#### Continuous Random Variables

STAT 330 - Iowa State University

#### **Outline**

In this lecture, students will be introduced to continuous random variables. We look at how to use the various functions associated with continuous R.Vs to answer questions.

# Continuous Random Variables

#### Discrete vs. Continuous R.Vs

#### Discrete Random Variable

Sample space  $(\Omega)$  maps to finite or countably infinite set in  $\mathbb{R}$  Ex:  $\{1,2,3\}, \{1,2,3,4,\ldots\}$ 

#### Continuous Random Variable

Sample space  $(\Omega)$  maps to an uncountable set in  $\mathbb{R}$ . Ex:  $(0, \infty)$ , (10, 20)

• We have already learned about discrete R.Vs (Lectures 5-9)

- All properties of discrete R.Vs have direct counterparts for continuous R.Vs
- Summations (Σ) used for discrete R.V's are replaced by integrals (∫) for continuous R.V's.

#### **CDF** of Continuous Random Variables

#### **Definition**

Let X be a continuous random variable. The *cumulative* distribution function (cdf) of X is

$$F_X(t) = \mathbb{P}(X \leq t)$$

- All cdf properties discussed earlier still hold
  - 1.  $0 \le F_X(t) \le 1$
  - 2.  $F_X$  is non-decreasing (if  $a \le b$ , then  $F_X(a) \le F_X(b)$ ).
  - 3.  $\lim_{t\to-\infty} F_X(t) = 0$  and  $\lim_{t\to\infty} F_X(t) = 1$
  - 4.  $F_X$  is right-continuous with respect to t
- The cdf for continuous R.V is also continuous (not a step function like in discrete case)

#### $PDF \longleftrightarrow CDF$

#### Definition

For a continuous variable X with cdf  $F_X$ , the *probability density* function (pdf) of X is defined as:

$$f_X(x) = F_X'(x) = \frac{d}{dx}F_X(x)$$

### Properties of pdf:

- 1.  $f_X(x) \ge 0$  for all x,
- $2. \int_{-\infty}^{\infty} f_X(x) dx = 1.$

Additionally, for continuous R.V X,

- $F_X(t) = \mathbb{P}(X \le t) = \int_{-\infty}^t f_X(x) dx$  for any  $t \in \mathbb{R}$
- $\mathbb{P}(a \le X \le b) = \int_a^b f_X(x) dx$  for any  $a, b \in \mathbb{R}$
- $\mathbb{P}(X = a) = \mathbb{P}(a \le X \le a) = \int_a^a f_X(x) dx = 0$  for any  $a \in \mathbb{R}$

# **Examples**

#### Continuous R.V Example

Example 1: Let Y be the time (in yrs) until the first major failure of a new disk drive. Suppose the probability density function (pdf) of Y is given by

$$f_Y(y) = \begin{cases} 0 & y \le 0 \\ e^{-y} & y > 0 \end{cases}$$

1. Check whether  $f_Y(y)$  is a *valid* density function.

We need to check the 2 properties of pdfs.

- (1)  $f_Y(y)$  is non-negative function on  $\mathbb{R}$
- (2)  $\int_{-\infty}^{\infty} f_Y(y) dy = 1$

$$\int_{-\infty}^{\infty} f_{Y}(y) dy =$$

2. What is the probability that the  $1^{st}$  major disk drive failure occurs within the first year?

$$\mathbb{P}(Y \leq 1) =$$

3. What is the probability that the 1<sup>st</sup> major disk drive failure occurs before the first year?

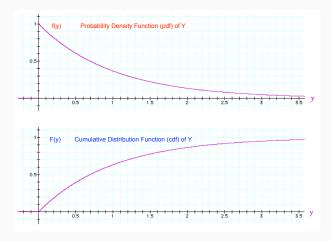
$$\mathbb{P}(Y < 1) =$$

4. What is the probability that the 1<sup>st</sup> major disk drive failure occurs after the first year?

5. What is the probability that the  $1^{st}$  major disk drive failure occurs after first year but before second year?

6. What is the cumulative distribution function (cdf) of Y?

For Example 1, the pdf and cdf of Y are shown below.



SHORT CUT: Use the cdf to calculate desired probabilities instead of integrating the pdf for each problem.

- Only need to integrate the pdf once to obtain the cdf
- Write any probability in terms of the cdf and plug in to solve

#### Back to Example 1:

- $\mathbb{P}(Y \leq 1) =$
- $\mathbb{P}(Y > 1) =$
- $\mathbb{P}(1 < Y < 2) =$

## **Additional Example**

Example 2: Let X be a random variable with the following probability density function (pmf):

$$f_X(x) = \begin{cases} 0 & \text{for } x < 0\\ \frac{1}{2} & \text{for } 0 \le x \le \frac{1}{2}\\ 2x & \text{for } \frac{1}{2} < x < 1\\ 0 & \text{for } x \ge 1 \end{cases}$$

### **Additional Example Cont.**

1. Give the cumulative distribution function (cdf) of X



### **Additional Example Cont.**

2. What is the probability that X is less that 0.75?

# Summary of Discrete & Continuous R.V.

#### Discrete R.V.

- Im(X) finite or countable infinite
- CDF:  $F_X(t) = \mathbb{P}(X \le t)$ =  $\sum_{x \le t} p_X(x)$
- PMF:  $p_X(x) = \mathbb{P}(X = x)$
- $\mathbb{E}(h(X)) = \sum_{x} h(x) p_X(x)$
- $\mathbb{E}(X) = \sum_{x} x p_X(x)$
- $Var(X) = \mathbb{E}(X^2) [\mathbb{E}(X)]^2$

#### Continuous R.V.

- Im(X) uncountable
- CDF:  $F_X(t) = \mathbb{P}(X \le t)$ =  $\int_{-\infty}^t f_X(x) dx$
- PDF:  $f_X(x) = \frac{d}{dx} F_X(x)$
- $\mathbb{E}(h(X)) = \int_X h(x) f_X(x) dx$
- $\mathbb{P}(X) = \int_{-\infty}^{\infty} x f_X(x) dx$
- $Var(X) = \mathbb{E}(X^2) [\mathbb{E}(X)]^2$

#### Recap

Students should now be familiar with continuous random variables. They should know how to obtain a cdf and use the pdf or cdf to calculate probabilities for a continuous random variable. They should be aware of how to calculate expectation and variance in the continuous case.