

# CSE400 — Fundamentals of Probability in Computing

## Lecture L11/L12: Transformation of Random Variables

February 10–12, 2026

### Outline (as given)

Transformation of Random Variables

Learning transformation techniques for random variables.

Function of Two Random Variables

Joint transformations and derived distributions.

Illustrative Example

Detailed derivation for the case:

$$Z = X + Y$$

L12\_S1\_A (2)

## 1. Transformation of a Single Random Variable

Let  $X$  be a known random variable with known PDF.

Let a new random variable be defined as

$$Y = g(X)$$

The goal is to find:

CDF of  $Y$ :

$$F_Y(y)$$

PDF of  $Y$ :

$$f_Y(y)$$

### Step-by-Step Method (as presented)

**Step 1 — Start from the CDF definition**

$$F_Y(y) = \Pr(Y \leq y)$$

**Step 2 — Substitute  $Y = g(X)$**

$$F_Y(y) = \Pr(g(X) \leq y)$$

**Step 3 — Convert inequality into an equivalent condition on  $X$** 

This produces an interval (or union of intervals) in terms of  $X$ :

$$= \Pr(X \in \text{corresponding region})$$

**Step 4 — Express using the CDF of  $X$** 

$$F_Y(y) = F_X(\text{boundary})$$

or equivalently

$$F_Y(y) = 1 - F_X(\text{boundary})$$

depending on monotonicity.

**Step 5 — Differentiate to obtain the PDF**

$$f_Y(y) = \frac{d}{dy} F_Y(y)$$

This yields the required PDF of  $Y$ .

These steps are explicitly shown in the slides as:

Write CDF of  $Y$

Replace  $Y$  by function of  $X$

Convert probability to  $X$ -domain

Use known CDF/PDF of  $X$

Differentiate to obtain  $f_Y(y)$

L12\_S1\_A (2)

**Worked Example 1 (Single RV)**

Given

$X$  is uniformly distributed on  $(-1, 1)$

Hence,

$$f_X(x) = \frac{1}{2}, \quad -1 < x < 1$$

Define

$$Y = \sin(\pi X)$$

Required

Find the PDF of  $Y$ .

**Solution (exact lecture flow)****Step 1 — Start with CDF of  $Y$** 

$$F_Y(y) = \Pr(Y \leq y)$$

**Step 2 — Substitute transformation**

$$= \Pr(\sin(\pi X) \leq y)$$

**Step 3 — Convert inequality to bounds on  $X$** 

This produces inverse-sine type limits on  $X$  (shown graphically/analytically in slides).

**Step 4 — Use PDF of uniform  $X$** 

Since

$$f_X(x) = \frac{1}{2}$$

the probability becomes an integral over the corresponding  $X$ -interval.

**Step 5 — Differentiate CDF to obtain PDF**

Final result (as written in slides):

$$f_Y(y) = \frac{1}{\pi\sqrt{1-y^2}}, \quad |y| < 1$$

and

$$f_Y(y) = 0 \text{ otherwise.}$$

L12\_S1\_A (2)

## 2. Function of Two Random Variables

Let two random variables  $X, Y$  be given.

Define a new RV:

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

Goal: find CDF/PDF of  $Z$ .

**General Definition**

$$F_Z(z) = \Pr(Z \leq z) = \Pr(g(X, Y) \leq z)$$

This is evaluated by integrating the joint PDF over the appropriate region:

$$F_Z(z) = \iint_{g(x,y) \leq z} f_{X,Y}(x, y) \, dx \, dy$$

L12\_S1\_A (2)

**Illustrative Example:  $Z = X + Y$** **Step 1 — Write CDF**

$$F_Z(z) = \Pr(X + Y \leq z)$$

**Step 2 — Convert to double integral**

$$F_Z(z) = \iint_{x+y \leq z} f_X(x) f_Y(y) \, dx \, dy$$

(Independence is assumed in the lecture examples.)

**Step 3 — Express limits explicitly**

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

**Step 4 — Differentiate to obtain PDF**

Using Leibniz rule for differentiation under the integral sign (explicitly shown):  
If

$$G(z) = \int h(x, z) dx$$

then

$$\frac{d}{dz}G(z) = \int \frac{\partial}{\partial z}h(x, z) dx$$

Applying this,

$$f_Z(z) = \frac{d}{dz}F_Z(z)$$

leads to the convolution form:

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

L12\_S1\_A (2)

## Example: Exponential Random Variables

Given:

$X, Y$  are exponential RVs

PDFs shown in slides as:

$$f_X(x) = \lambda e^{-\lambda x}, \quad x \geq 0$$

$$f_Y(y) = \lambda e^{-\lambda y}, \quad y \geq 0$$

Define:

$$Z = X + Y$$

Using convolution:

$$f_Z(z) = \int_0^z \lambda e^{-\lambda x} \lambda e^{-\lambda(z-x)} dx$$

Step-by-step (as written)

Substitute PDFs

Combine exponentials

Take constants outside

Integrate over  $x$  from 0 to  $z$

Final expression in slides:

$$f_Z(z) = \lambda^2 z e^{-\lambda z}, \quad z \geq 0$$

and

$$f_Z(z) = 0 \text{ otherwise.}$$

L12\_S1\_A (2)

## Additional Slide Examples (listed)

Linear combinations such as:

$$Z = X - 2Y$$

CDF written as:

$$F_Z(z) = \Pr(X - 2Y \leq z)$$

and converted into joint-probability regions in the  $x$ - $y$  plane, following the same steps:

Write CDF

Express inequality

Set integration region

Integrate joint PDF

L12\_S1\_A (2)

## Key Exam Takeaways (from slides)

Always start with CDF definition.

Substitute transformation explicitly.

Convert probability into integration region.

Use known PDF(s).

Apply Leibniz rule when differentiating integrals.

For  $Z = X + Y$ , final PDF is obtained via convolution.

For single-RV transformations, use CDF  $\rightarrow$  differentiate.