

# STK4011 and STK9011

Autumn 2016

## Order statistics

Covers (most of) the following material from chapter 5:

- Section 5.4: pages 226-231

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## Random sample

The random variables  $X_1, X_2, \dots, X_n$  are called a **random sample from the population**  $f_X(x)$  if  $X_1, X_2, \dots, X_n$  are mutually independent and the marginal pmf or pdf of each  $X_i$  is  $f_X(x)$

Alternatively, we may say that  $X_1, X_2, \dots, X_n$  are **independent and identically distributed (iid)** random variables with pmf or pdf  $f_X(x)$

The joint pmf or pdf is

$$f(x_1, x_2, \dots, x_n) = \prod_{i=1}^n f_X(x_i)$$

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## Order statistics

The **order statistics** of a random sample  $X_1, X_2, \dots, X_n$  are the sample values placed in ascending order

The order statistics are denoted  $X_{(1)}, X_{(2)}, \dots, X_{(n)}$  and they satisfy  $X_{(1)} \leq X_{(2)} \leq \dots \leq X_{(n)}$

Some statistics defined by the order statistics:

**Sample range:**  $R = X_{(n)} - X_{(1)}$

**Sample median:**  $M = \begin{cases} X_{((n+1)/2)} & \text{if } n \text{ is odd} \\ \frac{1}{2}(X_{(n/2)} + X_{(n/2+1)}) & \text{if } n \text{ is even} \end{cases}$

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**(100p)th sample percentile:** 
$$\begin{cases} X_{(np)} & \text{if } \frac{1}{2n} < p < 0.5 \\ X_{(n+1-\{n(1-p)\})} & \text{if } 0.5 < p < 1 - \frac{1}{2n} \end{cases}$$

{b} is b rounded to the nearest integer

Other definitions of sample percentiles exist (the command "quantile" in R has 9 options)

**Lower quartile:**  $Q_1 = 25\text{th percentile}$

**Upper quartile:**  $Q_3 = 75\text{th percentile}$

**Interquartile range:**  $Q_3 - Q_1$

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We will look at the distribution of the order statistics

Let  $X_1, \dots, X_n$  be a random sample from a continuous distribution with cdf  $F_X(x)$  and pdf  $f_X(x)$

We start out by considering  $X_{(n)} = \max X_i$   
and  $X_{(1)} = \min X_i$

The cumulative distribution of  $X_{(n)}$  is given by

$$\begin{aligned} F_{X_{(n)}}(x) &= P(\max X_i \leq x) = P\left(\bigcap_{i=1}^n \{X_i \leq x\}\right) \\ &= \prod_{i=1}^n P(X_i \leq x) = [F_X(x)]^n \end{aligned}$$

Hence the pdf becomes

$$f_{X_{(n)}}(x) = n f_X(x) [F_X(x)]^{n-1}$$

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The cumulative distribution of  $X_{(1)}$  is given by

$$\begin{aligned} F_{X_{(1)}}(x) &= P(\min X_i \leq x) = 1 - P(\min X_i > x) \\ &= 1 - P\left(\bigcap_{i=1}^n \{X_i > x\}\right) = 1 - \prod_{i=1}^n P(X_i > x) \\ &= 1 - \prod_{i=1}^n [1 - P(X_i \leq x)] = 1 - [1 - F_X(x)]^n \end{aligned}$$

Hence the pdf becomes

$$f_{X_{(1)}}(x) = n f_X(x) [1 - F_X(x)]^{n-1}$$

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We then look at the  $j$ -th order statistic

#### Theorem 5.4.4

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

Then the  $j$ -th order statistic is has cdf

$$F_{X_{(j)}}(x) = \sum_{k=j}^n \binom{n}{k} [F_X(x)]^k [1 - F_X(x)]^{n-k}$$

and pdf

$$f_{X_{(j)}}(x) = \frac{n!}{(j-1)!(n-j)!} f_X(x) [F_X(x)]^{j-1} [1 - F_X(x)]^{n-j}$$

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#### Example: uniform order statistic

Let  $X_1, \dots, X_n$  be iid and uniform(0,1)

Then  $f_X(x) = 1$  and  $F_X(x) = x$  for  $0 < x < 1$

Hence

$$\begin{aligned} f_{X_{(j)}}(x) &= \frac{n!}{(j-1)!(n-j)!} f_X(x) [F_X(x)]^{j-1} [1 - F_X(x)]^{n-j} \\ &= \frac{n!}{(j-1)!(n-j)!} 1 \cdot x^{j-1} (1-x)^{n-j} \\ &= \frac{\Gamma(n+1)}{\Gamma(j) \Gamma(n-j+1)} x^{j-1} (1-x)^{(n-j+1)-1} \end{aligned}$$

Thus  $X_{(j)} \sim \text{beta}(j, n-j+1)$

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A result on the joint pdf of two order statistics:

**Theorem 5.4.6**

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

Then the joint pdf of the  $i$ -th and  $j$ -th order statistic is given by ( $1 \leq i < j \leq n$ )

$$f_{X_{(i)}, X_{(j)}}(u, v) = \frac{n!}{(i-1)!(j-1-i)!(n-j)!} \times f_X(u) f_X(v) [F_X(u)]^{i-1} [F_X(v) - F_X(u)]^{j-1-i} [1 - F_X(v)]^{n-j}$$

(A formal proof is outlined in exercise 5.26)

**Range for a uniform distribution (example 5.4.7)**

Let  $X_1, \dots, X_n$  be iid and uniform(0,1)

Then  $f_X(x) = 1$  and  $F_X(x) = x$  for  $0 < x < 1$

The joint distribution of  $X_{(1)}$  and  $X_{(n)}$  is given by (for  $0 < x_1 < x_n < 1$ )

$$f_{X_{(1)}, X_{(n)}}(x_1, x_n) = n(n-1)(x_n - x_1)^{n-2}$$

Introduce the range

$$R = X_{(n)} - X_{(1)}$$

and the midrange

$$V = (X_{(1)} + X_{(n)}) / 2$$

The inverse transformation is given by

$$X_{(1)} = V - R/2 \quad X_{(n)} = V + R/2$$

The Jacobi equals -1 and the joint pdf of  $(R, V)$  becomes (for  $0 < r < 1, r/2 < v < 1 - r/2$ )

$$f_{R,V}(r, v) = n(n-1)r^{n-2}$$

The marginal pdf for  $R$  is given by

$$f_R(r) = \int_{r/2}^{1-r/2} n(n-1)r^{n-2} dv = n(n-1)r^{n-2}(1-r)$$

Thus  $R \sim \text{beta}(n-1, 2)$

We also have a result on the joint pdf of all  $n$  order statistics:

Let  $X_1, \dots, X_n$  be a random sample from a continuous distribution with pdf  $f_X(x)$

Then the joint pdf of all the order statistic is given by

$$f_{X_{(1)}, \dots, X_{(n)}}(x_1, \dots, x_n) = \begin{cases} n! f_X(x_1) \cdot \dots \cdot f_X(x_n) & \text{if } x_1 < \dots < x_n \\ 0 & \text{otherwise} \end{cases}$$