

## Chapter 3: Random Variables and Probability

### Distributions:

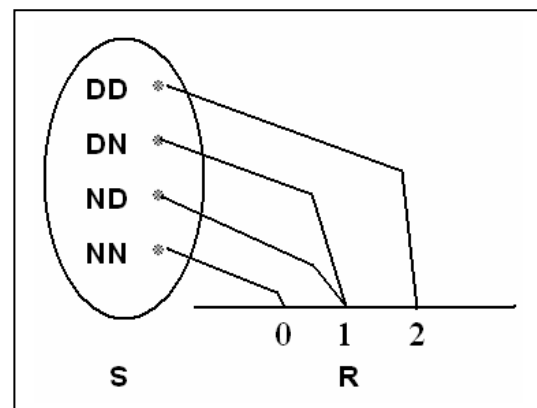
#### 3.1 Concept of a Random Variable:

- In a statistical experiment, it is often very important to allocate numerical values to the outcomes.

#### **Example:**

- Experiment: testing two components.  
(D=defective, N=non-defective)
- Sample space:  $S = \{DD, DN, ND, NN\}$
- Let  $X =$  number of defective components when two components are tested.
- Assigned numerical values to the outcomes are:

Sample point (Outcome)	Assigned Numerical Value (x)
DD	2
DN	1
ND	1
NN	0



- Notice that, the set of all possible values of the random variable  $X$  is  $\{0, 1, 2\}$ .

#### **Definition 3.1:**

A random variable  $X$  is a function that associates each element in the sample space with a real number (i.e.,  $X : S \rightarrow \mathbf{R}$ .)

**Notation:** "  $X$  " denotes the random variable .

"  $x$  " denotes a value of the random variable  $X$ .

#### **Types of Random Variables:**

- A random variable  $X$  is called a **discrete** random variable if its set of possible values is countable, i.e.,  
 $x \in \{x_1, x_2, \dots, x_n\}$  OR  $x \in \{x_1, x_2, \dots\}$
- A random variable  $X$  is called a **continuous** random variable if it can take values on a continuous scale, i.e.,  
 $x \in \{x: a < x < b; a, b \in \mathbf{R}\}$

- In most practical problems:
  - A discrete random variable represents count data, such as the number of defectives in a sample of  $k$  items.
  - A continuous random variable represents measured data, such as height.

### 3.2 Discrete Probability Distributions

- A discrete random variable  $X$  assumes each of its values with a certain probability.

#### **Example:**

- Experiment: tossing a non-balance coin 2 times independently.
- $H$ = head ,  $T$ =tail
- Sample space:  $S=\{HH, HT, TH, TT\}$
- Suppose  $P(H)=\frac{1}{2}P(T) \Leftrightarrow P(H)=\frac{1}{3}$  and  $P(T)=\frac{2}{3}$
- Let  $X$ = number of heads

Sample point (Outcome)	Probability	Value of X (x)
HH	$P(HH)=P(H) P(H)=\frac{1}{3}\times\frac{1}{3} = \frac{1}{9}$	2
HT	$P(HT)=P(H) P(T)=\frac{1}{3}\times\frac{2}{3} = \frac{2}{9}$	1
TH	$P(TH)=P(T) P(H)=\frac{2}{3}\times\frac{1}{3} = \frac{2}{9}$	1
TT	$P(TT)=P(T) P(T)=\frac{2}{3}\times\frac{2}{3} = \frac{4}{9}$	0

- The possible values of  $X$  are: 0, 1, and 2.
- $X$  is a discrete random variable.
- Define the following events:

Event ( $X=x$ )	Probability = $P(X=x)$
$(X=0)=\{TT\}$	$P(X=0) = P(TT)=\frac{4}{9}$
$(X=1)=\{HT,TH\}$	$P(X=1) =P(HT)+P(TH)=\frac{2}{9}+\frac{2}{9}=\frac{4}{9}$
$(X=2)=\{HH\}$	$P(X=2) = P(HH)= \frac{1}{9}$

- The possible values of  $X$  with their probabilities are:

x	0	1	2	Total
$P(X=x)=f(x)$	$\frac{4}{9}$	$\frac{4}{9}$	$\frac{1}{9}$	1.00

The function  $f(x)=P(X=x)$  is called the probability function (probability distribution) of the discrete random variable  $X$ .

#### **Definition 3.4:**

The function  $f(x)$  is a probability function of a discrete random variable  $X$  if, for each possible values  $x$ , we have:

1.  $f(x) \geq 0$
2.  $\sum_{all\ x} f(x) = 1$
3.  $f(x) = P(X=x)$

**Note:**

$$\bullet P(X \in A) = \sum_{\text{all } x \in A} f(x) = \sum_{\text{all } x \in A} P(X = x)$$

**Example:**

For the previous example, we have:

x	0	1	2	Total
f(x)= P(X=x)	4/9	4/9	1/9	$\sum_{x=0}^2 f(x) = 1$

$$P(X < 1) = P(X=0) = 4/9$$

$$P(X \leq 1) = P(X=0) + P(X=1) = 4/9 + 4/9 = 8/9$$

$$P(X \geq 0.5) = P(X=1) + P(X=2) = 4/9 + 1/9 = 5/9$$

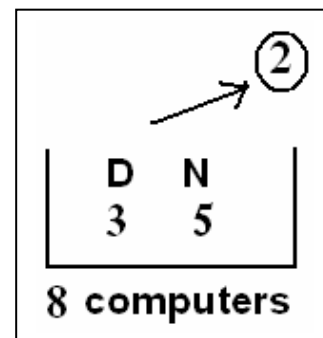
$$P(X > 8) = P(\phi) = 0$$

$$P(X < 10) = P(X=0) + P(X=1) + P(X=2) = P(S) = 1$$

**Example 3.3:**

A shipment of 8 similar microcomputers to a retail outlet contains 3 that are defective and 5 are non-defective.

If a school makes a random purchase of 2 of these computers, find the probability distribution of the number of defectives.

**Solution:**

We need to find the probability distribution of the random variable:  $X =$  the number of defective computers purchased.

Experiment: selecting 2 computers at random out of 8

$$n(S) = \binom{8}{2} \text{ equally likely outcomes}$$

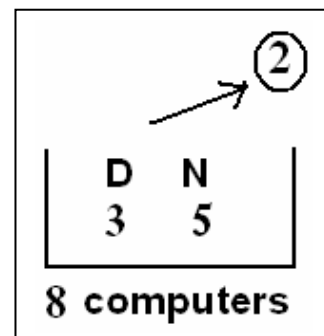
The possible values of  $X$  are:  $x=0, 1, 2$ .

Consider the events:

$$(X=0) = \{0D \text{ and } 2N\} \Rightarrow n(X=0) = \binom{3}{0} \times \binom{5}{2}$$

$$(X=1) = \{1D \text{ and } 1N\} \Rightarrow n(X=1) = \binom{3}{1} \times \binom{5}{1}$$

$$(X=2) = \{2D \text{ and } 0N\} \Rightarrow n(X=2) = \binom{3}{2} \times \binom{5}{0}$$



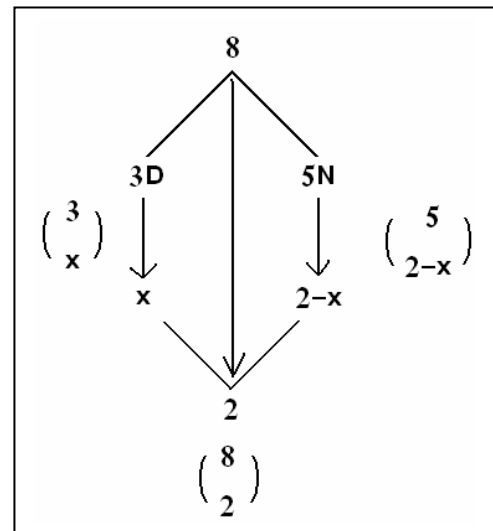
$$f(0)=P(X=0)=\frac{n(X=0)}{n(S)}=\frac{\binom{3}{0}\times\binom{5}{2}}{\binom{8}{2}}=\frac{10}{28}$$

$$f(1)=P(X=1)=\frac{n(X=1)}{n(S)}=\frac{\binom{3}{1}\times\binom{5}{1}}{\binom{8}{2}}=\frac{15}{28}$$

$$f(2)=P(X=2)=\frac{n(X=2)}{n(S)}=\frac{\binom{3}{2}\times\binom{5}{0}}{\binom{8}{2}}=\frac{3}{28}$$

In general, for  $x=0,1,2$ , we have:

$$f(x)=P(X=x)=\frac{n(X=x)}{n(S)}=\frac{\binom{3}{x}\times\binom{5}{2-x}}{\binom{8}{2}}$$



The probability distribution of  $X$  can be given in the following table:

$x$	0	1	2	Total
$f(x)=P(X=x)$	$\frac{10}{28}$	$\frac{15}{28}$	$\frac{3}{28}$	1.00

The probability distribution of  $X$  can be written as a formula as follows:

$$f(x)=P(X=x)=\begin{cases} \frac{\binom{3}{x}\times\binom{5}{2-x}}{\binom{8}{2}}; & x=0,1,2 \\ 0; & \text{otherwise} \end{cases}$$

Hypergeometric  
Distribution

**Definition 3.5:**

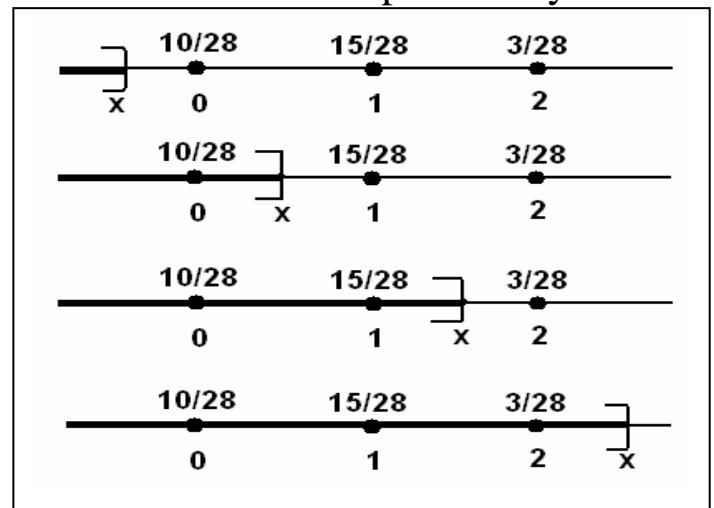
The cumulative distribution function (CDF),  $F(x)$ , of a discrete random variable  $X$  with the probability function  $f(x)$  is given by:

$$F(x) = P(X \leq x) = \sum_{t \leq x} f(t) = \sum_{t \leq x} P(X = t) ; \text{ for } -\infty < x < \infty$$

**Example:**

Find the CDF of the random variable  $X$  with the probability function:

$x$	0	1	2
$f(x)$	$\frac{10}{28}$	$\frac{15}{28}$	$\frac{3}{28}$

**Solution:**

$$F(x) = P(X \leq x) \text{ for } -\infty < x < \infty$$

$$\text{For } x < 0: F(x) = 0$$

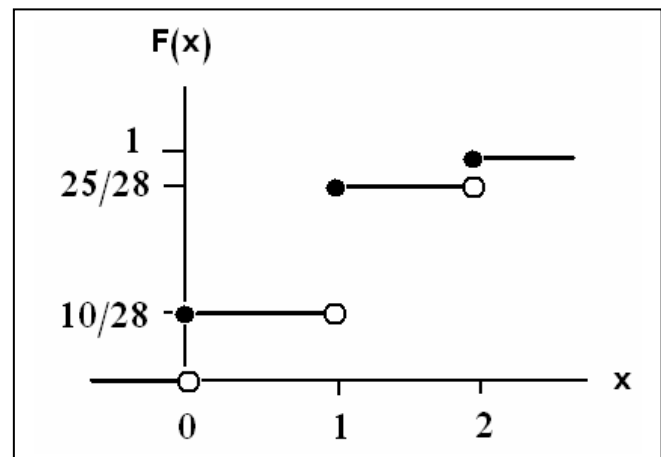
$$\text{For } 0 \leq x < 1: F(x) = P(X=0) = \frac{10}{28}$$

$$\text{For } 1 \leq x < 2: F(x) = P(X=0) + P(X=1) = \frac{10}{28} + \frac{15}{28} = \frac{25}{28}$$

$$\text{For } x \geq 2: F(x) = P(X=0) + P(X=1) + P(X=2) = \frac{10}{28} + \frac{15}{28} + \frac{3}{28} = 1$$

The CDF of the random variable  $X$  is:

$$F(x) = P(X \leq x) = \begin{cases} 0 & ; x < 0 \\ \frac{10}{28} & ; 0 \leq x < 1 \\ \frac{25}{28} & ; 1 \leq x < 2 \\ 1 & ; x \geq 2 \end{cases}$$



Note:

$$F(-0.5) = P(X \leq -0.5) = 0$$

$$F(1.5) = P(X \leq 1.5) = F(1) = \frac{25}{28}$$

$$F(3.8) = P(X \leq 3.8) = F(2) = 1$$

Another way to find  $F(x)$ :

x	< 0	0	1	2	> 2
f(x)		$\frac{10}{28}$	$\frac{15}{28}$	$\frac{3}{28}$	
F(x)	0	$\frac{10}{28}$	$\frac{25}{28}$	$\frac{28}{28}$	1

$$F(x) = \begin{cases} 0 & ; x < 0 \\ \frac{10}{28} & ; 0 \leq x < 1 \\ \frac{25}{28} & ; 1 \leq x < 2 \\ 1 & ; x \geq 2 \end{cases}$$

**Result:**

$$P(a < X \leq b) = P(X \leq b) - P(X \leq a) = F(b) - F(a)$$

$$P(a \leq X \leq b) = P(a < X \leq b) + P(X=a) = F(b) - F(a) + f(a)$$

$$P(a < X < b) = P(a < X \leq b) - P(X=b) = F(b) - F(a) - f(b)$$

**Result:**

Suppose that the probability function of  $X$  is:

x	$x_1$	$x_2$	$x_3$	...	$x_n$
f(x)	$f(x_1)$	$f(x_2)$	$f(x_3)$	...	$f(x_n)$

Where  $x_1 < x_2 < \dots < x_n$ . Then:

$$F(x_i) = f(x_1) + f(x_2) + \dots + f(x_i) \quad ; \quad i=1, 2, \dots, n$$

$$F(x_i) = F(x_{i-1}) + f(x_i) \quad ; \quad i=2, \dots, n$$

$$f(x_i) = F(x_i) - F(x_{i-1})$$

**Example:**

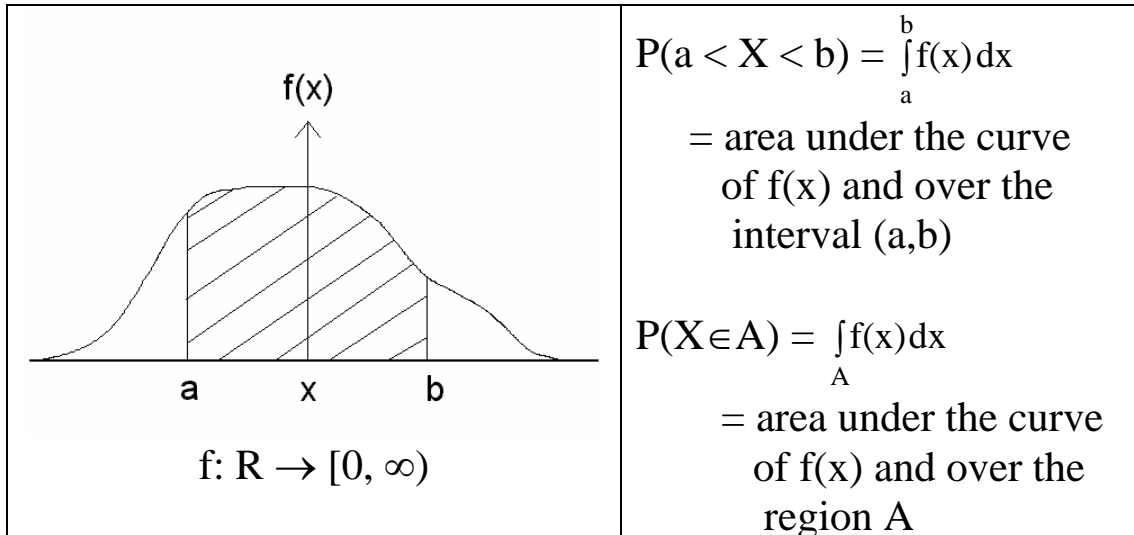
In the previous example,

$$P(0.5 < X \leq 1.5) = F(1.5) - F(0.5) = \frac{25}{28} - \frac{10}{28} = \frac{15}{28}$$

$$P(1 < X \leq 2) = F(2) - F(1) = 1 - \frac{25}{28} = \frac{3}{28}$$

### 3.3. Continuous Probability Distributions

For any continuous random variable,  $X$ , there exists a non-negative function  $f(x)$ , called the probability density function (p.d.f) through which we can find probabilities of events expressed in term of  $X$ .



#### Definition 3.6:

The function  $f(x)$  is a probability density function (pdf) for a continuous random variable  $X$ , defined on the set of real numbers, if:

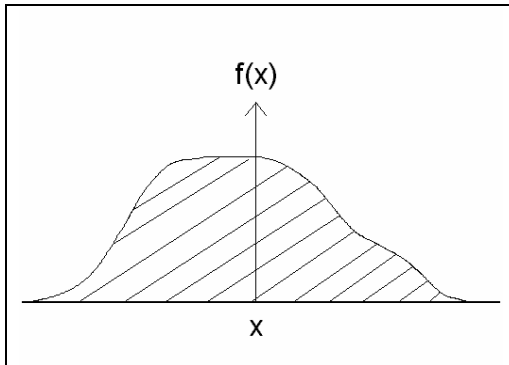
1.  $f(x) \geq 0 \quad \forall x \in \mathbb{R}$
2.  $\int_{-\infty}^{\infty} f(x) dx = 1$
3.  $P(a \leq X \leq b) = \int_a^b f(x) dx \quad \forall a, b \in \mathbb{R}; a \leq b$

Note:

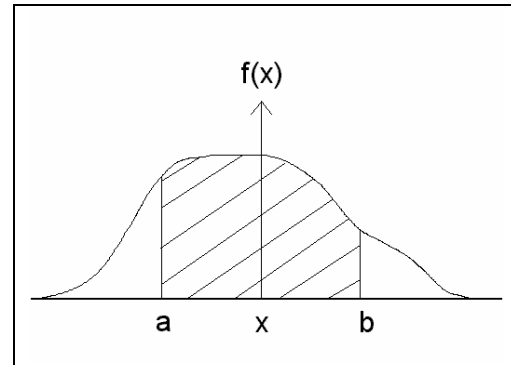
For a continuous random variable  $X$ , we have:

1.  $f(x) \neq P(X=x)$  (in general)
2.  $P(X=a) = 0$  for any  $a \in \mathbb{R}$
3.  $P(a \leq X \leq b) = P(a < X \leq b) = P(a \leq X < b) = P(a < X < b)$
4.  $P(X \in A) = \int_A f(x) dx$



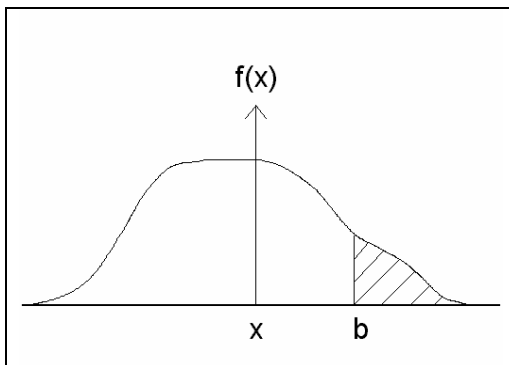


$$\text{Total area} = \int_{-\infty}^{\infty} f(x) dx = 1$$



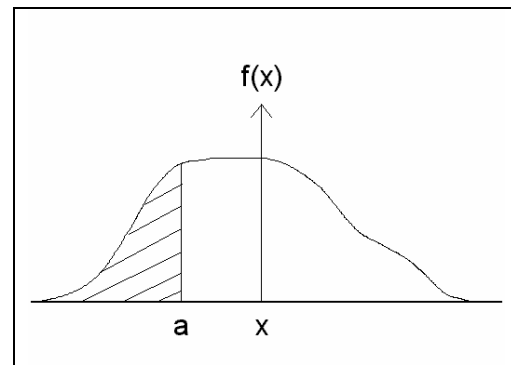
$$\text{area} = P(a \leq X \leq b)$$

$$= \int_a^b f(x) dx$$



$$\text{area} = P(X \geq b)$$

$$= \int_b^{\infty} f(x) dx$$



$$\text{area} = P(X \leq a)$$

$$= \int_{-\infty}^a f(x) dx$$

### Example 3.6:

Suppose that the error in the reaction temperature, in  $^{\circ}\text{C}$ , for a controlled laboratory experiment is a continuous random variable  $X$  having the following probability density function:

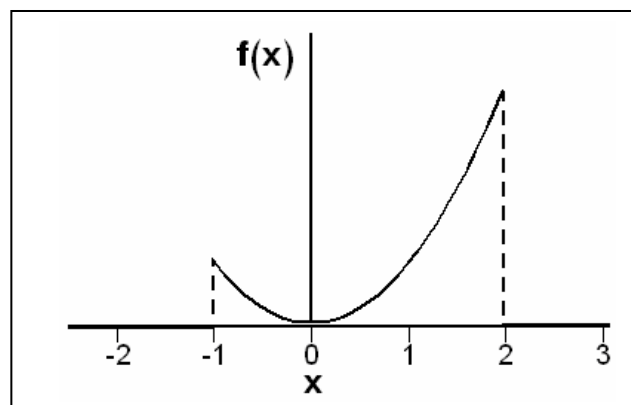
$$f(x) = \begin{cases} \frac{1}{3}x^2 & ; -1 < x < 2 \\ 0 & ; \text{elsewhere} \end{cases}$$

1. Verify that:

(a)  $f(x) \geq 0$

(b)  $\int_{-\infty}^{\infty} f(x) dx = 1$

2. Find  $P(0 < X \leq 1)$



### Solution:

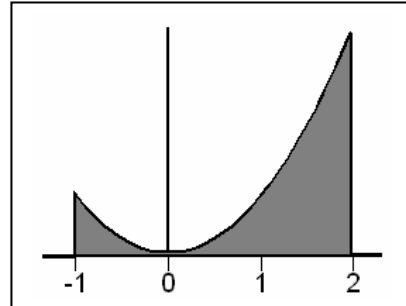
$X$  = the error in the reaction temperature in  $^{\circ}\text{C}$ .

$X$  is continuous r. v.

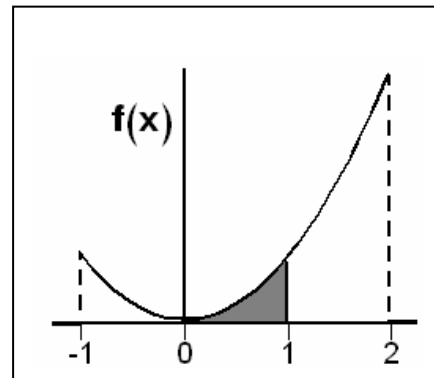
$$f(x) = \begin{cases} \frac{1}{3}x^2 & ; -1 < x < 2 \\ 0 & ; \text{elsewhere} \end{cases}$$

1. (a)  $f(x) \geq 0$  because  $f(x)$  is a quadratic function.

$$\begin{aligned} \text{(b)} \quad \int_{-\infty}^{\infty} f(x) dx &= \int_{-\infty}^{-1} 0 dx + \int_{-1}^2 \frac{1}{3}x^2 dx + \int_2^{\infty} 0 dx \\ &= \int_{-1}^2 \frac{1}{3}x^2 dx = \left[ \frac{1}{9}x^3 \right]_{x=-1}^{x=2} \\ &= \frac{1}{9}(8 - (-1)) = 1 \end{aligned}$$



$$\begin{aligned} 2. \quad P(0 < X \leq 1) &= \int_0^1 f(x) dx = \int_0^1 \frac{1}{3}x^2 dx \\ &= \left[ \frac{1}{9}x^3 \right]_{x=0}^{x=1} \\ &= \frac{1}{9}(1 - (0)) \\ &= \frac{1}{9} \end{aligned}$$



### Definition 3.7:

The cumulative distribution function (CDF),  $F(x)$ , of a continuous random variable  $X$  with probability density function  $f(x)$  is given by:

$$F(x) = P(X \leq x) = \int_{-\infty}^x f(t) dt ; \text{ for } -\infty < x < \infty$$

### Result:

$$P(a < X \leq b) = P(X \leq b) - P(X \leq a) = F(b) - F(a)$$

### Example:

in Example 3.6,

1. Find the CDF
2. Using the CDF, find  $P(0 < X \leq 1)$ .

### Solution:

$$f(x) = \begin{cases} \frac{1}{3}x^2 & ; -1 < x < 2 \\ 0 & ; \text{elsewhere} \end{cases}$$

(1) Finding  $F(x)$ :

$$F(x) = P(X \leq x) = \int_{-\infty}^x f(t) dt ; \text{ for } -\infty < x < \infty$$

For  $x < -1$ :

$$F(x) = \int_{-\infty}^x f(t) dt = \int_{-\infty}^x 0 dt = 0$$

For  $-1 \leq x < 2$ :

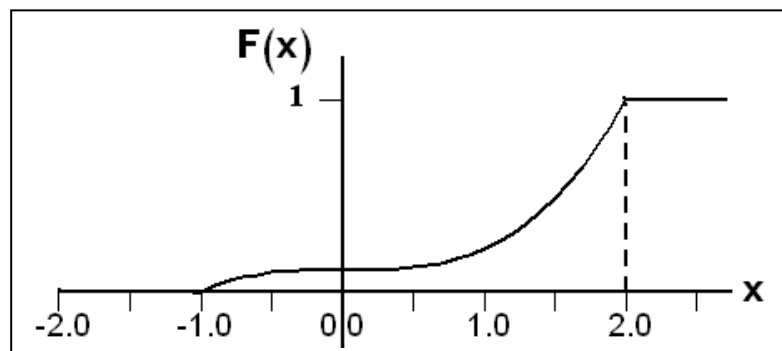
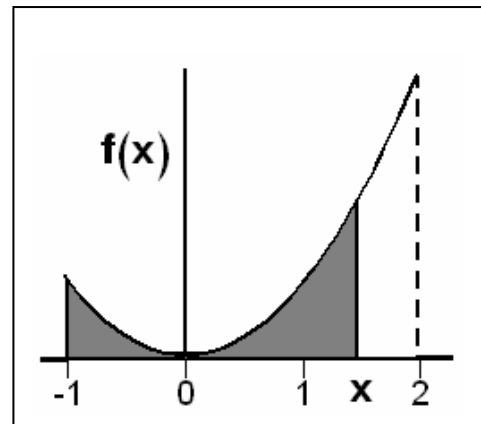
$$\begin{aligned} F(x) &= \int_{-\infty}^x f(t) dt = \int_{-\infty}^{-1} 0 dt + \int_{-1}^x \frac{1}{3} t^2 dt \\ &= \int_{-1}^x \frac{1}{3} t^2 dt \\ &= \left[ \frac{1}{9} t^3 \right]_{t=-1}^{t=x} = \frac{1}{9} (x^3 - (-1)) = \frac{1}{9} (x^3 + 1) \end{aligned}$$

For  $x \geq 2$ :

$$F(x) = \int_{-\infty}^x f(t) dt = \int_{-\infty}^{-1} 0 dt + \int_{-1}^2 \frac{1}{3} t^2 dt + \int_2^x 0 dt = \int_{-1}^2 \frac{1}{3} t^2 dt = 1.$$

Therefore, the CDF is:

$$F(x) = P(X \leq x) = \begin{cases} 0 & ; x < -1 \\ \frac{1}{9} (x^3 + 1) & ; -1 \leq x < 2 \\ 1 & ; x \geq 2 \end{cases}$$



2. Using the CDF,

$$P(0 < X \leq 1) = F(1) - F(0) = \frac{2}{9} - \frac{1}{9} = \frac{1}{9}$$