

Conditional Distributions in Continuous Case.

Suppose Y is a continuous r.v., s.t.

$$f_Y(y) > 0. \quad (\text{i.e. } y \in \text{domain (support)} \text{ of } Y.)$$

Assume that (X, Y) are jointly continuous. Then the conditional density

of $X | Y = y$ is

$$f_{X|Y}(x|y) = \frac{\underbrace{f(x,y)}_{\text{joint density}}}{\underbrace{f_Y(y)}_{\text{marginal density}}}$$

So

$$P(X \in A | Y = y) = \int_A f_{X|Y}(x|y) dx$$

(the cond'l distribution of $X | Y = y$
is a dist'n in x ; y acts as a parameter).

example $f(x,y) = \frac{15}{2} x(2-x-y), \quad 0 < x < 1$
 $0 < y < 1.$

$\Rightarrow X \not\perp Y$ (f does not factor
into a product $h(x)g(y)$)

into a producer [maximize]

$$\begin{aligned}
 f_Y(y) &= \int_{-\infty}^{\infty} f(x, y) dx \\
 &= \int_0^1 \frac{15}{2} x(2-x-y) dx \\
 &= \left[\frac{15x^2}{2} - \frac{15x^3}{6} - \frac{15yx^2}{4} \right]_0^1 \\
 &= 5 - \frac{15y}{4}, \quad 0 < y < 1.
 \end{aligned}$$

$$\begin{aligned}
 f_{X|Y}(x|y) &= \frac{\frac{15}{2} x(2-x-y)}{5 - \frac{15y}{4}} = \frac{6x(2-x-y)}{4-3y}, \\
 &\quad 0 < x < 1 \\
 &\quad (y \in (0, 1) \text{ fixed}). //
 \end{aligned}$$

example

$$\begin{aligned}
 f(x, y) &= \frac{e^{-x/y} e^{-y}}{y}, \quad x > 0, y > 0 \\
 \Rightarrow X &\not\perp Y.
 \end{aligned}$$

Want to compute $P(X > 1 | Y = y)$, $y > 0$.

$$\begin{aligned}
 f_Y(y) &= \int_0^{\infty} e^{-x/y} e^{-y} / y dx \\
 &= e^{-y} \cdot \left(-e^{-x/y} \Big|_0^{\infty} \right) \\
 &= e^{-y}, \quad y > 0.
 \end{aligned}$$

$$f_{X|Y}(x|y) = \frac{e^{-x/y} e^{-y}}{y} / e^{-y}$$

$$= \frac{1}{y} e^{-x/y}, \quad x > 0.$$

($y > 0$ acts as a parameter for this distribution).

$$\begin{aligned} P(X > 1 | Y = y) &= \int_1^{\infty} \frac{1}{y} e^{-x/y} dx \\ &= -e^{-x/y} \Big|_1^{\infty} = e^{-1/y}, \quad \underline{y > 0.} \\ &\qquad\qquad\qquad \text{fixed.} \end{aligned}$$

Note If $X \perp\!\!\!\perp Y$, jointly continuous

$$\begin{aligned} f_{X|Y}(x|y) &= \frac{f(x,y)}{f_Y(y)} = \frac{f_X(x) f_Y(y)}{f_Y(y)} \\ &= f_X(x). \end{aligned}$$

example Suppose $P \sim \text{Unif}[0,1]$

and $X|P=p \sim \text{Bin}(n, p)$.

Think of first selecting a number p for prob. of success uniformly over $[0,1]$.

Throw a coin n times with this p
= prob. of success.

Then $X = \# \text{ of successes among } n \text{ trials.}$

(X, P) neither jointly discrete nor
jointly continuous.

Marginal dist'n of X ?
 $0, \dots, n$ are values for X .

Fix $x \in \{0, \dots, n\}$

analogue of joint

$$\begin{aligned} P(X = x) &= \int_{-\infty}^{\infty} P(X = x | P = p) f_p(p) dp \\ &= \binom{n}{x} p^x (1-p)^{n-x} \cdot 1 dp \\ &= \binom{n}{x} \int_0^1 p^x (1-p)^{n-x} dp \quad \textcircled{=} \end{aligned}$$

Reminder:

$$a > -1, b > -1$$

$$\int_0^1 x^a (1-x)^b dx =$$

$$= \frac{\Gamma(a+1) \Gamma(b+1)}{\Gamma(a+b+2)}$$

If $a \in \mathbb{Z}_+$,

$$\Gamma(a+1) = a!$$

$$\text{If } x > 0 : \Gamma(x+1) = x \Gamma(x)$$

$$\begin{aligned} \textcircled{=} \binom{n}{x} &\frac{\Gamma(x+1) \Gamma(n-x+1)}{\Gamma(n+2)} \\ &= \binom{n}{x} \frac{x! (n-x)!}{(n+1)!} \\ &= \frac{n!}{\cancel{x!} \cancel{(n-x)!}} \frac{\cancel{x!} (n-x)!}{(n+1)!} = \\ &= \frac{1}{n+1} \quad , \quad 0 \leq x \leq n, \text{ integer} \end{aligned}$$

$\Rightarrow X \sim \text{Uniform on } \{0, 1, \dots, n\}$

Extending all notions to 3 or more random variables.

Def The joint cdf of n random variab-

Def The joint cdf of n random variables X_1, X_2, \dots, X_n is defined as a function $F: \mathbb{R}^n \rightarrow [0, 1]$ s.t.

$$F(x_1, \dots, x_n) = P(X_1 \leq x_1, \dots, X_n \leq x_n)$$

"and"

Def A collection of random variables X_1, X_2, \dots, X_n has jointly discrete distribution if (X_1, \dots, X_n) can take on at most countably many values in \mathbb{R}^n .

Def The joint pmf of jointly discrete r.v.'s X_1, X_2, \dots, X_n is $f: \mathbb{R}^n \rightarrow [0, 1]$

s.t. $f(x_1, \dots, x_n) = P(X_1 = x_1, \dots, X_n = x_n)$

$$A \subset \mathbb{R}^n$$

$$P((X_1, \dots, X_n) \in A) = \sum_{(x_1, x_2, \dots, x_n) \in A} f(x_1, \dots, x_n)$$

"n summations"

In vector notation

$$\underline{X} = (X_1, \dots, X_n), \quad \underline{x} = (x_1, \dots, x_n)$$

$$f(\underline{x}) = P(X = \underline{x}),$$

$$P(X \in A) = \sum_{\underline{x} \in A} f(\underline{x}), \quad A \subset \mathbb{R}^n.$$

Def A collection of random variables

X_1, \dots, X_n has jointly continuous dist'n if there exists a non-negative function $f : \mathbb{R}^n \rightarrow \mathbb{R}_+$ s.t.

$$P((X_1, \dots, X_n) \in A) = \int_{A} \dots \int f(x_1, \dots, x_n) dx_1 \dots dx_n$$

for all $A \subset \mathbb{R}^n$; say that f is the joint pdf (density) of X_1, X_2, \dots, X_n .

In vector notation,

$$P(X \in A) = \int_A f(\underline{x}) d\underline{x}, \quad A \subset \mathbb{R}^n$$

$$f(x_1, \dots, x_n) = \frac{\partial^n F(x_1, \dots, x_n)}{\partial x_1 \partial x_2 \dots \partial x_n}$$

- jointly continuous X_1, X_2, \dots, X_n .

Also, X_1, X_2, \dots, X_n are independent iff

$$\text{i) } f(x_1, \dots, x_n) = \prod_{i=1}^n f_i(x_i) \quad \left. \begin{array}{l} \text{pmf (discrete)} \\ \text{pdf (continuous)} \end{array} \right\} \text{marginal pmf/pdf for } X_i.$$

marginal pmf(pdf) for X_i .

2) $F(x_1, \dots, x_n) = \prod_{i=1}^n F_i(x_i)$ } for both cases.

3) Same principle applies when recognizing independence by looking at the joint pdf (pmf) as for d.r.v.'s. Need to factor joint fns into fns that involve one variable at a time.

$$(f(x_1, \dots, x_n) = \prod_{i=1}^n h_i(x_i), h_i \geq 0 \text{ and rectangular support})$$

Discrete case (jointly)

X_1, \dots, X_n

$$f_1(x_1) = \sum_{x_2} \sum_{x_3} \dots \sum_{x_n} f(x_1, \dots, x_n)$$

$$f_{12}(x_1, x_2) = \sum_{x_3} \dots \sum_{x_n} f(x_1, \dots, x_n)$$

Generally, $j \neq k$ n summation over variables

$$f_{jk}(x_j, x_k) = \sum_{\substack{x_i : i \neq j \\ \& i \neq k}} f(x_1, \dots, x_n).$$

More generally, $1 \leq r \leq n-1$,

$$f_{i_1, \dots, i_r}(x_{i_1}, \dots, x_{i_r}) = \sum_{\substack{x_i : i \notin \{i_1, \dots, i_r\}}} f(x_1, \dots, x_n)$$

summation over $n-r$ variables

Here $\{i_1, \dots, i_r\} \subset \{1, 2, \dots, n\}$.

Continuous case, jointly : X_1, \dots, X_n .

$$f_1(x_1) = \int_{-\infty}^{\infty} \dots \int_{-\infty}^{\infty} f(x_1, x_2, \dots, x_n) dx_2 \dots dx_n$$

Moreover, if $\{i_1, \dots, i_r\} \subset \{1, 2, \dots, n\}$

$$f_{i_1 \dots i_r}(x_{i_1}, \dots, x_{i_r}) = \int_{-\infty}^{\infty} \dots \int_{-\infty}^{\infty} f(x_1, \dots, x_n) dx_{j_1} \dots dx_{j_{n-r}}$$

$j_k \notin \{i_1, \dots, i_r\}$

In both jointly discrete & jointly continuous cases

$$F_1(x_1) = P(X_1 \leq x_1) = \\ = P(X_1 \leq x_1, X_2 < \infty, \dots, X_n < \infty)$$

$$= \lim_{\substack{x_2 \rightarrow \infty \\ \vdots \\ x_n \rightarrow \infty}} F(x_1, x_2, \dots, x_n)$$

More generally,

$$F_{\dots}(x_{i_1}, \dots, x_{i_r}) = \lim_{n \rightarrow \infty} F(x_1, \dots, x_n)$$

$$F_{i_1, \dots, i_r}(x_{i_1}, \dots, x_{i_r}) = \lim_{\substack{x_j \rightarrow \infty \\ \text{all } j \notin \{i_1, \dots, i_r\}}} F(x_1, \dots, x_n)$$

Def n random variables are said to be i.i.d. (independent identically distributed) if they are independent and all X_i 's have the same marginal distribution.

If X_1, X_2, \dots, X_n are jointly discrete (continuous) and iid then

$$f(x_1, \dots, x_n) = f_1(x_1) \dots f_n(x_n) \left\{ \begin{array}{l} \text{pmfs} \\ (\text{pdfs}) \end{array} \right.$$

here $f_1 \equiv f_2 \equiv \dots \equiv f_n \equiv f$.
(same marginals).

Def X_1, \dots, X_n is a random sample from a distribution with pdf (pmf) $f(x)$ if X_1, \dots, X_n are iid and have the common marginal with $f(x)$ as a pdf (pmf).

Continuous Case

$$X_1 | X_2 = x_2, \dots, X_n = x_n$$

pdf

$$f(x_1 | x_2, \dots, x_n) = \frac{f(x_1, \dots, x_n)}{f(x_2, \dots, x_n)}$$

More generally, if

\underline{X} is divided into α vectors

\underline{Y} and \underline{Z} , \underline{Y} - k-dim, \underline{Z} - $(n-k)$ -dim

$$f(\underline{y} | \underline{z}) = \frac{f(\underline{y}, \underline{z})}{f_z(\underline{z})}$$

example let X, Y, Z with joint pdf

$$f(x, y, z) = e^{-(x+y+z)}, x \geq 0, y \geq 0, z \geq 0$$

$$= e^{-x} \cdot \mathbb{1}_{\{x \geq 0\}} \cdot e^{-y} \cdot \mathbb{1}_{\{y \geq 0\}} \cdot e^{-z} \cdot \mathbb{1}_{\{z \geq 0\}}$$

$$F(x, y, z) = \int_{-\infty}^x \int_{-\infty}^y \int_{-\infty}^z f(a, b, c) da db dc$$

$$= \int_0^x \int_0^y \int_0^z e^{-(a+b+c)} da db dc$$

$$= (1 - e^{-x})(1 - e^{-y})(1 - e^{-z}),$$

$$x, y, z \geq 0.$$

Distribution of $U = X + Y$?

$$\begin{aligned} F_{X,Y}(x,y) &= \lim_{z \rightarrow \infty} F(x,y,z) = \\ &= (1 - e^{-x})(1 - e^{-y}), \quad \underbrace{x, y \geq 0.}_{\text{ }} \end{aligned}$$

$$U = X + Y, \quad u \geq 0.$$

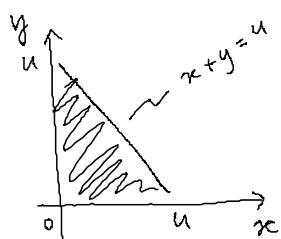
Fix $u \geq 0$.

$$\text{cdf } F_U(u) = P(U \leq u) =$$

$$= P(X + Y \leq u) \quad \text{①}$$

$$f_{X,Y}(x,y) = \frac{\partial^2 F_{X,Y}(x,y)}{\partial x \partial y} = e^{-x-y}, \quad x, y \geq 0.$$

$$\begin{aligned} \text{②} \quad & \iint_{x+y \leq u} e^{-x-y} dx dy = \\ &= \int_0^u \int_0^{u-x} e^{-x} e^{-y} dy dx \\ &= \int_0^u e^{-x} (-e^{-y}) \Big|_0^{u-x} dx = \\ &= \int_0^u e^{-x} (1 - e^{-(u-x)}) dx \\ &= -e^{-x} \Big|_0^u - e^{-u} \cdot u \\ &= 1 - e^{-u} - ue^{-u}, \quad u \geq 0. \end{aligned}$$



p.d.f. of $U = X + Y$

$$\frac{\partial F_u(u)}{\partial u} = e^{-u} - e^{-u} + ue^{-u}, \quad u \geq 0$$

$$= ue^{-u}, \quad u \geq 0.$$

Dist'n of $X | (Y, z) = ?$

Same as marginal X :

$$F_x(x) = \lim_{y \rightarrow \infty} F_{xy}(x, y) = 1 - e^{-x}, \quad x \geq 0.$$

Dist'n of $X | U = u$? $U = X + Y$

$$f_{X|U}(x|u) = \frac{f_{x,u}(x, u)}{f_u(u)}$$

Can we get $f_{x,u}$ from $f_{x,y}$? No

$$F_{x,u}(x, u) = P(X \leq x, U \leq u) \\ = P(X \leq x, X+Y \leq u) =$$

$$= \iint_{\substack{a \leq x \\ a+b \leq u}} f_{xy}(a, b) da db$$

= ... (calculus exercise)

$$= \begin{cases} 1 - e^{-x} - xe^{-u}, & 0 \leq x \leq u \\ 0, & \text{otherwise} \end{cases}$$

$$= \begin{cases} 1 - e^{-u} - xe^{-u}, & 0 \leq x \leq u \\ 1 - e^{-u} - ue^{-u}, & x > u \end{cases}$$

$$f_{X|U}(x|u) = \frac{\partial^2}{\partial x \partial u} F_{X|U}(x|u) = \\ = e^{-u}, \quad 0 \leq x \leq u$$

$$f_{X|U}(x|u) = \frac{f_{X|U}(x|u)}{f_u(u)} \\ = \frac{e^{-u}}{ue^{-u}} = \frac{1}{u}, \quad 0 \leq x \leq u$$

We can write

$$X|U=u \sim \text{Unif}[0, u].$$

Functions of Jointly Distributed Random Variables

r.v. $X : S \rightarrow \mathbb{R}$ & $f : \mathbb{R} \rightarrow \mathbb{R}$
 $\Rightarrow Y = f(X) : S \rightarrow \mathbb{R}$ is also
 a r.v.

Sums and convolutions.

but X, Y jointly discrete

$Z = X + Y$ and know $f(x, y)$

pmf $Z?$

$$f_Z(z) = P(X + Y = z)$$

$$= \sum_{x+y=z} \sum f(x, y)$$

$$= \sum_x \sum_{y=z-x} f(x, y)$$

$$= \underbrace{\sum_x f(x, z-x)}_{\text{pmf of } Z}$$

Suppose X, Y are jointly continuous

and $Z = X + Y$.

$$F_Z(z) = P(X + Y \leq z)$$

$$= \iint_{x+y \leq z} f(x, y) dx dy$$

$$= \int_{-\infty}^{\infty} \int_{y \leq z-x} f(x, y) dy dx \quad \textcircled{1}$$

Change of variables,

$$\text{let } v = y + x$$

$$y = v - x$$

$$dy = dv$$

$$\{y \leq z-x\} = \{v \leq z\}$$

$$\textcircled{1} = \int_{-\infty}^{\infty} \int_{v \leq z} f(x, v-x) dv dx$$

$$= \int_{-\infty}^{\infty} \int_{-\infty}^z f(x, v-x) dv dx$$

$$f_Z(z) = \int_{-\infty}^z \int_{-\infty}^{\infty} f(x, v-x) dx dv$$

cdf of $z \nearrow$

\Rightarrow pdf is $f_z(z) = \frac{\partial}{\partial z} F_z(z)$

$$= \int_{-\infty}^{\infty} f(x, z-x) dx$$

$$z = X+Y \quad \nearrow$$

If $X \perp Y$ and still jointly continuous

$$f(x, y) = f_x(x) f_y(y)$$

then for $z = X+Y$ the density

$$f_z(z) = \int_{-\infty}^{\infty} f_x(x) f_y(z-x) dx$$

f_z is the convolution of f_x and f_y .

General Case of Functions of Random Variables.

Suppose we know the dist'n of X .

We want to find the dist'n of

$$Y = g(X), \text{ a r.v. itself.}$$

$$\begin{aligned} \text{Then } F_Y(y) &= P(Y \leq y) = \\ &= P(g(X) \leq y). \end{aligned}$$

example $X \sim \text{Unif } [-1, 1]$

$$Y = X^2, \quad y \in [0, 1]$$

$$\begin{aligned}
 P(Y \leq y) &= P(X^2 \leq y) = \\
 &= P(-\sqrt{y} \leq X \leq \sqrt{y}) \\
 &= \int_{-\sqrt{y}}^{\sqrt{y}} \frac{1}{2} dx = \sqrt{y}, \quad 0 \leq y \leq 1.
 \end{aligned}$$

$$f_Y(y) = \frac{dF_Y(y)}{dy} = \frac{1}{2\sqrt{y}}, \quad 0 < y \leq 1.$$
