

## EEL 5544 Lecture 19

GENERATING RANDOM VARIABLES

- To generate a random variable with an arbitrary distribution, we would like to:
  1. Generate a Uniform random variable on  $(0, 1]$ ,  $U$
  2. Apply a function  $g$  to  $U$  such that if  $X = g(U)$ , then  $X$  has the desired distribution
- We begin by making an observation: Suppose  $X$  is a random variable with distribution function  $F_X(x)$

Then what is the distribution of  $Y = F_X(X)$ ?

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

$$=$$

$$=$$

$$=$$

$$=$$

and

$$F_Y(y) = \begin{cases} 0, & y \leq 0 \\ 1, & y \geq 1 \end{cases}$$

By inspection  $Y$  is a \_\_\_\_\_ random variable!

- Thus to generate a random variable  $X$  with distribution function  $F_X(x)$ , we can use the following procedure:
- **Transformation Method**

To generate a RV  $X$  with a **continuous distribution**:

1. Generate a random variable  $U$  that is distributed uniform on  $[0, 1]$  using commonly available methods.
2. Let  $X = F_X^{-1}(U)$

*Proof:*

It is a notational nightmare if we straight away let  $X = F_X^{-1}(U)$  so instead, let's first just let  $Z = F_X^{-1}(U)$

Then

$$\begin{aligned} F_Z(z) &= P(F_X^{-1}(U) \leq z) \\ &= \\ &= \\ &= \end{aligned}$$

because

So  $Z$  has the desired distribution. Replacing  $Z$  with  $X$  finishes the proof.

**Example:** Generate a random variable  $X$  that has an exponential distribution with parameter  $\lambda$

To generate a RV  $X$  with a **discrete distribution** on a consecutive subset of the integers:

1. Generate a random variable  $U$  that is distributed uniform on  $[0, 1]$  using commonly available methods.

2. Let  $X = k$  if  $F_X(k-1) < U \leq F_X(k)$ .

*Proof:*

Again, in order to avoid confusing notation, let's let  $Z = k$  if  $F_X(k-1) < U \leq F_X(k)$ .

$$P(Z = k) =$$

$$=$$

which is the desired probability mass at point  $k$

Again, replace  $Z$  with  $X$ , and the proof is complete.

### FUNCTIONS OF MULTIPLE RANDOM VARIABLES:

#### ONE FUNCTION OF SEVERAL RANDOM VARIABLES

- We often have situations in which we are interested in a function that involves two or more random variables
- For instance, if  $X$  and  $Y$  are random variables, then we may be interested in the following:
  - The signal  $X$  is received in the presence of additive noise  $Y$ ,  $Z = X + Y$
  - A device has two identical components. Let  $X$  and  $Y$  be the time until each component fails. Let  $Z$  be the time until the device stops working, which can be:
    - \* Only when both components fail:  $Z = \max(X, Y)$
    - \* When either component fails:  $Z = \min(X, Y)$
  - A random signal is modulated by another signal,  $Z = XY$ .
  - The Euclidean distance of a point in a plane is  $Z = \sqrt{X^2 + Y^2}$
- I'll use a more general notation than the book's notation at this point. Let the random variables that are input to the function  $g$  be denoted by

$$X_1, X_2, \dots, X_n = \mathbf{X}_n$$

Then  $Z = g(\mathbf{X}_n)$

- The solutions to problems of the form  $Z = g(\mathbf{X}_n)$  are not fundamentally different from the solutions to problems of the form  $Z = g(X)$ .

We just have to be a little more careful.

Consider the distribution function for  $Z$ ,

$$F_Z(z) = P[g(\mathbf{X}_n) \leq z]$$

Let  $R_z = \{\mathbf{x}_n | g(\mathbf{x}_n) \leq z\}$ . Then

$$F_Z(z) = P[\mathbf{X}_n \in R_z]$$

The problem is that the region  $R_z$  is not necessarily rectangular, in which case the probability of  $\mathbf{X}_n \in R_z$  cannot be directly calculated from the distribution function

However, the probability of any region can be calculated by integrating the density over that region:

$$F_Z(z) = \int \cdots \int_{\mathbf{x}_n \in R_z} f_{\mathbf{X}_n}(x_1, x_2, \dots, x_n) dx_1 dx_2 \cdots dx_n$$

This is best illustrated by an example:

**Example:** Let  $Z = X + Y$ . Find the distribution and density functions for  $Z$  in terms of the joint density function for  $X$  and  $Y$ .

- Note that if  $X$  and  $Y$  are s.i., then

$$\begin{aligned}f_Z(z) &= \int_{-\infty}^{\infty} f_X(x)f_Y(z-x)dx \\ &= [f_X * f_Y](z),\end{aligned}$$

where  $*$  represents the \_\_\_\_\_ operator

**Example:**  $X$  and  $Y$  are independent exponential random variables, each with parameter  $\lambda = 1$

- Applying this technique repeatedly for sums of multiple random variables would be difficult. We will later investigate more powerful techniques to deal with sums of multiple random variables.

### USING CONDITIONAL PDFS TO FIND THE PDF OF

#### A FUNCTION OF SEVERAL RVs

- Let  $Z = g(X, Y)$
- If we condition on  $Y = y$ , then  $g(X, y)$  is a function of only one RV, so we can use the techniques from the previous sections to find

$$f_{Z|Y}(z|Y = y)$$

- Then

$$f_Z(z) = \int_{-\infty}^{\infty} f_{Z|Y}(z|y) f_Y(y) dy$$

by the Law of Total Probability