

STAT 345/445 Lecture 19

Order Statistics – Section 5.4

Order Statistics

Let X_1, X_2, \dots, X_n be a random sample. The statistics

- $X_{(1)} = \min\{X_1, X_2, \dots, X_n\}$
- $X_{(2)} = \text{second smallest}\{X_1, X_2, \dots, X_n\}$
- ...
- $X_{(n-1)} = \text{second largest}\{X_1, X_2, \dots, X_n\}$
- $X_{(n)} = \max\{X_1, X_2, \dots, X_n\}$

are called **order statistics**.

Examples:

- Weight of smallest kitten in a litter
- Highest score on an exam

We want to find the (*sampling*) *distributions* of order statistics

Get used to the
 $X_{(j)}$ notation
and $X_j \neq X_{(j)}$

Could have used
 $U_1 = \min(X_1, \dots, X_n)$
 \vdots
 $U_n = \max(X_1, \dots, X_n)$

Order Statistics - Example

- Random sample X_1, X_2, \dots, X_{10}
- Order statistics $X_{(1)}, X_{(2)}, \dots, X_{(10)}$

Observed random sample:

X_1	X_2	X_3	X_4	X_5	X_6	X_7	X_8	X_9	X_{10}
1.9	3.5	9.0	2.8	3.3	2.1	6.1	1.1	0.7	7.9

Corresponding observed order statistics:

$X_{(1)}$	$X_{(2)}$	$X_{(3)}$	$X_{(4)}$	$X_{(5)}$	$X_{(6)}$	$X_{(7)}$	$X_{(8)}$	$X_{(9)}$	$X_{(10)}$
0.7	1.1	1.9	2.1	2.8	3.3	3.5	6.1	7.9	9.0

Distributions of order statistics

Theorem: distribution of order statistics

Let X_1, X_2, \dots, X_n be a random sample from a continuous distribution, with pdf $f(x)$ and cdf $F(x)$. Then the cdf and pdf of the j th order statistic $X_{(j)}$ are

$$F_{(j)}(x) = \sum_{k=j}^n \binom{n}{k} F(x)^k (1 - F(x))^{n-k} \quad \square$$

$$f_{(j)}(x) = \frac{n!}{(j-1)!(n-j)!} f(x) F(x)^{j-1} (1 - F(x))^{n-j}$$

proof... Lets take $j = 1$ and $j = n$ first

$$f_{(j)}(x) = \frac{d}{dx} \sum_{k=j}^n \binom{n}{k} F(x)^k (1 - F(x))^{n-k}$$

Order Statistics as transformations

- Transformations from a random sample to order statistics

$$(X_1, X_2, \dots, X_n) \mapsto X_{(j)}$$

or $(X_1, X_2, \dots, X_n) \mapsto (X_{(1)}, X_{(2)}, \dots, X_{(n)})$

are not one-to-one transformations

- Example:** $n = 3$ and $(X_1, X_2, X_3) \mapsto (X_{(1)}, X_{(2)}, X_{(3)})$
 - The following observations of (X_1, X_2, X_3)

$$(3, 5, 8), (3, 8, 5), (5, 3, 8), (5, 8, 3), (8, 3, 5), (8, 5, 3)$$

are all mapped to $(X_{(1)}, X_{(2)}, X_{(3)}) = (3, 5, 8)$

- So we can't use out transformation formulas to obtain $f_{(j)}(x)$
 - So use the "cdf-method" to find $F_{(j)}(x) = P(X_{(j)} \leq x)$

Untangling expressions in the Theorem

- Recall that by the Binomial formula

$$\sum_{k=0}^n \binom{n}{k} F(x)^k (1 - F(x))^{n-k} = (F(x) + 1 - F(x))^{\overset{n}{\bullet}} = 1$$

- So the cdf $F_{(1)}(x)$ can be written as

$$\begin{aligned} F_{(1)}(x) &= \sum_{k=1}^n \binom{n}{k} F(x)^k (1 - F(x))^{n-k} \quad n-1 \text{ terms} \\ &= \sum_{k=0}^n \binom{n}{k} F(x)^k (1 - F(x))^{n-k} - \binom{n}{0} F(x)^0 (1 - F(x))^{n-0} \\ &= 1 - (1 - F(x))^n \end{aligned}$$

add and
subtract
term for
 $k=0$

and

$$\begin{aligned} f_{(1)}(x) &= \frac{n!}{(1-1)!(n-1)!} f(x) F(x)^{1-1} (1 - F(x))^{n-1} \\ &= n f(x) (1 - F(x))^{n-1} = \frac{d}{dx} F_{(1)}(x) \end{aligned}$$

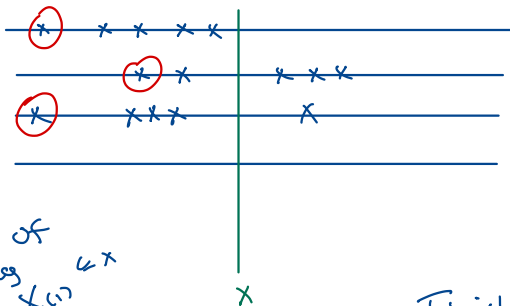
Distribution of $X_{(1)}$

- Have the joint pdf of X_1, X_2, \dots, X_n
- How can we relate $X_{(1)}$ to (X_1, X_2, \dots, X_n) ?
- "Smallest value is bigger than x " is the same event as "all values are bigger than x "

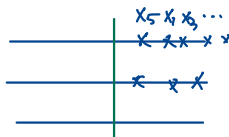
$$X_{(1)} \leq x$$

$$X_{(1)} > x?$$

$n=5$



etc.
many
ways of
getting
 $X_{(1)} \leq x$



all x 's
to be $> x$

Finished on the
white board

Distribution of $X_{(n)}$

- Theorem for $j = n$:

$$\begin{aligned} \frac{d}{dx} F_{(n)}(x) &= \frac{d}{dx} F(x)^n \\ &= n F(x)^{n-1} \cdot f(x) \end{aligned}$$

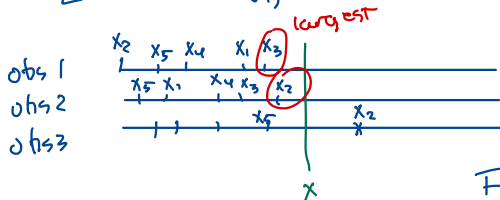
(chain rule.)

$$F_{(n)}(x) = \binom{n}{n} F(x)^n (1 - F(x))^{n-n} = F(x)^n$$

$$f_{(n)}(x) = \frac{n!}{(n-1)!(n-n)!} f(x) F(x)^{n-1} (1 - F(x))^{n-n} = n f(x) F(x)^{n-1}$$

- How can we relate $X_{(n)}$ to (X_1, X_2, \dots, X_n) ?
- "Largest value is less than x " is the same event as "all values are less than x "

Event $X_{(n)} \leq x$ $n=5$

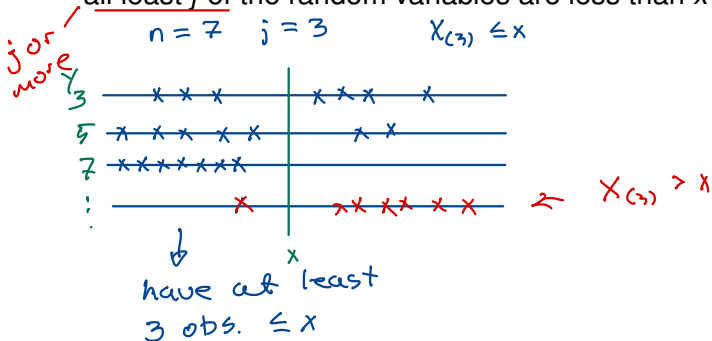


here $X_{(n)} > x$

Finished on the whiteboard

Distribution of $X_{(j)}$, $j = 2, \dots, n - 1$

- How can we relate $X_{(j)}$ to (X_1, X_2, \dots, X_n) ?
- " $X_{(j)} \leq x$ " is the same event as
"all least j of the random variables are less than x "



$F_{(j)}(x)$ on the whiteboard

$$F_{(j)}(x) = \sum_{k=j}^n \binom{n}{k} F(x)^k (1-F(x))^{n-k}$$

showed last time.
 $(uv)' = u'v + uv'$

$$f_{(j)}(x) = \frac{n!}{(j-1)!(n-j)!} f(x) F(x)^{j-1} (1-F(x))^{n-j}$$

$$f_{(j)}(x) = \frac{d}{dx} \sum_{k=j}^n \binom{n}{k} F(x)^k (1-F(x))^{n-k}$$

$$= \sum_{k=j}^n \left[\binom{n}{k} k F(x)^{k-1} f(x) (1-F(x))^{n-k} \right.$$

$$\left. - \binom{n}{k} F(x)^k (n-k) (1-F(x))^{n-k-1} f(x) \right]$$

= ... some algebra, term cancel...

$$= \frac{n!}{(j-1)!(n-j)!} f(x) F(x)^{j-1} (1-F(x))^{n-j}$$

support of $f_{(j)}(x)$?
 same as X_1, \dots, X_n

Example: Order stats for Uniform

- Let $X_{(1)}, X_{(2)}, \dots, X_{(n)}$ be order statistics from a random sample X_1, X_2, \dots, X_n from the $\text{Uniform}(0, 1)$. What are the (marginal) pdfs of $X_{(1)}, X_{(2)}, \dots, X_{(n)}$?

On the whiteboard...

Example: Order stats for Uniform

- Let $X_{(1)}, X_{(2)}, \dots, X_{(n)}$ be order statistics from a random sample X_1, X_2, \dots, X_n from the $\text{Uniform}(0, 1)$.
- Then

$$f_{(j)}(x) = \frac{\Gamma(n+1)}{\Gamma(j)\Gamma(n-j+1)!} x^{j-1} (1-x)^{(n-j+1)-1}$$

So $X_{(j)} \sim \text{Beta}(j, n-j+1)$

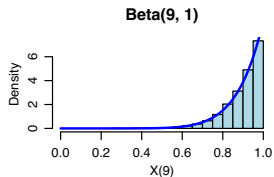
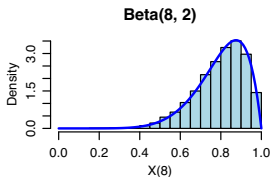
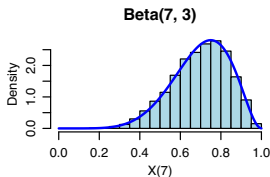
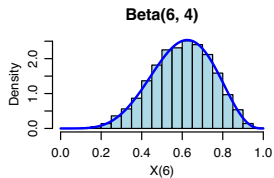
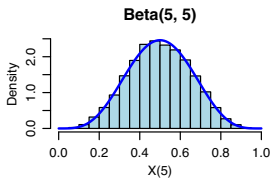
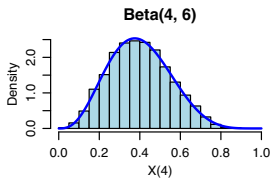
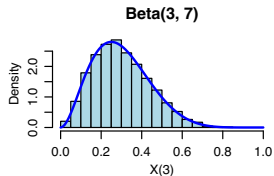
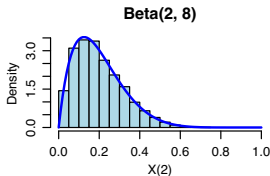
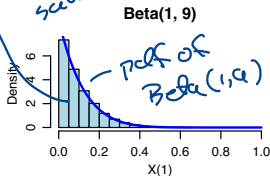
- R simulation:

```
x.order <- c() ← makes an empty vector
for(i in 1:10000){
  sample → x <- runif(n=9)      n=9
  x.order <- rbind(x.order, x[order(x)])
}
```

1st col. of x.order are samples of $X_{(i)}$ etc. ordering elements of x

Example

list of samples.



Joint distribution of order statistics

e.g. range: $X_{(n)} - X_{(1)}$

Joint pdf of two order statistics

Let $X_{(1)}, X_{(2)}, \dots, X_{(n)}$ be order statistics from a random sample X_1, X_2, \dots, X_n with pdf $f(x)$ and cdf $F(x)$.

The joint pdf of $(X_{(i)}, X_{(j)})$, $1 \leq i < j \leq n$ is

$$f_{(i,j)}(\underline{u}, v) = \frac{n!}{(i-1)!(j-1-i)!(n-j)!} f(u) f(v) \\ \times F(u)^{i-1} (F(v) - F(u))^{j-1-i} (1 - F(v))^{n-j}$$

for $-\infty < u < v < \infty$

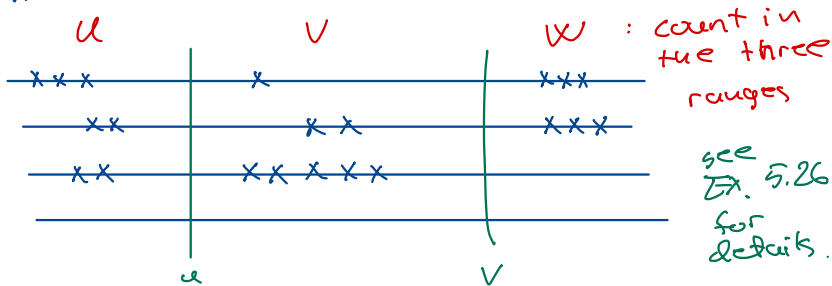
Think: 3-dimensional Multinomial distribution

very similar argument but with multinomial instead of Binomial

"cdf method"
 Joint cdf of $X_{(i)}, X_{(j)}$ $i < j$:

$$F_{(i,j)}(u,v) = P(X_{(i)} \leq u, X_{(j)} \leq v) \quad u < v$$

e.g. $n=7$ $i=2, j=4$ $= P(u \geq i, u+v \geq j)$



$X_{(i)} \leq u$ and $X_{(j)} \leq v$: same as
 at least i of X_1, \dots, X_n are $\leq u$ and
 at least j of X_1, \dots, X_n are $\leq v$

$$(u, v, w) \sim \text{Multinomial}(n+3, p = (F(u), F(v)-F(u), 1-F(v)))$$

Joint distribution of order statistics

Let X_1, X_2, \dots, X_n be a random sample from a continuous distribution, with pdf $f(x)$ and cdf $F(x)$.

Joint pdf for all n order statistics $(X_{(1)}, X_{(2)}, \dots, X_{(n)})$:

$$f_{(1), \dots, (n)}(u_1, \dots, u_n) = n! f(u_1) f(u_2) \cdots f(u_n)$$

for $-\infty < u_1 < u_2 < \cdots < u_n < \infty$ *proof in book...*

Can get any marginal and joint marginal pdfs by integrating out other variables.

see book for discrete order stats.