## STAT 512 su 2021 Lec 02 slides

# Transformations of multiple random variables 

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These slides are an instructional aid; their sole purpose is to display, during the lecture, definitions, plots, results, etc. which take too much time to write by hand on the blackboard.

They are not intended to explain or expound on any material.

We often wish to find the distribution of a function of two (or more) rvs:
Exercise: Let $X_{1}$ and $X_{2}$ be indep. Exponential $(\lambda)$ rvs. Let $Y=X_{1} /\left(X_{1}+X_{2}\right)$.
(1) Find the cdf of $Y$.
(2) Find the pdf of $Y$.

## Theorem (Bivariate transformation method)

Let $\left(X_{1}, X_{2}\right)$ be a pair of cont. rvs with joint pdf $f_{X_{1}, X_{2}}$ on $\mathcal{X}$ and

$$
Y_{1}=g_{1}\left(X_{1}, X_{2}\right) \quad \text { and } \quad Y_{2}=g_{2}\left(X_{1}, X_{2}\right),
$$

where $g_{1}$ and $g_{2}$ define a $1: 1$ transformation of $\mathcal{X}$ onto $\mathcal{Y}$ (define these).
Let $g_{1}^{-1}$ and $g_{2}^{-1}$ be the functions satisfying

$$
\begin{aligned}
& y_{1}=g_{1}\left(x_{1}, x_{2}\right) \\
& y_{2}=g_{2}\left(x_{1}, x_{2}\right)
\end{aligned} \Longleftrightarrow \begin{aligned}
& x_{1}=g_{1}^{-1}\left(y_{1}, y_{2}\right) \\
& x_{2}=g_{2}^{-1}\left(y_{1}, y_{2}\right)
\end{aligned}
$$

for all $\left(x_{1}, x_{2}\right) \in \mathcal{X}$ and $\left(y_{1}, y_{2}\right) \in \mathcal{Y}$.
Then the joint pdf of $\left(Y_{1}, Y_{2}\right)$ is given by

$$
f_{Y_{1}, Y_{2}}\left(y_{1}, y_{2}\right)=f_{X_{1}, x_{2}}\left(g_{1}^{-1}\left(y_{1}, y_{2}\right), g_{2}^{-1}\left(y_{1}, y_{2}\right)\right)\left|J\left(y_{1}, y_{2}\right)\right|, \quad \text { for }\left(y_{1}, y_{2}\right) \in \mathcal{Y} \text {, }
$$

where $J\left(y_{1}, y_{2}\right)$ is the Jacobian (next slide), if $J\left(y_{1}, y_{2}\right)$ is not always 0 on $\mathcal{Y}$.

## Jacobian

In the setup of the previous slide, the Jacobian of the transformation is defined as

$$
J\left(y_{1}, y_{2}\right)=\left|\begin{array}{ll}
\frac{\partial}{\partial y_{1}} g_{1}^{-1}\left(y_{1}, y_{2}\right) & \frac{\partial}{\partial y_{2}} g_{1}^{-1}\left(y_{1}, y_{2}\right) \\
\frac{\partial}{\partial y_{1}} g_{2}^{-1}\left(y_{1}, y_{2}\right) & \frac{\partial}{\partial y_{2}} g_{2}^{-1}\left(y_{1}, y_{2}\right)
\end{array}\right| .
$$

For real numbers $a, b, c, d$,

$$
\left|\begin{array}{ll}
a & b \\
c & d
\end{array}\right|=a d-b c .
$$

This is called the determinant.

Exercise: Let $X_{1}, X_{2}$ be independent $\operatorname{Normal}(0,1)$ rvs.
(1) Find the joint pdf of $Y_{1}=X_{1} / X_{2}$ and $Y_{2}=X_{2}$.
(2) Find the marginal pdf of $Y_{1}$.

Exercise: Let $X_{1}, X_{2}$ have joint pdf

$$
f_{X_{1}, X_{2}}\left(x_{1}, x_{2}\right)=\frac{1}{\lambda^{2}} \exp \left[-\frac{x_{1}+x_{2}}{\lambda}\right] \mathbf{1}\left(x_{1}>0, x_{2}>0\right) .
$$

(1) Find the joint pdf of $Y_{1}=X_{1} /\left(X_{1}+X_{2}\right)$ and $Y_{2}=X_{1}+X_{2}$.
(2) Find the marginal pdf of $Y_{1}$.

Exercise: Let $X_{1} \sim \operatorname{Beta}(1,1)$ and $X_{2} \sim \operatorname{Beta}(2,1)$ be independent rvs.
(1) Find the joint pdf of $Y_{1}=X_{1} X_{2}$ and $Y_{2}=X_{2}$.
(2) Find the marginal pdf of $Y_{1}$.

Exercise: Let $Z_{1}, Z_{2}$ have the bivariate Normal distribution, with joint pdf

$$
f_{z_{1}, Z_{2}}\left(z_{1}, z_{2}\right)=\frac{1}{2 \pi} \frac{1}{\sqrt{1-\rho^{2}}} \exp \left[-\frac{1}{2} \frac{z_{1}^{2}-2 \rho z_{1} z_{2}+z_{2}^{2}}{1-\rho^{2}}\right] .
$$

(1) Find the joint pdf of $U_{1}=Z_{1}+Z_{2}$ and $U_{2}=Z_{1}-Z_{2}$.
(2) Find the marginal pdfs of $U_{1}$ and $U_{2}$.

Exercise: Let $Z_{1}, Z_{2}$ have the bivariate Normal distribution, with joint pdf

$$
f_{z_{1}, z_{2}}\left(z_{1}, z_{2}\right)=\frac{1}{2 \pi} \frac{1}{\sqrt{1-\rho^{2}}} \exp \left[-\frac{1}{2} \frac{z_{1}^{2}-2 \rho z_{1} z_{2}+z_{2}^{2}}{1-\rho^{2}}\right] .
$$

(1) Find the joint pdf of $U_{1}=\min \left\{Z_{1}, Z_{2}\right\}$ and $U_{2}=\max \left\{Z_{1}, Z_{2}\right\}$ ? (Not 1:1).
(2) Find the marginal pdf of $U_{2}$ (can skip this example; see notes if curious).

Theorem (mgf method for sums of independent rvs)
Let $X_{1}, \ldots, X_{n}$ be ind. rvs with mgfs $M_{X_{1}}, \ldots, M_{X_{n}}$, resp. Let $Y=X_{1}+\cdots+X_{n}$.
The mgf of $Y$ is given by

$$
M_{Y}(t)=\prod_{i=1}^{n} M_{X_{i}}(t)
$$

Moreover, if $X_{1}, \ldots, X_{n}$ are ind. and all have $m g f M_{X}$ (are iid), then

$$
M_{Y}(t)=\left[M_{X}(t)\right]^{n} .
$$

Exercise: Prove the above.

Exercise: Let $X_{1}, \ldots, X_{n}$ be ind. chi-squared rvs with dfs $\nu_{1}, \ldots, \nu_{n}$, resp. Find the distribution of $Y=X_{1}+\cdots+X_{n}$.

Exercise: Let $X_{1}, \ldots, X_{n}$ be ind. Normal rvs with means $\mu_{1}, \ldots, \mu_{n}$ and variances $\sigma_{1}^{2}, \ldots, \sigma_{n}^{2}$, resp.
(1) Find the distribution of $Y=X_{1}+\cdots+X_{n}$.
(2) Find the distribution of $V=a_{1} X_{1}+\cdots+a_{n} X_{n}$, for $a_{1}, \ldots, a_{n} \in \mathbb{R}$.
(0) Find the distribution of $\bar{X}_{n}=\left(X_{1}+\cdots+X_{n}\right) / n$.

