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► Probability in data mining

Order statistics

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► Finding the distribution of extremes:

ORDER STATISTICS

PROBABILITY THEORY - L4

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PROBABILITY THEORY - L4

2 / 8

### DISTRIBUTION OF THE MAXIMUM

 $\Gamma_{
m H}$  The distribution of the maximum  $X_{(n)}$ 

$$F_{X_{(n)}}(x) = P(X_1 \le x, X_2 \le x, ..., X_n \le x)$$
  
=  $\prod_{i=1}^{n} P(X_i \le x) = [F(x)]^n$ .

The density of the maximum  $X_{(n)}$ 

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

Let 
$$X_1, ..., X_n \sim L(a)$$
. Find  $F_{X_{(n)}}(x)$ . Solution: If  $X \sim L(a)$  then 
$$F(x) = \begin{cases} \frac{1}{2} \exp\left(\frac{x}{b}\right) & \text{if } x < 0\\ 1 - \frac{1}{2} \exp\left(-\frac{x}{b}\right) & \text{if } x \geq 0 \end{cases}$$

so 
$$F_{X_{(n)}}(x) = [F(x)]^n = \begin{cases} \frac{1}{2^n} \exp\left(\frac{nx}{b}\right) & \text{if } x < 0\\ \left[1 - \frac{1}{2} \exp\left(-\frac{x}{b}\right)\right]^n & \text{if } x \ge 0 \end{cases}$$

Even if the original sample  $X_1, X_2, ..., X_n$  are independent, their order

statistics  $X_{(1)}, X_{(2)}, ..., X_{(n)}$  are not clearly not.

DEF The order statistics:  $X_{(1)} \le X_{(2)} \le \cdots \le X_{(n)}.$ 

 $X_{(k)}=$  the kth smallest of  $X_1,X_2,...,X_n$ 

• Range:  $R = \max(X_1, X_2, ..., X_n) - \min(X_1, X_2, ..., X_n)$ .

DEF The kth order variable

▶ Median:  $Pr(X \le m) = 1/2$ .

•  $min(X_1, X_2, ..., X_n)$ •  $max(X_1, X_2, ..., X_n)$ .

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 ${\rm TH}~{\rm The~distribution~of~the~minimum~}\chi_{(1)}$ 

$$F_{X_{(1)}}(x) = 1 - P\left(X_{(1)} > x\right)$$
  
= 1 - P(X<sub>1</sub> > x, X<sub>2</sub> > x, ..., X<sub>n</sub> > x)  
= 1 - \prod\_{i=1}^{n} P(X\_i > x) = 1 - [1 - F(x)]^n.

The density of the minimum  $X_{(n)}$ 

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

lacksquare Let  $X_1,...,X_n\sim Exp(a)$ . What is  $f_{X_{(1)}}(x)$  and  $E(X_{(1)})$ ? Solution: We have

$$F(x) = 1 - e^{-ax}$$

S

$$f_{X_{(1)}}(x) = n \left[ e^{-ax} \right]^{n-1} a e^{-ax} = a n e^{-anx}$$

 $\sup_{\text{MATTIAS VILLAN!}} \langle X_{11} \rangle \sim Exp(an) \text{ and } E(X_{(1)}) = \frac{1}{3n}. \text{ [Serial electric circuits]}$ 

 $F_{Y_i}(y) = [F(y)]^3 = \left(\frac{y-a}{b-a}\right)^3$  $F_{X_{(k)}}(x) = F_{\beta(k,n+1-k)}[F(x)]$ three jumps for the *i*th athlete, for i = 1, ..., n: Then derive  $Y_{(n-1)}$ 

# JOINT DISTRIBUTION OF THE EXTREMES AND RANGE

► So far: only marginal distributions of order statistics.

TH The joint density of  $X_{(1)}$  and  $X_{(n)}$ 

$$f_{X_{(1)},X_{(n)}}(x,y) = \begin{cases} n(n-1) \left( F(y) - F(x) \right)^{n-2} f(y) f(x) & \text{if } x < y \\ 0 & \text{otherwise} \end{cases}$$

From  $f_{X_{(1)},X_{(n)}}(x,y)$  we can derive the distribution of the Range  $R_n=X_{(n)}-X_{(1)}$  by the transformation theorem.

TH The distribution of the Range  $R_n = X_{(n)} - X_{(1)}$  is

$$f_{R_n}(r) = n(n-1) \int_{-\infty}^{\infty} \left( F(u+r) - F(u) \right)^{n-2} f(u+r) f(u) du$$

### Marginal distribution of $X_{(k)}$

TH The distribution of the kth order variable  $X_{(k)}$  from a random sample

where  $F_{eta(k,n+1-k)}(\cdot)$  is the cdf of a Beta(k,n+1-k) variable.

**Solution**: First, calculate the distribution of  $Y_i$  = longest jump out of  $\bigcirc$  Let the individual jumps of n athletes in a long jump tournament be independently U(a,b) distributed. What is the probability that the recorded score of the silver medalist is longer than c meters?

$$F_{Y_{(n-1)}}(y) = F_{\beta(n-1,2)}\left(\left(rac{y-a}{b-a}
ight)^3
ight)$$

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5 / 8

8/9

## JOINT DISTRIBUTION OF ORDER STATISTICS

TH The joint density of the order statistics is

$$f_{X_{(1)},\dots,X_{(n)}}(y_1,\dots,y_n) = \begin{cases} n! \prod_{k=1}^n f(y_k) & \text{if } y_1 < y_2 < \dots < y_n \\ 0 & \text{otherwise} \end{cases}$$

► The marginal densities of any order variable can be derived by integrating  $f_{X_{(1)},\dots,X_{(n)}}(y_1,\dots,y_n)$  in the usual fashion.