

CONDITIONAL DISTRIBUTIONS

Let (X, Y) be a pair of discrete r.v.s with support \mathcal{X} and \mathcal{Y} , respectively, joint pmf p and marginal pmfs p_X and p_Y .

Consider the events $\{X \in A\}$ and $\{Y \in B\}$ for some $A \subset \mathcal{X}$ and $B \subset \mathcal{Y}$.

The conditional probability of $\{X \in A\}$ given $\{Y \in B\}$ is given by

$$\begin{aligned} P(X \in A | Y \in B) &= \frac{P(\{X \in A\} \cap \{Y \in B\})}{P_Y(Y \in B)} \\ &= \frac{P(\{s \in S : (X(s), Y(s)) \in A \times B\})}{P(\{s \in S : Y(s) \in B\})}. \end{aligned}$$

We may now define the conditional pmf of X given Y :

Defn: For any value y for which $p_Y(y) > 0$, the conditional pmf of X given $Y=y$ is the function given by

$$p(x|y) = P(X=x | Y=y) = \frac{\underbrace{p(x,y)}_{\substack{\text{joint pmf of } (x,y)}}}{\underbrace{p_Y(y)}_{\substack{\text{marginal pmf of } Y}}}$$

for all $x \in \mathbb{R}$.

Likewise, for any value x for which $p_X(x) > 0$, the conditional pmf of Y given that $X=x$ is the function given by

$$p(y|x) = P(Y=y | X=x) = \frac{\underbrace{p(x,y)}_{\substack{\text{marginal pmf of } X}}}{\underbrace{p_X(x)}_{\substack{\text{marginal pmf of } X}}}$$

for all $y \in \mathbb{R}$.

Remark: The conditional pmfs are legitimate pmfs:

- $\sum_{x \in \mathcal{X}} p(x|y) = \sum_{x \in \mathcal{X}} \frac{p(x,y)}{p_Y(y)} = \frac{p_Y(y)}{p_Y(y)} = 1$

- $p(x|y) \geq 0$ for all $x \in \mathbb{R}$ since $p(x,y) \geq 0$ and $p_y(y) > 0$.
 (likewise for $p(y|x)$)

E.g. Roll two dice. Let $X = \text{sum of rolls}$, $Y = \text{maximum of rolls}$.

Get conditional pmfs of X given $Y=4$ and Y given $X=7$.

Consider single column or single row of joint pmf table:

	Y						$p_X(x)$	marginal pmf of X
	1	2	3	4	5	6		
2	$\frac{1}{36}$						$\frac{1}{36}$	
3		$\frac{2}{36}$					$\frac{2}{36}$	
4		$\frac{1}{36}$	$\frac{2}{36}$				$\frac{3}{36}$	
5			$\frac{2}{36}$	$\frac{3}{36}$			$\frac{4}{36}$	
6			$\frac{1}{36}$	$\frac{2}{36}$	$\frac{3}{36}$		$\frac{5}{36}$	
7				$\frac{1}{36}$	$\frac{1}{36}$	$\frac{1}{36}$	$\frac{6}{36}$	
8					$\frac{1}{36}$	$\frac{1}{36}$	$\frac{1}{36}$	
9					$\frac{1}{36}$	$\frac{1}{36}$	$\frac{1}{36}$	
10					$\frac{1}{36}$	$\frac{1}{36}$	$\frac{1}{36}$	
11						$\frac{1}{36}$	$\frac{1}{36}$	
12						$\frac{1}{36}$	$\frac{1}{36}$	
$p_Y(y)$	$\frac{1}{36}$	$\frac{2}{36}$	$\frac{3}{36}$	$\frac{5}{36}$	$\frac{6}{36}$	$\frac{10}{36}$	1	

marginal pmf of Y
 $\rightarrow p_Y(y)$

Joint probabilities
 $p(x,y)$

Conditional pmf of X given that $Y=4$
 \downarrow
 $p(x|4)$

$\frac{p(5,4)}{p_Y(4)} = \frac{2/36}{4/36} = \frac{1}{2}$

Conditional pmf of Y given that $X=7$

Let (X, Y) be a pair of continuous r.v.s with joint pdf f and marginal pdfs f_X and f_Y .

We define conditional pdfs much like we have defined conditional pmfs.

Defn: For any $y \in \mathbb{R}$ such that $f_Y(y) > 0$, the conditional pdf of X given that $Y=y$ is

$$f(x|y) = \frac{f(x,y)}{f_Y(y)} \quad \text{for all } x \in \mathbb{R}.$$

Likewise, for any $x \in \mathbb{R}$ such that $f_X(x) > 0$, the conditional pdf of Y given that $X=x$ is

$$f(y|x) = \frac{f(x,y)}{f_X(x)} \quad \text{for all } y \in \mathbb{R}.$$

Remark: The conditional pdfs are legitimate pdfs:

- $\int_{\mathbb{R}} f(x|y) dx = \int_{\mathbb{R}} \frac{f(x,y)}{f_Y(y)} dx = \frac{f_Y(y)}{f_Y(y)} = 1$
- $f(x|y) \geq 0$ for all $x \in \mathbb{R}$ since $f(x,y) \geq 0$ and $f_Y(y) > 0$.

(Likewise for $f(y|x)$)

Eg: Let (X, Y) have joint pdf given by

$$f(x,y) = y(1-x)^{y-1} e^{-y} \mathbf{1}(0 < x < 1, 0 < y < \infty).$$

- a) Find the conditional pdf $f(x|y)$ of $X|Y=y$.

We have

$$f(x|y) = \frac{f(x,y)}{f_Y(y)},$$

so we first need to find the marginal pdf of Y :

$$\begin{aligned} f_Y(y) &= \int_0^1 y(1-x)^{y-1} e^{-y} dx \cdot \mathbb{1}(0 < y < \infty) \\ &= y e^{-y} \cdot \mathbb{1}(0 < y < \infty) \cdot \underbrace{\int_0^1 (1-x)^{y-1} dx}_{= \frac{1}{y}} \\ &= e^{-y} \cdot \mathbb{1}(0 < y < \infty). \end{aligned}$$

So, for $0 < x < 1$ and $y > 0$, we have

$$\begin{aligned} f(x|y) &= \frac{y(1-x)^{y-1} e^{-y}}{e^{-y}} \\ &= y(1-x)^{y-1}. \end{aligned}$$

If we study this pdf carefully, we can see that

$$X|Y=y \sim \text{Beta}(2, y).$$

(b) Find $P(X < \frac{1}{2} | Y=2)$.

We have

$$\begin{aligned} P(X < \frac{1}{2} | Y=2) &= \int_0^{\frac{1}{2}} 2(1-x)^{2-1} dx \\ &= 2 \left(x - \frac{x^2}{2} \right) \Big|_0^{\frac{1}{2}} \\ &= 2 \left(\frac{1}{2} - \frac{1}{4} \right) \\ &= \frac{3}{4}. \end{aligned}$$

Eg: let (x, y) have joint pdf given by

$$f(x, y) = \frac{1}{2\pi} x^{-\frac{3}{2}} \exp\left[-\frac{1}{2x}(y^2+1)\right] \mathbf{1}(0 < x < \infty, -\infty < y < \infty).$$

(a) Find the conditional pdf $f(y|x)$ of $y|x=x$.

Since

$$f(y|x) = \frac{f(x,y)}{f_X(x)},$$

we must first find the marginal pdf f_X of X .

$$\begin{aligned} f_X(x) &= \int_{-\infty}^{\infty} \frac{1}{2\pi} x^{-\frac{3}{2}} \exp\left[-\frac{1}{2x}(y^2+1)\right] \mathbf{1}(0 < x < \infty) dy \\ &= \frac{x^{-\frac{3}{2}}}{\sqrt{2\pi}} e^{-\frac{1}{2x}} \underbrace{\sqrt{x} \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} \frac{1}{\sqrt{x}} e^{-\frac{y^2}{2x}} dy}_{=1} \mathbf{1}(0 < x < \infty) \\ &= \frac{1}{\sqrt{2\pi}} \frac{1}{x} e^{-\frac{1}{2x}} \mathbf{1}(0 < x < \infty). \end{aligned}$$

So, for $0 < x < \infty$ and $-\infty < y < \infty$, we have

$$\begin{aligned} f(y|x) &= \frac{\frac{1}{2\pi} x^{-\frac{3}{2}} \exp\left[-\frac{1}{2x}(y^2+1)\right]}{\frac{1}{\sqrt{2\pi}} \frac{1}{x} e^{-\frac{1}{2x}}} \\ &= \frac{1}{\sqrt{2\pi}} \frac{1}{\sqrt{x}} \exp\left[-\frac{y^2}{2x}\right], \end{aligned}$$

which we recognize as the pdf of the $\text{Normal}(0, \sigma^2)$

$$Y|X=x \sim \text{Normal}(0, \sigma^2).$$

(b) Find $P(Y > 1 | X=1)$.

We have

$$\begin{aligned} P(Y > 1 | X=1) &= P\left(\frac{Y-1}{\sigma} > \frac{1-\mu}{\sigma} | X=1\right) \\ &= P(Z > \frac{1}{\sigma}), \quad Z \sim \text{Normal}(0, 1) \\ &= 1 - \underbrace{\Phi\left(\frac{1}{\sigma}\right)}_{\text{pnorm}(0.5)} \\ &= 0.3085375 \end{aligned}$$

E.g. Let (U, V) have joint pdf given by

$$f(u, v) = 6(v-u) \mathbf{1}(0 < u < v < 1).$$

(a) Find the conditional density $f(u|v)$ of $U|V=v$.

Since

$$f(u|v) = \frac{f(u, v)}{f_V(v)}$$

when $f_V(v) > 0$, we must first find the marginal pdf $f_V(v)$ of V .

We have, for $0 < v < 1$,

$$\begin{aligned}
f_v(v) &= \int_{-\infty}^{\infty} 6(v-u) \mathbf{1}(0 < u < v) du \\
&= \int_0^v 6(v-u) du \\
&= 6 \left(vu - \frac{u^2}{2} \right) \Big|_0^v \\
&= 6 \cdot \frac{v^2}{2} \\
&= 3v^2.
\end{aligned}$$

So we have, for $0 < v < 2$,

$$\begin{aligned}
f(u|v) &= \frac{6(v-u)}{3v^2} \mathbf{1}(0 < u < v) \\
&= 2 \left(\frac{1}{v} - \frac{u}{v^2} \right) \mathbf{1}(0 < u < v).
\end{aligned}$$

(b) Find $P(U < \frac{1}{2} | V = \frac{1}{2})$.

We have

$$\begin{aligned}
P(U < \frac{1}{2} | V = \frac{1}{2}) &= \int_0^{\frac{1}{2}} 2 \left(\frac{1}{\frac{1}{2}} - \frac{u}{(\frac{1}{2})^2} \right) du \\
&= 4 \left(u - \frac{u^2}{2} \right) \Big|_0^{\frac{1}{2}}
\end{aligned}$$

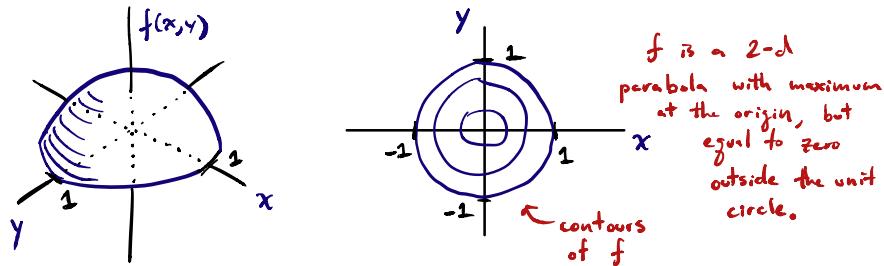
$$= 4 \left[\frac{1}{9} - \left(\frac{1}{9} \right)^2 \right]$$

$$= 1 - \frac{1}{9}$$

$$= \frac{8}{9}.$$

E.g. (2-dimensional Epanechnikov Kernel)

Let $(x, y) \sim f(x, y) = \frac{2}{\pi} [1 - (x^2 + y^2)]_+$, where $[\cdot]_+ = \max(0, \cdot)$.



(i) Get conditional pdf of X given that $Y=y$ for some $y \in (-1, 1)$.

By the definition,

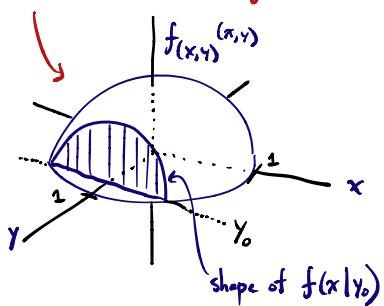
$$f(x|y) = \frac{f(x, y)}{f_y(y)}.$$

We need the marginal pdf of Y

$$\begin{aligned} \text{First get } f_y(y) &= \int_{-\infty}^{\infty} \frac{2}{\pi} \underbrace{[1 - (x^2 + y^2)]_+}_{\text{is positive when } x^2 < 1-y^2 \Leftrightarrow |x| < \sqrt{1-y^2}} dx \\ &= \int_{-\sqrt{1-y^2}}^{\sqrt{1-y^2}} \frac{2}{\pi} [(1-y^2) - x^2] dx \end{aligned}$$

$$= \frac{2}{\pi} \left[(1-y^2)x - \frac{x^3}{3} \right] \Bigg|_{-\sqrt{1-y^2}}^{\sqrt{1-y^2}}$$

Slice the joint pdf at some y , then rescale so that the cross-section integrates to 1



$$= \frac{2}{\pi} \left[\left((1-y^2)\sqrt{1-y^2} - \frac{(1-y^2)^{3/2}}{3} \right) - \left((1-y^2)(-\sqrt{1-y^2}) + \frac{(1-y^2)^{3/2}}{3} \right) \right]$$

$$= \frac{2}{\pi} \left[\frac{6}{3} (1-y^2)^{3/2} - \frac{2}{3} (1-y^2)^{3/2} \right]$$

$$= \frac{4}{3\pi} (1-y^2)^{3/2}.$$

Then

$$f(x|y) = \frac{\frac{2}{\pi} [1-(x^2+y^2)]_+}{\frac{8}{3\pi} (1-y^2)^{3/2}}$$

We can use an indicator function instead of $[\cdot]_+$

$$= \frac{\frac{3}{4} [1-(x^2+y^2)]}{(1-y^2)^{3/2}} \mathbb{1}(-\sqrt{1-y^2} < x < \sqrt{1-y^2}) \quad \text{for all } x \in \mathbb{R}.$$

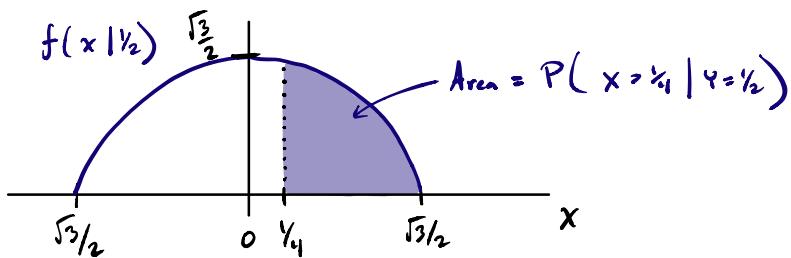
$$(ii) \text{ Find } P(X > y_1 | Y = y_2).$$

The conditional pdf of X given that $Y = y_2$ is

$$f(x|y_2) = \frac{\frac{3}{4} [1 - (x^2 + (y_2)^2)]}{(1 - (y_2)^2)^{3/2}} \mathbb{1}(-\sqrt{1-(y_2)^2} < x < \sqrt{1-(y_2)^2})$$

$$\left(\sqrt{1-(y_2)^2} = \sqrt{3/4} = \frac{\sqrt{3}}{2}\right)$$

$$= \frac{2}{\sqrt{3}} \left(\frac{3}{4} - x^2\right) \mathbb{1}(-\sqrt{3}/2 < x < \sqrt{3}/2)$$



$$P(X > y_1 | Y = y_2) = \int_{y_1}^{\infty} \frac{2}{\sqrt{3}} \left(\frac{3}{4} - x^2\right) \mathbb{1}(-\sqrt{3}/2 < x < \sqrt{3}/2) dx$$

$$\left(\frac{1}{4} < \frac{\sqrt{3}}{2}\right) = \int_{y_1}^{\sqrt{3}/2} \frac{2}{\sqrt{3}} \left(\frac{3}{4} - x^2\right) dx$$

$$= \frac{2}{\sqrt{3}} \left[\frac{3}{4}x - \frac{x^3}{3} \right]_{y_1}^{\sqrt{3}/2}$$

$$= \frac{2}{\sqrt{3}} \left[\left(\frac{3}{4} \frac{\sqrt{3}}{2} - \left(\frac{\sqrt{3}/2}{3} \right)^3 \right) - \left(\frac{3}{4} \frac{1}{4} - \left(\frac{y_1}{3} \right)^3 \right) \right]$$

$$= 2 \left(\frac{3}{8} - \frac{1}{8} \right) - \frac{2}{\sqrt{3}} \left(\frac{3 \cdot 3 \cdot 4 - 1}{192} \right)$$

$$= \frac{1}{2} - \frac{1}{\sqrt{3}} \frac{35}{96}$$

CONDITIONAL EXPECTATION

Let (X, Y) be a pair of discrete or continuous r.v.s.

Defn: For any function $g: \mathbb{R} \rightarrow \mathbb{R}$ and any value y such that $p_y(y) > 0$ or $f_y(y) > 0$, the conditional expectation of $g(X)$ given that $Y=y$ is

$$\mathbb{E}[g(X)|Y=y] = \begin{cases} \sum_{x \in X} g(x) \cdot p(x|y) & \text{if } (X, Y) \text{ is discrete} \\ \int_{\mathbb{R}} g(x) \cdot f(x|y) dx & \text{if } (X, Y) \text{ is continuous.} \end{cases}$$

And likewise for $\mathbb{E}[g(Y)|X=x]$.

Remark: The conditional expectation, $\mathbb{E}[g(X)|Y=y]$ is a function of y , since X is summed / integrated out.

Set aside the function $g: \mathbb{R} \rightarrow \mathbb{R}$ for a moment:

If we write $\mathbb{E}[X|Y]$, without choosing a value y for Y , then $\mathbb{E}[X|Y]$ is a random variable, as its value depends on the value of the random variable Y .

E.g. Roll two dice. Let $X = \text{sum of rolls}$, $Y = \text{maximum of rolls}$.

Get conditional expectation of X given $Y=4$ and Y given $X=7$.

From before:

$$\begin{array}{c|ccccc} x & 5 & 6 & 7 & 8 \\ \hline P_{X|Y}(x|4) & \frac{1}{4} & \frac{1}{4} & \frac{1}{4} & \frac{1}{4} \end{array} \quad \text{gives } \mathbb{E}[X|Y=4] = \frac{2(5) + 2(6) + 2(7) + 1(8)}{7} = \frac{44}{7}$$

$$\begin{array}{c|ccc} y & 4 & 5 & 6 \\ \hline P_{Y|X}(y|7) & \frac{1}{6} & \frac{1}{6} & \frac{1}{6} \end{array} \quad \text{gives } \mathbb{E}[Y|X=7] = \frac{2(4) + 2(5) + 2(6)}{6} = 5$$

E.g. Let (U, V) have joint pdf given by

$$f(u, v) = 6(v-u) \mathbf{1}(0 < u < v < 1).$$

(a) Find $\mathbb{E}[U | V=v]$.

We have

$$\begin{aligned} \mathbb{E}[U | V=v] &= \int_0^v u \cdot 2 \left(\frac{1}{v} - \frac{u}{v^2} \right) du \\ &= 2 \left(\frac{u^2}{2v} - \frac{u^3}{3v^2} \right) \Big|_0^v \\ &= \frac{v}{3}. \end{aligned}$$

(b) Give $\mathbb{E}[U | V=\frac{1}{2}]$.

We have

$$\mathbb{E}[U | V=\frac{1}{2}] = \frac{1}{2} - \left(\frac{2}{3}\right)\left(\frac{1}{2}\right)^2 = \frac{1}{2} - \frac{1}{6} = \frac{1}{3}.$$

E.g. Let $(X, Y) \sim \frac{2}{\pi} [1 - (x^2 + y^2)]_+$.

(i) Find $\mathbb{E}[X | Y=y]$ for any $y \in (-1, 1)$.

$$\mathbb{E}[X | Y=y] = \int_{-\infty}^{\infty} x \cdot f(x|y) dx$$

$$\begin{aligned}
&= \int_{\mathbb{R}} x \cdot \frac{\frac{3}{4} [1 - (x^2 + y^2)]}{(1-y^2)^{3/2}} \mathbb{I}(-\sqrt{1-y^2} < x < \sqrt{1-y^2}) dx \\
&= \frac{3/4}{(1-y^2)^{3/2}} \int_{-\sqrt{1-y^2}}^{\sqrt{1-y^2}} [(1-y^2)x - x^3] dx \\
&= \frac{3/4}{(1-y^2)^{3/2}} \left[\frac{(1-y^2)x^2}{2} - \frac{x^4}{4} \right] \Big|_{-\sqrt{1-y^2}}^{\sqrt{1-y^2}} \\
&= \frac{3/4}{(1-y^2)^{3/2}} \left[\left(\frac{(1-y^2)^2}{2} - \frac{(1-y^2)^2}{4} \right) - \left(\frac{(1-y^2)^2}{2} - \frac{(1-y^2)^2}{4} \right) \right] \\
&= 0.
\end{aligned}$$

↑
We might have guessed: A cross-section of
the joint pdf at any y is symmetric
about 0.

(ii) Find $\mathbb{E}[x^2 | y=y]$.

$$\begin{aligned}
\mathbb{E}[x^2 | y=y] &= \frac{3/4}{(1-y^2)^{3/2}} \int_{-\sqrt{1-y^2}}^{\sqrt{1-y^2}} [(1-y^2)x^2 - x^4] dx \\
&= \frac{3/4}{(1-y^2)^{3/2}} \left[(1-y^2) \frac{x^3}{3} - \frac{x^5}{5} \right] \Big|_{-\sqrt{1-y^2}}^{\sqrt{1-y^2}}
\end{aligned}$$

$$\begin{aligned}
&= \frac{3/4}{(1-y^2)^{3/2}} \left[\left(\left(\frac{1-y^2}{3}\right)^{5/2} - \left(\frac{1-y^2}{5}\right)^{5/2} \right) - \left(-\left(\frac{1-y^2}{3}\right)^{5/2} + \left(\frac{1-y^2}{5}\right)^{5/2} \right) \right] \\
&= (1-y^2) \frac{3}{4} \left[\frac{2}{3} - \frac{2}{5} \right] \\
&= \frac{1}{5} (1-y^2).
\end{aligned}$$

Defn: The conditional variance of X given that $Y=y$ is

$$\text{Var}[X|Y=y] = \mathbb{E}\left[\left(X - \mathbb{E}[X|Y=y]\right)^2 \mid Y=y\right].$$

Remark: If we write $\text{Var}[X|Y]$, without choosing a value y for Y , then $\text{Var}[X|Y]$ is a random variable, as it depends on the value of the random variable Y .

Useful expression: $\text{Var}[X|Y] = \mathbb{E}[X^2|Y] - (\mathbb{E}[X|Y])^2$

E.g.: Let $(X, Y) \sim \frac{2}{\pi} [1 - (x^2 + y^2)]_+$.

From before: $\mathbb{E}[X|Y] = 0$ and $\mathbb{E}[X^2|Y] = \frac{1}{5} (1-Y^2)$, so

$$\text{Var}[X|Y] = \frac{1}{5} (1-Y^2).$$

Then, for example, $\text{Var}[X \mid Y=\frac{1}{2}]$ is given by

$$\begin{aligned}
\text{Var}[X \mid Y=\frac{1}{2}] &= \frac{1}{5} \left[1 - \left(\frac{1}{2}\right)^2 \right] \\
&= \frac{3}{20}
\end{aligned}$$

E.g. Let (X, Y) have joint pdf given by

$$f(x, y) = y(1-x)^{y-1} e^{-y} \mathbf{1}(0 < x < 1, 0 < y < \infty).$$

(a) Find $\mathbb{E}[X | Y=y]$.

Use the fact that $X|Y=y \sim \text{Beta}(z, y)$, $\mathbb{E}[X|Y=y] = \frac{z}{z+y}$.

We can also obtain by the calculation

$$\begin{aligned}\mathbb{E}[X | Y=y] &= \int_0^1 x \cdot y(1-x)^{y-1} dx \\ &= y \cdot \underbrace{\int_0^1 x^{z+1} (1-x)^{y-1} dx}_{B(z, y) = \frac{\Gamma(z)\Gamma(y)}{\Gamma(z+y)}} \\ &= y \cdot \frac{\Gamma(z)\Gamma(y)}{\Gamma(z+y)} \\ &= \frac{y \Gamma(y)}{(y+1)y \Gamma(y)} \\ &= \frac{1}{1+y}.\end{aligned}$$

(b) Find $\text{Var}[X | Y=y]$.

We have, since $X|Y=y \sim \text{Beta}(z, y)$, that $\text{Var}[X|Y=y] = \frac{y}{(y+1)^2(y+2)}$.

We can also find this as $\mathbb{E}[X^2 | Y=y] - (\mathbb{E}[X | Y=y])^2$. We have

$$\begin{aligned}\mathbb{E}[X^2 | Y=y] &= \int_0^1 x^2 \cdot y(1-x)^{y-1} dx \\ &= y \int_0^1 x^{3-1} (1-x)^{y-1} dx\end{aligned}$$

$$\begin{aligned}
 &= \gamma \cdot \frac{P(z) P(y)}{P(z+y)} \\
 &= \frac{\gamma \cdot 2 \cdot P(y)}{(y+2)(y+1) y P(y)} \\
 &= \frac{2}{(y+2)(y+1)}
 \end{aligned}$$

so that

$$\begin{aligned}
 V_{\nu} [X = |Y=y] &= \frac{2}{(y+2)(y+1)} - \left[\frac{1}{y+1} \right]^2 \\
 &= \frac{(y+1)^2 - (y+2)}{(y+1)^2 (y+2)} \\
 &= \frac{y}{(y+1)^2 (y+2)}.
 \end{aligned}$$

ADDENDUM: To find C such that $C [1 - (x^2 + y^2)]_+$ is a joint pdf, we do the following:

$$\begin{aligned}
 1 &= \iint_{\mathbb{R}^2} C [1 - (x^2 + y^2)]_+ dx dy \\
 &= C \iint_{\mathbb{R}^2} [1 - (x^2 + y^2)] \mathbf{1}(x^2 + y^2 < 1) dx dy \\
 &\quad \left(\text{switch to polar coordinates: } r^2 = x^2 + y^2, x = r \cos \theta, y = r \sin \theta \right. \\
 &\quad \qquad \qquad \qquad \left. dr dy = r dr d\theta, 0 < r < 1, 0 < \theta < 2\pi \right. \\
 &= C \int_0^{2\pi} \int_0^1 (1 - r^2) r dr d\theta \\
 &= C \int_0^{2\pi} \left(\frac{r^2}{2} - \frac{r^4}{4} \right) \Big|_0^1 d\theta \\
 &= C \int_0^{2\pi} \frac{1}{4} d\theta \\
 &= C \frac{2\pi}{4} \quad \Rightarrow \quad C = \frac{2}{\pi}
 \end{aligned}$$