

Self - Understanding

→ discrete r.v. everything is good

$$\textcircled{1} \begin{cases} \text{pmf} & f_X(x) := \mathbb{P}(X=x) \\ \text{cdf} & F_X(x) := \mathbb{P}(X \leq x) \end{cases} \quad \underline{f_X(x) = F_X(x) - F_X(x-1)}$$

② Conditional distribution

$$\begin{cases} \text{pmf} & f_{X|Y}(x|y) := \mathbb{P}(X=x|Y=y) = \mathbb{P}(\tilde{X}_y=x) \\ \text{cdf} & F_{X|Y}(x|y) := \mathbb{P}(X \leq x|Y=y) = \mathbb{P}(\tilde{X}_y \leq x) \end{cases} \quad \text{where } \tilde{X}_y = X|Y=y$$

→ continuous r.v. something breaks down $f_X(x) \Delta x = \mathbb{P}(x < X \leq x+\Delta x)$

$$\textcircled{1} \begin{cases} \text{pdf} & \boxed{f_X(x) := \frac{1}{\Delta x} \mathbb{P}(X=x)} \\ \text{cdf} & F_X(x) = \mathbb{P}(X \leq x) \end{cases} \rightarrow \begin{aligned} f_X(x) &:= \lim_{\Delta x \rightarrow 0} \frac{\mathbb{P}(x < X \leq x+\Delta x)}{\Delta x} \\ &= \lim_{\Delta x \rightarrow 0} \frac{F(x+\Delta x) - F(x)}{\Delta x} \\ &= F'(x) \end{aligned}$$

② Conditional distribution

$$\begin{cases} \text{pdf} & \boxed{f_{X|Y}(x|y)} \\ \text{cdf} & F_{X|Y}(x|y) := \mathbb{P}(X \leq x|Y=y) \end{cases}$$

derivative

$$\begin{aligned} F_{X|Y}(x|y) &:= \lim_{\Delta y \rightarrow 0} \mathbb{P}(X \leq x | y \leq Y \leq y+\Delta y) \\ &= \frac{\int_{-\infty}^x f(u,y) dy}{f_Y(y)} \end{aligned}$$

not well defined
since $\frac{\mathbb{P}(X \leq x, Y=y)}{\mathbb{P}(Y=y)} = \frac{0}{0}$

Since $f_{X|Y}(x|y)$ is the derivative of $F_{X|Y}(x|y)$

$$\text{then } f_{X|Y}(x|y) := \lim_{\Delta x \rightarrow 0} \frac{F(x+\Delta x|y) - F(x|y)}{\Delta x}$$

$$= \lim_{\Delta x \rightarrow 0} \lim_{\Delta y \rightarrow 0} \frac{\mathbb{P}(x < X \leq x+\Delta x | y \leq Y \leq y+\Delta y)}{\Delta x}$$

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

$$X | Y = y_0$$

a) cdf $P(X \leq x_0 | Y = y_0) := \lim_{\Delta y \rightarrow 0} P(X \leq x_0 | Y \in [y_0, y_0 + \Delta y]) = \frac{\int_{-\infty}^{x_0} f(u, y_0) du}{f_Y(y_0)}$

note that $P_X(x) := \lim_{\Delta x \rightarrow 0} \frac{P(X \in [x, x + \Delta x])}{\Delta x}$

$$= \frac{dP(X \leq x)}{dx}$$

b) pdf $P_{X|Y}(x_0 | y_0) = \frac{dP(X \leq x | Y = y)}{dx} \Big|_{x=x_0} = \frac{f(x_0, y_0)}{f_Y(y_0)}$

pf: $P_{X|Y}(x_0 | y_0) = \frac{P_{X,Y}(x_0, y_0)}{P_Y(y_0)}$

$$\rightarrow P(X \leq x | Y = y_0) = \lim_{\Delta y \rightarrow 0} \frac{P(X \leq x, Y \in [y_0, y_0 + \Delta y])}{P(Y \in [y_0, y_0 + \Delta y])}$$

$$= \lim_{\Delta y \rightarrow 0} \frac{F(x, y_0 + \Delta y) - F(x, y_0)}{\Delta y}$$

$$= \lim_{\Delta y \rightarrow 0} \frac{F_Y(y_0 + \Delta y) - F_Y(y_0)}{\Delta y}$$

$$= \frac{\frac{\partial F(x, y)}{\partial y}}{\frac{dF_Y(y)}{dy}}$$

$$= \frac{\int_{-\infty}^x f(u, y) du}{f_Y(y)}$$

c) from b), we have

$$P_{X|Y}(x_0 | y_0) = \frac{P_{X,Y}(x_0, y_0)}{P_Y(y_0)} \quad \text{pdf}$$

$$P_{X,Y}(x_0, y_0) = P_{X|Y}(x_0 | y_0) P_Y(y_0)$$

SS

$$P(X = x_0, Y = y_0) = P(X = x_0 | Y = y_0) P(Y = y_0)$$

d) $P(x) = \int_{-\infty}^{\infty} P(x, y) dy$ & $P(y = x) = \int P(X = x, Y = y)$

①

$$\boxed{\text{Pf:}} \quad P(X \leq x) = \sum_{\text{partition}} P(X \leq x, Y \in \text{Partition } i)$$

$$\begin{aligned} \textcircled{2} \quad & P(X \leq x) \\ &= P(X \leq x, Y \leq +\infty) \\ &= \int_{-\infty}^x \int_{-\infty}^{+\infty} p_{X,Y}(u,v) dv du \\ \Rightarrow & \underline{p_X(x) = \int_{-\infty}^{+\infty} p_{X,Y}(x,v) dv} \end{aligned}$$

$$\begin{aligned} &= \sum_{\text{part}} \underbrace{P(X \leq x | Y \in \text{Partition})}_{\text{fun defn}} \cdot \underbrace{P(Y \in \text{Partition})}_{\downarrow P(Y \in [y, y+\Delta y])} \\ &\stackrel{\Delta y \rightarrow 0}{=} \int_{\mathbb{R}} \underbrace{P(X \leq x | Y = y)}_{\text{defn}} d \underbrace{P(Y \leq y)}_{\downarrow p_Y(y)} \\ &= \int_{-\infty}^{+\infty} \underbrace{p_{X|Y}(x|y)}_{\text{defn}} \underbrace{p_Y(y)}_{\downarrow} dy \\ &= \underline{\int_{-\infty}^{+\infty} p_{X,Y}(x,y) dy} \end{aligned}$$

$$\begin{aligned} \textcircled{2} \quad & E[X] = E\left[\sum_i X \cdot \mathbb{1}_{\{X \in P_i\}}\right] \\ &= \sum_i E[X \cdot \mathbb{1}_{\{X \in P_i\}}] \\ &\stackrel{\Delta x \rightarrow 0}{=} \int x P(X \in P_i) \\ &= \int_{\mathbb{R}} x dP(X \leq x) \end{aligned}$$

Change Of Variable Summary

① 1-variate case

$$\begin{aligned} &\rightarrow X \sim P_X \quad \underline{Y = f(X)} \\ &\rightarrow P_Y ? \end{aligned}$$

(stronger condition)

Answer: Condition (Assump.) \rightarrow $f(\cdot)$ is monotone

then $P(Y \leq y)$

$$= P(f(X) \leq y) = \underline{P(X \leq f^{-1}(y))}$$

$$= \int_X p(x) dx \quad X := \{x : f(x) \leq y\}$$

$$\stackrel{y=f(x)}{=} \int_Y p(f^{-1}(y)) \left| \frac{df^{-1}(y)}{dy} \right| dy \quad Y := \{y : y \leq y\}$$

we need $f'(x) \neq 0$

$$y = f(x)$$

$$\rightarrow y - f(x) = 0$$

$$\rightarrow \begin{cases} \textcircled{1} x = f^{-1}(y) \end{cases}$$

$$\begin{cases} \textcircled{2} f^{-1}(y)' = - \frac{F_y}{F_x} \end{cases}$$

$$= \frac{1}{f'(x)}$$

Implicit Function

② Multi-variate case

$$\rightarrow (X, Y) \sim P_{X,Y}$$

$$\begin{cases} U = f_1(X, Y) \\ V = f_2(X, Y) \end{cases}$$

$$\rightarrow P_{U,V} ?$$

Answer: Similarly, $P(U \leq u, V \leq v)$

$$= P(f_1(X, Y) \leq u, f_2(X, Y) \leq v)$$

$$= \int_S P_{X,Y}(x, y) dx dy \quad S = \{(x, y) : \begin{matrix} f_1(x, y) \leq u \\ f_2(x, y) \leq v \end{matrix}\}$$

change of variable:

$$\begin{cases} f_1(x, y) = a \\ f_2(x, y) = b \end{cases}$$

$$\begin{cases} x = g_1(a, b) \\ y = g_2(a, b) \end{cases}$$

change of variable

$$\int_S P_{X,Y}(g_1(a, b), g_2(a, b)) \left| \begin{matrix} g_{1,a} & g_{1,b} \\ g_{2,a} & g_{2,b} \end{matrix} \right| da db$$

implicit function theorem

$$\begin{cases} u - f_1(x, y) = 0 \\ v - f_2(x, y) = 0 \end{cases}$$

$$S = \{(a, b) : \begin{matrix} a \leq u \\ b \leq v \end{matrix}\}$$

$$\Rightarrow F(x, y; u, v) := \begin{pmatrix} u - f_1(x, y) \\ v - f_2(x, y) \end{pmatrix}$$
