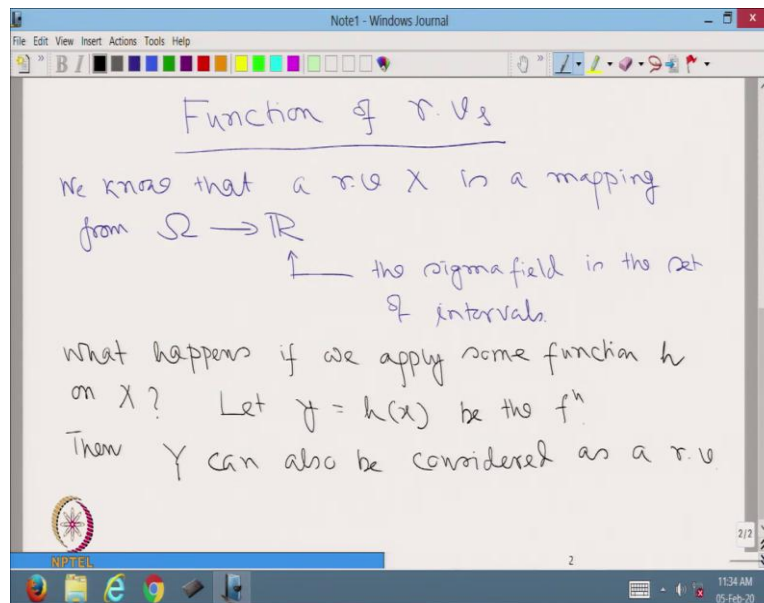


Advanced Probability Theory
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Lecture 19

Welcome students to the MOOCS lecture series on Advanced Probability Theory, this is lecture number 19.

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As I said at the end of the last class that from today, I shall start function of random variables. We know that a random variable X is a mapping from Ω to \mathbb{R} on which the sigma field is the set of intervals. Now, what happens if, we apply some function h on X ? Let, y is equal to $h(x)$ be the function. Then Y can also be considered as a random variable.

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Y : a mapping from Ω to \mathbb{R} as well.

i.e.

$\Omega \xrightarrow{X} R \xrightarrow{Y} \mathbb{R}$

Question is how to find the distribution of Y from the distribution of X !

We have already seen such things.

For example $V(x) = E(x^2) - (E(x))^2$

↑ This is a function of x & we obtained its Expectation.

Because, Y is a mapping from Ω to \mathbb{R} as well. That is, we have Ω from there we get X which maps to R and from there we have the mapping Y that is also on R . Question is, how to find the distribution of Y from the distribution of X ? That is the question.

We have already seen such things. For example, variance of X is equal to the expected value of X square minus expected value of X whole square. Now, obviously this is a function of x and we obtained its expectation. This type of problem we have already seen where from that distribution of X we are trying to find the expectation of X square or we can think of getting the PDF of x square that we have also seen.

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How to do it in general?

Discrete Case

Let Ω be not $h(x_{i1}), h(x_{i2}), \dots, h(x_{ik}) \rightarrow y_i$
 $i = 1 \text{ to } m$

Then $P(Y = y_i) = \sum_{j=1}^k P(X = x_{ij})$

We already have an X is random variable whose pmf is known.

How to do it in general? So, consider discrete case. Suppose, this is Ω and X is such that all these points are mapping to say x_1 some other set of Ω s are mapping to x_2 and some other set of points are mapping to x_n and that is how X behaves, what we are saying, we are now, looking at the mapping of Y and some points from here are mapping to say y_1 , some other points are mapping to say y_2 and some other points say mapping to y_m .

And suppose, this is the function h . Therefore, let h be such that h of x_{i1} , comma h of x_{i2} , h of x_{ik} they all map to y_i , i is equal to 1 to m . Then, probability Y is equal to y_i is equal to \sum , probability X is equal to x_{ij} , j is equal to 1 to k and do we know that, these probabilities we already have as X is a random variable whose PMF is known.

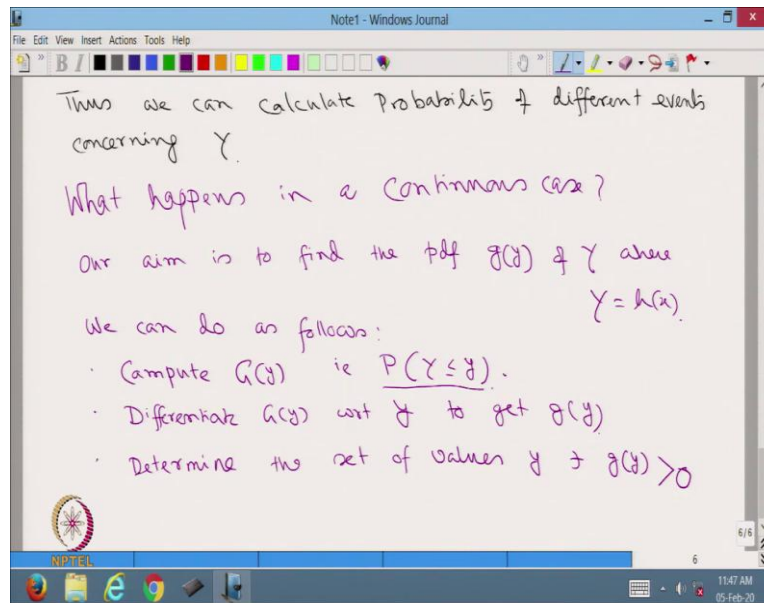
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The image shows a digital whiteboard with handwritten mathematical notes. At the top, it says 'Ex $X \sim \text{Bin}(10, \frac{1}{2})$ '. Below that, it says 'Let $h(x) = 3x + 1$ $\therefore Y = h(X)$ is a r.v.'. Then it lists 'Y takes values 1 4 7 - - - 31'. The next line is ' $P(Y=1) = P(X=0) = \binom{10}{0}(\frac{1}{2})^{10}$ '. Below that is 'etc.'. Then it asks 'What is the probability of the event $Y \leq 10$ '. The next line says 'We know that this is $P(X=0) + P(X=1) + P(X=2) + P(X=3)$ '. The final line is a calculation: ' $= (\frac{1}{2})^{10} + 10(\frac{1}{2})^{10} + 45(\frac{1}{2})^{10} + 120(\frac{1}{2})^{10} = \frac{176}{1024} = \frac{11}{64}$ '. The whiteboard has a toolbar at the top with various drawing tools and a taskbar at the bottom with icons for NPTEL, a folder, and a clock showing 11:44 AM on 05-Feb-20.

So, let us give an example, consider X to binomial 10, comma half. Let h of x is equal to $3x$ plus 1 therefore, Y is equal to hx is a random variable where Y takes values 1 4 7 up to 31 with the probabilities that we can understand that probability Y is equal to 1 is equal to probability X is equal to 0 is equal to $10c0$ half to the power 10 etc. So, what is the probability of the event Y less than equal to 10?

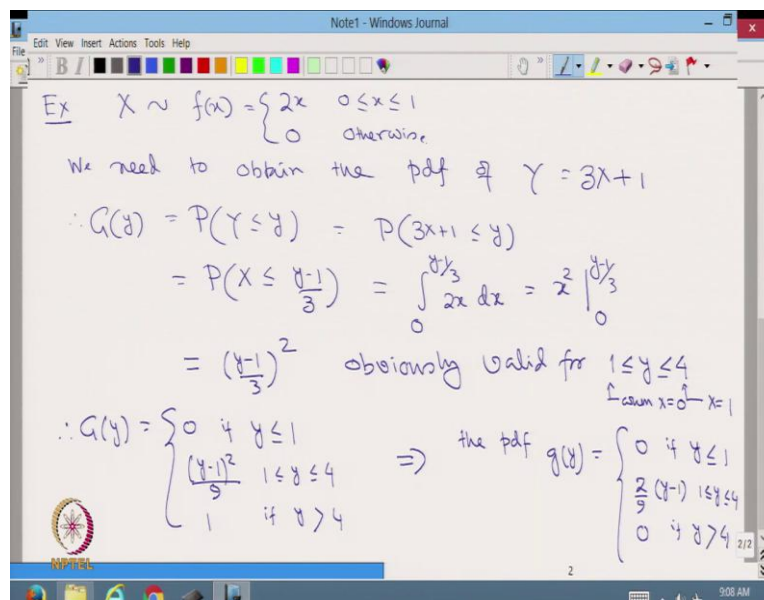
We know that this is probability X is equal to 0 plus probability X is equal to 1 plus probability X is equal to 2 plus probability X is equal to 3 if x takes value beyond 3 then Y will be greater than equal to 10 is equal to half to the power 10 plus 10 half to the power 10 plus $10c2$ which is going to be 45 half to the power 10 plus $10c3$ which is going to be 120 into half to the power 10 is equal to 176 upon 1024 which is equal to 11 upon 64 .

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Thus, we can calculate probability of different events concerning Y . What happens in a continuous case? So, our aim is to find the pdf $g(y)$ of Y where Y is equal to h of x , one way of doing it is first compute G of y that is probability Y less than equal to y . Differentiate G_y with respect to y to get g_y . Determine the set of values of y such that, g_y is greater than 0. But, the main question is how to get probability capital Y is less than equal to small y ?

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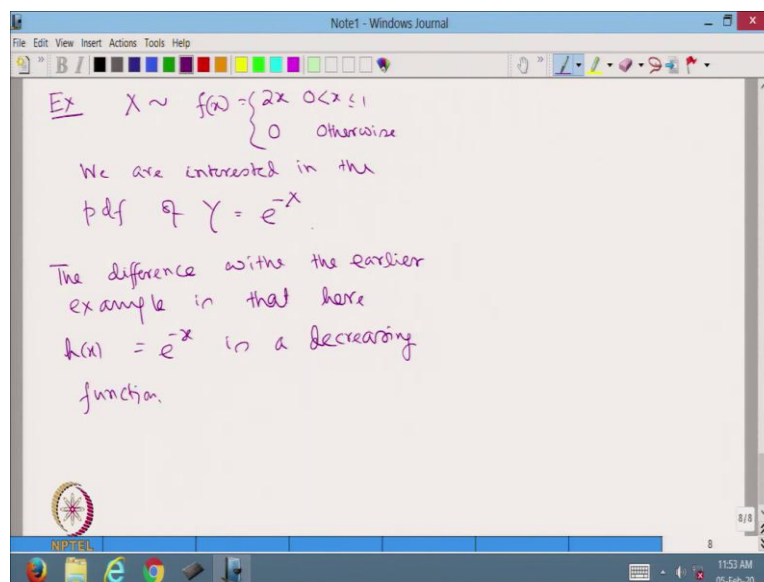


Let us start with an example, X is distributed as follows that f of x is equal to $2x$ when 0 less than equal to x less than equal to 1 and it is 0 otherwise. Therefore, we need to obtain the pdf of Y is equal to $3x$ plus 1 . Therefore, G of y is equal to probability Y less than equal to small

y is equal to probability $3x$ plus 1 less than equal to small y is equal to probability X less than equal to y minus 1 upon 3 is equal to integration 0 to y minus 1 by $3 \cdot 2x \, dx$ is equal to x^2 from 0 to y minus 1 upon 3 is equal to y minus 1 upon 3 whole square, obviously valid for 1 less than equal to y less than equal to 4.

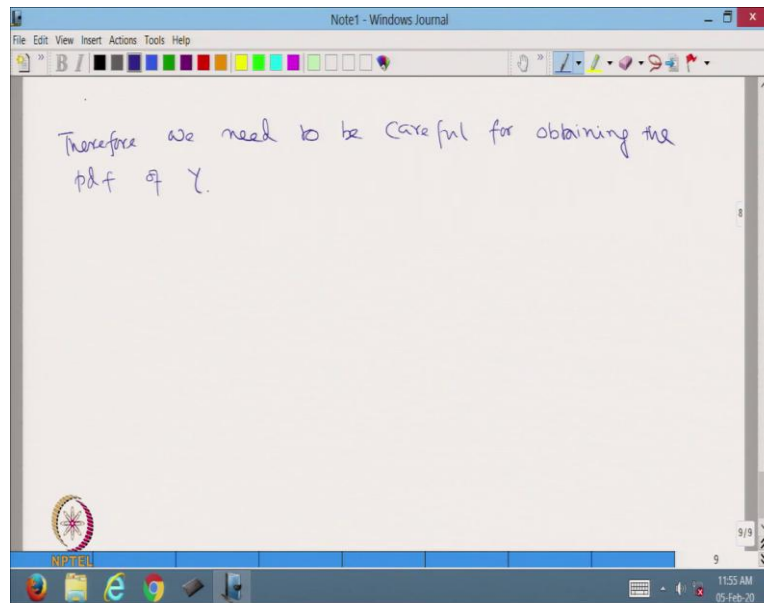
So, y takes the value 1 when X is equal to 0 and it takes the value 4 when X is equal to 1. Therefore, G of y is equal to 0 if y less than equal to 1, y minus 1 whole square by 9 if 1 less than equal to y less than equal to 4 and is equal to 1 if y greater than 4, implies that the pdf g(y) is going to be 0 if y less than equal to 1 is equal to $\frac{2}{9}(y-1)$ for 1 less than equal to y less than equal to 4 and is equal to 0 if y greater than 4.

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Let us consider another example, the same X which is f_X is equal to $2x$ for 0 less than x less than equal to 1 and it is 0 otherwise. And we are interested in the pdf of Y is equal to e to the power minus X. The difference with the earlier example is that here, h_X is equal to e to the power minus x is a decreasing function.

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Therefore, we need to be careful for obtaining the pdf of Y.

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$$\begin{aligned} G(y) &= P(Y \leq y) = P(e^X \leq y) = P(-X \leq \ln y) \\ &= P(X \geq -\ln y) \\ &= 1 - F(-\ln y) \\ &= 1 - \int_0^{-\ln y} 2x \, dx \quad (\text{integrating the pdf of } X) \\ &= 1 - x^2 \Big|_0^{-\ln y} \\ &= 1 - (-\ln y)^2 = 1 - (\ln y)^2 \\ \therefore g(y) &= \frac{d}{dy} [1 - (\ln y)^2] = -2 \ln(y) \cdot \frac{1}{y} \end{aligned}$$

Note that in $[0, 1]$ $\ln(y)$ is negative \therefore The expression is positive

So, G of y is equal to probability Y less than equal to y is equal to probability e to the power minus X less than equal to y is equal to probability minus x less than equal to log y is equal to probability X greater than minus log y is equal to 1 minus F of minus log y is equal to 1 minus integration 0 to minus log y of 2x dx this is the pdf of X is equal to 1 minus x square from 0 to minus log of y is equal to 1 minus, minus log y whole square is equal to 1 minus log y square. Therefore, g of y this is the pdf of Y is equal to minus 2 log y multiplied by 1 upon y. Note that, in 0 to 1 log y is negative. Therefore, the expression is positive.

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Let us look at the general problem in another way:

For example 1 : $h(x) = 3x + 1 = y$

$$\therefore G(y) = P(Y \leq y) = P(3x + 1 \leq y) = P\left(x \leq \frac{y-1}{3}\right)$$

$$\therefore g(y) = \frac{dG(y)}{dy} = \frac{dG(y)}{d\left(\frac{y-1}{3}\right)} \cdot \frac{d\left(\frac{y-1}{3}\right)}{dy} = \frac{F\left(\frac{y-1}{3}\right)}{d\left(\frac{y-1}{3}\right)} \cdot \frac{d\left(\frac{y-1}{3}\right)}{dy}$$

$$= f\left(\frac{y-1}{3}\right) \cdot \frac{1}{3} = \frac{2(y-1)}{3} \times \frac{1}{3}$$

$$= \frac{2(y-1)}{9}$$

Let us look at the general problem in another way. Say, for example 1, h of x is equal to $3x$ plus 1 is equal to y . Therefore, G_y is equal to probability y less than equal to Y is equal to probability $3x$ plus 1 less than equal to y is equal to probability X less than equal to y minus 1 upon 3 .

Therefore, G_y is equal to dG_y dy is equal to dG_y dy minus 1 upon 3 multiplied by dy minus 1 upon 3 dy which is is equal to F at y minus 1 upon 3 d y minus 1 upon 3 into d y minus 1 upon 3 dy is equal to F at y minus 1 upon 3 multiplied by 1 upon 3 is equal to 2 into y minus 1 upon 3 multiplied 1 upon 3 is equal to 2 into y minus 1 upon 9 .

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Ex $X \sim f(x) = \begin{cases} 2x & 0 \leq x \leq 1 \\ 0 & \text{otherwise} \end{cases}$

We need to obtain the pdf of $Y = 3X + 1$

$$\begin{aligned} \therefore G(y) &= P(Y \leq y) = P(3X + 1 \leq y) \\ &= P(X \leq \frac{y-1}{3}) = \int_0^{\frac{y-1}{3}} 2x \, dx = x^2 \Big|_0^{\frac{y-1}{3}} \\ &= \left(\frac{y-1}{3}\right)^2 \quad \text{obviously valid for } 1 \leq y \leq 4 \end{aligned}$$

$\therefore G(y) = \begin{cases} 0 & \text{if } y \leq 1 \\ \frac{(y-1)^2}{9} & 1 \leq y \leq 4 \\ 1 & \text{if } y > 4 \end{cases} \Rightarrow \text{the pdf } g(y) = \begin{cases} 0 & \text{if } y \leq 1 \\ \frac{2}{9}(y-1) & 1 \leq y \leq 4 \\ 0 & \text{if } y > 4 \end{cases}$

Assum $x=0 \rightarrow y=1$

So, let us see, what we have got all your and we find that we have got the same pdf there as well.

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Let us look at the general problem in another way:

For example 1 $\therefore h(x) = 3x + 1 = y$

$$\begin{aligned} \therefore G(y) &= P(Y \leq y) = P(3X + 1 \leq y) = P(X \leq \frac{y-1}{3}) \\ \therefore g(y) &= \frac{dG(y)}{dy} = \frac{dG(y)}{d(\frac{y-1}{3})} \cdot \frac{d(\frac{y-1}{3})}{dy} = \frac{F(\frac{y-1}{3})}{d(\frac{y-1}{3})} \cdot \frac{d(\frac{y-1}{3})}{dy} \\ &= f\left(\frac{y-1}{3}\right) \cdot \frac{1}{3} = 2\left(\frac{y-1}{3}\right) \times \frac{1}{3} \\ &= \frac{2(y-1)}{9} \end{aligned}$$

Thus we get the same pdf as obtained before

Thus, we get the same pdf as obtained before.

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Note that the crucial step in obtaining the pdf of Y or getting the probability of the event $(Y \leq y)$ is that we are replacing the above event by an equivalent event with respect to the random variable X .

And this we are doing with the help of the inverse mapping of h .

When $h = 3x + 1$ we used $h^{-1} = \frac{y-1}{3}$
When $h = e^{-x}$ we used $h^{-1} = -\ln y$

Hence it is critical to have h^{-1} defined unambiguously

Note that, the crucial step in obtaining the pdf of Y or getting the probability of the event Y less than equal to y is that we are replacing the above event by an equivalent event with respect to the random variable X . And this we are doing with the help of the inverse mapping of h . So, when h is equal to $3x$ plus 1 we used h inverse is equal to y minus 1 upon 3 , when h is equal to e to the power minus x we used h inverse is equal to minus log of y . Hence, it is critical to have h inverse defined unambiguously.

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Hence the major requirement is h is invertible i.e. h^{-1} exists.

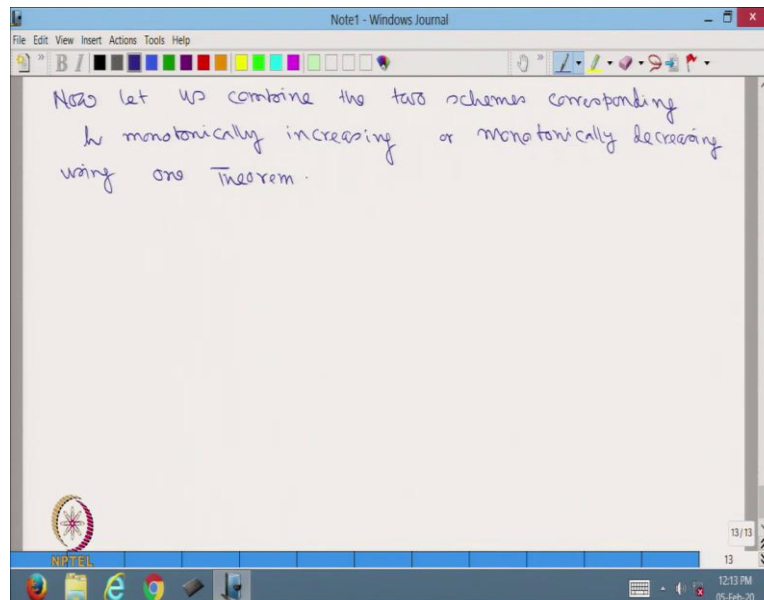
For a continuous function this means that h is strictly monotonic i.e. either strictly increasing or strictly decreasing.

If that is not the case we shall have to use other mechanism.

This we shall see later

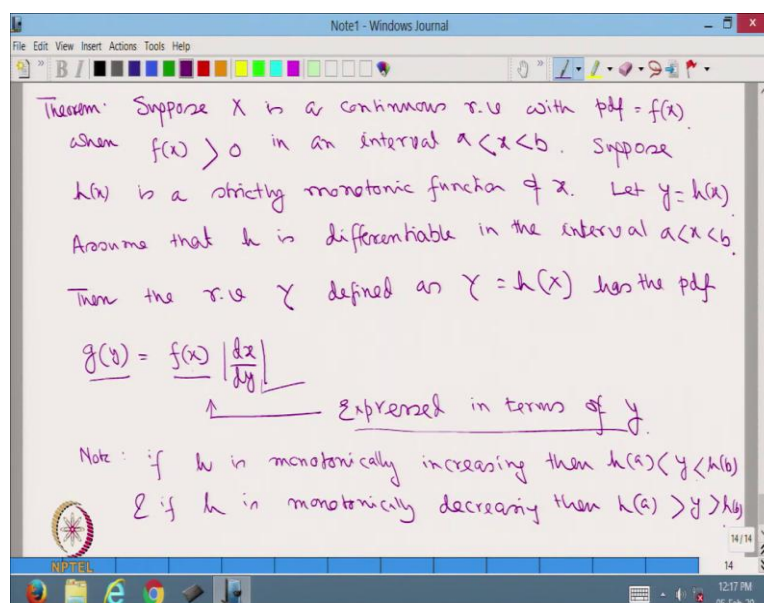
Hence, the measures requirement is h is invertible that is h inverse exists for a continuous function, this means that h is strictly monotonic that is either strictly increasing or strictly decreasing. If that is not the case we shall have to use other mechanism as we will see later.

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Now, let us combine the two schemes corresponding to h monotonically increasing or monotonically decreasing using one theorem.

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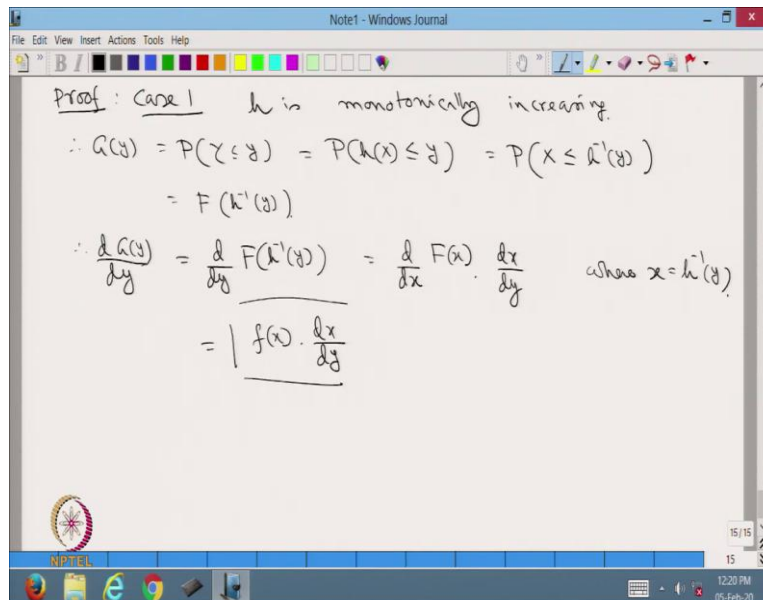


Theorem, suppose X is a continuous random variable with pdf equal to f of x when, f of x is greater than 0 in an interval a less than x less than b . Suppose, h is a strictly monotonic

function of x . Let, y is equal to h of x . Assume that h is differentiable in the interval a less than x less than b . Then the random variable y defined as Y is equal to h of x has the pdf, g of y is equal to f_x multiplied by dx dy absolute value of that one the whole thing expressed in terms of y . Note, if h is monotonically increasing then h a less than y less than hb and if, h is monotonically decreasing then ha is greater than y greater than hb .

So, that is the theorem we are obtaining the pdf of Y by looking at the pdf of X but, this we are expressing in terms of Y that means, we are using h inverse here and we are looking at the derivative of X with respect to Y and that also is express to Y .

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Proof: Case 1 h is monotonically increasing.

$$\begin{aligned} \therefore G(y) &= P(X \leq y) = P(h(X) \leq y) = P(X \leq h^{-1}(y)) \\ &= F(h^{-1}(y)) \\ \therefore \frac{dG(y)}{dy} &= \frac{d}{dy} F(h^{-1}(y)) = \frac{dF(x)}{dx} \cdot \frac{dx}{dy} \quad \text{where } x = h^{-1}(y) \\ &= \left| f(x) \cdot \frac{dx}{dy} \right| \end{aligned}$$

Proof, case 1 h is strictly increasing or monotonically increasing. Therefore, G of y is equal to probability Y less than equal to y is equal to probability h of X less than equal to y is equal to probability X less than equal to h inverse y is equal to F at h inverse y therefore, dG/dy is equal to d/dy of F of h inverse y is equal to d/dx of F_x multiplied by dx/dy where x is equal to h inverse y is equal to $f(x)$ into dx upon dy .

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Case-2 : Assume h is monotonically decreasing

$$\therefore G(y) = P(X \leq y) = P(h(X) \leq y) = P(X \geq h^{-1}(y))$$

$$= 1 - P(X \leq h^{-1}(y)) = 1 - F(h^{-1}(y))$$

$$\therefore g(y) = \frac{dG(y)}{dy} = - \frac{d}{dy} F(h^{-1}(y)) = - \frac{d}{dx} F(x) \cdot \frac{dx}{dy}$$

$$= f(x) \left(- \frac{dx}{dy} \right)$$

$$= f(x) \cdot \left| \frac{dx}{dy} \right|$$

Note that in this case the derivative is negative

Case 2, h is monotonically decreasing. Therefore, G of y is equal to probability Y less than equal to y is equal to probability h of X less than equal to y is equal to probability X greater than equal to h inverse y is equal to 1 minus probability X less than H inverse y in a continuous case we can write with equality also is equal to 1 minus F of h inverse y .

Therefore, g of y is equal to dG/dy is equal to minus d/dy of F of h inverse y is equal to minus d/dx of $F(x)$ into dx/dy is equal to $f(x)$ into minus dx/dy . Note that, in this case the derivative is negative. Therefore, we can write it as a $f(x)$ times modulus of dx/dy .

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Proof: Case 1 h is monotonically increasing.

$$\therefore G(y) = P(X \leq y) = P(h(X) \leq y) = P(X \leq h^{-1}(y))$$

$$= F(h^{-1}(y))$$

$$\therefore \frac{dG(y)}{dy} = \frac{d}{dy} F(h^{-1}(y)) = \frac{d}{dx} F(x) \cdot \frac{dx}{dy} \quad \text{where } x = h^{-1}(y)$$

$$= \left| f(x) \cdot \frac{dx}{dy} \right|$$

here $\frac{dx}{dy}$ is $(+)$

Note, in the earlier case here dx/dy is positive.

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Case-2 : Assume h is monotonically decreasing
 $\therefore G(y) = P(X \leq y) = P(h(X) \geq y) = P(X \geq h^{-1}(y))$
 $= 1 - P(X \leq h^{-1}(y)) = 1 - F(h^{-1}(y))$
 $\therefore g(y) = \frac{dG(y)}{dy} = - \frac{d}{dy} F(h^{-1}(y)) = - \frac{d}{dx} F(x) \cdot \frac{dx}{dy}$
 $= f(x) \left(- \frac{dx}{dy}\right)$ Note that in this case the derivative is negative.
 $= f(x) \cdot \left| \frac{dx}{dy} \right|$
 Therefore, by combining the two cases we have
 $g(y) = f(x) \cdot \left| \frac{dx}{dy} \right|$ expressed in

Therefore, by combining the two cases we have g of y is equal to $f(x)$ modulus of dx/dy expressed in y .

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Let us look at the two examples:
 1. $h(x) = 3x + 1 = y \quad \therefore \frac{dy}{dx} = 3 \quad \therefore \frac{dx}{dy} = \frac{1}{3}$
 $\therefore g(y) = f(h^{-1}(y)) \cdot \frac{1}{3} = f\left(\frac{y-1}{3}\right) \cdot \frac{1}{3}$
 $= \frac{2}{3} \cdot \frac{(y-1)}{3} = \frac{2}{9} (y-1)$
 (2) $h(x) = e^x$
 $\therefore g(y) = f(x) \left| \frac{dx}{dy} \right|$ Now $\frac{dx}{dy} = \frac{d(\ln y)}{dy} = \frac{1}{y} \therefore$
 2. $f(x) = 2x \quad \therefore$ Together we get $2(\ln y) \cdot \frac{1}{y} \therefore \left| \frac{1}{y} \right| = \frac{1}{y}$
 Thus we get the same result.

So, let us look at the two examples. 1, h of x is equal to $3x$ plus 1 is equal to y . Therefore, dy/dx is equal to 3, therefore, dx/dy is equal to 1 upon 3, therefore, g of y is equal to f at h inverse y into 1 by 3 is equal to f at y minus 1 upon 3 into 1 by 3 is equal to 2 by 3 into y minus 1 upon 3 is equal to 2 into y minus 1 by 9.

In a similar way case 2, h of x is equal to e to the power minus x g of y is equal to $f(x)$ into modulus of dx/dy . Now, dx/dy is equal to $d(\ln y)/dy$ is equal to $1/y$ and f of x is equal to $2x$ therefore, together we get 2 into \ln value into $1/y$ because modulus of $1/y$ is equal to $1/y$. Thus, we get the same result.

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What happens if h is Not strictly Monotonic?

For example consider $Y = X^2$

$$\therefore G(y) = P(X^2 \leq y) = P(-\sqrt{y} \leq X \leq \sqrt{y})$$

$$= F(\sqrt{y}) - F(-\sqrt{y})$$

$$\therefore g(y) = \frac{d}{dy} (F(\sqrt{y}) - F(-\sqrt{y})) = f(\sqrt{y}) \cdot \frac{1}{2\sqrt{y}} - f(-\sqrt{y}) \cdot \frac{d(-\sqrt{y})}{dy}$$

$$> f(\sqrt{y}) \cdot \frac{1}{2\sqrt{y}} + f(-\sqrt{y}) \cdot \frac{1}{2\sqrt{y}}$$

$$= (f(\sqrt{y}) + f(-\sqrt{y})) \cdot \frac{1}{2\sqrt{y}} = \frac{f(\sqrt{y}) + f(-\sqrt{y})}{2\sqrt{y}}$$

What happens if h is not strictly monotonic? That is the question. For example, consider Y is equal to X square. Therefore, G of y is equal to probability X square less than equal to y is equal to probability minus root to y less than equal to X less than equal to root y is equal to F of root y minus F of minus root y .

Therefore, g of y is equal to d/dy of F of root y minus F of minus root y is equal to f root y multiplied by half root y minus f at minus root y into $d(\text{minus root } y)/dy$ is equal to f of root y $1/2\sqrt{y}$ this minus minus makes it plus f at minus root y upon $1/2\sqrt{y}$ is equal to f of root y plus f of minus root y into $1/2\sqrt{y}$ is equal to f of root y plus F of minus root y upon $2\sqrt{y}$.

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We have done the same thing when we were finding the distⁿ of X^2 when $X \sim N(0,1)$.

We had $\text{pdf}(X^2) =$

$$\frac{1}{2\sqrt{y}} [f(\sqrt{y}) + f(-\sqrt{y})] = \frac{1}{2\sqrt{y}} \left(\frac{1}{\sqrt{2\pi}} e^{-\frac{y}{2}} + \frac{1}{\sqrt{2\pi}} e^{-\frac{y}{2}} \right)$$

$$= \frac{1}{\sqrt{2\pi}\sqrt{y}} e^{-\frac{y}{2}}$$

$$= \frac{1}{\sqrt{2\pi}} e^{-\frac{y}{2}} y^{\frac{1}{2}-1} = \frac{\left(\frac{1}{2}\right)^{\frac{1}{2}}}{\frac{1}{2}} e^{-\frac{1}{2}y} y^{\frac{1}{2}-1}$$

which is the pdf of $\Gamma(\frac{1}{2})$ or χ^2_1 .

if you remember, we have done the same thing when we were finding the distribution of X square when X is distributed as normal $0, 1$ we had pdf of x square is equal to 1 over 2 root y into f of root y plus f of minus root y is equal to 1 over 2 root y into 1 over root over 2π e to the power minus y by 2 plus 1 over root over 2π e to the power minus y by 2 is equal to 1 over root over 2π root over y e to the power minus y by 2 is equal to 1 over root over 2 root over π e to the power minus y by 2 y to the power half minus 1 is equal to half to the power half gamma half e to the power minus half y y to the power half minus 1 which is the pdf of gamma half, half or Chi square with 1 degree of freedom.

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Let us consider another problem of same nature.

Suppose $X \sim C(1,0)$ i.e. $f(x) = \frac{1}{\pi} \cdot \frac{1}{1+x^2}$
 $-\infty < x < \infty$

What is the pdf of X^2 ?

We have seen $f(y)$, when $Y = X^2$ is

$$\frac{1}{2\sqrt{y}} (f(\sqrt{y}) + f(-\sqrt{y}))$$

\therefore In the above case it is $\frac{1}{2\sqrt{y}} \left(\frac{1}{\pi} \cdot \frac{1}{1+y} + \frac{1}{\pi} \cdot \frac{1}{1+y} \right)$

$$= \frac{2}{2\sqrt{y}} \left(\frac{1}{\pi} \cdot \frac{1}{1+y} \right) = \frac{1}{\pi} \cdot \frac{y^{\frac{1}{2}-1}}{1+y}$$

Let us consider another problem of same nature. Suppose, X is distributed as Cauchy with 1, comma 0 that is f of x is equal to 1 over π into 1 over 1 plus x square minus infinity less than x less than infinity. What is the pdf of X square?

We have already seen f of y , when Y is equal to x square is 1 over 2 times root y into f at root y plus f at minus root y . Therefore, in the above case it is 1 over 2 root y into 1 over π into 1 upon 1 plus y because root y square is y plus 1 by π into 1 upon 1 plus y is equal to 2 upon 2 root y multiplied by 1 over π into 1 upon 1 plus y is equal to 2 and 2 gets cancelled. Therefore, we get 1 over π into y to the power half minus 1 we transfer it like that upon 1 plus y .

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The image shows a handwritten derivation in a Notepad window titled "Note2 - Windows Journal". The derivation starts with an expression for the probability density function (pdf) of $Y = X^2$ where X is Cauchy(1, 0). The expression is:

$$= \frac{1}{\pi} \cdot \frac{y^{\frac{1}{2}-1}}{(1+y)^{\frac{1}{2}+\frac{1}{2}}} \quad (*)$$

Below this, it states: "We know pdf of $\text{Beta}_2(m, n) = \frac{1}{B(m, n)} \frac{x^{m-1}}{(1+x)^{m+n}}$ ".

Hence, $(*)$ appears to be $\text{Beta}(\frac{1}{2}, \frac{1}{2})$ but we need to check the constant.

The constant is calculated as:

$$B(\frac{1}{2}, \frac{1}{2}) = \frac{\Gamma(\frac{1}{2}) \Gamma(\frac{1}{2})}{\Gamma(\frac{1}{2} + \frac{1}{2})} = \frac{\sqrt{\pi} \sqrt{\pi}}{1} = \pi$$

Therefore, the final result is:

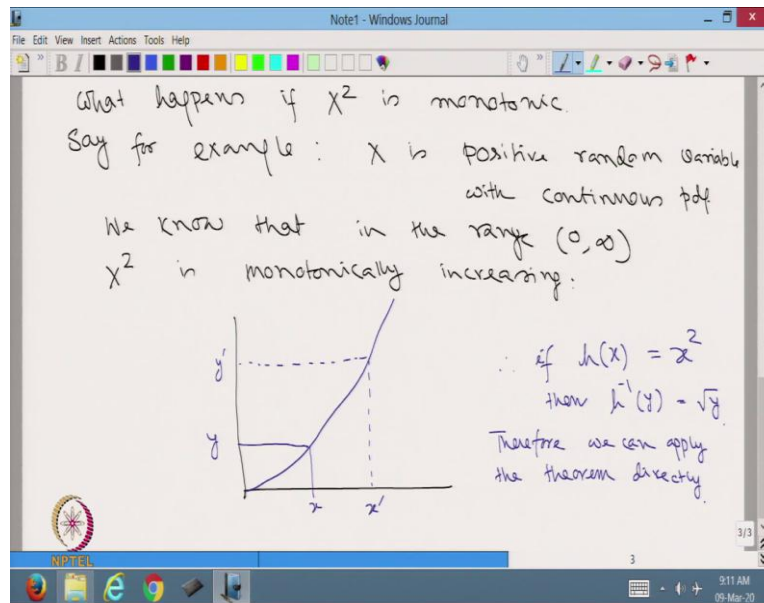
$$\therefore (*) \frac{1}{B(\frac{1}{2}, \frac{1}{2})} \frac{y^{\frac{1}{2}-1}}{(1+y)^{\frac{1}{2}+\frac{1}{2}}} \Rightarrow Y = X^2 \sim \text{Beta}_2(\frac{1}{2}, \frac{1}{2})$$

when $X \sim \text{Cauchy}(1, 0)$

Is equal to 1 upon π y to the power half minus 1 upon 1 plus y whole to the power half plus half so this part looks like beta 2 distribution. We know pdf of beta 2 with m , comma n is equal to 1 upon beta m , comma n x to the power m minus 1 upon 1 plus x whole to the power m plus n thus, we can notice a very similar pattern.

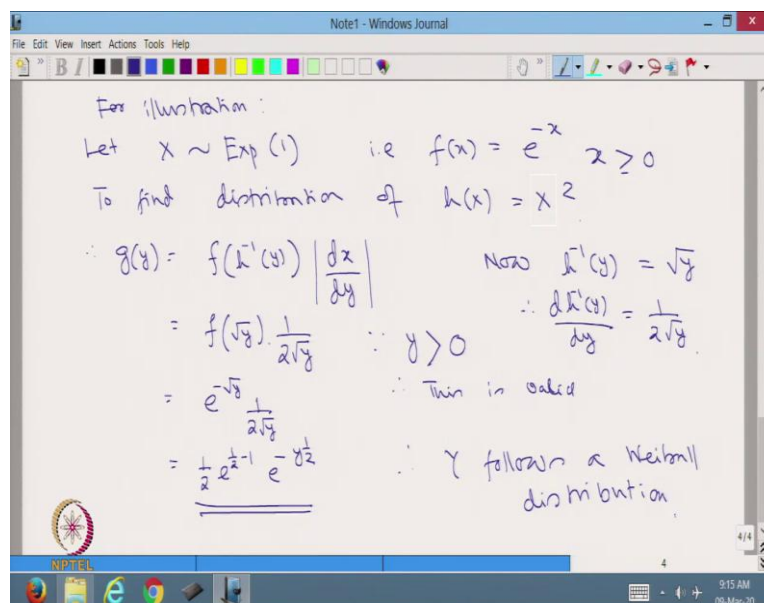
Hence, let us call it star, star appears to be beta with half and half but we need to check the constant. Now, beta half, comma half is equal to gamma half gamma upon gamma half plus half is equal to root π root π upon 1 is equal to π . Therefore, star is 1 upon beta half comma half y to the power half minus 1 upon 1 plus y whole to the power half plus half implies Y is equal to X square is distributed as beta 2 with half comma half when y is Cauchy with 1 and 0.

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What happens if X square is monotonic? Say, for example, X is a positive random variable with continuous pdf. Now, we know that in the range 0 to infinity X square is monotonically increasing. So, if this is the graph X is monotonically increasing therefore, given any x we can find y uniquely and conversely if given any y y prime say we can get corresponding x prime uniquely. Therefore, if $h(x)$ is equal to x square then $h^{-1}(y)$ is equal to root over y . Therefore, we can apply the theorem directly.

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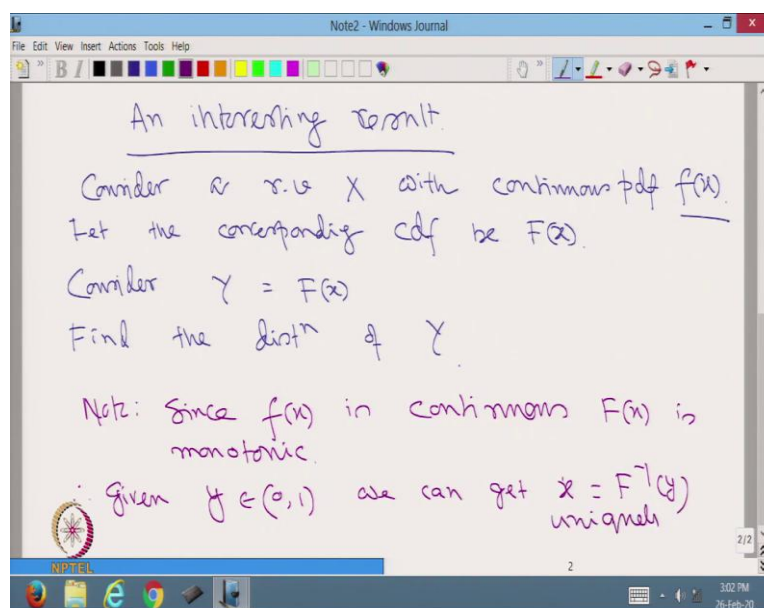


For illustration, let X distributed as exponential with parameter 1 that is f of x is equal to e to the power minus x when x is greater than equal to 0. To find distribution of h of x is equal to

x square. Therefore, g of y is equal to f at h inverse y multiplied by dx/dy . Now, h inverse of y is equal to \sqrt{y} therefore, dh inverse of y dy is equal to $\frac{1}{2\sqrt{y}}$.

Therefore, this equation becomes, f at \sqrt{y} multiplied by $\frac{1}{2\sqrt{y}}$ since, y is greater than 0 therefore, this is valid is equal to $e^{-\sqrt{y}}$ $\frac{1}{2\sqrt{y}}$ \sqrt{y} is equal to $\frac{1}{2} e^{-\sqrt{y}}$ $\frac{1}{2}$ into $e^{-\sqrt{y}}$ to the power $\frac{1}{2}$. Do you remember this density function, yes? Therefore, y follows a weibull distribution.

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So, let us consider an interesting result which comes from the application of the above theorem, consider a random variable X with continuous pdf f_x . Let the corresponding cumulative distribution function be capital F_x . Consider Y is equal to F_x find the distribution of Y .

So, that is a very interesting question because I am looking at any arbitrary continuous density random variable and Y is equal to its cumulative distribution function. Note that since, f_x is continuous capital F_x is monotonic. Therefore, given y belonging to $(0, 1)$ we can get x is equal to F inverse y uniquely.

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The above theorem is applicable.
 Y is r.v taking values in $[0,1]$
 \therefore pdf of $Y = g(y) = f(x) \cdot \left| \frac{dx}{dy} \right|$ expressed in y
Since $y = F(x)$ $x = F^{-1}(y)$
 $\therefore \left| \frac{dx}{dy} \right| = \frac{1}{\left| \frac{dy}{dx} \right|} = f(x)$ (when $f(x) > 0$)
 $\therefore g(y) = f(x) \cdot \frac{1}{f(x)} = 1$

Therefore, the above theorem is applicable and y is a random variable taking values in $0, 1$. Therefore, pdf of Y is equal to $g(y)$ is equal to by the above theorem $f(x)$ into modulus of dx/dy expressed in y . But, we do not do that, we write it as follows since, y is equal to $F(x)$, x is equal to F inverse of y . Therefore, modulus of dx/dy is equal to 1 upon modulus of dy/dx is equal to $f(x)$. Therefore, $g(y)$ is equal to $f(x)$ into 1 upon $f(x)$ is equal to 1 . Obviously, we can write when $f(x)$ greater than 0 .

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Thus $Y \in [0,1]$ & its pdf is 1
 $\therefore Y$ is $U[0,1]$
This is an important result as it is true for all $f(x)$ irrespective of its form.
Aliter:
 $G(y) = P[Y \leq y] = P[X \leq F^{-1}(y)] = F(F^{-1}(y))$
 $= y$
 $\therefore g(y) = G'(y) = \frac{d}{dy} y = 1$

Thus, Y belongs to $0, 1$ and its pdf is 1 . Therefore, Y is uniformly distributed in the interval $0, 1$. This is an important result as it is true for all $f(x)$ irrespective of its form. Now, some of u

may doubt that whether this result is correct or not so, let me prove it in a different way. G of Y is equal to say probability Y less than equal to y is equal to probability X less than equal to $F^{-1}(y)$ is equal to $F(F^{-1}(y))$ since, f is the CDF of X is equal to y . Therefore, g of y is equal to $G'(F^{-1}(y))$ is equal to $\frac{d}{dy} F(F^{-1}(y))$ is equal to 1. So, we get the same result.

Okay, friends I stop here today. In the next class I shall look at functions of two random variables and from there we shall try to see the pdf of different interesting functions and also we shall arrive at two very important distributions namely T and F. Okay friends. Thank you so much.