

# 18.440: Lecture 22

## Joint distributions functions

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Distributions of functions of random variables

Joint distributions

Independent random variables

Examples

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# Distribution of function of random variable

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- ▶ This is a general principle. If  $X$  is a continuous random variable and  $g$  is a strictly increasing function of  $x$  and  $Y = g(X)$ , then  $F_Y(a) = F_X(g^{-1}(a))$ .

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- ▶ How can we use this to compute the probability density function  $f_Y$  from  $f_X$ ?
- ▶ If  $Z = X^2$ , then what is  $P\{Z \leq 16\}$ ?

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## Joint probability mass functions: discrete random variables

- ▶ If  $X$  and  $Y$  assume values in  $\{1, 2, \dots, n\}$  then we can view  $A_{ij} = P\{X = i, Y = j\}$  as the entries of an  $n \times n$  matrix.

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- ▶ In general, when  $X$  and  $Y$  are jointly defined discrete random variables, we write  $p(x, y) = p_{X,Y}(x, y) = P\{X = x, Y = y\}$ .

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- ▶ Question: if I tell you the two parameter function  $F$ , can you use it to determine the marginals  $F_X$  and  $F_Y$ ?
- ▶ Answer: Yes.  $F_X(a) = \lim_{b \rightarrow \infty} F(a, b)$  and  $F_Y(b) = \lim_{a \rightarrow \infty} F(a, b)$ .

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$$\int_{-\infty}^b \int_{-\infty}^a \frac{\partial}{\partial x} \frac{\partial}{\partial y} F(x, y) dx dy = \int_{-\infty}^b \frac{\partial}{\partial y} F(a, y) dy = F(a, b).$$

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- ▶ From this, we can show that it works for strips, rectangles, general open sets, etc.

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- ▶ When  $X$  and  $Y$  are continuous, they are independent if  $f(x, y) = f_X(x)f_Y(y)$ .

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- ▶ Using polar coordinates, we want 
$$\int_0^1 (2\pi r) \frac{1}{2\pi} e^{-r^2/2} dr = -e^{-r^2/2} \Big|_0^1 = 1 - e^{-1/2} \approx .39.$$

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- ▶ If  $j \geq 1$ , then

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- ▶ Can we get the marginals from that?



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- ▶ Are all of the  $T_i$  and  $A_i$  independent of each other? What are their probability distributions?

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- ▶ Exponential  $\lambda_{\text{tiger}} = .2/\text{hour}$ . So  $P\{T_{\text{tiger}} > a\} = e^{-.2a}$ .

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- ▶ Draw the box  $[0, 1] \times [0, \pi]$  on which  $(X, \theta)$  is uniform. What's the area of the subset where  $X \geq 1 - \sin \theta$ ?