

$(X, Y)$  - a pair of real r.v.s.

Definition: A function  $\Psi$  defined on  $\overset{\text{support of } X}{S_X} \times \mathcal{B}$  satisfying:

(a)  $\forall x \in S_X$ ,  $\Psi(x, \cdot)$  is a probability on  $\mathcal{B}$ .

(b)  $\forall B \in \mathcal{B}$ ,  $P(X \in A, Y \in B) = E(\Psi(X, B) \cdot \mathbf{1}_A(X))$

$\forall$  borel sets  $A$ .

is called the

conditional distribution of  $Y$ , given  $X$ .

More specifically,  $\Psi(x, B) = P(Y \in B | X = x)$ .

Result: Such a  $\Psi$  always exists.

(can't prove this now. Out of scope).

Two special cases where we have explicit formula for such a  $\Psi$ :

Case-I:  $X$  is a discrete r.v. with values in the countable set  $D_X$ .

For every  $x \in D_X$  &  $B \in \mathcal{B}$ ,

define  $\Psi(x, B) = P(Y \in B | X = x)$

$$= \frac{P(Y \in B, X = x)}{P(X = x)}.$$

[Exercise:  
Check that  $\Psi$  satisfies (a) & (b).]

Case-II:  $(X, Y)$  has a joint density.

Let  $f$  represent the joint density, and

$f_X$  be the (marginal) density of  $X$ .  $\left[ f_X(x) = \int f(x, y) dy \right]$

Let  $g(y|x) = \underline{f(x, y)}$  for  $x$  s.t.  $f(x) > 0$ ,

Let  $g(y|x) = \frac{f(x,y)}{f_x(x)}$  for  $x$  s.t.  $f_x(x) > 0$ ,  
 $g(\cdot|x)$  is a joint density.

$$\text{Define } \Psi(x, B) = \int_B g(y|x) dy$$

Here, property (a) is trivially satisfied, as  
 $\Psi(x, B)$  is just the integral of a density over  $B$ .

Exc.: check that it follows (b).

Example:  $(x, y)$  has joint density.

$$f(x, y) = 2\lambda x y^{-2} e^{-\lambda y}, \quad 0 < x < y < \infty.$$

first, let's verify that this is a joint density.

$$\begin{aligned} \iint f(x, y) dx dy &= \int_{y=0}^{\infty} \int_{x=0}^y 2\lambda x y^{-2} e^{-\lambda y} dx dy = \int_{y=0}^{\infty} \lambda y^{-2} x^2 \Big|_0^y dy \\ &= \int_{y=0}^{\infty} \lambda e^{-\lambda y} dy = 1 \quad \checkmark \end{aligned}$$

Now, • conditional distribution of  $X|Y$ :

$$F_Y(y) = \int_{x=0}^y f(x, y) dx = \lambda e^{-\lambda y}$$

$$\therefore g(x|y) = \frac{f(x,y)}{f_x(x)} = \frac{2\lambda x y^{-2} e^{-\lambda y}}{\lambda e^{-\lambda y}} = \frac{2x}{y^2}$$

• Conditional expectation of  $E(X^2|Y=y) = \int_0^y x^2 \cdot g(x|y) dx$

$$= \int_0^y x^2 \cdot \frac{2x}{y^2} dx$$

$$= \frac{1}{y^2} \cdot \frac{x^4}{2} \Big|_0^y$$

$$= \frac{y^2}{2}$$

• Conditional distribution of  $Y|X$ :

Example:

$U, V$  are independent  $\text{Exp}(\lambda)$ .

Find conditional distribution of  $U-V$ , given  $U+V$ .

$$X = U+V$$

$$Y = U-V$$

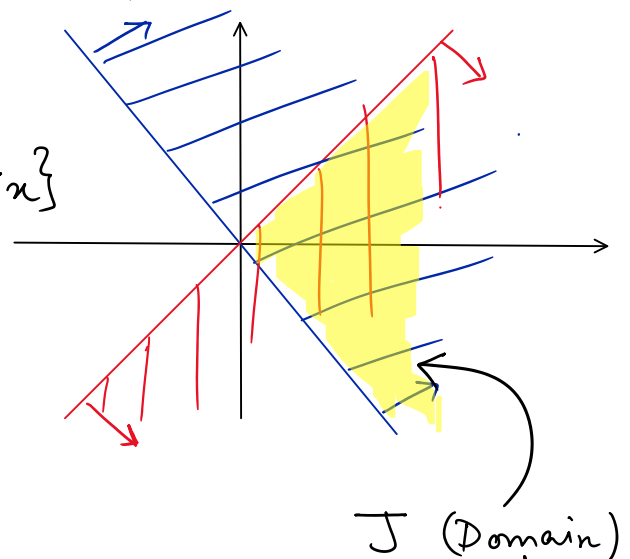
$(U, V)$  takes values on

$$I = (0, \infty) \times (0, \infty)$$

$$\therefore U = \frac{X+Y}{2}, \quad V = \frac{X-Y}{2} \quad \therefore X+Y > 0, \quad X-Y > 0.$$

$$(u, v) \mapsto (x, y)$$

$$I \rightarrow J = \{(x, y) : -x < y < x\}$$



$$\therefore u = \frac{x+y}{2}, \quad v = \frac{x-y}{2}$$

J (Domain)  
obtained,  
after applying all  
constraints.

$\therefore$  The Jacobian,

$$J = \left| \det \begin{pmatrix} \frac{\partial u}{\partial x} & \frac{\partial u}{\partial y} \\ \frac{\partial v}{\partial x} & \frac{\partial v}{\partial y} \end{pmatrix} \right| = \left| \det \begin{pmatrix} \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & -\frac{1}{2} \end{pmatrix} \right| = \left( \frac{1}{2} \right)$$

$$\begin{aligned} f_{u,v}(u,v) &= f_u(u) \times f_v(v) \\ &= \lambda e^{-\lambda u} \times \lambda e^{-\lambda v} \\ &= \lambda^2 e^{-\lambda(u+v)} \end{aligned}$$

$$\therefore f_{x,y}(x,y) = \frac{1}{2} \lambda^2 e^{-\lambda x}$$

• Marginal density of X:  $f_X(x) = \int_{y=-x}^x f_{x,y}(x,y) dy = \lambda^2 x e^{-\lambda x}$

$\therefore X \sim \text{Gamma}(\lambda, 2)$

• Conditional density of Y|X:  $g(y|x) = \frac{f_{x,y}(x,y)}{f_X(x)} = \frac{1}{2x}, \quad -x < y < x$

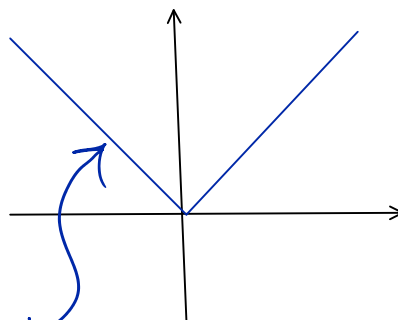
Example:

$$X \sim N(0,1).$$

$$Y = |X|.$$

Find the

Here,  
the  
dist<sup>n</sup>  
 $Y=|X|$   
lies  
only on  
this  $y=|x|$



Find the conditional distribution of  $X$ , given  $Y$ .

this  $y=|x|$  ✓

line (1D).

∴ This is a continuous dist<sup>n</sup>,

∴ mass over lower dimension = 0.

∴  $X$  &  $Y$  have **no** joint density.

$$\Psi(y, B) = \frac{1}{2} \delta_{\{y\}}(B) + \frac{1}{2} \delta_{\{-y\}}(B)$$

↑  
Dirac mass  
at  $y$

↑  
Dirac mass  
at  $-y$ .

( if  $Y = |X| = 5$ ,

$X = 5$  or  $-5$ , with

equal probability =  $\frac{1}{2}$

$$*\mathbb{1}_{\{y \in B\}} = \delta_y(B)$$

is,  $P(Y=y) = 1$ .

So, if  $y \in B$ ,  $\delta_y(B) = 1$

$y \notin B$ ,  $\delta_y(B) = 0$ .

why? Since,  $N(0,1)$  is symmetric about  $y$ -axis.

Example:

$U, V$  - independent Uniform  $(0,1)$

$X = \max\{U, V\}$ .

$Y = U$

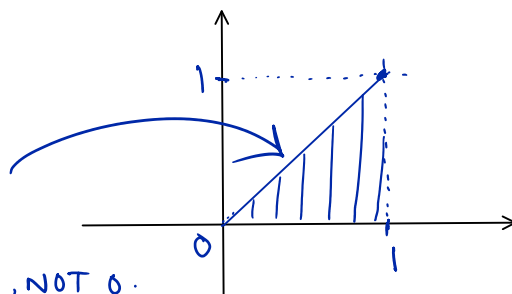
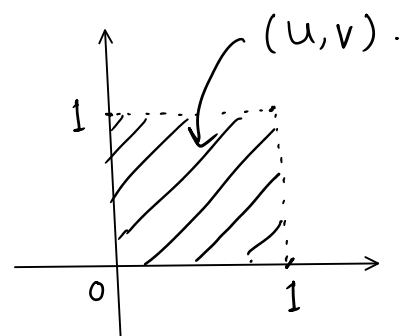
clearly,  $U \leq \max\{U, V\}$ .

i.e,  $Y \leq X$ .

$$P(X=Y) = \frac{1}{2}$$

Here, this line (with 0 area) has positive prob. mass, NOT 0.

So,  $(X, Y)$  cannot have a joint density.



a joint density.

Still, we want the conditional distribution of  $Y$ , given  $X$ .  
For  $x \in (0, 1)$ .

$$\Psi(x, (0, a)) = P(Y \leq a, X = x)$$

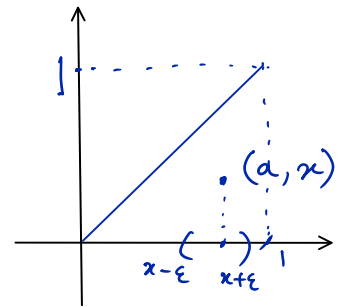
(intuitively)  $= \begin{cases} \frac{a}{2x} & , a < x \\ 1 & , a \geq x. \end{cases} \quad (*)$

↓  
we think of the  
distribution of  
Unif(0,1).

Let's check this result formally.

Fix  $0 < x < 1$ , &  $0 < a < x$ .

$$P(Y \leq a | X = x) = P(Y \leq a | X \in (x-\varepsilon, x+\varepsilon))$$



Can't calculate  
wrt a point.

Hence, we take a

small interval around  $X = x$ ,  $\therefore$  that small interval has +ve mass.

$$= \frac{P(Y \leq a, X \in (x-\varepsilon, x+\varepsilon))}{P(X \in (x-\varepsilon, x+\varepsilon))}$$

( $\because u, v$   
ind.)  $\rightarrow$

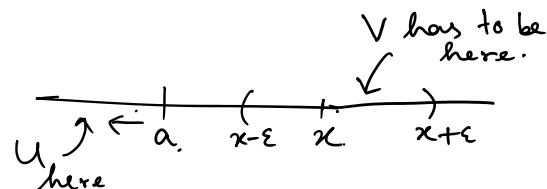
$$= \frac{P(U \leq a) \cdot P(V \in (x-\varepsilon, x+\varepsilon))}{(x+\varepsilon)^2 - (x-\varepsilon)^2}$$

$$\begin{aligned} P(X \leq x) &= P(\max\{U, V\} \leq x) \\ &= P(U \leq x, V \leq x) \\ &= P(U \leq x) \cdot P(V \leq x) \\ &= x \times x = x^2 \end{aligned}$$

$$\begin{aligned} \therefore P(X \leq x) &= x^2 \\ \therefore P(X \in (x-\varepsilon, x+\varepsilon)) &= P(X < x+\varepsilon) - P(X \leq x-\varepsilon) \end{aligned}$$

$$= \frac{a \cdot (x+\varepsilon - (x-\varepsilon))}{4x\varepsilon}$$

$$= \frac{2a\varepsilon}{4x\varepsilon} = \frac{a}{2x}$$



$\therefore$  Our initial (\*) intuition was wrong. We have  $\frac{1}{2}$  mass

.. Our initial (\*) intuition was wrong. We have  $\frac{1}{2}$  mass over  $y=x$  line, & remaining half mass is distributed over the region.

So, it's  $\frac{a}{2x}$ , not  $\frac{a}{x}$ .