

X Lecture 2, Sept. 15, 04

Random Variables

A random variable X is a function that assigns a real number ~~to each~~ $X(\omega)$ to each outcome of a random experiment $\omega \in S$.

S is the domain and $S_X = \{X(\omega) : \omega \in S\}$ is the range of the random variable X .

Example:

A binary communication system with

$$S = \{0, 1\} \text{ and } X(\omega) \in \{-A, +A\}$$

where $0 \rightarrow -A$ and $1 \rightarrow +A$, i.e.,

$$S_X = \{-A, A\}$$

CDF (Cumulative Distribution Function)

$$F_X(x) = P[X \leq x] \quad -\infty < x < \infty$$

Properties of $F_X(x)$

- 1) $0 \leq F_X(x) \leq 1$
- 2) $\lim_{x \rightarrow \infty} F_X(x) = 1$
- 3) $\lim_{x \rightarrow -\infty} F_X(x) = 0$

4) $F_X(x)$ is a nondecreasing function of x , i.e.,
if $a < b$ then $F_X(a) \leq F_X(b)$

5) $F_X(x)$ is continuous from right, i.e.,

$$\text{for } h > 0 \quad F_X(b) = \lim_{h \rightarrow 0} F_X(b+h) = F_X(b^+).$$

6)

$$P(a < X \leq b) = F_X(b) - F_X(a) \quad \text{for } b \geq a$$

Proof:

$$\{X \leq a\} \cup \{a < X \leq b\} = \{X \leq b\}$$

and

$$\{X \leq a\} \cap \{a < X \leq b\} = \emptyset$$

So

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

$$F_X(a) + P(a < X \leq b) = F_X(b)$$

$$\boxed{P(a < X \leq b) = F_X(b) - F_X(a)} \quad \text{Q.E.D.}$$

7) $P(X = b) = F_X(b) - F_X(b^-)$

~~~~~

Example: Consider throwing a dice and let

$$S = \{1, 2, \dots, 6\} = S_x$$

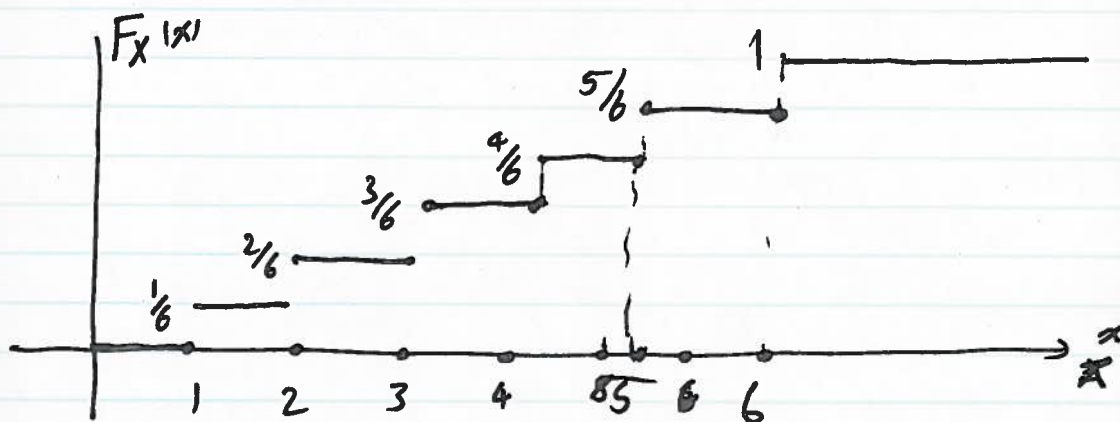
~~$$F_x(x) = 0 \quad x < 1$$~~

$$F_x(x) = \frac{1}{6} \quad 1 \leq x < 2$$

$$F_x(x) = \frac{2}{6} \quad 2 \leq x < 3$$

⋮

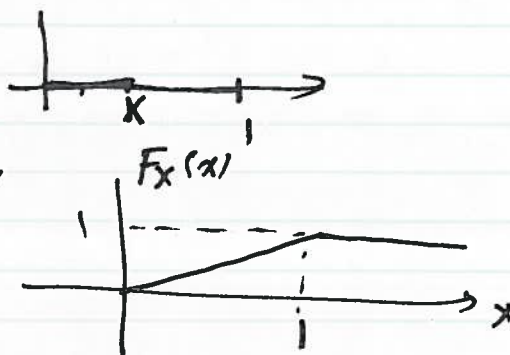
$$F_x(x) = 1 \quad x \geq 6$$



Example:

Picking a number between 0 and 1 uniformly (without prejudice).

$$F_x(x) = P(X \leq x) = \frac{x}{1} = x$$



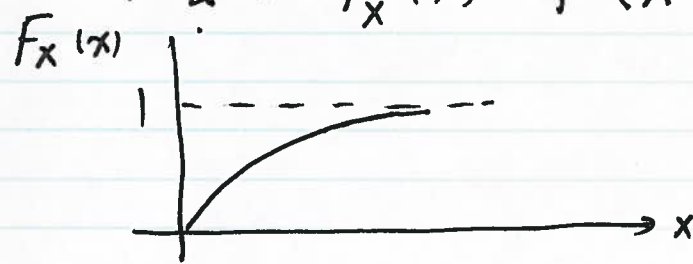
Example:

~~The transmission time of a message is distributed~~

The life<sup>time</sup> of a device, say, a light bulb, is distributed exponentially, i.e.,

$$P[X > x] = e^{-\lambda x}$$

So  ~~$F_X(x)$~~   $F_X(x) = P(X \leq x) = 1 - e^{-\lambda x}$



Discrete, Continuous and Mixed Random Variables

Discrete random variables take values from a finite or at most countably infinite set.

e.g.,  $\{1, 2, \dots, 6\}$ ,  $\{0, 1\}$ ,  $\{1, 2, \dots, \dots\}$

i.e.,

$$S_X = \{x_0, x_1, \dots\}$$

The CDF of a discrete random variable is given as

$$F_X(x) = \sum_k P_X(x_k) u(x - x_k)$$

where  $P_X(x_k)$  is the probability mass function of  $X$

given as

$$P_X(x_k) = P(X=x_k)$$

as an example for  $S_X = \{1, 2, \dots, 6\}$  with

$$P_X(x_k) = \frac{1}{6} \quad \forall k \in \{1, \dots, 6\}$$

we have

~~$$F_X(x) = \sum_{k=1}^n P_X(x_k)$$~~

$$F_X(x) = \frac{1}{6} \mathbb{1}(x-1) + \frac{1}{6} \mathbb{1}(x-2) + \dots + \frac{1}{6} \mathbb{1}(x-6)$$

A Continuous Random Variable has a CDF that is continuous everywhere.

A mixed R.V. has a cdf that is continuous over certain values but jumps at certain points.

The CDF for a mixed R.V. can be written as

$$F_X(x) = p F_1(x) + (1-p) F_2(x)$$

where  $0 \leq p < 1$  and  $F_1(x)$ , and  $F_2(x)$  are CDF's that are discrete, and continuous, respectively.



The probability density function :

pdf is defined as

$$f_x(x) = \frac{dF_x(x)}{dx}$$

by definition

$$F_x(x) = \int_{-\infty}^x f_x(t) dt$$

$$P[x < X \leq x+h] = P[X \leq x+h] - P[X \leq x]$$

$$= \cancel{F_x(x+h)} + \cancel{F_x(x)} = F_x(x+h) - F_x(x)$$

$$= \int_{-\infty}^{x+h} f_x(t) dt - \int_{-\infty}^x f_x(t) dt$$

$$= \int_x^{x+h} f_x(t) dt = f_x(x)h \quad \text{for small } h$$

So

$$P[x \leq X \leq x+h] = f_x(x)h$$

Thus  $f_x(x)$  represents the "density" of the probability at point  $X=x$ .

Properties of pdf:

$$1) f_x(x) \geq 0$$

$$2) P[a \leq X \leq b] = \int_a^b f_X(x) dx$$

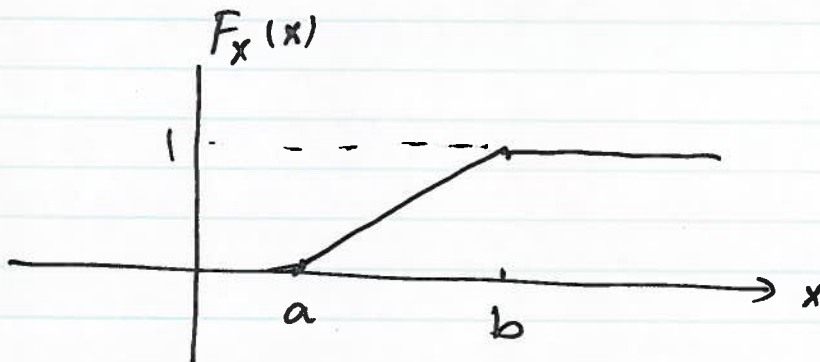
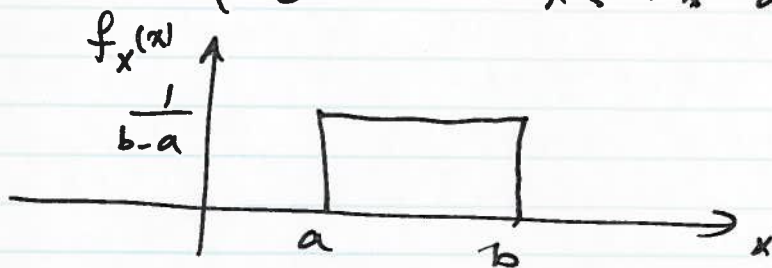
$$3) F_X(x) = \int_{-\infty}^x f_X(t) dt$$

$$4) \int_{-\infty}^{\infty} f_X(x) dx = 1$$

- Examples of some random variables and their pdf

1) Uniform random variable

$$f_X(x) = \begin{cases} \frac{1}{b-a} & a \leq x \leq b \\ 0 & x < a \text{ and } x > b \end{cases}$$



$$F_X(x) = \begin{cases} 0 & x < a \\ \frac{x-a}{b-a} & a \leq x \leq b \\ 1 & x > b \end{cases}$$

The pdf of the speech samples ~~are~~<sup>is</sup> given by:

$$f_x(x) = c e^{-\alpha|x|} \quad -\infty < x < \infty$$

a) Find the constant  $c$ .

b) Find  $P[|X| < v]$

$$\int_{-\infty}^{\infty} f_x(x) dx = 2 \int_0^{\infty} c e^{-\alpha x} dx = \frac{2c}{\alpha} = 1$$

$$c = \frac{\alpha}{2}$$

$$P[|X| < v] = \frac{\alpha}{2} \int_{-v}^v e^{-\alpha|x|} dx = \frac{\alpha}{2} \int_0^v e^{-\alpha x} dx$$

$$= 1 - e^{-\alpha v}$$

For a discrete random variable, the CDF is not continuous at some points  $\Rightarrow$  the derivative does not exist. To overcome this problem, we define the delta function  $\delta(x)$



which is the derivative of  $u(x)$

$$u(x) = \begin{cases} 0 & x < 0 \\ 1 & x \geq 0 \end{cases}$$

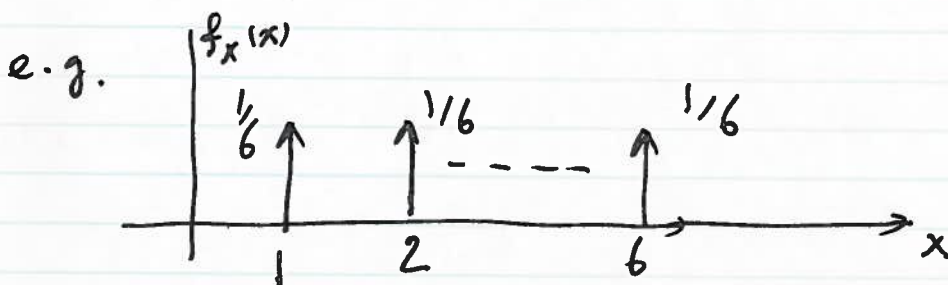
$$u(x) = \int_{-\infty}^{\infty} \delta(x) dt$$

For a discrete r.v., we have:

$$F_X(x) = \sum_k P_X(x_k) u(x - x_k)$$

So

$$f_X(x) = \sum_k P_X(x_k) \delta(x - x_k)$$



example:  $X = \#$  of heads in three coin tosses.

$$P_X(0) = \frac{1}{8} \quad TTT \quad P_X(1) = \frac{3}{8}, \quad P_X(2) = \frac{3}{8}, \quad P_X(3) = \frac{1}{8}$$

then:

$$f_X(x) = \frac{1}{8} \delta(x) + \frac{3}{8} \delta(x-1) + \frac{3}{8} \delta(x-2) + \frac{1}{8} \delta(x-3)$$

since:

$$F_X(x) = \frac{1}{8} u(x) + \frac{3}{8} u(x-1) + \frac{3}{8} u(x-2) + \frac{1}{8} u(x-3)$$

$$P[KX \leq 2] = \int_{1^-}^{2^+} f_X(x) dx = \frac{3}{8}$$

Conditional CDF and pdf

$$F_x(x|A) = P[X \leq x | A] = \frac{P[\{X \leq x\} \cap A]}{P[A]} \quad \text{if } P[A] > 0$$

$$f_x(x) = \frac{d}{dx} F_x(x|A)$$

example:

The life time of a device (a machine, a bulb) has CDF  $F_x(x)$ . Find the condition CDF and pdf given the event  $A = \{X > t\}$ , i.e., given that the device has survived until  $x = t$  <sup>assess</sup> ~~find~~ its future survivability.

$$\begin{aligned} F_x(x | X > t) &= P[X \leq x | X > t] \\ &= \frac{P[\{X \leq x\} \cap \{X > t\}]}{P[\{X > t\}]} \end{aligned}$$

a) if  $x < t \Rightarrow \{X \leq x\} \cap \{X > t\} = \emptyset \Rightarrow F_x(x | X > t) = 0$

b) if  $x \geq t \Rightarrow \{X \leq x\} \cap \{X > t\} = \{t < X \leq x\}$

$$\Rightarrow F_x(x) = \frac{F_x(x) - F_x(t)}{1 - F_x(t)}$$

So:

$$F_x(x | X > t) = \begin{cases} 0 & x < t \\ \frac{F_x(x) - F_x(t)}{1 - F_x(t)} & x > t \end{cases}$$

and

$$f_X(x|X>t) = \frac{f_X(x)}{1-F_X(t)} \quad x \geq t$$

### Well-known Random variables

$S_X = \{0, 1\}$     1) Bernoulli R.V.

$$P_X(0) = 1-p \quad \text{and} \quad P_X(1) = p$$

it can model any binary experiment, e.g.,  
tossing of a coin, value of a random bit,  
state of a device (working/broken).

let  $S$  be a sample space and  $A \subset S$   
be an event. Define

$$I_A(\omega) = \begin{cases} 0 & \text{if } \omega \notin A \\ 1 & \text{if } \omega \in A \end{cases} \quad \left| \quad I_A \text{ is called the} \right. \\ \left. \text{indicator of event } A. \right.$$

$I_A$  is ~~an~~ a random variable, since it  
assigns a number to each outcome of  $S$ .

$$P(A) = P(\omega \in A) = P(I=1) = p$$

So  $P_I(1) = p$      $P_I(0) = 1-p$

Memory

## 2) Binomial Distribution:

Assume we repeat a <sup>random</sup> experiment  $n$  times and

Let  $X$  be the number of times the event  $A$  occurs.

So,

$$S_X = \{0, 1, 2, \dots, n\}$$

Let  $I_j$  denote the indicator function of the event  $A$  in  $j$ -th trial  $0 \leq j \leq n$ . Then:

$$X = I_1 + I_2 + \dots + I_n$$

is the sum of Bernoulli trials.

$$P[X=k] = \binom{n}{k} p^k (1-p)^{n-k} \quad k=0, 1, \dots, n.$$

$X$  is called ~~the~~ <sup>the</sup> Binomial random variable.

## 3) Geometrical R. V.

example:

In an ARQ system, one sends a packet, waits for an Ack and resends if a NAK is received or in the case of time-out. Let the probability of successful transmission (no error in the packet) be  $p$ . Then the probability ~~that the message is~~ <sup>of</sup> ~~sent~~ number of ~~attempts~~ of transmission be  $k$  is

$$P[X=k] = (1-p)^{k-1} p \quad k=1, 2, \dots$$



This is called a Geometric R.V.

X Lecture 3: Sept. 22, 2004

4) Poisson Random Variable

$$P[N=k] = \frac{\alpha^k}{k!} e^{-\alpha} \quad k = 0, 1, 2, \dots$$

This models, e.g., the number of phone calls in a certain time interval or the number of packets needed to be transmitted in certain interval.  $\alpha$  is the average number of occurrences of that event in that time interval.

The Poisson probabilities can be derived from Binomial Distribution if we assume  $\alpha = np$  and keep  $\alpha$  fixed while letting  $n \rightarrow \infty$ , i.e.

$$p_k = \binom{n}{k} p^k (1-p)^{n-k} \rightarrow \frac{\alpha^k}{k!} e^{-\alpha} \quad k = 0, 1, \dots$$