

Stat 2911 Lecture Notes

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Beta distribution, Density of a
quotient, Distribution of a function
of a random point (Rice 2.2, 3.6)

Let $Z = \bar{X} + Y$ where \bar{X} and Y have joint pdf $f_{\bar{X}Y}$.
 Z is a cont. RV with density

$$f_Z(z) = \int_{-\infty}^{\infty} f_{\bar{X}Y}(x, z-x) dx = \int_{-\infty}^{\infty} \underset{\substack{\uparrow \\ \text{ex. change of variable}}}{f_{\bar{X}Y}(z-y, y)} dy.$$

If \bar{X} and Y are ind. then

$$f_Z(z) = \int_{-\infty}^{\infty} f_{\bar{X}}(x) f_Y(z-x) dx, \text{ or}$$

$$f_Z = f_{\bar{X}} * f_Y .$$

Example. If $\bar{X} \sim \Gamma(\alpha, \lambda)$ ind. of $Y \sim \Gamma(\beta, \lambda)$ then

$$\bar{X} + Y \sim \Gamma(\alpha + \beta, \lambda)$$

Cor. 1

$$\beta(a, b) = \frac{\Gamma(a)\Gamma(b)}{\Gamma(a+b)}$$

Cor. 2 If $\bar{X}_1, \dots, \bar{X}_n$ are iid $\exp(\lambda)$ RVs then

$$\sum_{i=1}^n \bar{X}_i \sim \Gamma(n, \lambda).$$

Hence, for large n $\Gamma(n, \lambda) \approx \text{normal}$.

Def. The **Beta dist.** is a 2-parameters family of dists defined by the density

$$f_{ab}(u) = \frac{1}{\beta(a,b)} u^{a-1} (1-u)^{b-1} \mathbf{1}_{u \in (0,1)},$$

where $a, b > 0$ are parameters.

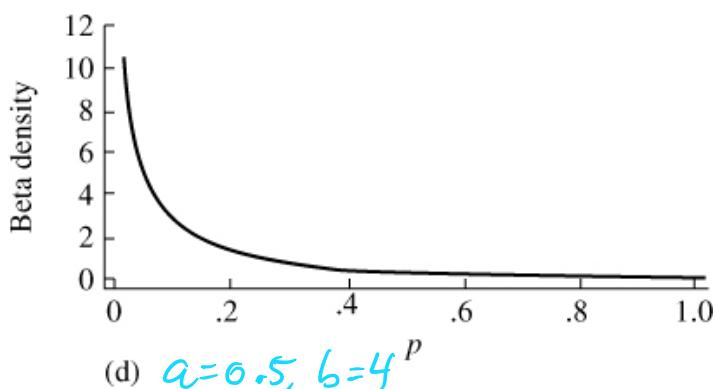
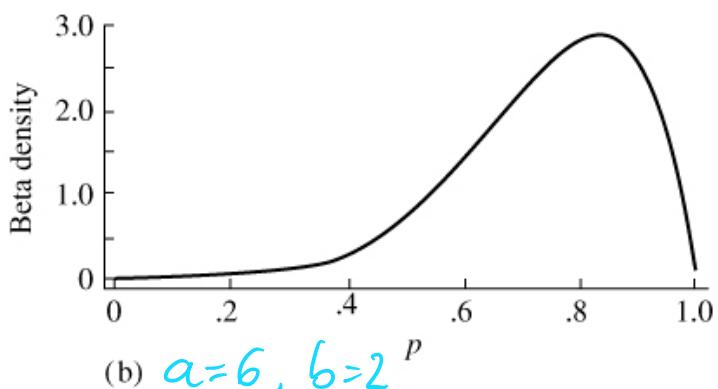
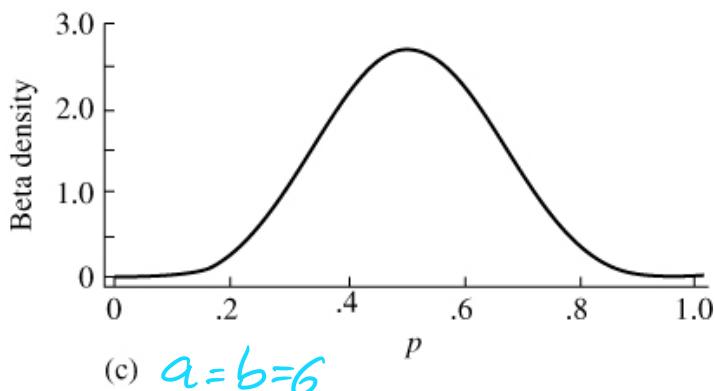
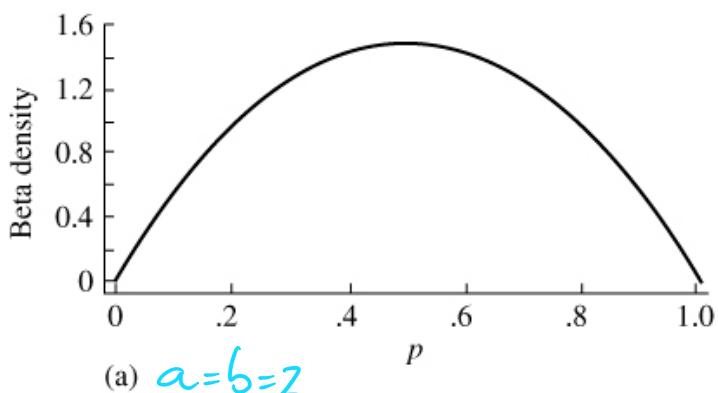
Note that $\int_0^1 f_{ab}(u) du = 1$ by def'n of $\beta(a,b)$.

Taking $a=b=1$,

$$f(u) = c \cdot \mathbf{1}_{u \in (0,1)} = \mathbf{1}_{u \in (0,1)},$$

so the Beta($1, 1$). dist is a $U(0,1)$.

Examples of Beta densities:



Quotients

Suppose X and Y have a joint density f_{XY} and let $Z = Y/X$. What is the dist. of Z ? Is it ?
 (since $P(X=0)=0$, Z is well defined)

Let $A_3 = \{(x,y) \in \mathbb{R}^2; y/x \leq 3\}$, then

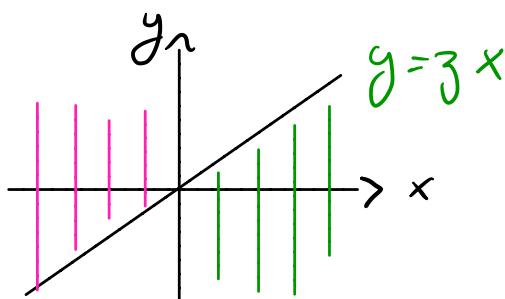
$$F_Z(z) = P(Y/X \leq z) = \iint_{A_3} f_{XY}(x,y) dx dy.$$

How does A_3 look like?

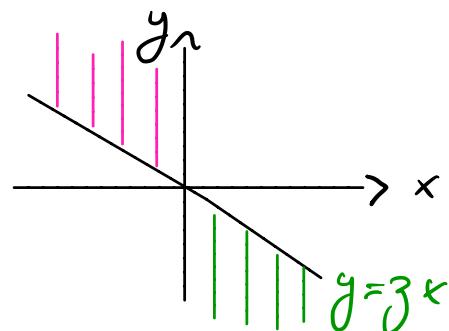
$$(i) x > 0: (x,y) \in A_3 \Leftrightarrow y/x \leq z \Leftrightarrow y \leq z \cdot x$$

$$(ii) x < 0: (x,y) \in A_3 \Leftrightarrow y/x \leq z \Leftrightarrow y \geq z \cdot x$$

$z > 0:$



$z < 0:$



$$F_Z(z) = \int_{x=-\infty}^0 \int_{y=-\infty}^{y=3x} f_{XY}(x,y) dy dx + \int_{x=0}^{\infty} \int_{y=-\infty}^{y=3x} f_{XY}(x,y) dy dx$$

$$x > 0: \int_{y=-\infty}^{y=3x} f_{XY}(x,y) dy = \int_{-\infty}^3 f_{XY}(x, u \cdot x) x du$$

$|x|$

$$x < 0 : \int_{y=3x}^{\infty} f_{\bar{X}Y}(x,y) dy = \int_3^{\infty} f_{\bar{X}Y}(x, \omega \cdot x) x d\omega \\ = \int_{-\infty}^3 f_{\bar{X}Y}(x, \omega \cdot x) (-x) d\omega \quad || \\ |x|$$

Therefore,

$$F_Z(z) = \iint_{-\infty - \infty}^{0 \ 3} f_{\bar{X}Y}(x, \omega x) |x| d\omega dx + \iint_{0 - \infty}^{\infty \ 3} f_{\bar{X}Y}(x, \omega x) |x| d\omega dx \\ = \int_{-\infty}^{\infty} \int_{-\infty}^3 f_{\bar{X}Y}(x, \omega x) |x| d\omega dx \\ = \int_{-\infty}^{\infty} \left[\int_{-\infty}^{\infty} f_{\bar{X}Y}(x, \omega x) |x| dx \right] d\omega$$

$\Rightarrow Z = Y/X$ is a cont. RV with

$$f_Z(z) = \int_{-\infty}^{\infty} f_{\bar{X}Y}(x, zx) |x| dx \\ y \Rightarrow y/x = z$$

Compare with

$$f_{\bar{X}+Y}(z) = \int_{-\infty}^{\infty} f_{\bar{X}Y}(x, z-x) dx \\ y \Rightarrow x+y = z$$

Note the extra $|x|$ term for Y/X : density \times length/area
= prob.

If X and Y are ind. then

$$f_{Y/X}(z) = \int_{-\infty}^{\infty} f_X(x) f_Y(z \cdot x) |x| dx .$$

Example. X, Y are iid $N(0, 1)$ RVs, $Z = Y/X$

$$\begin{aligned} f_Z(z) &= \int_{-\infty}^{\infty} \underbrace{\frac{1}{\sqrt{2\pi}} e^{-x^2/2}}_{\text{even fn.}} \underbrace{\frac{1}{\sqrt{2\pi}} e^{-(zx)^2/2}}_{|x|} |x| dx \\ &= 2 \cdot \int_0^{\infty} \frac{1}{2\pi} e^{-\frac{x^2}{2}(1+z^2)} \cdot x dx \\ &= \frac{1}{\pi} \left[-\frac{e^{-\frac{x^2}{2}(1+z^2)}}{1+z^2} \right]_0^{\infty} \\ &= \frac{1}{\pi} \frac{1}{1+z^2} \end{aligned}$$

This is the density of the **Cauchy dist.**
Decays slowly (heavy tail).

Functions of jointly cont. RVs

Recall: if g is strictly monotone and diff. and if \underline{X} is a cont. RV, then $\underline{Y} = g(\underline{X})$ is a cont. RV and

$$f_y(y) = f_{\underline{X}}(g^{-1}(y)) / |g'(g^{-1}(y))|.$$

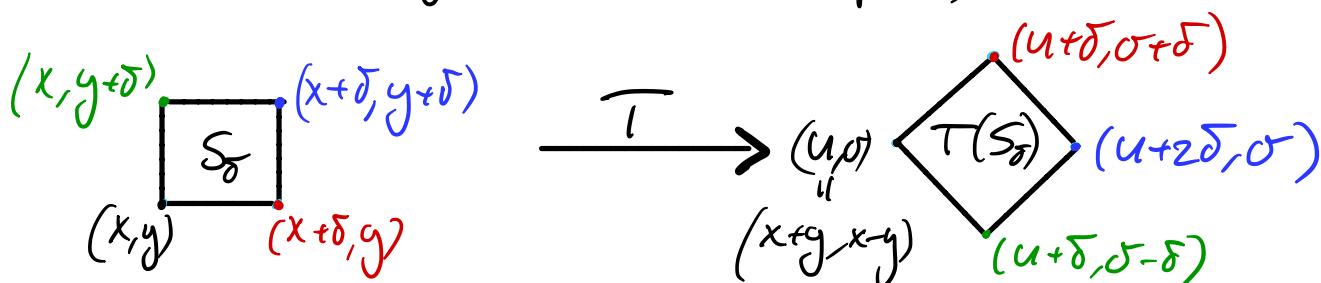
Suppose \underline{X} and \underline{Y} have a joint density $f_{\underline{X}\underline{Y}}$, and $T: \mathbb{R}^2 \rightarrow \mathbb{R}^2$ is 1:1, onto, and diff., and define the random point

$$(U, V) := T(\underline{X}, \underline{Y}).$$

For example,

$$\begin{aligned} T(x, y) &= (x+y, x-y) \\ \Rightarrow (U, V) &= (\underline{X} + \underline{Y}, \underline{X} - \underline{Y}). \end{aligned}$$

Assume, for now, that (U, V) has a joint density f_{UV} , then with $(u, v) = T(x, y)$ in our example, and $\delta > 0$



S_δ is a square with $|S_\delta| = \delta^2$.

$T(S_\delta)$ is a square with each side of length $\sqrt{\delta^2 + \delta^2} = \sqrt{2}\delta$, so $|T(S_\delta)| = 2\delta^2$.

Assume both f_{xy} and f_{uv} are cont. fns. It follows that

$$\begin{aligned} P((\bar{x}, \bar{y}) \in S_\delta) &= \iint_{S_\delta} f_{xy}(s, t) \, ds \, dt \\ &= f_{xy}(x, y) / |S_\delta| + R_\delta \cdot |S_\delta|, \end{aligned}$$

where

$$R_\delta = R_\delta(x, y) = \frac{1}{|S_\delta|} \iint_{S_\delta} (f_{xy}(s, t) - f_{xy}(x, y)) \, ds \, dt \xrightarrow[\delta \rightarrow 0]{} 0. \quad \text{exercise}$$

Similarly,

$$\begin{aligned} P((U, V) \in T(S_\delta)) &= \iint_{T(S_\delta)} f_{uv}(q, r) \, dq \, dr \\ &= f_{uv}(u, v) / |T(S_\delta)| + R'_\delta / |T(S_\delta)|, \end{aligned}$$

where $R'_\delta \xrightarrow[\delta \rightarrow 0]{} 0$.

Since T is 1:1,

$$(\bar{x}, \bar{y}) \in S_\delta \Leftrightarrow (U, V) = T(\bar{x}, \bar{y}) \in T(S_\delta)$$

$$\Rightarrow P((\bar{x}, \bar{y}) \in S_\delta) = P((U, V) \in T(S_\delta))$$

$$\Rightarrow (f_{xy}(x, y) + R_\delta) \delta^2 = (f_{uv}(u, v) + R'_\delta) \cdot 2\delta^2$$

Divide by δ^2 and let $\delta \rightarrow 0$ to get:

$$f_{xy}(x, y) = 2f_{uv}(u, v), \quad \text{or}$$

$$f_{uv}(u, v) = \frac{1}{2} f_{xy}(T(u, v))$$

The unit area in the $x-y$ coordinates is 2x larger when mapped to the $u-v$ coords. Therefore, the density in the $u-v$ coords is 2x smaller.

The factor of 2 is the "blow up" factor:

$$|\mathcal{T}(S_\delta)| / |S_\delta| = 2$$

By how much does a small unit square in the $x-y$ coords grow, when it is mapped under \mathcal{T} .

It is given by the **Jacobian determinant**, or the determinant of the Jacobian/Derivative matrix:

$$\mathcal{J}_T = |\mathcal{D}_T| = \left| \begin{pmatrix} \frac{\partial u}{\partial x} & \frac{\partial u}{\partial y} \\ \frac{\partial v}{\partial x} & \frac{\partial v}{\partial y} \end{pmatrix} \right|,$$

which in our case with $u=x+y$, $v=x-y$ is

$$\mathcal{J}_T = \left| \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix} \right| = -2 \quad \text{constant!}$$

and $|\mathcal{J}_T| = 2$.

A feature of linear transformation.

$\Rightarrow f_{\bar{x}\bar{y}}(x,y) = P_{uv}(\mathcal{T}(x,y)) |\mathcal{J}_T|$ or equivalently,

$$f_{uv}(u,v) = f_{\bar{x}\bar{y}}(\mathcal{T}^{-1}(u,v)) |\mathcal{J}_T|^{-1}$$

$$= f_{\bar{x}\bar{y}}(\mathcal{T}^{-1}(u,v)) |\mathcal{J}_{T^{-1}}(u,v)|.$$

Note that if $T: D \subset \mathbb{R}^n \rightarrow \mathbb{R}^n$ where D is an open set and T is diff., then if $J_T \neq 0$ at $\underline{x} \in \mathbb{R}^n$, then T is invertible in the neighborhood of $T(\underline{x})$ and by the chain rule $J_T^{-1} = J_{T^{-1}}^{-1}$.

Our result generalizes as follows.

Thm. (density of function of a random point)

Suppose $T: \underset{\text{open}}{D} \subset \mathbb{R}^2 \rightarrow \underset{\text{open}}{R} \subset \mathbb{R}^2$ is diff and 1:1 on D with $J_T \neq 0$, and it is onto R .

Suppose (\underline{x}, y) is jointly cont. with density $f_{\underline{x}y}$ which vanishes outside of D , and let $(u, v) = T(\underline{x}, y)$.

Then (u, v) is jointly cont. and $T(u, v) \in R$

$$f_{uv}(u, v) = f_{\underline{x}y}(T^{-1}(u, v)) |J_{T^{-1}}(u, v)|.$$

The theorem can be proved with the help of the following general theorem in calculus.

Thm. If T is as above then for any open $A \subset D$, $B = T(A)$ is an open subset of R , and for any integrable $\psi: A \rightarrow \mathbb{R}$ we have

$$\iint_A \psi(x, y) dx dy = \iint_B \psi(T^{-1}(u, v)) |J_{T^{-1}}(u, v)| du dv$$

The first (density) thus follows from the second (change of variable):

Take $\gamma = f_{xy}$, then for any open $B \subset R$

$$P((U, V) \in B) = P((X, Y) \in A = T(B)) \quad (\text{T is 1:1})$$

$$= \iint_A f_{xy}(x, y) dx dy$$

$$= \iint_{B=T(A)} f_{xy}(T^{-1}(u, v)) |J_{T^{-1}}(u, v)| du dv$$

Since this holds for all open $B \subset R$, (U, V) has a density which is again:

$$f_{uv}(u, v) = f_{xy}(T^{-1}(u, v)) |J_{T^{-1}}(u, v)|.$$