

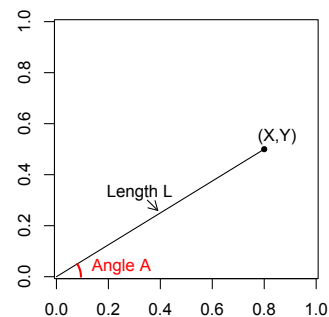
## Joint distributions (part 4)

Lecture 7a (STAT 24400 F24)

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## Example: Joint distribution, functions of r.v. pair

Suppose that  $(X, Y)$  is drawn uniformly from the unit square. Drawing the segment from  $(0, 0)$  to  $(X, Y)$ , let  $A$  = angle &  $L$  = length



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## Joint distribution example (cont.)

- Express  $A$  and  $L$  in terms of  $X$  and  $Y$ .

$$L = \sqrt{X^2 + Y^2}$$

$$\tan(A) = \frac{Y}{X} \Rightarrow A = \arctan \frac{Y}{X}$$

- Are  $A$  and  $L$  independent?

No. Possible values of  $L$  depend on the value of  $A$ .

— e.g., if  $A = 0$  then we must have  $L \leq 1$ ,  
but if  $A = \frac{\pi}{4}$  then we can have  $L > 1$ .

(Note this approach is simpler than trying to prove via joint density.)

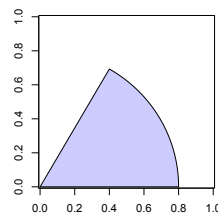
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## Joint distribution example (cont.)

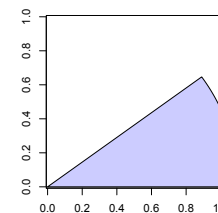
- What is the joint distribution of  $A$  and  $L$ ?

$$\begin{aligned} F_{A,L}(s, t) &= \mathbb{P}(A \leq s, L \leq t) \\ &= \mathbb{P}(\arctan(Y/X) \leq s, \sqrt{X^2 + Y^2} \leq t) \\ &= \text{Area}(\{(x, y) \in [0, 1]^2 : \arctan(y/x) \leq s, \sqrt{x^2 + y^2} \leq t\}) \end{aligned}$$

E.g., for  $s = \pi/3$  and  $t = 0.8$ :



for  $s = \pi/5$  and  $t = 1.1$ :



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## Review: Independence

- If events  $A$  and  $B$  are independent, the probability of their intersection

$$\mathbb{P}(AB) = \mathbb{P}(A)\mathbb{P}(B)$$

- If two r.v.'s  $X$  and  $Y$  are independent, their joint CDF factors:

$$F_{X,Y}(x,y) = F_X(x)F_Y(y)$$

In addition,

- If  $X$  and  $Y$  are discrete and independent, their joint PMF

$$p_{X,Y}(x,y) = p_X(x)p_Y(y)$$

- If  $X$  and  $Y$  are continuous and independent, their joint density

$$f_{X,Y}(x,y) = f_X(x)f_Y(y)$$

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## Joint distribution for i.i.d. r.v.'s

Suppose  $X_1, \dots, X_n$  are independent and identically distributed ("i.i.d.") r.v.'s drawn from a distribution with CDF  $F$ .

- Independence  $\Rightarrow F_{X_1, \dots, X_n}(x_1, \dots, x_n) = \prod_{i=1}^n F(x_i)$
- If  $F$  is discrete (or continuous), then the joint PMF (or joint density) is a product also:

$$p_{X_1, \dots, X_n}(x_1, \dots, x_n) = \prod_{i=1}^n p(x_i)$$

or

$$f_{X_1, \dots, X_n}(x_1, \dots, x_n) = \prod_{i=1}^n f(x_i)$$

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## Order statistics

For an  $n$ -tuple  $X_1, \dots, X_n$ , the **order statistics** are the ranked values:

- $X_{(1)}$  denotes the smallest value, i.e.,  $X_{(1)} = \min\{X_1, \dots, X_n\}$
- $X_{(2)}$  denotes the next-smallest value
- ...
- $X_{(n)}$  denotes the largest value, i.e.,  $X_{(n)} = \max\{X_1, \dots, X_n\}$

Note: if there are ties, then the same value appears multiple times (e.g., if we observe 3, 5, 3, then  $X_{(1)} = X_{(2)} = 3$  and  $X_{(3)} = 5$ )

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## Order statistics

Order statistics is an important tool in statistics.

The probability laws of order statistics can be used for questions such as:

- What is the probability that the largest draw among  $n$  draws will exceed some value?  
(This will help us to quantify and control extreme events.)
- How accurately does the sample median estimate the true median of the distribution?
- ... and many more

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## Order statistics for i.i.d. data — distribution of $\min_i X_i$

Suppose  $X_1, \dots, X_n$  are i.i.d. draws from a distribution with CDF  $F$ .

What is the distribution of  $X_{(1)} = \min_i X_i$ ?

$$\begin{aligned} F_{X_{(1)}}(x) &= \mathbb{P}(X_{(1)} \leq x) = 1 - \mathbb{P}(X_{(1)} > x) \\ &= 1 - \mathbb{P}(\min\{X_1, \dots, X_n\} > x) \\ &= 1 - \mathbb{P}(X_i > x \text{ for all } i = 1, \dots, n) \\ &= 1 - \prod_{i=1}^n \mathbb{P}(X_i > x) = 1 - (1 - F(x))^n \end{aligned}$$

If the original distribution is continuous with density  $f = F'$ , then

$$f_{X_{(1)}}(x) = \frac{d}{dx} F_{X_{(1)}}(x) = n(1 - F(x))^{n-1} \cdot \frac{d}{dx} F(x) = n(1 - F(x))^{n-1} \cdot f(x).$$

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## Order statistics for i.i.d. data — distribution of $\max_i X_i$

Suppose  $X_1, \dots, X_n$  are i.i.d. draws from a distribution with CDF  $F$

What is the distribution of  $X_{(n)} = \max_i X_i$ ?

$$\begin{aligned} F_{X_{(n)}}(x) &= \mathbb{P}(X_{(n)} \leq x) \\ &= \mathbb{P}(X_i \leq x \text{ for all } i = 1, \dots, n) \\ &= \prod_{i=1}^n \mathbb{P}(X_i \leq x) = F(x)^n \end{aligned}$$

If the original distribution is continuous with density  $f = F'$ :

$$f_{X_{(n)}}(x) = \frac{d}{dx} F_{X_{(n)}}(x) = nF(x)^{n-1} \cdot \frac{d}{dx} F(x) = nF(x)^{n-1} \cdot f(x).$$

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## Joint distribution of $(\min_i X_i, \max_i X_i)$ for i.i.d. data

Suppose  $X_1, \dots, X_n$  are i.i.d. draws from a distribution with CDF  $F$

What is the joint distribution of  $X_{(1)}$  and  $X_{(n)}$ ?

$$\begin{aligned} F_{X_{(1)}, X_{(n)}}(x, y) &= \mathbb{P}(X_{(1)} \leq x, X_{(n)} \leq y) \\ &= \mathbb{P}(X_{(n)} \leq y) - \mathbb{P}(X_{(1)} > x, X_{(n)} \leq y) \\ &= \mathbb{P}(X_{(n)} \leq y) - \mathbb{P}(x < X_i \leq y \text{ for all } i = 1, \dots, n) \\ &= F(y)^n - (F(y) - F(x))^n \end{aligned}$$

**Remarks** Distributions for other order statistics (e.g.  $\mathbb{P}(X_{(3)} < x)$ ) and their joint distributions (e.g. for  $X_{(2)}$  and  $X_{(3)}$ ) can also be obtained analogously.

The derivations involve detailed counting this are more technical (omitted).

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## Order statistics: Exponential distribution

### Example

Suppose  $X_1, \dots, X_n$  are i.i.d.  $\text{Exponential}(\lambda)$ .

What are the CDF and density of  $X_{(n)}$ ?

$$\begin{aligned} F_{X_{(n)}}(x) &= F(x)^n = (1 - e^{-\lambda x})^n \\ f_{X_{(n)}}(x) &= nF(x)^{n-1} \cdot f(x) = n(1 - e^{-\lambda x})^{n-1} \cdot \lambda e^{-\lambda x} \end{aligned}$$

What are the CDF and density of  $X_{(1)}$ ?

$$\begin{aligned} F_{X_{(1)}}(x) &= 1 - (1 - F(x))^n = 1 - e^{-n\lambda x} \\ f_{X_{(1)}}(x) &= n(1 - F(x))^{n-1} \cdot f(x) \\ &= n(1 - (1 - e^{-\lambda x}))^{n-1} \cdot \lambda e^{-\lambda x} = (n\lambda)e^{-(n\lambda)x} \end{aligned}$$

$$\Rightarrow X_{(1)} \sim \text{Exponential}(n\lambda)$$

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