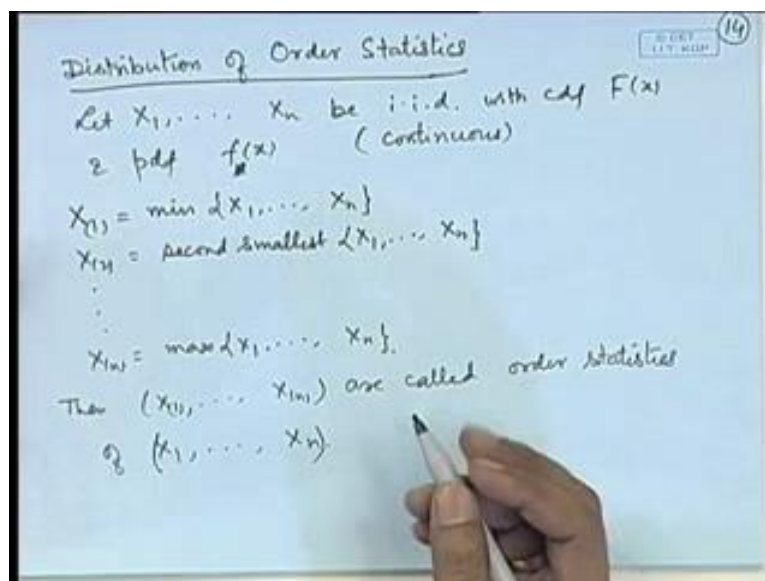


Probability and Statistics
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Lecture - 44
Distribution of Order Statistics

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What are order statistics? Let X_1, X_2, \dots, X_n be i. i. d with some cdf say $F(x)$ and pdf $f(x)$ and let me assume it to be continuous. We define $X_{(1)}$ to be the minimum of X_1, X_2, \dots, X_n ; $X_{(2)}$ to be second smallest of X_1, X_2, \dots, X_n . So, like that $X_{(3)}$ will be third smallest and so on; $X_{(n)}$ will be the maximum of X_1, X_2, \dots, X_n . Then $X_{(1)}, X_{(2)}, \dots, X_{(n)}$ these are called order statistics of X_1, X_2, \dots, X_n in many practical aspects, the order statistics are quite important. For example, the raw observations may be something; suppose the raw observations denote the marks by the students, but we may be interested in the ordered observation that who is getting the highest marks; who is a second highest etcetera.

If we are selecting certain candidates on the basis of certain scores, so we will be interested in selecting the best 10. So, we will be interested in say $X_{(n)}, X_{(n-1)}, \dots, X_{(n-9)}$ say; so the top 10 students. So, in general we are interested in order statistics and therefore, the distributions of the order statistics. Suppose you want to find out the distribution of $X_{(n)}$, so we may use a direct approach.

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The image shows handwritten mathematical derivations on a whiteboard. The derivations are as follows:

$$F_{X_{(n)}}(y_n) = P(X_{(n)} \leq y_n)$$

$$= P(X_1 \leq y_n, \dots, X_n \leq y_n)$$

$$= \prod_{i=1}^n P(X_i \leq y_n) = [F(y_n)]^n$$

pdf of $X_{(n)}$ as

$$f_{X_{(n)}}(y_n) = n [F(y_n)]^{n-1} f(y_n)$$

$$F_{X_{(1)}}(y_1) = P(X_{(1)} \leq y_1) = 1 - P(X_{(1)} > y_1)$$

$$= 1 - P(X_1 > y_1, \dots, X_n > y_1)$$

$$= 1 - \prod_{i=1}^n P(X_i > y_1) = 1 - [1 - F(y_1)]^n$$

$$f_{X_{(1)}}(y_1) = n [1 - F(y_1)]^{n-1} f(y_1)$$

For example we look at F of X_n ; let me use a notation here, say Y_i is equal to x_i ; for i is equal to 1 to n . So, the distribution of the largest that is probability of; X_n less than or equal to y_n , now notice here this is maximum being less than or equal to y_n ; this event is equivalent to that each of the X_i is less than or equal to y_n . Now we make use of the fact that the random variables are independent and identically distributed all of them have the cdf; F_x . So, each of these values is F of y_n and therefore, we get this as F of y_n to the power N ; if we are assuming the distributions to be a continuous then we can differentiate it and get the pdf of X_n as n times F of y_n to the power n minus one F of y_n .

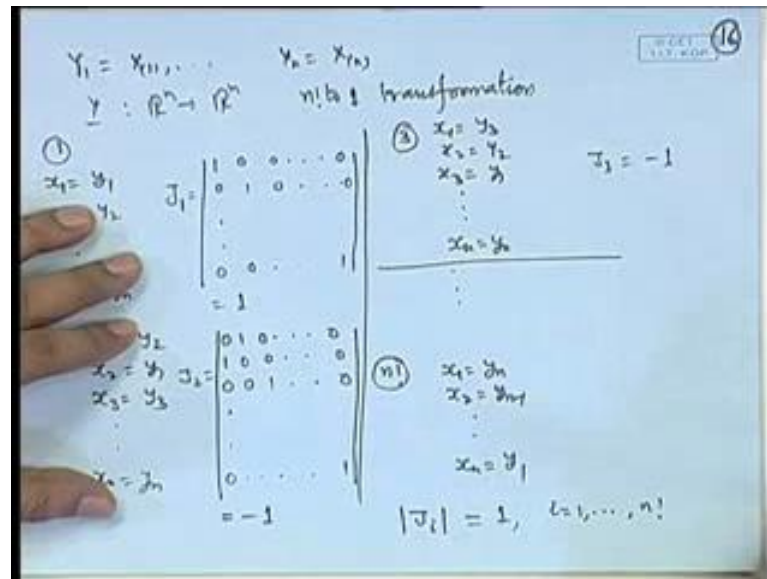
So, using a direct cdf approach the distributions of the largest can be determined. In a similar way we can obtain the distribution of the minimum also; that is probability of X_1 less than or equal to say y_1 ; this we can write as 1 minus probability of X_1 greater than y_1 . Once again if we are saying that the minimum is greater than y_1 ; it means that each of the observations is greater than y_1 .

At this stage we can use the independence 1 minus product X_i greater than y_1 and since each of this is identically same. So, it is 1 minus 1 minus F of y_1 to the power n , so the cdf of by smallest can be determined and the pdf of the smallest can be determined by differentiating this with respect to y_1 ; that gives n times 1 minus f of y_1 to the power n minus 1; f of y_1 . So, the distribution of the largest and by smallest order statistics can be determined using the cdf approach; however, this approach will be complicated; suppose

we want to determine the distribution of third order statistics or say the joint distribution of fourth and seventh order statistics.

So, in that case if we are assuming the continuous random variables; we make use of the Jacobean method. So, let us consider the joint distribution of Y_1, Y_2, Y_n .

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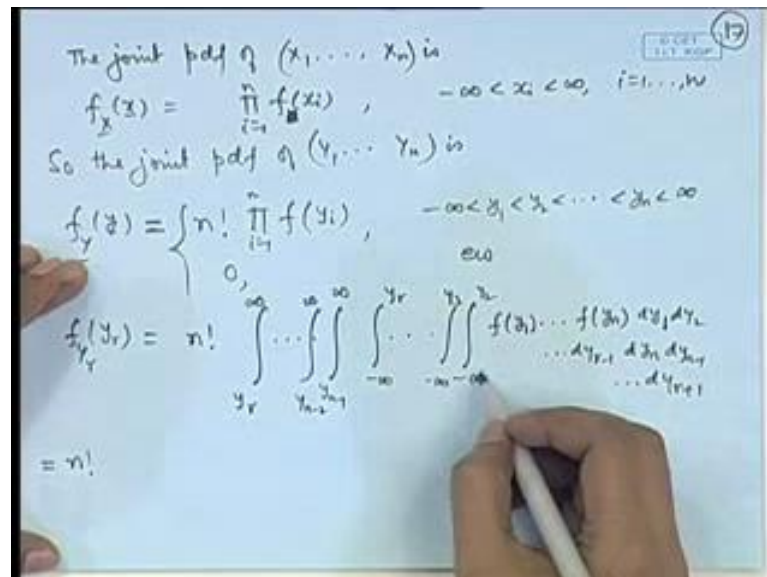
So, here the Function Y_1 is equal to X_1 , Y_n is equal X_n ; now this is a transformation from r_n to r_n , but it is a many one transformation. In fact, n factorial combinations of X_1, X_2, X_n give the same values of Y_1, Y_2, Y_n . Let us consider say n is equal to 2; suppose I say X_1 is equal to 1, X_2 is equal to 1.5, then Y_1 will be 1, Y_2 will be 1.5. We can take X_1 is equal to 1.5 and X_2 is equal to 1; once again Y_1 and Y_2 will remain the same; that means, two sets of X_1 and X_2 give the same value of minimum and maximum. In a similar way, suppose I have three values X_1, X_2, X_3 then six different combinations of X_1, X_2, X_3 will give the same values of Y_1, Y_2 and Y_3 .

So, in general Y is \mathbb{R}^n to \mathbb{R}^n ; this is n to n factorial 2, 1 transformation. So, we can partition r_n into n factorial different regions so that from each region it is a 1 to 1 transformation. In the region 1 for example, you may have x_1 is equal to y_1 , x_2 is equal to y_2 and so on x_n is equal y_n . In the region 2; we may have x_1 is equal to say y_2 ; x_2 is equal to y_1 , x_3 is equal to y_3 and so on x_n is equal to y_n . In the region 3; it may be x_1 is equal to say y_3 , x_2 is equal to y_2 , x_3 is equal to y_1 and so on; x_n is

equal y_n and so on. In the n factorial region, we may have x_1 is equal to say y_1 , x_2 is equal to y_{n-1} and so on; x_n is equal to y_1 .

Let us look at the Jacobean in each case, the Jacobean here is the determinant of the identity matrix 1, 0, 0, 0, 0, 1, 0 and so on 0, 0, 1 which is simply equal to 1. If you look at the Jacobean in the second case; this is 0, 1, 0 and so on 1, 0, 0 and so on 0, 0, 1 and so on. Note here that this is obtained from j_1 by interchanging the first and second row or first and second column, so the value of this will be minus 1. So, likewise the Jacobean in each case will be either plus 1 or minus 1 because it is obtained by permuting rows of identity matrix. So, Jacobean absolute values for each of them will be 1 only.

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Now, the joint distribution of X_1, X_2, X_n is product of $f(x_i)$; i is equal to 1 to n . So, in general the regions are, so the joint pdf of Y_1, Y_2, Y_n that is order statistics f_y . Now in the first region, we will substitute x_i is equal to y_i multiplied by 1, in the second region we will multiply by 1 and substitute X_1 is equal to Y_2 , X_2 is equal to Y_n etcetera. Notice here that in each of the cases its only a permutation of y_1, y_2, y_n , since it is a product of all these terms; it will always give total product of f of y_1, f of y_2, f of y_n .

So, this becomes product of f of y_i ; i is equal to 1 to n and we have n factorial regions; so it is n factorial times and the region is y_1 is less than y_2 less than y_n less than; we may have the cases where y_1 is equal to y_2 or say Y_s is equal to Y_{s+1} for certain

s. Since we are assuming the continuous distributions, the probability of two variables being equal is 0 and we can ignore this. So, we can see here that the joint distribution of y_1, y_2, \dots, y_n can be retained, now suppose we are interested in the distribution of say r th order statistics say; what is $f(y_r)$; then we need to integrate y_1, y_2, \dots, y_{r-2} and y_{r+1}, y_{r+2} up to y_n .

So, we can devise a scheme for integration it will be integral f of y_1, y_2, \dots, f of y_n . So, we can integrate with respect to Y_1 from minus infinity to Y_2 with respect to Y_2 from minus infinity to Y_3 and so on; up to y_{r-1} ; minus infinity to y_r . Then we can integrate y_n from y_{n-1} to infinity; y_{n-1} from y_{n-2} to infinity and so on; $d y_{r+1}$ will be from y_r to infinity.

So let us look at the evaluation here; if we integrate with respect to y_1 ; f of small f of y_1 gives capital F of Y_1 and when we substitute, the limits here at minus infinity it is 0 and at Y_2 ; it is capital F of Y_2 .

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The image shows a whiteboard with handwritten mathematical derivations. At the top, the expression $\int_{-\infty}^{y_2} (F(y_2) - f(y_2)) f(y_2) dy_2$ is written. Below it, an arrow points to $\int_{-\infty}^{y_2} \frac{1}{2} F^2(y_2) f(y_2) dy_2$. Further down, the equation $F(y_2) \int_{y_{n-1}}^{\infty} = \int_{y_{n-1}}^{\infty} ((1 - F(y_{n-1})) f(y_{n-1}))$ is written, with the term $(1 - F(y_{n-1})) f(y_{n-1})$ circled. Below this, the expression $+ \frac{1}{2} [1 - F(y_{n-2})]^2 f(y_{n-2})$ is written. A hand is visible at the bottom left, pointing towards the equations, and another hand is at the bottom right, holding a white marker.

So, this gives us capital F of y_2 in to small f of y_2 in the next stage, this will be integrated from minus infinity to y_3 . Now notice here is that if we integrate this will get half F square y_2 and again if we substitute the limits; I will get it as half F square Y_3 and at minus infinity this will be 0. So, at the next stage the integrand will be this term from minus infinity to y_4 .

So, next stage it will give F^q of y_4 and here it will be 1 by 2 in to 3. Now this step will be continue up to F of y_{r-1} . So, the last step will give us up to $r-1$ factorial F of y_r to the power $r-1$ because in the first stage; it is capital F of y_2 , in the second stage it is F^2 by 2, in the next stage it is F^3 by 3 factorial. So, at the $r-1$ of the stage; this will give F of y_r to the power $r-1$ by $r-1$ factorial.

Now, let us look at the other set of variables to be integrated; when we integrate f of y_n then that will gives us capital F of y_n . Now the region of integration is from y_n-1 to infinity, now at infinity this is 1, so it is 1 minus F of y_n-1 . So, at the next stage the integrand is 1 minus F of y_n-1 into f of y_n-1 . This we are integrating from y_n-2 to infinity. Again we notice here is that the integral will be 1 minus F of y_n-1 is square by 2 with a minus sign. So, at infinity this will become 0 and at y_n-1 , this will become this term.

So, that the next stage again and this will give us cube divided by 3 into 1.

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So the joint pdf of (y_1, \dots, y_n) is

$$f_y(y) = \begin{cases} n! \prod_{i=1}^n f(y_i), & -\infty < y_1 < y_2 < \dots < y_n < \infty \\ 0, & \text{else} \end{cases}$$

$$f(y_r) = n! \int_{y_r}^{\infty} \dots \int_{y_{r-1}}^{\infty} \int_{-\infty}^{y_r} \dots \int_{-\infty}^{y_2} f(y_1) \dots f(y_n) dy_1 dy_2 \dots dy_{r-1} dy_{r+1} \dots dy_n$$

$$= \frac{n!}{(r-1)!(n-r)!} [F(y_r)]^{r-1} [1-F(y_r)]^{n-r} f(y_r)$$

Like that we have to continue up to $d y_r + 1$, so this will give us n minus r factorial; 1 minus F of y_r to the power n minus r ; multiplied by f of y_r . So, we are able to determine the distribution of the r th order statistics, here the particular case we can see, suppose the random variables are uniformly distributed on the interval $0, 1$; then this will become y_r , this will become 1 minus y_r to the power n minus r ; which is nothing but a beta

distribution. So, that is one of the origins of or you can say applications of beta distribution.

Likewise if you want to integrate living variables say Y_r and Y_s ; so the joint pdf of say two order statistics.

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The joint pdf of Y_r and Y_s is

$$f_{Y_r, Y_s}(y_r, y_s) = \frac{n!}{(r-1)!(s-r-1)!(n-s)!} [F(y_r)]^{r-1} [F(y_s) - F(y_r)]^{s-r-1} [1 - F(y_s)]^{n-s} f(y_r) f(y_s)$$

$r < s, \quad -\infty < y_r < y_s < \infty$

Say Y_r and Y_s , so that is determined as f of $Y_r Y_s$; y_r ; y_s that is equal to n factorial divided by r minus 1 factorial; s minus r minus 1 factorial n minus s factorial; F of y_r to the power r minus 1; F of y_s minus f of y_r to the power s minus r minus 1; 1 minus F of y_s to the power n minus s ; F of $y_r y_s$; here I am taking r to be less than s . So, y_r will be less than y_s . In particular, we may write the joint distribution of by smallest and the largest from there we can determine the distribution of the range that is y_n minus y_1 etcetera. So, in the next lecture we will be considering various applications of the transformation see you.

Thank you.