

## 2.1 The Random variable Concept

- a real random variable is a real function of the elements of a sample space  $S$ .
- Thus, this function maps every element in the sample space to a real number.
- We denote the random variable function by a capital letter (such as  $W$ ,  $X$  or  $Y$ )
- Any particular ~~number~~ value of the random variable is denoted by a small letter ( ~~$w$ ,  $x$~~  or  $y$ )

- Example 2.1-1

- Example 2.1-2

\* Conditions for a function to be a Random Variable.

- 1- Every point in  $S$  must correspond to only one value of the random variable.
- 2- The set  $\{X \leq x\}$  should be an event for any real number  $x$   
 $\Rightarrow P\{X \leq x\}$  exist
- 3-  $P\{X = -\infty\} = 0$

- The Random Variables can be Discrete or Continuous

\* A Discrete random variable ~~is~~ has discrete values only

- See example 2.1-3

\* A continuous random variable has a continuous range of values.

- See example 2.1-4

Fact: the probability of occurrence of any discrete value of a continuous random variable is zero.

(why? see example 2.1-4)

\* A mixed R.V. has discrete and continuous values.

## 2.2 and 2.3 Distribution and Density Functions

- Cumulative Prob. Distribution Functions  
"Distribution Function"  
"CDF"

$$F_X(x) = P\{X \leq x\} \quad ; \quad \begin{array}{l} \text{where} \\ x \text{ is} \\ \text{a real} \\ \text{number} \end{array}$$

↑ Random variable

- Properties of  $F_X(x)$ :

- ①  $F_X(-\infty) = 0$
- ②  $F_X(\infty) = 1$
- ③  $0 \leq F_X(x) \leq 1$
- ④  $F_X(x_1) \leq F_X(x_2)$   
if  $x_1 < x_2$  "non decreasing"
- ⑤  $P\{x_1 < X \leq x_2\}$   
 $= F_X(x_2) - F_X(x_1)$
- ⑥  $F_X(x^+) = F_X(x)$   
"continuous from the right"

- Probability Density Function (pdf)

$$f_X(x) = \frac{dF_X(x)}{dx}$$

- Properties of  $f_X(x)$

- ①  $0 \leq f_X(x)$  all  $x$   
"nonnegative"
- ②  $\int_{-\infty}^{\infty} f_X(x) dx = 1$   
"Unit Area"
- ③  $F_X(x) = \int_{-\infty}^x f_X(\tau) d\tau$
- ④  $P\{x_1 < X \leq x_2\}$   
 $= \int_{x_1}^{x_2} f_X(x) dx$

— Discrete random variable

$$\text{Let } X = \{x_1, x_2, \dots, x_N\}$$

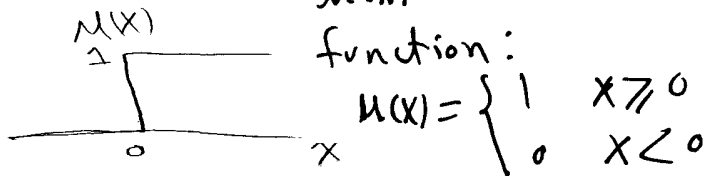
be a discrete random variable,

~~then~~ and let  $P\{X=x_i\} = P(x_i)$

then, the discrete CDF will be:

$$\begin{aligned} F_X(x) &= \sum_{i=1}^N P\{X=x_i\} u(x-x_i) \\ &= \sum_{i=1}^N P(x_i) u(x-x_i) \end{aligned}$$

where  $u(x)$  is the unit-step function:



then, the pdf will be

$$f_X(x) = \frac{dF_X(x)}{dx}$$

$$= \sum_{i=1}^N P(x_i) \delta(x-x_i)$$

where  $\delta(x)$  is unit-impulse function  
"Delta function"

$$\delta(x) = \frac{du(x)}{dx}$$

and

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

See Example 2-1 Page 46

— Continuous random variable

Let  $f_X(x)$  be the pdf

of a continuous random variable, then

the CDF will be

$$F_X(x) = \int_{-\infty}^x f_X(\eta) d\eta$$

or we can say that

$$f_X(x) = \frac{dF_X(x)}{dx}$$

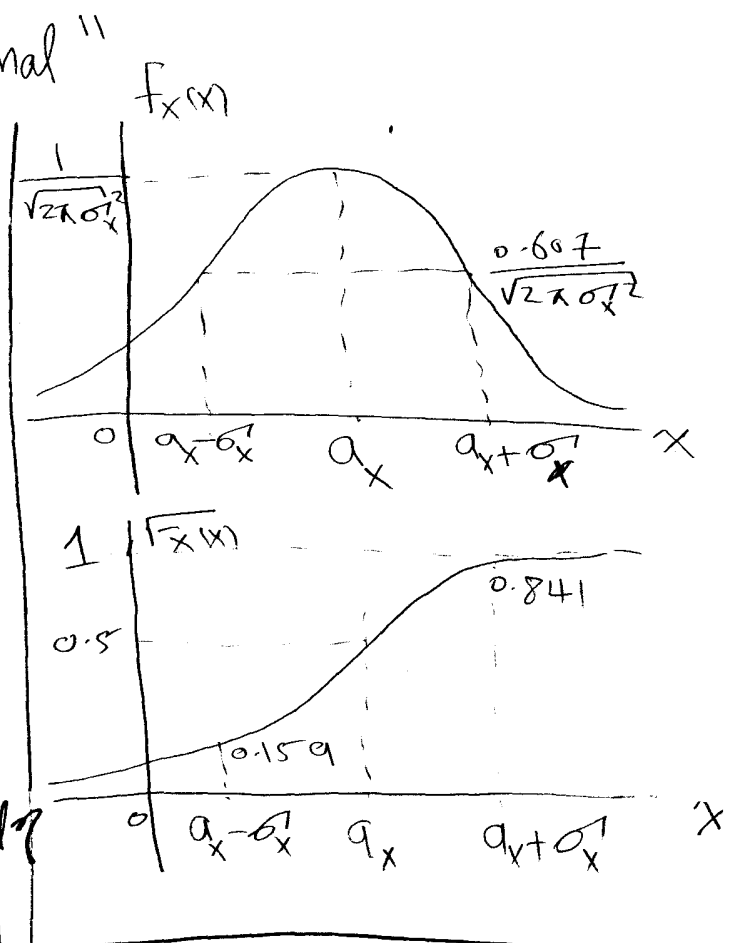
See Example 2-2-2  
Page 47

2.4 The Gaussian "Normal" Random Variable.

pdf  $f_X(x) = \frac{1}{\sqrt{2\pi\sigma_x^2}} e^{-\frac{(x-\mu_x)^2}{2\sigma_x^2}}$

Def:  $\mu_x \equiv$  the mean  
 $\sigma_x \equiv$  standard deviation  
 $\sigma_x^2 \equiv$  variance

CDF  $F_X(x) = \frac{1}{\sqrt{2\pi\sigma_x^2}} \int_{-\infty}^x e^{-\frac{(\eta-\mu_x)^2}{2\sigma_x^2}} d\eta$



- Can we evaluate the integral and find a closed-form solution?

- The answer is NO. there are no closed-form solutions for this integral.

- ~~Thus~~, we use numerical or approximation methods to evaluate the CDF.

- In Appendix B, there is a numerical table for the normalized CDF which is

$$F(x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^x e^{-\eta^2/2} d\eta$$

where  $\mu_x = 0$

Then, to find  $F_X(x) = \frac{1}{\sqrt{2\pi\sigma_x^2}} \int_{-\infty}^x e^{-\frac{(\eta-\mu_x)^2}{2\sigma_x^2}} d\eta$

Let  $u = \frac{\eta - \mu_x}{\sigma_x} \Rightarrow du = \frac{d\eta}{\sigma_x}$

$$\Rightarrow F_X(x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\frac{x-\mu_x}{\sigma_x}} e^{-\frac{u^2}{2}} du = F\left(\frac{x-\mu_x}{\sigma_x}\right)$$

See Example 2.4-1 and Example 2.4-2

Note:  $F(-x) = 1 - F(x)$

What is the relation between the Gaussian CDF and the Q-Function?

A famous function in probability theory and Mathematics is the Q-Function.

$$Q(x) = \frac{1}{\sqrt{2\pi}} \int_x^{\infty} e^{-\eta^2/2} d\eta$$

Note that

$$F(x) = 1 - Q(x)$$

The Q-Function has the approximation

$$Q(x) \approx \left[ \frac{1}{(1-a)x + a\sqrt{x^2+b}} \right] \frac{e^{-x^2/2}}{\sqrt{2\pi}} \quad x \gg 0$$

where  $a = 0.339$   
and  $b = 5.510$

See Example 2.4-3

2.5 Other Distribution and Density Functions

① Binomial

Let  $0 < p < 1$  and  $N = 1, 2, \dots$   
 then the Binomial pdf is

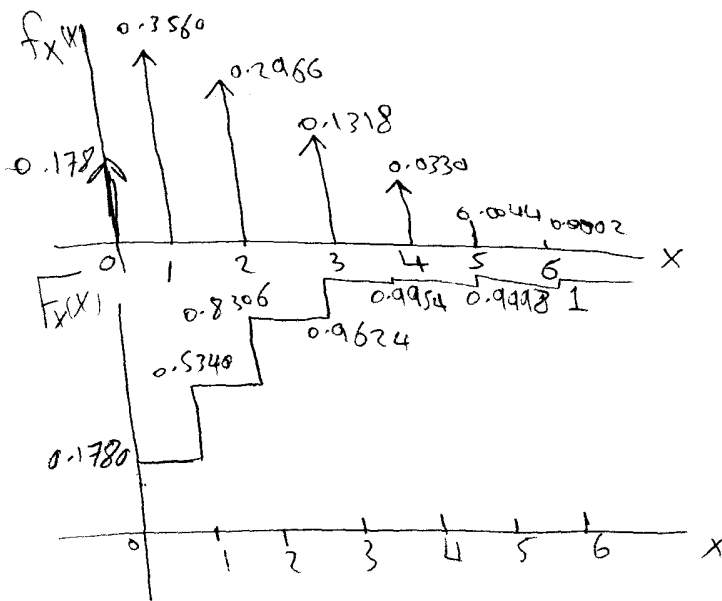
$$f_X(x) = \sum_{k=0}^N \binom{N}{k} p^k (1-p)^{N-k} \delta(x-k)$$

Binomial coefficient  
 $\binom{N}{k} = \frac{N!}{k!(N-k)!}$

the CDF is

$$F_X(x) = \sum_{k=0}^N \binom{N}{k} p^k (1-p)^{N-k} u(x-k)$$

Example at  $N=6$  and  $p=0.25$



② Poisson

pdf  $f_X(x) = e^{-b} \sum_{k=0}^{\infty} \frac{b^k}{k!} \delta(x-k)$

CDF  $F_X(x) = e^{-b} \sum_{k=0}^{\infty} \frac{b^k}{k!} u(x-k)$

where  $b > 0$  is a real constant.

- Binomial Approximation using Poisson

when  $N$  is large and  $p$  is very small,

we can approximate the Binomial R.V. with a Poisson R.V. with  $b = Np$

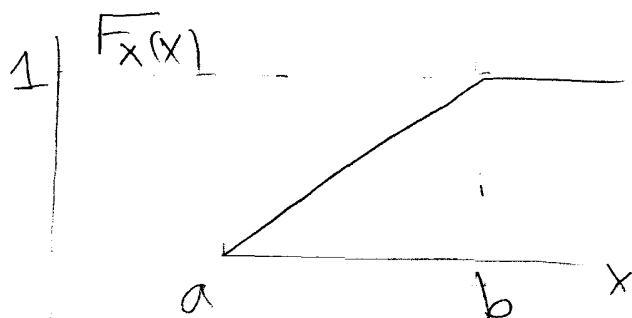
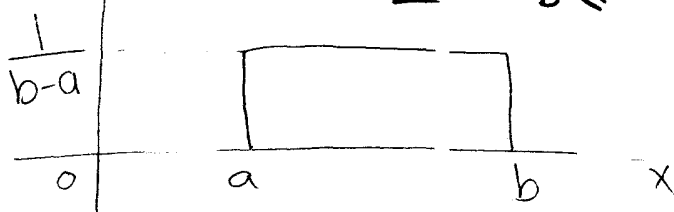
- Time & Arrival Problems.

See Example 2.5-1.

③ Uniform

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

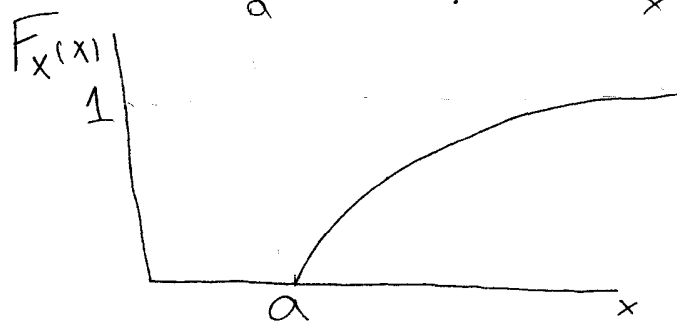
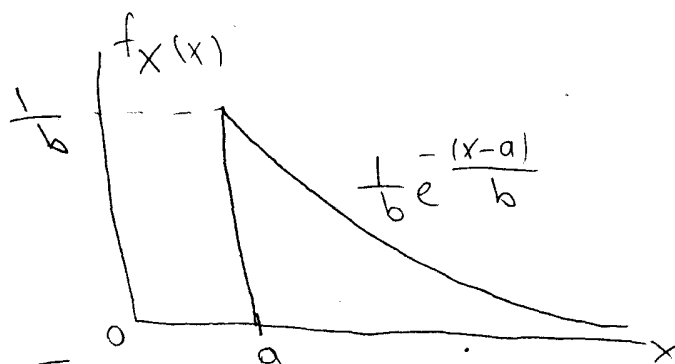
cdf  $F_X(x) = \begin{cases} 0 & x < a \\ \frac{x-a}{b-a} & a \leq x < b \\ 1 & b \leq x \end{cases}$

④ Exponential

pdf  $f_X(x) = \begin{cases} \frac{1}{b} e^{-\frac{(x-a)}{b}} & x > a \\ 0 & x < a \end{cases}$

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

where  $-\infty < a < \infty$   
and  $b > 0$



Solve Example 2.5-2

### ⑤ Rayleigh

pdf  $f_X(x) = \begin{cases} \frac{2}{b}(x-a)e^{-\frac{(x-a)^2}{b}} & x \geq a \\ 0 & x < a \end{cases}$

cdf  $F_X(x) = \begin{cases} 1 - e^{-\frac{(x-a)^2}{b}} & x \geq a \\ 0 & x < a \end{cases}$

where  $-\infty < a < \infty$  and  $b > 0$

