# **Def 3.1 Cumulative Distribution Function**



- The CDF of a RV X is  $F_X(x) = P[X \le x]$
- Properties of a CDF are the same for all kinds of random variables
  - Discrete,
  - Continuous, and
  - Mixed.

#### Theorem 3.1

For any random variable X,

(a) 
$$F_X(-\infty) = 0$$
  
(b)  $F_X(\infty) = 1$   $f(X \le \infty)$   
(c)  $P[x_1 < X \le x_2] = F_X(x_2) - F_X(x_1)$ 

## CDF Discrete vs. Continuous

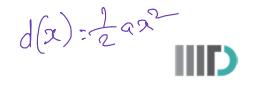


The CDF of a discrete RV contained jumps

• The size of the jump (change) was the *probability* of the RV at the point of jump (given by the PMF)

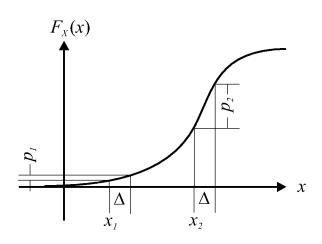
 The larger the jump, the larger the accumulation of probability at the point in the range space

# CDF Discrete vs. Continuous



The CDF of a continuous RV is continuous

## Figure 3.2



The graph of an arbitrary CDF  $F_X(x)$ .

## **Quiz 3.1**

The cumulative distribution function of the random variable Y is

$$F_Y(y) = \begin{cases} 0 & y < 0, \\ y/4 & 0 \le y \le 4, \\ 1 & y > 4. \end{cases}$$
 (3.9)

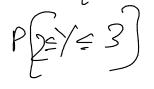
Sketch the CDF of *Y* and calculate the following probabilities:

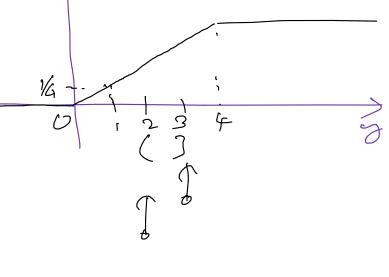
(1) 
$$P[Y \le -1]$$

(2) 
$$P[Y \le 1]$$

(3) 
$$P[2 < Y \le 3] = P\left( \cancel{Y} \le \cancel{S} \right) - P\left( \cancel{Y} \le \cancel{S} \right)$$

(4) 
$$P[Y > 1.5]$$

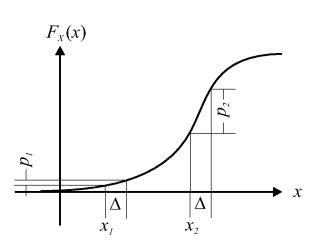




# Example CDF of a continuous RV



## Figure 3.2



$$P\left(x_2 < X \leq x_2 + D\right)$$

$$P\left(x_2 < X \leq x_3 + D\right)$$

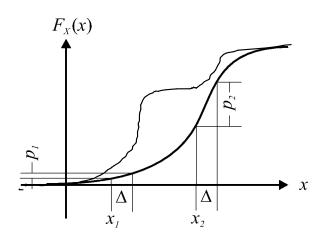
The graph of an arbitrary CDF  $F_X(x)$ .

- No jumps anymore
- Instead, we have rates of accumulation of probability

## Information in a CDF



## Figure 3.2



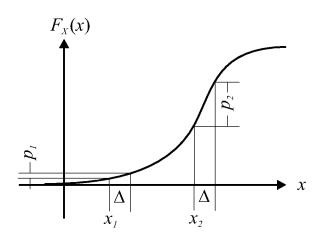
The graph of an arbitrary CDF  $F_X(x)$ .

$$F_X(x_2+\Delta)-F_X(x_2)>F_X(x_1+\Delta)-F_X(x_1)$$
• What is the LHS and the RHS?

## Information in a CDF



## Figure 3.2



The graph of an arbitrary CDF  $F_X(x)$ .

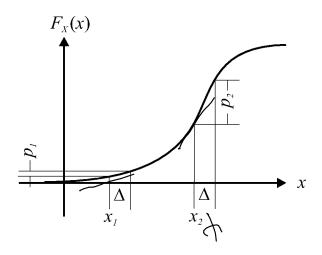
$$F_X(x_2 + \Delta) - F_X(x_2) > F_X(x_1 + \Delta) - F_X(x_1)$$

• Rate of accumulation of probability in the region right of  $x_2$  is greater than in the region right of  $x_1$ 

## Information in a CDF



## Figure 3.2



The graph of an arbitrary CDF  $F_X(x)$ .

• As the interval  $\Delta$  becomes smaller the rates of accumulation -> slope of the CDF

# **Probability Density Function**



 Greater the slope of the CDF at any point x, the more the likelihood of occurrence of the interval around x

## **Probability Density Function**

## Definition 3.3 (PDF)

The probability density function (PDF) of a continuous random variable X is

$$f_X(x) = \frac{dF_X(x)}{dx}.$$
Note the lowercase f



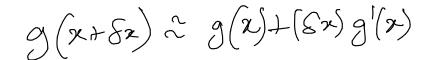
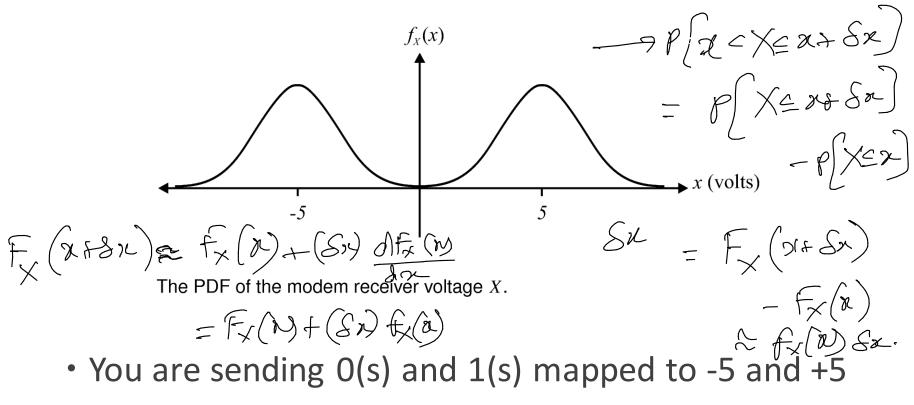




Figure 3.3



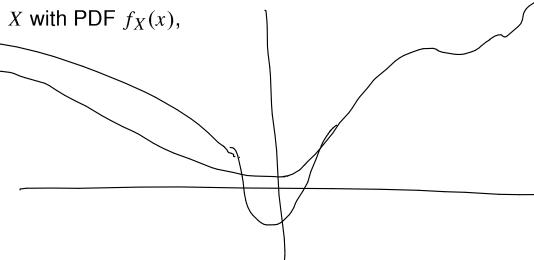
- Receiver sees the above PDF
- Which regions are most likely to be seen by the receiver and why?



#### **Theorem 3.2**

For a continuous random variable X with PDF  $f_X(x)$ ,

- (a)  $f_X(x) \ge 0$  for all x,
- (b)  $F_X(x) = \int_{-\infty}^x f_X(u) du$ ,
- (c)  $\int_{-\infty}^{\infty} f_X(x) \, dx = 1.$



· Why?



#### **Theorem 3.2**

For a continuous random variable X with PDF  $f_X(x)$ ,

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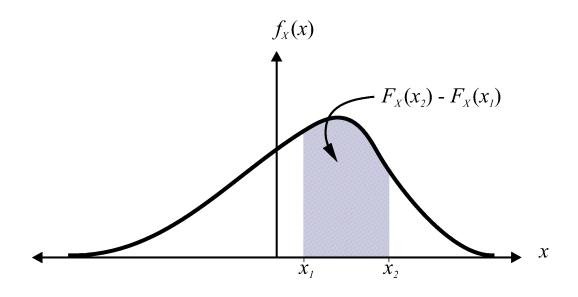
(c) 
$$\int_{-\infty}^{\infty} f_X(x) \, dx = 1.$$

- Why is (a) true?
  - CDF is a non-decreasing function and PDF is its slope
- Why (b)?
  - From definition of a PDF
- Why (c)?
  - The integral is  $P[X < \infty] = F_x(\infty) = 1$ .



## Theorem 3.3

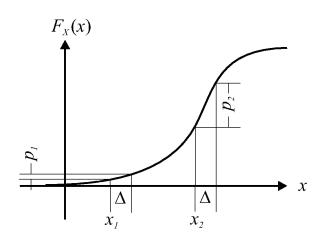
$$P[x_1 < X \le x_2] = \int_{x_1}^{x_2} f_X(x) \ dx.$$



The PDF and CDF of X.



## Figure 3.2



The graph of an arbitrary CDF  $F_X(x)$ .

As the interval  $\Delta$  becomes smaller:

$$P[x_2 < X \le x_2 + \Delta] \stackrel{\mathcal{F}_X(x_2 + \Delta) - F_X(x_2)}{\Delta} \Delta$$

# Intervals and CRV(s)



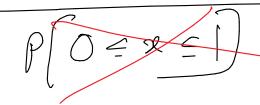
Consider the four different events

• 
$$C = [0,1)$$

• D = [0,1] 
$$\longrightarrow$$
  $P(0) = P(0 \leq X \leq 1)$ 

- They belong to the range space of a continuous RV
- What can we say about P[A], P[B], P[C], and P[D]?





## Problem on PDF



### **Quiz 3.2**

Random variable *X* has probability density function

$$f_X(x) = \begin{cases} cxe^{-x/2} & x \ge 0, \\ 0 & \text{otherwise.} \end{cases}$$
 (3.25)

Sketch the PDF and find the following:

- (1) the constant c
- (2) the CDF  $F_X(x)$
- (3)  $P[0 \le X \le 4]$
- (4)  $P[-2 \le X \le 2]$
- (5) What is the mode of the random variable X?

## **Problem 3.1.3**

The CDF of random variable W is

$$F_{W}(w) = \begin{cases} 0 & w < -5, & \text{if } w < -3, & \text{if } w <$$

b) What is 
$$P[-2 < W \le 2]$$
?  $F_{\mathcal{W}}(2) - F_{\mathcal{W}}(-2)$ 

- (c) What is P[W > 0]?  $| F_{\mathcal{W}}(\sigma)$
- (d) What is the value of  $\underline{a}$  such that  $P[W \le a] = 1/2$ ?

## Problem 3.2.4

For a constant parameter a > 0, a Rayleigh random variable X has PDF

$$f_X(x) = \begin{cases} a^2 x e^{-a^2 x^2/2} & x > 0, \\ 0 & \text{otherwise.} \end{cases}$$
f X?

What is the CDF of *X*?

$$f_X(x) = \begin{cases} a^2 x e^{-a^2 x^2/2} & x > 0, \\ 0 & \text{otherwise.} \end{cases}$$
of  $X$ ?
$$f_X(x) = \begin{cases} a^2 x e^{-a^2 x^2/2} & x > 0, \\ 0 & \text{otherwise.} \end{cases}$$

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# **Expected Value**



 Def 3.4 The expected value of a continuous random variable X is

$$E[X] = \int_{-\infty}^{\infty} x f_X(x) \, dx$$

- Summation for the discrete case is replaced by integration for the continuous case
- Theorem 3.4 The expected value of a function g(X) of a continuous random variable X is

$$E[X] = \int_{-\infty}^{\infty} g(x) f_X(x) dx$$

# These Equalities are Valid for Both Continuous and Discrete RV(s)



## **Theorem 3.5**

For any random variable X,

(a) 
$$E[X - \mu_X] =$$

(b) 
$$E[aX + b] =$$

(c) 
$$Var[X] =$$

(d) 
$$Var[aX + b] =$$

$$E((X-f_X)^T)$$

$$g(X)=(X-f_X)$$