

Math 3235 Probability Theory  
3/16/23

$X$   $Y$

$$u(X, Y) = U$$

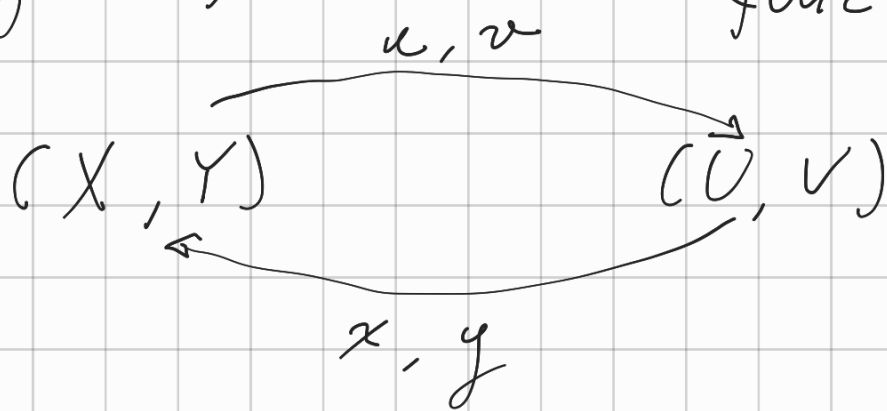
$$v(X, Y) = V$$

$f_{X,Y}(x,y)$  joint p.d.f. of  $X$  and  $Y$ . Find the joint p.d.f. of  $U$  and  $V$ . (Chapter 6.5 of the Textbook).

$$X = x(U, V)$$

$$Y = y(U, V)$$

Inverse function



For example:

$$u = x + y$$

$$v = \frac{x}{x + y}$$

in Term of random variables

$$U = X + Y$$

$$V = \frac{X}{X + Y}$$

$\Downarrow$

$$X = UV$$

$$Y = U(1 - V)$$

$$x = uv$$

$$y = u(1 - v)$$

$$f(u, v) = \begin{pmatrix} \frac{\partial x(u, v)}{\partial u} & \frac{\partial x(u, v)}{\partial v} \\ \frac{\partial y(u, v)}{\partial u} & \frac{\partial y(u, v)}{\partial v} \end{pmatrix}$$

Jacobian  
matrix

$f(u, v)$  is a  $2 \times 2$  matrix  
whose entries are functions of  
 $u, v$ .

$$x = uv$$

$$y = u(1-v)$$

$$\frac{\partial x}{\partial u} = v$$

$$\frac{\partial x}{\partial v} = u$$

$$\frac{\partial y}{\partial u} = (1-v)$$

$$\frac{\partial y}{\partial v} = -u$$

$$J(u, v) = \begin{pmatrix} v & u \\ 1-v & -u \end{pmatrix}$$

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o

$$J(u, v) = \left| \det J(u, v) \right|$$

Reminder:

$$\text{if } A = \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix}$$

$$\det A = a_{11} a_{22} - a_{21} a_{12}$$

$$J(u, v) = \begin{vmatrix} \frac{\partial x}{\partial u} & \frac{\partial y}{\partial v} \\ \frac{\partial x}{\partial v} & \frac{\partial y}{\partial u} \end{vmatrix}$$

In our example we get

$$J(u, v) = \begin{vmatrix} -uv & -u(1-v) \end{vmatrix} = |u|$$

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Theorem:

$X, Y$  are cont. r.v. with joint p.d.f.  $f_{X,Y}(x, y)$ .

$T$  is a mapping from  $\mathbb{R}^2 \rightarrow \mathbb{R}^2$

$T(x, y) = (u(x, y), v(x, y))$  and

$T$  is bijective. Then  $I$

can write  $T^{-1}(u, v) = (u(x, y), v(x, y))$

and

$$U = u(X, Y) \quad V = v(X, Y)$$

$$f_{u,v}(u, v) = f_{x,y}(x(u, v), y(u, v)) |J(u, v)|$$

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Assume that  $X$  and  $Y$  have joint p. d. f. given by

$$f_{X,Y}(x, y) = \begin{cases} e^{-x-y} & x, y > 0 \\ 0 & \text{otherwise} \end{cases}$$

$$U = X + Y$$

$$V = \frac{X}{X + Y}$$

Observe that

$$f_X(x) = \int_0^{\infty} e^{-x-y} dy = e^{-x}$$

$$f_Y(y) = \int_0^{\infty} e^{-x-y} dx = e^{-y}$$

$X$  and  $Y$  are indep. exp with  
par  $\lambda$ .

$$e^{-x-y} = \int_{X,Y} f(x,y) = \int_X f(x) \int_Y f(y)$$

$$U = X+Y \quad V = \frac{X}{X+Y}$$

$$J(u,v) = u$$

$$f_{X,Y}(x(u,v), y(u,v)) = e^{-u}$$

$$f_{U,V}(u,v) = u e^{-u}$$

$u = x+y$  since both  $x$   
and  $y$  are positive  
Then  $x+y \geq 0$ .

$$0 \leq \frac{x}{x+y} = v \leq 1$$

Finally The joint p.d.f of  $U$  and  $V$  is

$$f_{U,V}(u,v) = \begin{cases} u e^{-u} & u > 0, 0 < v < 1 \\ 0 & \text{otherwise} \end{cases}$$

Marginals:

$$f_U(u) = \int_0^1 u e^{-u} dv = u e^{-u} \quad u > 0$$

$$f_V(v) = \int_0^{\infty} u e^{-u} du = 1 \quad 0 \leq v \leq 1$$

$U$  is Gamma  $(2, 1)$  while

$V$  is uniform in  $[0, 1]$ .

Moreover  $U$  and  $V$  are

independent.

Idea of the proof of the Theorem.

$B \in \mathbb{R}^2$  and we want to

know

$$P((u, v) \in B) =$$

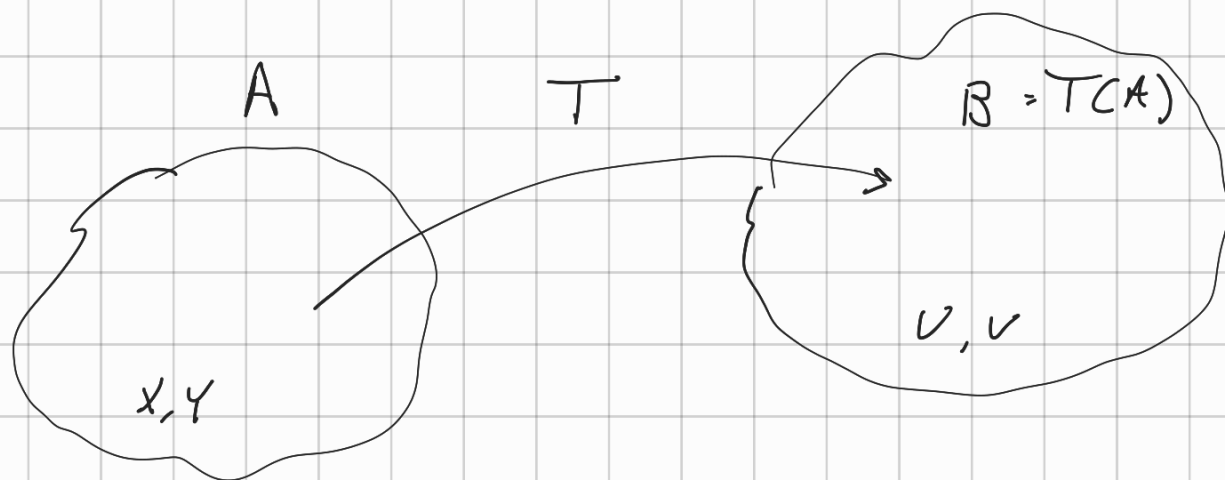
$$\iint_B f_{u,v}(u, v) \, du \, dv$$

If  $B = T(A)$  that is

$$A = T^{-1}(B)$$

it follows that

$$P((u, v) \in B) = P((x, y) \in A)$$



$$P((x, y) \in A) = \iint_A f_{x,y}(x, y) \, dx \, dy =$$



(change of variable  $u = u(x, y)$   
 $v = v(x, y)$ )

$$= \iint_{B=T^{-1}(A)} f_{X,Y}(x(u,v), y(u,v)) J(u,v) du dv$$

For every subset  $B \subset \mathbb{R}^2$  we  
have

$$\iint_B f_{u,v}(u,v) du dv = \iint_B f_{X,Y}(x(u,v), y(u,v)) J(u,v) du dv$$

$\Downarrow$

$$f_{u,v}(u,v) = f_{X,Y}(x(u,v), y(u,v)) J(u,v)$$

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Example

X Y

$$f_{X,Y}(x,y) = \begin{cases} e^{-y} & y > x > 0 \\ 0 & \text{otherwise} \end{cases}$$

$$f_X(x) = \int_x^{\infty} e^{-y} dy = -e^{-y} \Big|_x^{\infty} = e^{-x}$$

$$f_Y(y) = \int_0^y e^{-y} dx = y e^{-y}$$

$$U = X \quad V = Y - X$$

$$X = U \quad Y = U + V$$

$$\begin{aligned} f_{U,V}(u,v) &= e^{-y(u,v)} J(u,v) \\ &= e^{-(u+v)} J(u,v) \end{aligned}$$

$$J(u,v) = \begin{pmatrix} 1 & 0 \\ 1 & 1 \end{pmatrix}$$

$$J(u,v) = 1$$

$$f_{U,V}(u,v) = \begin{cases} e^{-u-v} & u > 0 \quad v > 0 \\ 0 & \text{otherwise} \end{cases}$$

Expected Values of  $g$ . dist. r.v.

If  $X, Y$  are continuous r.v.

and  $g: \mathbb{R}^2 \rightarrow \mathbb{R}$ , we can

consider

$$Z = g(X, Y)$$

$Z$  is a new continuous r.v.

Theorem:

$$\mathbb{E}(Z) = \mathbb{E}(g(X, Y)) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(x, y) f_{X, Y}(x, y) dx dy$$

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$$g(X, Y) = X$$

$$\mathbb{E}(X) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} x f_{X, Y}(x, y) dx dy =$$

$$= \int_{-\infty}^{\infty} x \left( \int_{-\infty}^{\infty} f_{X, Y}(x, y) dy \right) dx =$$

$$= \int_{-\infty}^{\infty} x f_X(x) dx$$

$$E(Y) = \int_{-\infty}^{\infty} y f_Y(y) dy$$

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$$E(aX + bY) =$$

$$\iint (ax + by) f_{X,Y}(x,y) dx dy =$$

$$a \iint x f_{X,Y}(x,y) dx dy +$$

$$b \iint y f_{X,Y}(x,y) dx dy =$$

$$a E(X) + b E(Y)$$

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What about The product  
if  $X$  and  $Y$  are independ.

$$\begin{aligned} E(XY) &= \iint xy f_{X,Y}(x,y) dx dy = \\ &= \iint xy f_X(x) f_Y(y) dx dy = \end{aligned}$$

$$= \int_{-\infty}^{\infty} y f_Y(y) dy \int_{-\infty}^{\infty} x f_X(x) dx$$

$$= E(X) E(Y)$$

$$\text{If } X \perp\!\!\!\perp Y \Rightarrow E(XY) = E(X) E(Y)$$

$$\Leftarrow$$

We define

$$\text{cov}(X, Y) = E(XY) - E(X) E(Y)$$

$$\text{corr}(X, Y) = \rho_{X, Y} = \frac{\text{cov}(X, Y)}{\sqrt{\text{var}(X) \text{var}(Y)}}$$

$$\text{cov}(aX + b, cY + d) =$$

$$ac \text{cov}(X, Y)$$

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$$E(XY) = E(X) E(Y) \not\Rightarrow X \perp\!\!\!\perp Y$$

Theorem: if for every functions  
 $g: \mathbb{R} \rightarrow \mathbb{R}$  and  $h: \mathbb{R} \rightarrow \mathbb{R}$

$$\mathbb{E}(g(X)h(Y)) = \mathbb{E}(g(X))\mathbb{E}(h(Y))$$

Then  $X$  and  $Y$  are independent.

Proof:

$$g(x) = \begin{cases} 1 & x < a \\ 0 & \text{otherwise} \end{cases}$$

$$h(y) = \begin{cases} 1 & y < b \\ 0 & \text{otherwise} \end{cases}$$

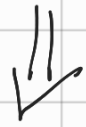
$$\mathbb{E}(g(X)h(Y)) =$$

$$\int_{-\infty}^a \int_{-\infty}^b f_{X,Y}(x,y) dx dy =$$

$$\mathbb{P}(X < a \text{ \& } Y < b)$$

$$E(g(x)) = P(X \leq a)$$

$$E(h(Y)) = P(Y \leq b)$$



$$P(X \leq a \text{ \& } Y \leq b) =$$

$$P(X \leq a) P(Y \leq b)$$

$$F(a, b) = F(a) F(b).$$