

Stat 2911 Lecture Notes

Class 22, 2016

Uri Keich

© Uri Keich, The University of
Sydney

Examples of functions of RVs
(Rice 2.3), Joint distribution
function/density, Marginal density
(Rice 3.1, 3.3)

Examples 1) $U \sim U(0,1)$ and $V = \frac{1}{U}$, find f_V .

$V = g(U)$ where $g(u) = \frac{1}{u}$ is smooth and strictly \downarrow on $(0,1) = U(\Omega)$.

Therefore, by the theorem $f_V(v) = f_U(g^{-1}(v)) |g'(v)|$.

$U = \frac{1}{u} \Rightarrow u = \frac{1}{v}$ so $g^{-1}(v) = \frac{1}{v}$ and

$$f_V(v) = f_U\left(\frac{1}{v}\right) \left| -\frac{1}{v^2} \right| = \begin{cases} \frac{1}{v^2} & v > 1 \\ 0 & v \leq 1 \end{cases} .$$

Alternatively, for $v > 1$,

$$\begin{aligned} F_V(v) &= P(V \leq v) \\ &= P\left(\frac{1}{U} \leq v\right) \quad (\text{if } v > 0) \\ &= 1 - P(U < \frac{1}{v}) \\ &= 1 - \frac{1}{v} . \end{aligned}$$

$$\begin{aligned} \Rightarrow f_V(v) &= \frac{d}{dv} \left(1 - \frac{1}{v} \right) \\ &= \frac{1}{v^2} \end{aligned}$$

2) $Z \sim N(0,1)$, $Y = Z^2 = g(Z)$ where $g(z) = z^2$.

$$\Rightarrow g'(y) = \sqrt{y}$$

$$\begin{aligned} \Rightarrow f_Y(y) &= f_Z(\sqrt{y}) \left| \frac{1}{2\sqrt{y}} \right| \\ &= \frac{1}{\sqrt{2\pi}} e^{-y/2} \cdot \frac{1}{2\sqrt{y}} \\ &= \frac{1}{2} \frac{1}{\sqrt{2\pi y}} e^{-y/2}. \end{aligned}$$

Sanity check:

$$\begin{aligned} \int_{-\infty}^{\infty} f_Y(y) dy &= \frac{1}{2} \int_0^{\infty} \frac{1}{\sqrt{2\pi y}} e^{-y/2} dy \quad \left[\begin{array}{l} \sqrt{y} \neq \\ \frac{dy}{2\sqrt{y}} = dt \end{array} \right] \\ &= \int_0^{\infty} \frac{1}{\sqrt{2\pi}} e^{-t^2/2} dt \\ &= \frac{1}{2} \end{aligned}$$

What has gone wrong??

$g(z) = z^2$ is not strictly monotone on $\mathbb{R} = \mathbb{Z}(\Omega)$.

$$z \sim N(0,1) \quad Y = Z^2$$

Instead of using the formula we compute it directly.

$$\begin{aligned} \text{For } y > 0: \quad F_Y(y) &= P(Y \leq y) \\ &= P(Z^2 \leq y) \\ &= P(-\sqrt{y} \leq Z \leq \sqrt{y}) \\ &= P(-\sqrt{y} < Z \leq \sqrt{y}) \\ &= \Phi(\sqrt{y}) - \Phi(-\sqrt{y}), \end{aligned}$$

$$\text{where } \Phi(z) = \int_{-\infty}^z \frac{1}{\sqrt{2\pi}} e^{-t^2/2} dt.$$

$\Rightarrow F_Y$ is diff. and by chain rule, for $y > 0$

$$\begin{aligned} f_Y(y) &= \Phi'(\sqrt{y}) \frac{1}{2\sqrt{y}} - \Phi'(-\sqrt{y}) \left(-\frac{1}{2\sqrt{y}}\right) \\ &= \frac{1}{\sqrt{2\pi}} e^{-y/2} \cdot \frac{1}{2\sqrt{y}} + \frac{1}{\sqrt{2\pi}} e^{-y/2} \frac{1}{2\sqrt{y}} \\ &= \frac{(1/2)^{1/2}}{\sqrt{\pi}} y^{1/2-1} e^{-y/2}, \end{aligned}$$

Recall the $\Gamma(\alpha, \lambda)$ density: $\frac{\lambda^\alpha}{\Gamma(\alpha)} x^{\alpha-1} e^{-\lambda x} 1_{x>0}$

$\Rightarrow Y \sim \chi^2 \sim \Gamma(1/2, 1/2)$ aka X_1^2 .

$$\Rightarrow \Gamma(1/2) = \sqrt{\pi}.$$

Joint Distributions (discrete analogue: joint part)

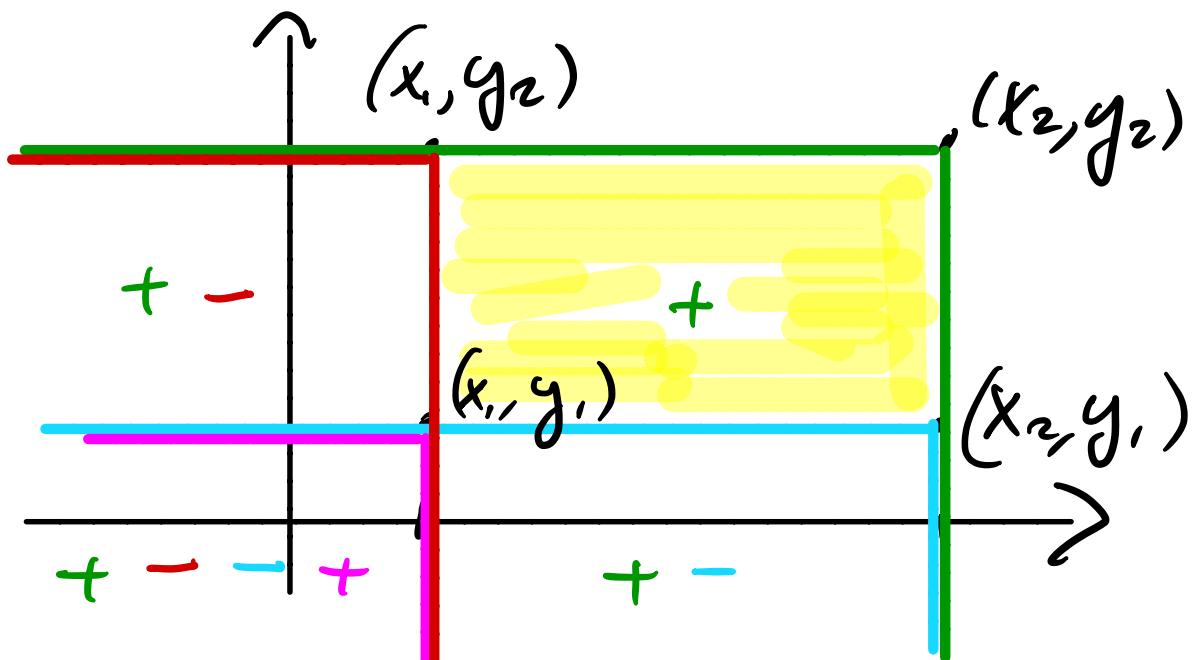
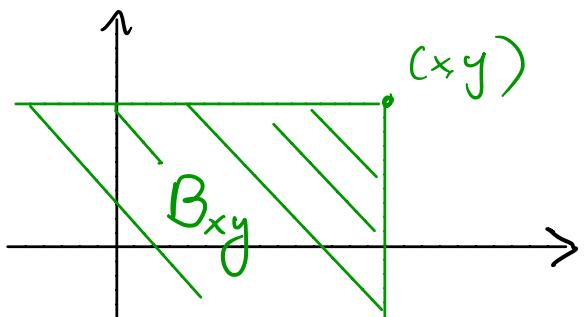
The joint dist. of two RVs (a random point in \mathbb{R}^2) is determined by the joint dist. function:

$$F_{xy}(x, y) = P(\bar{X} \leq x, Y \leq y)$$

$$\mathcal{B}_{xy} = \{(s, t) : s \leq x, t \leq y\}$$

$$F_{xy}(x, y) = P((\bar{X}, Y) \in \mathcal{B}_{xy})$$

F_{xy} specifies the probability that $(\bar{X}, Y) \in [x_1, x_2] \times [y_1, y_2]$:



$$P((\bar{X}, Y) \in [x_1, x_2] \times [y_1, y_2]) = F(x_2, y_2) - F(x_2, y_1) - F(x_1, y_2) + F(x_1, y_1)$$

It follows (with effort) that for any Borel measurable $B \subset \mathbb{R}^2$, $P((\bar{X}, Y) \in B)$ is specified by $F_{xy} = F$.

Def. X and Y have a **joint density** function if
 $\exists f: \mathbb{R}^2 \rightarrow \mathbb{R}^+$ s.t. $f(x,y) \in \mathbb{R}^+$:

$$F_{X,Y}(x,y) = \int_{-\infty}^x \int_{-\infty}^y f(s,t) dt ds = \iint_{B_{xy}} f(s,t) ds dt$$

If X and Y have a joint density then they are **jointly continuous**.

Comments.

(i) If $f: \mathbb{R}^2 \rightarrow \mathbb{R}^+$ is integrable with $\iint_{\mathbb{R}^2} f = 1$,
then $F(x,y) := \int_{-\infty}^x \int_{-\infty}^y f(s,t) dt ds$ is the joint CDF
of jointly cont. RVs with the joint pdf f .

Example. $f(x,y) = \lambda^2 e^{-\lambda(x+y)} 1_{x>0} 1_{y>0}$ ($\lambda > 0$).
 $\iint_{-\infty}^{\infty} f(x,y) dx dy = \int_0^{\infty} \lambda e^{-\lambda x} dx \int_0^{\infty} \lambda e^{-\lambda y} dy$
 $= 1$

$\Rightarrow f$ is a joint pdf of RVs with joint CDF

$$\begin{aligned} F(x,y) &= \int_{-\infty}^x \int_{-\infty}^y f(s,t) dt ds & x, y > 0 \\ &= \int_0^x \lambda e^{-\lambda s} ds \int_0^y \lambda e^{-\lambda t} dt \\ &= (1 - e^{-\lambda x})(1 - e^{-\lambda y}) \end{aligned}$$

(ii) By Tonelli's Theorem iterated/repeated integral of a non-negative function, is the same as the double integral:

$$\int_{-\infty}^x \int_{-\infty}^y f(s,t) dt ds = \iint_{B_{xy}} f(s,t) ds dt = \int_{-\infty}^y \int_{-\infty}^x f(s,t) ds dt$$

Fubini's extension: if for any integrable f one of the above three integrals is finite, when f is replaced by $|f|$, then the equalities still hold.

(iii) From $P(\xi, y) \in B_{xy}) = F_{\xi y}(x, y) = \iint_{B_{xy}} f ds dt \quad \forall (x, y) \in \mathbb{R}^2$

it follows that for any (Borel) set $B \subset \mathbb{R}^2$,

$$P(\xi, y) \in B) = \iint_B f ds dt .$$

(iv) By the fundamental theorem of calculus, $\forall (x, y) \in \mathbb{R}^2$ where f is cont. at (x, y) :

$$f(x, y) = \frac{\partial^2}{\partial x \partial y} F_{\xi y}|_{(x,y)} = \frac{\partial^2}{\partial y \partial x} F_{\xi y}|_{(x,y)}$$

(v) ξ and y can be cont. RVs but not jointly cont. (exercise)

(vi) If \underline{X} is a cont. RV with pdf $f_{\underline{X}}$ then if $f_{\underline{X}}$ is cont. at x and $\delta > 0$ is small:

$$\begin{aligned} P(\underline{X} \in (x - \delta/2, x + \delta/2)) &= \int_{x - \delta/2}^{x + \delta/2} f_{\underline{X}}(s) ds \\ &\approx f_{\underline{X}}(x) \cdot \delta. \end{aligned}$$

Similarly, if (\underline{X}, y) are jointly cont. with pdf $f_{\underline{X}y}$, then if $f_{\underline{X}y}$ is cont. at (x, y) and $\delta, \varepsilon > 0$ are small, then

$$\begin{aligned} P[(\underline{X}, y) \in \underbrace{(x - \delta/2, x + \delta/2) \times (y - \varepsilon/2, y + \varepsilon/2)}_{R_{xy}}] &= \iint_{R_{xy}} f_{\underline{X}y} ds dt \\ &\approx f_{\underline{X}y}(x, y) |R_{xy}| \\ &= f_{\underline{X}y}(x, y) \cdot \delta \cdot \varepsilon \end{aligned}$$

Marginal distribution

Recall: $P_{\underline{X}}(x) = \sum_y P_{\underline{X}Y}(x,y)$

Claim. If \underline{X} and y have a joint pdf $f_{\underline{X}Y}$, then \underline{X} is a cont. RV and

$$f_{\underline{X}}(x) = \int_{\mathbb{R}} f_{\underline{X}Y}(x,y) dy$$

marginal density
(Same for f_Y)

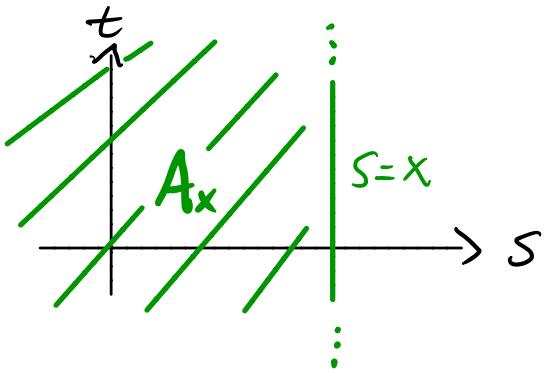
Proof.

$$F_{\underline{X}}(x) = P(\underline{X} \leq x)$$

$$= P(\underline{X} \leq x, y < \infty)$$

$$= P((\underline{X}, Y) \in A_x),$$

where $A_x = \{(s,t) \in \mathbb{R}^2 : s \leq x\}$:



$$\Rightarrow F_{\underline{X}}(x) = \iint_{A_x} f_{\underline{X}Y}(s,t) ds dt \quad (\text{double integral})$$

$$= \underbrace{\int_{-\infty}^x \int_{-\infty}^{\infty} f_{\underline{X}Y}(s,t) dt ds}_{g(s)} \quad (\text{Tonelli: repeated integral})$$

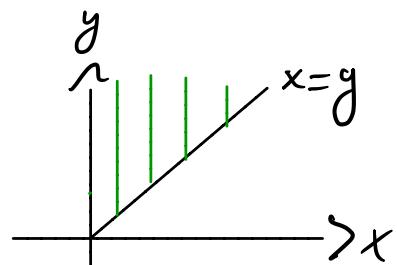
$$\Rightarrow F_{\underline{X}}(x) = \int_{-\infty}^x g(s) ds$$

$\Rightarrow \underline{X}$ is a cont. RV with pdf $f_{\underline{X}} = g$.

□

Examples

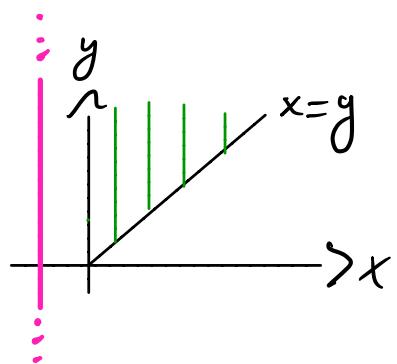
1) $f_{\bar{X}Y}(x,y) = \lambda^2 e^{-\lambda y}$ $y > x \geq 0$
 $\lambda > 0$



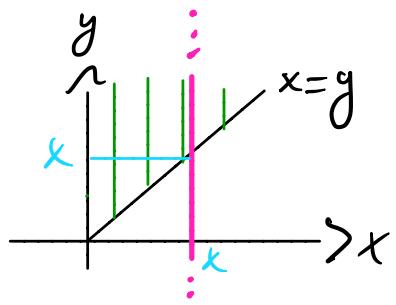
$$f_{\bar{X}}(x) = \int_{-\infty}^{\infty} f_{\bar{X}Y}(x,y) dy$$

Be particularly careful when plugging in the formula!

If $x < 0$, clearly $f_{\bar{X}}(x) = 0$:



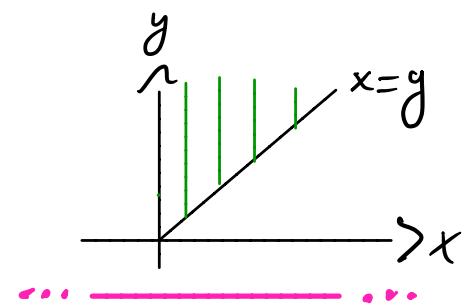
If $x \geq 0$ we have:



$$\begin{aligned} f_{\bar{X}}(x) &= \int_x^{\infty} \lambda^2 e^{-\lambda y} dy \\ &= -\lambda e^{-\lambda y} \Big|_x^{\infty} \\ &= \lambda e^{-\lambda x}. \end{aligned}$$

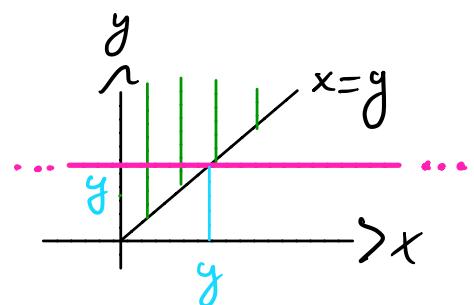
$\Rightarrow \bar{X} \sim \exp(\lambda)$ (not obvious from $f_{\bar{X}Y}$!)

$$f_y(y) = \int_{-\infty}^{\infty} f_{xy}(x,y) dx$$



If $y \leq 0$, clearly $f_y(y) = 0$:

If $y > 0$,



so

$$\begin{aligned} f_y(y) &= \int_0^y \lambda^2 e^{-\lambda y} dx \\ &= y \cdot \lambda^2 e^{-\lambda y} \end{aligned}$$

$\Rightarrow y \sim \Gamma(\alpha=2, \lambda)$

© Uri Keich, The University of Sydney