

## Review:

Cumulative distribution function.

Def. cdf of  $X$  is denoted by  $F_X$  and defined as  $F_X(c) = P(X \leq c)$  for any  $c \in \mathbb{R}$ .

main properties:

F.1. increasing:  $a < b \Rightarrow F_X(a) < F_X(b)$

F.2.  $\lim_{c \rightarrow \infty} F_X(c) = 1$ ,  $\lim_{c \rightarrow -\infty} F_X(c) = 0$

F.3.  $F_X$  is right continuous, i.e.,  $\lim_{u \rightarrow c^+} F_X(u) = F_X(c)$

Proposition: any function that satisfies above property is cdf of some random variable.

Important implications:

(i)  $\Delta F_X(c) = F_X(c) - F_X(c-) = P(X=c)$

(ii)  $F_X(c-) = \lim_{u \rightarrow c^-} F_X(u) = P(X < c)$

(iii) For any  $a < b$ ,  $F_X(b) - F_X(a) = P(a < X \leq b)$

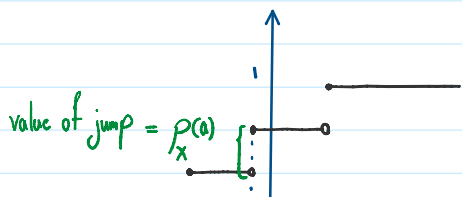
## Today:

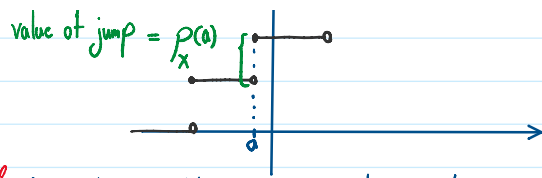
① Continuous random variable

② Uniform random variable

① Continuous random variable

For a discrete random variable, we have  $F_X(c) = \sum_{u: u \leq c} p_X(u)$





**Def.** A random variable  $X$  is a continuous type random variable if there exist a function  $f_x$ , called probability density function (pdf) of  $X$  such that

$$F_x(c) = \int_{-\infty}^c f_x(u) du. \quad \text{for any } c \in \mathbb{R}$$

**Def.** support of a pdf  $f_x$  is the set of  $u$  for which  $f_x(u) > 0$ .

**Important remarks:**

- (i) Suppose that  $f_x$  is continuous at point  $c$ . By fundamental theorem of calculus,  $f_x(c) = F_x'(c)$
- (ii) For a continuous-type random variable  $X$ , the cdf  $F_x$  is continuous.
- (iii) If  $F_x$  is continuous at point  $c$ , then  $F_x(c-) = F_x(c) = F_x(c+) \Rightarrow P(X=c) = \Delta F_x(c) = 0$ .

In particular,  $P(X=c) = 0$  for a continuous-type random variable  $X$ , and we have

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

(iv) Notice that by axioms of probability, if  $A_i \cap A_j = \emptyset$  for all  $i \neq j$ , then

$$P\left(\bigcup_{i=1}^{\infty} A_i\right) = \sum_{i=1}^{\infty} P(A_i)$$

Hence, for any sequence of numbers  $x_1, x_2, x_3, \dots$ , for a continuous-type random variable  $X$ , we have:

$$P(X \in \{x_1, x_2, \dots\}) = \sum_{i=1}^n P(X=x_i) = 0$$

In particular,  $P(X \text{ is a natural number}) = 0$

$P(X \text{ is a rational number}) = 0$

**Interpretation of pdf:**

Suppose that  $f_x$  is continuous at point  $c$ . We have,

$$\lim_{h \rightarrow 0} \frac{F_x(c+h) - F_x(c-h)}{2h} = \lim_{h \rightarrow 0} \frac{F_x(c+h) - F_x(c)}{2h} + \frac{F_x(c) - F_x(c-h)}{2h}$$

$$\begin{aligned} \lim_{h \rightarrow 0} \frac{F_X(c+h) - F_X(c-h)}{2h} &= \lim_{h \rightarrow 0} \frac{F_X(c+h) - F_X(c)}{2h} + \frac{F_X(c) - F_X(c-h)}{2h} \\ &= \frac{f_X(c)}{2} + \frac{f_X(c)}{2} = f_X(c) \end{aligned}$$

where we used the fact that  $F'_X(c) = f_X(c)$ . Hence, we can write

$$\frac{F_X(c+h) - F_X(c-h)}{2h} = f_X(c) + o(h)$$

where  $\lim_{h \rightarrow 0} o(h) = 0$ . Equivalently, we can write

$$F_X(c+h) - F_X(c-h) = 2h f_X(c) + o(h)$$

where  $\lim_{h \rightarrow 0} \frac{o(h)}{h} = 0$ , i.e.,  $o(h)$  converges to zero faster than  $h$ .

Hence,  $f_X(c)$  is related to the probability of  $X$  being in a small neighborhood of  $c$ :

$$P(c-h < X < c+h) = 2h f_X(c) + o(h)$$

properties of  $f_X$ :

(i) For any  $a < b$ :  $P(a < X \leq b) = F_X(b) - F_X(a) = \int_a^b f_X(u) du \in [0, 1] \Rightarrow f_X$  is non negative.

(ii)  $1 = \lim_{b \rightarrow +\infty} \lim_{a \rightarrow -\infty} F_X(b) - F_X(a) = \lim_{b \rightarrow +\infty} \lim_{a \rightarrow -\infty} \int_a^b f_X(u) du \Rightarrow \int_{-\infty}^{+\infty} f_X(u) du = 1$

Mean, variance, LOTUS

We can recycle all we had for discrete random variables by replacing summation with integration, & pmf with pdf, i.e., for a continuous-type random variable  $X$

$$\mu_X = E[X] = \int_{-\infty}^{+\infty} u f_X(u) du$$

$$\text{Var}(X) = E[(X - \mu_X)^2] = \int_{-\infty}^{+\infty} (u - \mu_X)^2 f_X(u) du \rightsquigarrow \text{measures how spread out } X \text{ is}$$

$$\text{Var}(X) = E[X^2] - (E[X])^2$$

$$\sigma_x = \sqrt{\text{Var}(X)} \rightsquigarrow \text{same unit as } X$$

$$\text{LOTUS: } E[g(X)] = \int_{-\infty}^{+\infty} g(u) f_X(u) du.$$

$$\begin{aligned} \text{Var}(g(X)) &= E[g(X)^2] - (E[g(X)])^2 \\ &= \int_{-\infty}^{+\infty} g(u)^2 f_X(u) du - \left( \int_{-\infty}^{+\infty} g(u) f_X(u) du \right)^2 \end{aligned}$$

$$\text{Standardized version: } \frac{X - \mu_X}{\sigma_X} \rightsquigarrow \text{dimensionless}$$

and all interpretations are valid as before.

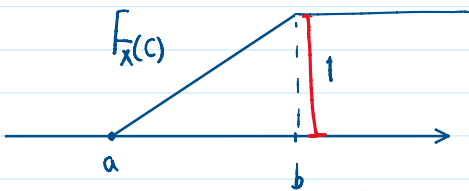
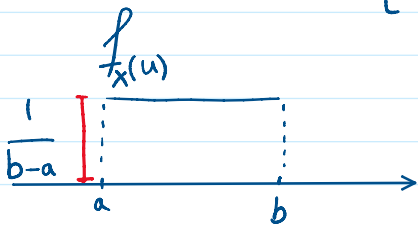
$$E[aX+b] = aE[X] + b$$

$$\text{Var}(aX+b) = a^2 \text{Var}(X)$$

② Uniform random variable:

**Def.** We say  $X$  is uniformly distributed over  $[a, b]$  if

$$f_X(u) = \begin{cases} \frac{1}{b-a} & a \leq u \leq b \\ 0 & \text{o.w.} \end{cases}$$



$$k\text{th moment: } E[X^k] = \int_a^b u^k \cdot \frac{1}{b-a} du = \frac{1}{b-a} \cdot \frac{u^{k+1}}{k+1} \Big|_a^b = \frac{b^{k+1} - a^{k+1}}{b-a} \cdot \frac{1}{k+1}$$

$$\mu_X = E[X] = \frac{b^2 - a^2}{1} \cdot \frac{1}{2} = \frac{b+a}{2}$$

$$\mu_X = E[X] = \frac{b^2 - a^2}{b - a} \cdot \frac{1}{2} = \frac{b + a}{2}$$

$$\text{var}(X) = E[X^2] - (E[X])^2 = \frac{b^3 - a^3}{b - a} \cdot \frac{1}{3} - \frac{(b + a)^2}{4} = \frac{b^2 + ab + a^2}{3} - \frac{b^2 + 2ab + a^2}{4} = \frac{(b - a)^2}{12}$$