties that it is 2-increasing as well as it is grounded, then we can say that it is non decreasing at each argument also. So, which is obvious because their marginal's are also one $c_d f$ which is non decreasing. So, that their joint function that is the property of the, property of the, copula function should be 2-increasing as well as it should be grounded.

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Basic Terminologies

- A copula (2D) is a function with domain is \mathbf{I}^2 and following properties:
 1. For every u, v in I ,
 - $C(u,0)=0=C(v,0)$ (Grounded)
 - And
 - $C(u,1)=u; C(1,v)=v$
 2. For every u_1, u_2, v_1, v_2 in I such that $u_1 \leq u_2$ and $v_1 \leq v_2$,
 - $C(u_2, v_2) - C(u_2, v_1) - C(u_1, v_2) + C(u_1, v_1) \geq 0$ (2-increasing)



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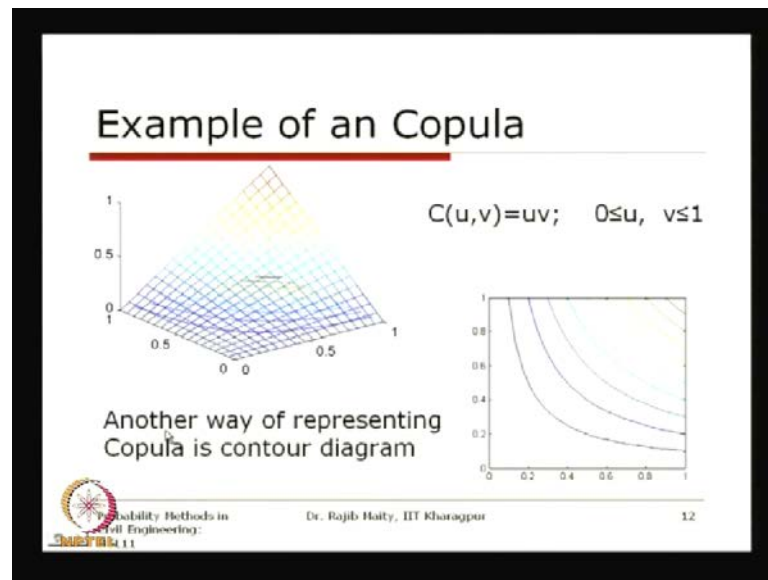
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So, a copula, and here, it is the two-dimensional case is a function with the domain is I square here. Now, I square just a few minutes before I have discussed. So, it is a two-dimensional plain bounded by 0 and 1, that is, vertices are 00, 01, 10 and 11. So, it should have the following properties - one is that this is grounded as I told; it is $C(u, 0)$ equals to 0 and $C(v, 0)$ equals to 0, and other one is that, when it is reaching one of the argument is reaching to its maximum value, that is, $C(u, 1)$ equals to u and $C(1, v)$ equals to v .

For every $u_1 \leq u_2, v_1 \leq v_2$ in I such that u_1 is less than equals to u_2 and v_1 is less than equals to v_2 , then the copula function at u_2, v_2 minus the value at u_2, v_1 minus the value at u_1, v_2 plus the value at u_1, v_1 should be greater than equal to 0, which is nothing but it is ensuring that the copula function is 2-increasing and this one this whole statement is nothing but the 4-volume.

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So, one example of this copula which is, you know that we use it also earlier, that is, that independent. If the two random variables are independent, then we know that their joint distribution is nothing but the product of their individual marginal distribution. So, say that if, so, where that so, the one so that particular copula function will be is equals to your $c(u,v)$ equals to u multiplied by v . So, when we can say that their joint distribution is their product of their individual marginal distribution, then we know that this is their, **their**, joint distribution is the product of their individual, individual, marginal distribution.

Now, this is also then, this is also one, also one, one, copula and which is known as the independent copula. So, this function is known as the independent copula where this u range of this u and v both are varying from 0 to 1. So, and if we represent this one how this copula function looks like, it will look like this; it will start. You see that both the functions are, **are**, it is grounded. Both the functions are starting from this line segment. Over this line segment, the value is 0 which you can also cross check from here that just, if we put that one value is equals to 0, then the value of the function is becomes, value of the function becomes 0.

Then, if one of the function reaches to its maximum value, that is, if u equals to 1 and v is in any value, then the value of the function is v and vice versa. So, here, that is reflected through this line which is a, which is a straight line, 45 degree straight line

through this plain. This is that direction of this copula function; this is one of this random variables. So, this line is a 45 degree line over this plain. Similarly, this line is also a 45 degree line over this plain which is the other random variable verses that function of this copula.

So, at 0.5 and this 0.5 you can see that if one is reaching one and other one is 0.5, this value of the function is 0.5. This just for one example and this is true for entire this straight line and which is also can be cross check from, [from](#), here. So, if we just put one value equals to one, the value of the function is the equal to that value of the other attribute.

So, this is also a surface. So, this is a copula function which is satisfying the, it is required property and that function looks like this surface that you can see here and also the another way of representing the copula is the contour diagram and the contour diagram is shown like this where you can see that this different contours are shown to with this different color bar. Here, the color bar is not mentioned here, but you see that this blue one is 0 and this red one is 1. So, in this way, it is gradually increasing; obviously, the copula this values of this contour should be mentioned here, [here](#), to properly represent what is its, what is height above the ground of this surface.

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Fréchet-Hoeffding Bounds


□ Let C' be a copula. Then for every (u,v) in $\text{Dom } C'$

$$\max(u+v-1, 0) \leq C'(u,v) \leq \min(u,v)$$

or

$$W(u,v) \leq C'(u,v) \leq M(u,v)$$

where $W(u,v) = \max((u+v-1, 0))$ and $M(u,v) = \min(u,v)$



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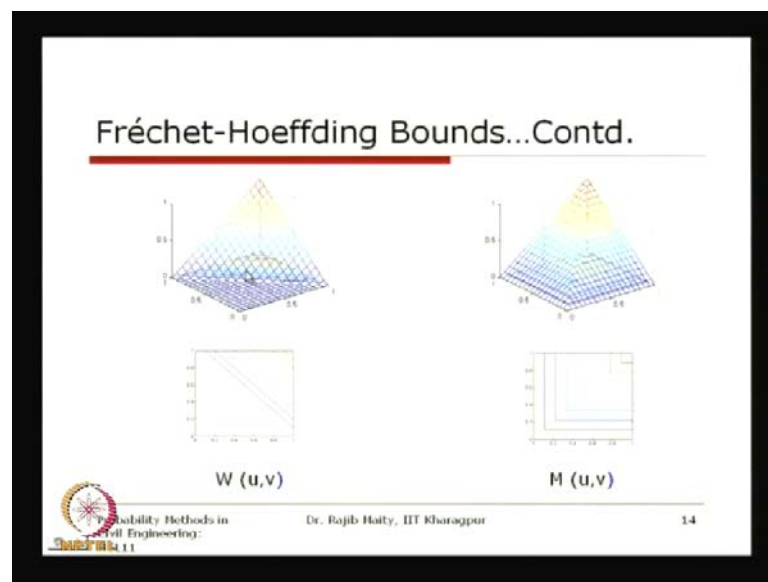
Now, there are one more property, [property](#), related to the bound of this each copula that we should discuss now. So, this is known as that Fréchet-Hoeffding bound - f h bound -

abbreviated form that this f_h bound it says that the any copula function should be bounded by, by two, two functions and those functions are that the minimum the lower bound is that maximum of u plus v minus 1 comma 0. So, i, so, in this 2, u plus v minus 1 or 0 whichever is maximum. So, this is the lower bound of that function and minimum of v this is the value of the, the, upper bound.

So, this is generally represented by this $w(u, v)$ and this maximum one is represented by this, the upper limit. Upper limit is the minimum of u and v ; the upper limit is represented by this $m(u, v)$. So, why this is w and why this is m that is no, these are just the notation, but you can just for the sake that this is the lower bound. That is the lower side the, the, way the word w is written that the lower side is just bounded and the way the m is written, it is upper side is bounded. This is just for the notational sake only. So, this is that your lower bound and this is your upper bound of the copula function.

So, that this lower bound is the maximum of this u plus v minus 1 and 0 and the upper bound is the minimum of u and v . So, how does this, how does this bounds look like? That we will see now. So, first we will take this lower bound here.

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So, these are the two surfaces that we have drawn, is that this is the lower bound and this is the upper bound and these are their corresponding contour diagram. So, now, you see that whenever we are talking about this lower bound on this floor of this 01 and 01 on this floor if I just draw the diagonal here, then up to that diagonal, the surface is on the

ground, and from there, from this diagonal of this base, we are just connecting to this upper side other side corner. So, this is a, this is a, this is a, straight surface.

So, this is the bound the, this surface and this one is the lower bound. This we can, we can also cross check, that is, when we are just talking about say that u equals to 0.5 and v equals to 0.5, that time the value of the surface should be at 0. So, if we just see one or two such values, then u equals to 0.5, v equals to point five which is equals to 1 minus 1 is 0. So, maximum of 0 and 0, so which is 0. So, in this way, if we take whatever the combination that we take which is coming to this, to the, to the, this side of this diagonal of any combination of this u and v , then we will see that this value will become 0.

So, which indicates that this surface is basically on the ground up to this diagonal, and after that, it is going and touching this values where this side you can see that this is also means, when one is reaching to this maximum one, the value itself is that value of the other attribute. So, that, suppose that just put one of this value as 1. So, the v equals to one. So, this 11 is cancel. So, the maximum of $u, 0$ and you know that u is then u is always greater than equal to 0. So, maximum value is u . So, so that is, sorry. So, that is what is reflected here that when one attribute is 1, the value of the surface is at the value of the other one, other attribute.

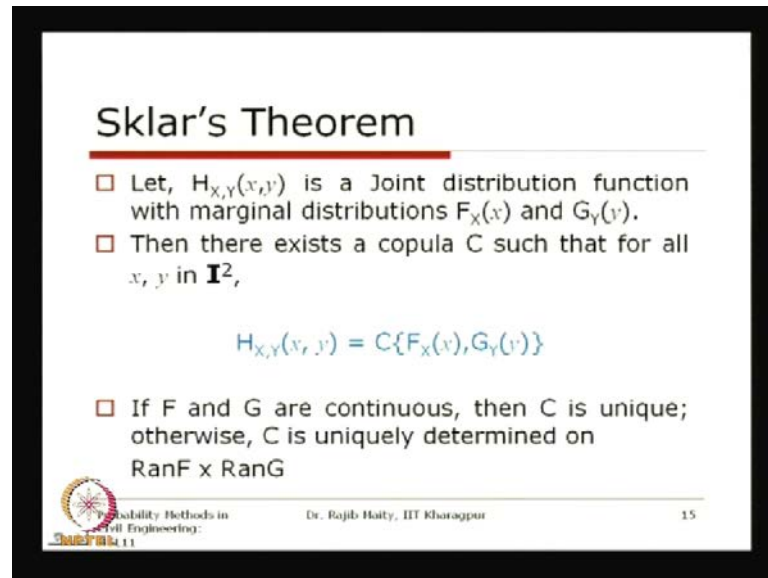
So, this is the lower bound, and similarly, the upper bound also we can see this is the minimum of u and v . So, just if we take any, any value, say that one is say that point five and other one is say 0. So, this is the minimum of this 2; that means, at this point, the value will be 1, and if I take that 0.5 and 0.5 both, so, it is the minimum of both the values. So, it will go and this height there at this point, 0.5 0.5 also will be 0, which will be more clear through this contour diagram. You can take any combination of these two and their point will be the, will be the minimum of the combination.

So, on this line, the minimum value is always 0.1. So, that is why what of irrespective of the value of this axis, this value is always 0.1. Similarly, in this way, we can just say this is one straight this is just like a half of a pyramid that kind of surface you can see that this is 1 surface of the pyramid and this is another surface of the pyramid. So, that is the way this upper bound is defined.

So, what is important here for the copula function is that. So, now, you can see that this is what we are talking about is a lower bound and this is what we are talking about the

upper bound. So, any surface which is representing a copula function should lie in between these two surface. So, basically I can place this surface over this and you can imagine that there will be a, there will be a, empty space between this two surface and all the copula function should lie in that space and that is what this upper bound and the lower bound of the copula function pictorially represents.

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Sklar's Theorem

- Let, $H_{X,Y}(x,y)$ is a Joint distribution function with marginal distributions $F_X(x)$ and $G_Y(y)$.
- Then there exists a copula C such that for all x, y in \mathbf{I}^2 ,

$$H_{X,Y}(x,y) = C\{F_X(x), G_Y(y)\}$$

- If F and G are continuous, then C is unique; otherwise, C is uniquely determined on $\text{Ran}F \times \text{Ran}G$

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Now, the Sklar theorem which is the most important theorem in this theory of this copula, through which we can connect or we can join, we can couple their marginal distribution to have their joint distribution. Let that $h(x,y)$ is a joint distribution function with marginal distribution $f(x)$ and $g(y)$. Then there exist a copula c such that for all x,y in \mathbf{I}^2 , you know that what is this \mathbf{I}^2 we defined earlier. So, for all x,y combination, that $h(x,y)$ this joint distribution is equal to the, that c of, that is a function of that $f(x)$ and $g(y)$. If f and g are continuous, then c is unique; otherwise, c is uniquely determined on the range of f and range of y . So, this is consisting that surface over this range, over this area, this should be uniquely determined.


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Sklar's Theorem...Contd.

□ Conversely, if C is a copula and $F_X(x)$ and $G_Y(y)$ are CDFs, then

$$H_{XY}(x, y) = C\{F_X(x), G_Y(y)\}$$

is the joint CDF with margins $F_X(x)$ and $G_Y(y)$



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On the, so, this is the, **the**, Sklar theorem. In the other, in the conversely also we can state that if c is a copula and f_X and g_Y are the cdf's of their random variable, obviously the x and y , then their joint distribution is equals to the copula function of $f_X(x)$ and $g_Y(y)$. So, if we have that f_X and g_Y , then we can get that this, through this copula function, we can obtain their joint distribution which is a, which is H_{XY} represented through this x and y . So, this is the joint cdf of the random variable x and y with the margins means the marginal distribution $f_X(x)$ and $g_Y(y)$.

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Measure of dependence

□ Spearman's correlation coefficient OR Spearman's rho, ρ_s


$$\rho_s = 12 \iint uv dC(u, v) - 3$$


□ Kendall's correlation coefficient OR Kendall's tau, τ

$$\tau = 4 \int C(u, v) dC(u, v) - 1$$

□ τ and ρ_s follow the equality

$$-1 \leq 3\tau - 2\rho_s \leq 1$$





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Now, as I was telling that we will be taking the measure of dependence through which basically this copula function should be defined because there is a parameter in the copula function that generally preserve that scale free measure of dependence, so that the two scale free measure of dependence we will discuss here now.

Now, the first one is the Spearman's correlation coefficient or the Spearman's rho, ρ_s , this should be ρ_s , this is the ρ_s which is the, through this copula, it is expressed that it is twice of this double integration of $u v d c$. So, with respect to c , that copula function if we integrate this function minus 3, this is the form of the Spearman's rho which can be derived from this, from this copula.

Similarly, the Kendall's Tau, there is the Kendall's Correlation Coefficient, Kendall's Tau is equal to the four times the integration of this copula function with respect to $d c$ minus 1. So, this is representing that Kendall's Tau, and this Tau and ρ_s follow the inequality that this three times Tau minus 2 times ρ_s this is generally bounded by from this minus 1 to plus 1.

So, even though these functions that we have discussed in terms of this through this copula, **copula**, function, this 2, this 2 major of scale free dependence, but for the, for the real, for the real time, **time**, series or the data that we are having. If we want to know what is their scale free dependence, then we have to go for some sample estimate, and from the sample, we will first estimate what is the Kendall's Tau, and from that Kendall's Tau, we will first find out what is the parameter of the copula function that we are taking, and after that, we can obtain what is the complete definition of the copula for the data in hand.

So, in the next class, what we will do? We will just, we will discuss first how to estimate this measure of dependence from the samples and we will discuss a particular class of copula which is known as the Archimedean copula and that and there we will show that how we can estimate from the sample and get the complete form of this copula function, which can be used to, and ultimately, we will be, we will be deriving their complete form of this joint p.d.f. So, we will discuss these things in the next class. Thank you.