

# Math 461 Spring 2024

Renming Song

University of Illinois Urbana-Champaign

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# Outline

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- 1 **General Info**
- 2 6.4 Conditional distributions: discrete case
- 3 6.5 Conditional distributions: continuous case

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Solution to HW7 is on my homepage now.

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Let  $X$  and  $Y$  be discrete random variables with joint mass function  $p(\cdot, \cdot)$ . If  $y$  is a possible value of  $Y$  (i.e,  $p_Y(y) > 0$ ), then

$$P(X = x|Y = y) = \frac{P(X = x, Y = y)}{P(Y = y)} = \frac{p(x, y)}{p_Y(y)}.$$

The function  $x \mapsto \frac{p(x, y)}{p_Y(y)}$  is a mass function. It is called the conditional mass function of  $X$  given  $Y = y$ .

The function

$$p_{X|Y}(x|y) = \begin{cases} \frac{p(x, y)}{p_Y(y)}, & p_Y(y) > 0, \\ 0, & \text{otherwise} \end{cases}$$

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If  $X$  and  $Y$  are independent, then for any possible value  $y$  of  $Y$ ,

$$p_{X|Y}(x|y) = p_X(x), \quad x \in \mathbb{R}.$$

We always have

$$p(x, y) = p_Y(y)p_{X|Y}(x|y), \quad x, y \in \mathbb{R}.$$

We can similarly define the conditional mass function of  $Y$  given  $X$ :

$$p_{Y|X}(y|x) = \begin{cases} \frac{p(x, y)}{p_X(x)}, & p_X(x) > 0, \\ 0, & \text{otherwise} \end{cases}$$

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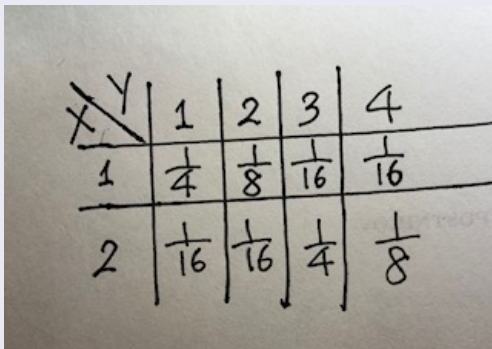
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## Example 1

The joint mass function of  $X$  and  $Y$  is given below. Find  $p_{X|Y}(x|2)$ .



A handwritten table representing the joint mass function of two discrete random variables, X and Y. The table is a 2x4 grid with X values (1, 2) in the rows and Y values (1, 2, 3, 4) in the columns. The diagonal cell (1,1) contains 'X \ Y'. The other cells contain the joint probabilities: (1,2) is 1/4, (1,3) is 1/8, (1,4) is 1/16, (2,1) is 1/16, (2,2) is 1/16, (2,3) is 1/4, and (2,4) is 1/8.

X \ Y	1	2	3	4
1	$\frac{1}{4}$	$\frac{1}{8}$	$\frac{1}{16}$	$\frac{1}{16}$
2	$\frac{1}{16}$	$\frac{1}{16}$	$\frac{1}{4}$	$\frac{1}{8}$

$p_Y(2) = 3/16$ . So

$$p_{X|Y}(x|2) = \begin{cases} 2/3, & x = 1, \\ 1/3, & x = 2, \\ 0, & \text{otherwise.} \end{cases}$$

### Example 1

Suppose  $X$  and  $Y$  are independent,  $X$  is a Poisson random variable with parameter  $\lambda_1$ ,  $Y$  is a Poisson random variable with parameter  $\lambda_2$ . For  $n \geq 1$ , find the conditional mass function of  $X$  given  $X + Y = n$ .

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We know that  $X + Y$  a Poisson random variable with parameter  $\lambda_1 + \lambda_2$ . If  $X + Y = n$ , then  $X$  can only take values  $0, 1, \dots, n$ . For any  $x = 0, 1, \dots, n$ ,

$$\begin{aligned} p_{X|X+Y}(x|n) &= \frac{P(X = x, X + Y = n)}{P(X + Y = n)} = \frac{P(X = x, Y = n - x)}{P(X + Y = n)} \\ &= \frac{P(X = x)P(Y = n - x)}{P(X + Y = n)} = \frac{e^{-\lambda_1} \frac{\lambda_1^x}{x!} e^{-\lambda_2} \frac{\lambda_2^{n-x}}{(n-x)!}}{e^{-(\lambda_1 + \lambda_2)} \frac{(\lambda_1 + \lambda_2)^n}{n!}} \\ &= \binom{n}{x} \left( \frac{\lambda_1}{\lambda_1 + \lambda_2} \right)^x \left( \frac{\lambda_2}{\lambda_1 + \lambda_2} \right)^{n-x}. \end{aligned}$$

Thus, given  $X + Y = n$ ,  $X$  is a binomial random variable with parameters  $(n, \lambda_1/(\lambda_1 + \lambda_2))$ .

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## Example 2

$X$  and  $Y$  are independent geometric random variables with parameter  $p$ . For  $n \geq 2$ , find the conditional mass function of  $X$  given  $X + Y = n$ .

$X + Y$  is a negative binomial random variable with parameters  $(2, p)$ :

$$p_{X+Y}(n) = \binom{n-1}{1} p^2 (1-p)^{n-2}, \quad n = 2, 3, \dots$$

Give  $X + Y = n$ ,  $X$  can only take values  $1, \dots, n-1$ . For,  $x = 1, \dots, n-1$ ,

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Thus the conditional mass function of  $X$  given  $X + Y = n$  is

$$p_{X|X+Y}(x|n) = \begin{cases} \frac{1}{n-1}, & x = 1, \dots, n-1, \\ 0, & \text{otherwise.} \end{cases}$$

### Example 3

A number  $Y$  is chosen randomly from  $\{1, 2, \dots, 100\}$  and then another number  $X$  is randomly chosen from  $\{1, 2, \dots, Y\}$ . Find the joint mass function of  $X$  and  $Y$ .

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$$p_Y(y) = \begin{cases} \frac{1}{100}, & y = 1, \dots, 100, \\ 0, & \text{otherwise.} \end{cases}$$

For any  $y = 1, \dots, 100$ ,

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Thus the joint mass function of  $X$  and  $Y$  is

$$p(x, y) = p_Y(y)p_{X|Y}(x|y) = \begin{cases} \frac{1}{100y}, & y = 1, \dots, 100; x = 1, \dots, y, \\ 0, & \text{otherwise.} \end{cases}$$

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Suppose that  $X$  and  $Y$  are jointly absolutely continuous with joint density  $f(\cdot, \cdot)$ . For any  $y$  with  $f_Y(y) > 0$ , the function

$$x \mapsto \frac{f(x, y)}{f_Y(y)}, \quad x \in \mathbb{R}$$

is a probability density function. It is called the conditional density of  $X$  given  $Y = y$ .

More generally, the function

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For any  $y$  with  $f_Y(y) > 0$ , the conditional density  $f_{X|Y}(x|y)$  allows us to define the conditional probability  $P(X \in A | Y = y)$ . For example, for any  $a < b$ ,

$$P(X \in (a, b) | Y = y) = \int_a^b f_{X|Y}(x|y) dx.$$

$$\begin{aligned} P(X \in (a, b) | Y = y) &= \lim_{h \downarrow 0} P(X \in (a, b) | Y \in (y - h, y + h)) \\ &= \lim_{h \downarrow 0} \frac{P(X \in (a, b), Y \in (y - h, y + h))}{P(Y \in (y - h, y + h))} \\ &= \lim_{h \downarrow 0} \frac{\frac{1}{2h} \int_{y-h}^{y+h} \int_a^b f(x, v) dx dv}{\frac{1}{2h} \int_{y-h}^{y+h} f_Y(v) dv} = \int_a^b \frac{f(x, y)}{f_Y(y)} dx = \int_a^b f_{X|Y}(x|y) dx. \end{aligned}$$

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### Example 1

Suppose the joint density of  $X$  and  $Y$  is given by

$$f(x, y) = \begin{cases} \lambda^2 e^{-\lambda y}, & 0 < x < y, \\ 0, & \text{otherwise.} \end{cases}$$

Find  $f_{Y|X}(y|x)$  for  $0 < x < y$ .

For  $x > 0$ ,

$$f_X(x) = \int_x^\infty \lambda^2 e^{-\lambda y} dy = \lambda e^{-\lambda x}.$$

Thus for  $0 < x < y$ ,

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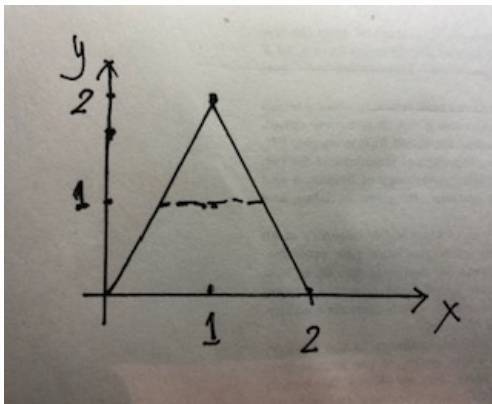
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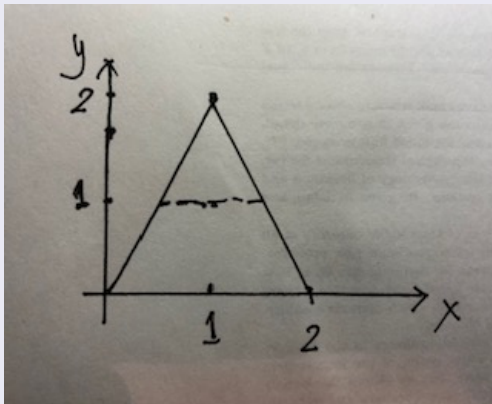
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Let  $X$  and  $Y$  be uniformly distributed in the triangle with vertices at  $(0, 0)$ ,  $(2, 0)$ ,  $(1, 2)$ . Find  $P(X \leq 1 | Y = 1)$ .



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$$f_Y(1) = \frac{1}{2} \int_{1/2}^{3/2} dx = \frac{1}{2}. \text{ Thus}$$

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Suppose that a point  $X$  is randomly chosen from the interval  $(0, 1)$ , and the a point  $Y$  is chosen randomly from  $(0, X)$ . Find the joint density of  $X$  and  $Y$ .

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