

Transformations of Jointly Distributed Random Variables¹

STA 256: Fall 2019

¹This slide show is an open-source document. See last slide for copyright information.

Overview

1 Convolutions

2 Jacobians

Transformations of Jointly Distributed Random Variables

Let $Y = g(X_1, \dots, X_n)$. What is the probability distribution of Y ?

For example,

- X_1 is the number of jobs completed by employee 1.
- X_2 is the number of jobs completed by employee 2.
- You know the probability distributions of X_1 and X_2 .
- You would like to know the probability distribution of $Y = X_1 + X_2$.

Convolutions of discrete random variables

- Let X and Y be discrete random variables.
- The standard case is where they are independent.
- Want probability mass function of $Z = X + Y$.

$$\begin{aligned}p_Z(z) &= P(Z = z) \\&= P(X + Y = z) \\&= \sum_x P(X + Y = z | X = x) P(X = x) \\&= \sum_x P(x + Y = z | X = x) P(X = x) \\&= \sum_x P(Y = z - x | X = x) P(X = x) \\&= \sum_x P(Y = z - x) P(X = x) \text{ by independence} \\&= \sum_x p_X(x) p_Y(z - x)\end{aligned}$$

Summarizing

Convolutions of discrete random variables

Let X and Y be *independent* discrete random variables, and $Z = X + Y$.

$$p_Z(z) = \sum_x p_X(x)p_Y(z - x)$$

Two Important results

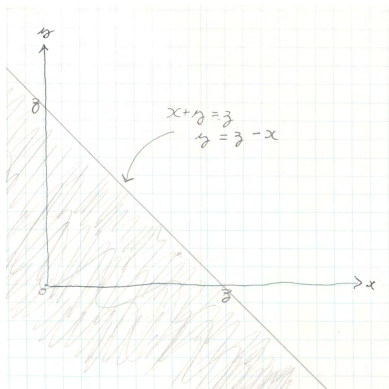
Proved using the convolution formula

- Let $X \sim \text{Poisson}(\lambda_1)$ and $Y \sim \text{Poisson}(\lambda_2)$ be independent. Then $Z = X + Y \sim \text{Poisson}(\lambda_1 + \lambda_2)$.
- Let $X \sim \text{Binomial}(n_1, \theta)$ and $Y \sim \text{Binomial}(n_2, \theta)$ be independent. Then $Z = X + Y \sim \text{Binomial}(n_1 + n_2, \theta)$

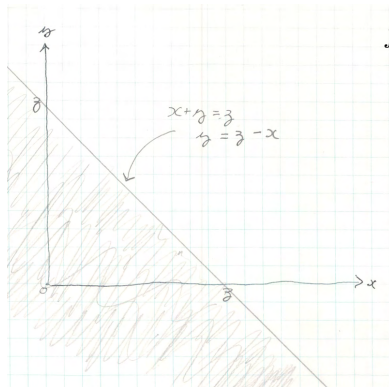
Convolutions of *continuous* random variables

- Let X and Y be continuous random variables.
- The standard case is where they are independent.
- Want probability density function of $Z = X + Y$.

$$\begin{aligned}
 f_Z(z) &= \frac{d}{dz} P(Z \leq z) \\
 &= \frac{d}{dz} P(X + Y \leq z)
 \end{aligned}$$



Continuing ...



$$\begin{aligned}
 f_Z(z) &= \frac{d}{dz} P(X + Y \leq z) \\
 &= \frac{d}{dz} \int_{-\infty}^{\infty} \int_{-\infty}^{z-x} f_{X,Y}(x, y) dy dx
 \end{aligned}$$

$$t = y + x \quad y = t - x \quad dy = dt$$

y	$t = y + x$
$z - x$	z
$-\infty$	$-\infty$

$$\int_{-\infty}^z f_{X,Y}(x, t - x) dt$$

Still continuing, have

$$\begin{aligned}f_Z(z) &= \frac{d}{dz} \int_{-\infty}^{\infty} \int_{-\infty}^z f_{X,Y}(x, t-x) dt dx \\&= \frac{d}{dz} \int_{-\infty}^z \int_{-\infty}^{\infty} f_{X,Y}(x, t-x) dx dt \\&= \int_{-\infty}^{\infty} f_{X,Y}(x, z-x) dx \\&= \int_{-\infty}^{\infty} f_X(x) f_Y(z-x) dx \quad \text{if } X \text{ and } Y \text{ are independent.}\end{aligned}$$

Compare

For continuous random variables:

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

For discrete random variables:

$$p_Z(z) = \sum_x p_X(x) p_Y(z - x)$$

Of course you need to pay attention to the limits of integration or summation, because $f_X(x) f_Y(z - x)$ may be zero for some x .

Two Important results for continuous random variables

Proved using the convolution formula

- Let X and Y be independent exponential random variables with parameter $\lambda > 0$. Then
 $Z = X + Y \sim \text{Gamma}(\alpha = 2, \lambda)$.
- Let $X \sim \text{Normal}(\mu_1, \sigma_1^2)$ and $Y \sim \text{Normal}(\mu_2, \sigma_2^2)$ be independent. Then
 $Z = X + Y \sim \text{Normal}(\mu_1 + \mu_2, \sigma_1^2 + \sigma_2^2)$.

The Jacobian Method

- X_1 and X_2 are continuous random variables.
- $Y_1 = g_1(X_1, X_2)$ and $Y_2 = g_2(X_1, X_2)$.
- Want $f_{Y_1, Y_2}(y_1, y_2)$

Solve for x_1 and x_2 , obtaining $x_1(y_1, y_2)$ and $x_2(y_1, y_2)$. Then

$$f_{Y_1, Y_2}(y_1, y_2) = f_{X_1, X_2}(x_1(y_1, y_2), x_2(y_1, y_2)) \cdot \text{abs} \begin{vmatrix} \frac{\partial x_1}{\partial y_1} & \frac{\partial x_1}{\partial y_2} \\ \frac{\partial x_2}{\partial y_1} & \frac{\partial x_2}{\partial y_2} \end{vmatrix}$$

The determinant $\begin{vmatrix} a & b \\ c & d \end{vmatrix} = ad - bc$.

More about the Jacobian method

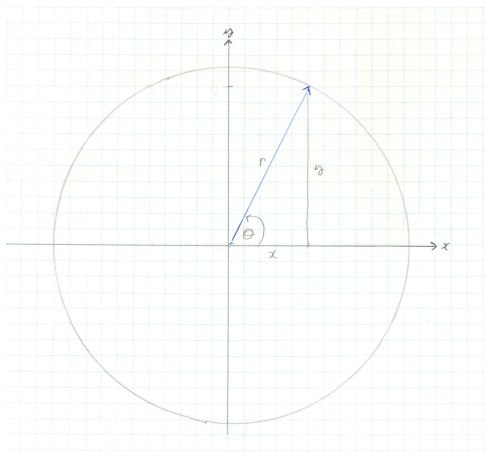
$$Y_1 = g_1(X_1, X_2) \text{ and } Y_2 = g_2(X_1, X_2)$$

- It follows directly from a change of variables formula in multi-variable integration. The proof is omitted.
- It must be possible to solve $y_1 = g_1(x_1, x_2)$ and $y_2 = g_2(x_1, x_2)$ for x_1 and x_2 .
- That is, the function $g : \mathbb{R}^2 \rightarrow \mathbb{R}^2$ must be one to one (injective).
- Frequently you are only interested in Y_1 , and $Y_2 = g_2(X_1, X_2)$ is chosen to make reverse solution easy.
- The partial derivatives must all be continuous, except possibly on a set of probability zero (they almost always are).
- It extends naturally to higher dimension.

Change from rectangular to polar co-ordinates

By the Jacobian method

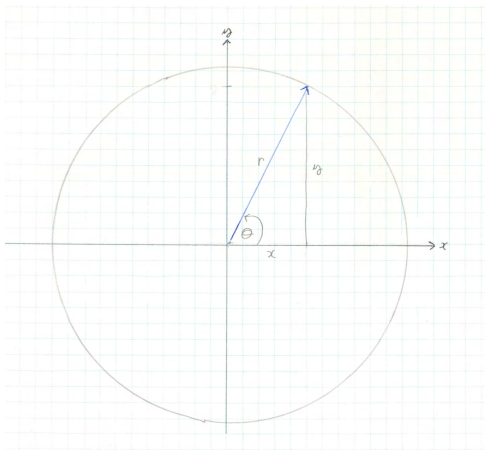
A point on the plane may be represented as (x, y) , or



An angle θ and a radius r .

Change of variables

From rectangular to polar coordinates



$$x = r \cos(\theta)$$

$$y = r \sin(\theta)$$

$$x^2 + y^2 = r^2$$

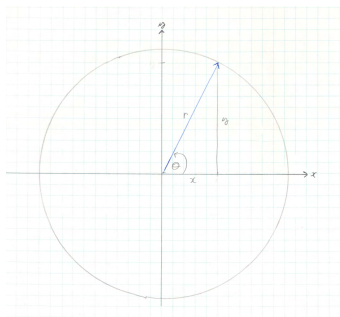
- As x and y range from $-\infty$ to ∞ ,
- r goes from 0 to ∞
- And θ goes from 0 to 2π .

$$\text{Integral } \int_0^\infty \int_0^\infty f_{x,y}(x, y) dx dy$$

Change of variables:

$$x = r \cos(\theta)$$

$$y = r \sin(\theta)$$



$$\begin{aligned} & \int_0^\infty \int_0^\infty f_{x,y}(x, y) dx dy \\ &= \int_0^{\pi/2} \int_0^\infty f_{x,y}(r \cos \theta, r \sin \theta) \text{abs} \begin{vmatrix} \frac{\partial x}{\partial r} & \frac{\partial x}{\partial \theta} \\ \frac{\partial y}{\partial r} & \frac{\partial y}{\partial \theta} \end{vmatrix} dr d\theta \end{aligned}$$

Evaluate the determinant

(with $x = r \cos(\theta)$ and $y = r \sin(\theta)$)

$$\begin{aligned} \begin{vmatrix} \frac{\partial x}{\partial r} & \frac{\partial x}{\partial \theta} \\ \frac{\partial y}{\partial r} & \frac{\partial y}{\partial \theta} \end{vmatrix} &= \begin{vmatrix} \frac{\partial r \cos(\theta)}{\partial r} & \frac{\partial r \cos(\theta)}{\partial \theta} \\ \frac{\partial r \sin(\theta)}{\partial r} & \frac{\partial r \sin(\theta)}{\partial \theta} \end{vmatrix} \\ &= \begin{vmatrix} \cos(\theta) & -r \sin(\theta) \\ \sin(\theta) & r \cos(\theta) \end{vmatrix} \\ &= r \cos^2 \theta - -r \sin^2 \theta \\ &= r(\sin^2 \theta + \cos^2 \theta) \\ &= r \end{aligned}$$

So the integral is

$$\int_0^{\infty} \int_0^{\infty} f_{x,y}(x, y) dx dy = \int_0^{\pi/2} \int_0^{\infty} f_{x,y}(r \cos \theta, r \sin \theta) r dr d\theta$$

- The standard formula for change from rectangular to polar co-ordinates is $dx dy = r dr d\theta$.
- It comes from a Jacobian.
- Other limits of integration are possible.
- $f(x, y)$ does not have to be a density.

Copyright Information

This slide show was prepared by **Jerry Brunner**, Department of Statistical Sciences, University of Toronto. It is licensed under a **Creative Commons Attribution - ShareAlike 3.0 Unported License**. Use any part of it as you like and share the result freely. The $\text{L}^{\text{A}}\text{T}_{\text{E}}\text{X}$ source code is available from the course website:

<http://www.utstat.toronto.edu/~brunner/oldclass/256f19>