

2. Consider the following joint probability distribution $p(x, y)$ of two discrete random variables X and Y :

		x		
		1	2	$p_Y(y)$
y	1	0.14	0.06	0.20
	2	0.12	0.18	0.30
	3	0.14	0.36	0.50
$p_X(x)$		0.40	0.60	1.00

Recall STAT 400: $P(A|B) = \frac{P(A \cap B)}{P(B)}$, provided $P(B) > 0$.

$$\Rightarrow P(Y=y|X=x) = \frac{P(X=x, Y=y)}{P(X=x)}, \quad \text{provided } P(X=x) > 0.$$

Def Conditional probability mass function of Y given $X=x$ is defined by

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

For a fixed x ,

- $p_{Y|X}(y|x) \geq 0 \quad \forall y.$
- $\sum_y p_{Y|X}(y|x) = \sum_y \frac{p(x, y)}{p_X(x)} = \frac{1}{p_X(x)} \sum_y p(x, y) = \frac{1}{p_X(x)} p_X(x) = 1.$

Similarly, $P(X=x|Y=y) = \frac{P(X=x, Y=y)}{P(Y=y)}$, provided $P(Y=y) > 0$.

Def Conditional probability mass function of X given $Y=y$ is defined by

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

For a fixed y ,

- $p_{X|Y}(x|y) \geq 0 \quad \forall x.$
- $\sum_x p_{X|Y}(x|y) = \sum_x \frac{p(x,y)}{p_Y(y)} = \frac{1}{p_Y(y)} \sum_x p(x,y) = \frac{1}{p_Y(y)} p_Y(y) = 1.$

Y given X = 1	
y	$p_{Y X}(y 1)$
1	$0.14 / 0.40 = 0.35$
2	$0.12 / 0.40 = 0.30$
3	$0.14 / 0.40 = 0.35$
	1.00

Y given X = 2	
y	$p_{Y X}(y 2)$
1	$0.06 / 0.60 = 0.10$
2	$0.18 / 0.60 = 0.30$
3	$0.36 / 0.60 = 0.60$
	1.00

For example, $P(Y \geq 2 | X = 1) = 0.65.$

$P(Y \leq 2 | X = 2) = 0.40.$

Def $E(Y|X=x) = \sum_y y P(Y=y|X=x) = \sum_y y p_{Y|X}(y|x).$

Y given X = 1		
y	$p_{Y X}(y 1)$	
1	0.35	0.35
2	0.30	0.60
3	0.35	1.05
	1.00	2.00

$$E(Y|X=1) = 2.00$$

Y given X = 2		
y	$p_{Y X}(y 2)$	
1	0.10	0.10
2	0.30	0.60
3	0.60	1.80
	1.00	2.50

$$E(Y|X=2) = 2.50$$

Denote by $E(Y|X)$ that function of the random variable X whose value at $X = x$ is $E(Y|X = x)$. Note that $E(Y|X)$ is itself a random variable, it depends on the (random) value of X that occurs.

$E(Y|X)$:

x	values $E(Y X = x)$	probabilities $p_X(x)$
1	2.00	0.40
2	2.50	0.60
		1.00

$$E[E(Y|X)] = 2.00 \cdot 0.40 + 2.50 \cdot 0.60 = 2.30 = E(Y).$$

Fact: $E[E(Y|X)] = E(Y)$.

Proof: $E(Y|X)$ is a function of X .

If $X = x$, then $E(Y|X) = E(Y|X = x)$.

Recall: $E[g(X)] = \sum_{\text{all } x} g(x) \cdot p_X(x)$.

$$\begin{aligned}
 E[E(Y|X)] &= \sum_{\text{all } x} E(Y|X = x) \cdot p_X(x) \\
 &= \sum_{\text{all } x} \left[\sum_{\text{all } y} y \cdot p_{Y|X}(y|x) \right] \cdot p_X(x) \\
 &= \sum_{\text{all } x} \left[\sum_{\text{all } y} y \cdot \frac{p(x, y)}{p_X(x)} \right] \cdot p_X(x) \\
 &= \sum_{\text{all } x} \sum_{\text{all } y} y \cdot p(x, y) \\
 &= E(Y).
 \end{aligned}$$

X given Y = 1	
x	$p_{X Y}(x 1)$
1	$\frac{0.14}{0.20} = 0.70$
2	$\frac{0.06}{0.20} = 0.30$
	1.00

$$E(X|Y=1) = 1.30$$

X given Y = 2	
x	$p_{X Y}(x 2)$
1	$\frac{0.12}{0.30} = 0.40$
2	$\frac{0.18}{0.30} = 0.60$
	1.00

$$E(X|Y=2) = 1.60$$

X given Y = 3	
x	$p_{X Y}(x 3)$
1	$\frac{0.14}{0.50} = 0.28$
2	$\frac{0.36}{0.50} = 0.72$
	1.00

$$E(X|Y=3) = 1.72$$

$E(X|Y)$:

y	values $E(X Y=y)$	probabilities $p_Y(y)$
1	1.30	0.20
2	1.60	0.30
3	1.72	0.50
		1.00

Def Conditional probability density function of Y given $X = x$ is defined by

$$f_{Y|X}(y|x) = \frac{f(x, y)}{f_X(x)}.$$

For a fixed x ,

- $f_{Y|X}(y|x) \geq 0 \quad \forall y.$
- $$\begin{aligned} \int_{-\infty}^{\infty} f_{Y|X}(y|x) dy &= \int_{-\infty}^{\infty} \frac{f(x, y)}{f_X(x)} dy = \frac{1}{f_X(x)} \int_{-\infty}^{\infty} f(x, y) dy \\ &= \frac{1}{f_X(x)} f_X(x) = 1. \end{aligned}$$

Def $E(Y|X=x) = \int_{-\infty}^{\infty} y \cdot f_{Y|X}(y|x) dy.$

Def Conditional probability density function of X given $Y = y$ is defined by

$$f_{X|Y}(x|y) = \frac{f(x, y)}{f_Y(y)}.$$

For a fixed y ,

- $f_{X|Y}(x|y) \geq 0 \quad \forall x.$
- $$\begin{aligned} \int_{-\infty}^{\infty} f_{X|Y}(x|y) dx &= \int_{-\infty}^{\infty} \frac{f(x, y)}{f_Y(y)} dx = \frac{1}{f_Y(y)} \int_{-\infty}^{\infty} f(x, y) dx \\ &= \frac{1}{f_Y(y)} f_Y(y) = 1. \end{aligned}$$

Def $E(X|Y=y) = \int_{-\infty}^{\infty} x \cdot f_{X|Y}(x|y) dx.$

1. Consider two continuous random variables X and Y with joint p.d.f.

$$f_{X,Y}(x,y) = \frac{4}{3} x^3 y, \quad 0 < x < 1, \quad 0 < y < 3x, \quad \text{zero otherwise.}$$

Recall: $f_X(x) = 6x^5, \quad 0 < x < 1.$

$$f_Y(y) = \frac{1}{3}y - \frac{1}{243}y^5, \quad 0 < y < 3.$$

h) Find $P(Y > 1 \mid X = \frac{2}{3})$.

$$f_X(x) = \int_0^{3x} \frac{4}{3} x^3 y \, dy = 6x^5, \quad 0 < x < 1.$$

$$f_{Y|X}(y|x) = \frac{\frac{4}{3} x^3 y}{6x^5} = \frac{2y}{9x^2}, \quad 0 < y < 3x.$$

$$f_{Y|X}(y|\frac{2}{3}) = \frac{y}{2}, \quad 0 < y < 2.$$

$$P(Y > 1 \mid X = \frac{2}{3}) = \int_1^2 \frac{y}{2} \, dy = \frac{y^2}{4} \Big|_1^2 = \frac{3}{4}.$$

i) Find $E(Y \mid X=x)$.

$$E(Y \mid X=x) = \int_0^{3x} y \cdot \frac{2y}{9x^2} \, dy = \frac{2}{3} \frac{(3x)^3}{9x^2} = 2x, \quad 0 < x < 1.$$

j) Find $P(X < \frac{2}{3} \mid Y = 1)$.

$$\begin{aligned} f_Y(y) &= \int_{y/3}^1 \frac{4}{3} x^3 y \, dx = \left(\frac{1}{3} x^4 y \right)_{x=y/3}^{x=1} \\ &= \frac{1}{3} y - \frac{1}{243} y^5, \quad 0 < y < 3. \end{aligned}$$

$$f_{X|Y}(x|y) = \frac{\frac{4}{3} x^3 y}{\frac{1}{3} y - \frac{1}{243} y^5} = \frac{4 x^3}{1 - \frac{y^4}{81}}, \quad \frac{y}{3} < x < 1.$$

$$f_{X|Y}(x|1) = \frac{81}{20} x^3, \quad \frac{1}{3} < x < 1.$$

$$P(X < \frac{2}{3} \mid Y = 1) = \int_{1/3}^{2/3} \frac{81}{20} x^3 \, dx = \frac{81}{80} x^4 \Big|_{1/3}^{2/3} = \frac{15}{80} = \frac{3}{16}.$$

k) Find $E(X \mid Y = y)$.

$$\begin{aligned} E(X \mid Y = y) &= \int_{y/3}^1 x \cdot \frac{4 x^3}{1 - \frac{y^4}{81}} \, dx = \frac{4}{5} \cdot \frac{1 - \frac{y^5}{243}}{1 - \frac{y^4}{81}} = \frac{4}{15} \cdot \frac{243 - y^5}{81 - y^4}, \\ &0 < y < 3. \end{aligned}$$