

3.1. RV of the continuous type

□ • Let $X =$ pos. of a point chosen "uniformly" at random from $[a, b]$.

$\Rightarrow$ For $a \leq x \leq b$, the CDF

$$F(x) = P(X \leq x)$$

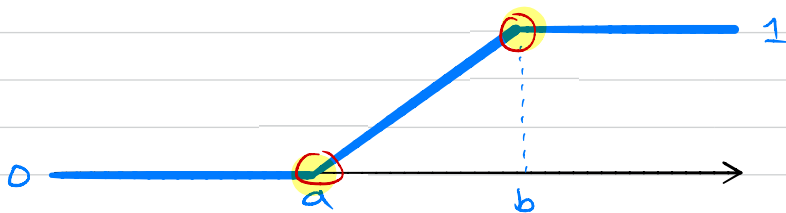
is proportional to the length of $[a, x]$,
so

$$F(x) = \frac{x - a}{b - a}$$

$\swarrow$ normalizes total prob. to 1.

$\Rightarrow$ For general $x \in \mathbb{R}$,

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



• We may write:

$$F(x) = \int_{-\infty}^x f(y) dy$$

for

$$f(x) = \begin{cases} \frac{1}{b - a}, \\ 0, \end{cases}$$

exact value of f at a & b is not important, b.c. F is not diff-ble at a & b .

$a \leq x \leq b$

elsewhere

Alternatively, $F'(x) = f(x)$.

- This $f(x)$ is called the **probability density function** (PDF) of X . This is an example of:

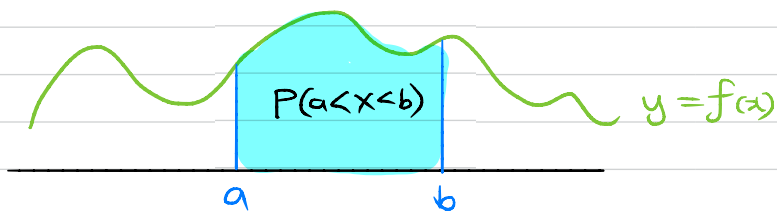
(a.k.a. continuous RV)

DEF The PDF of a **RV X of the continuous type**, with space = (union of) interval(s), is an integrable ftn s.t.

(a) $f(x) \geq 0$, $\forall x \in S$. ← for all

(b) $\int_S f(x) dx = 1$.

(c) $P(a < X < b) = \int_a^b f(x) dx$, $\forall (a, b) \subseteq S$.



Rmk 1) Can always extend f to $\mathbb{R}$ by setting

$$\text{"new } f(x) = \begin{cases} f(x), & x \in S \\ 0, & \text{elsewhere.} \end{cases}$$

Then (a) - (c) becomes:

$$(a') \quad f(x) \geq 0, \quad \forall x \in \mathbb{R}$$

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

$$(c) \quad P(a < X < b) = \int_a^b f(x) dx, \quad \forall (a, b) \subseteq \mathbb{R}.$$

Rmk2) PMF $f(x)$ is NOT a probability. In particular, a PMF can exceed 1.

Rmk3) If X is a conti. RV with PMF f , then

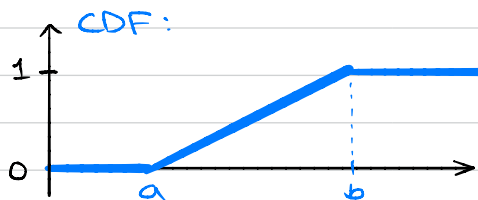
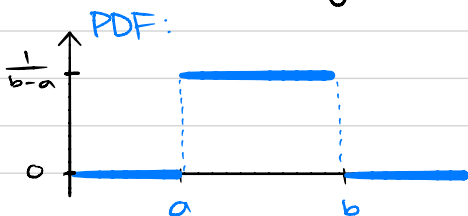
$$P(X=x) = \int_x^x f(y) dy = 0.$$

$$\text{So, } P(a \leq X \leq b) = P(a < X \leq b) = P(a \leq X < b) = P(a < X < b)$$

Ex (Uniform distribution) Both descriptions

$$\left[f(x) = \frac{1}{b-a}, \quad a < x < b \right] \text{ and } \left[f(x) = \begin{cases} \frac{1}{b-a}, & a < x < b \\ 0, & \text{e/w} \end{cases} \right]$$

refer to the PDF of the same dist., called the uniform distribution (or rectangular dist.), which is denoted by $U(a, b)$.



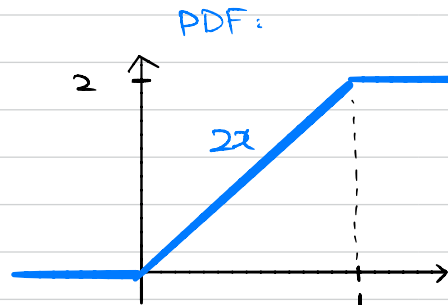
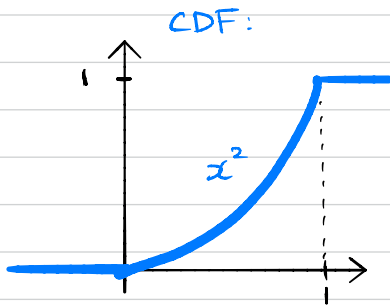
Ex Let X has CDF

$$F(x) = \begin{cases} 0, & x < 0, \\ x^2, & 0 \leq x < 1, \\ 1, & 1 \leq x. \end{cases}$$

□

Then its PDF is

$$f(x) = F'(x) = \begin{cases} 2x, & 0 < x < 1, \\ 0, & \text{elsewhere.} \end{cases}$$



Now,

$$P\left(\frac{1}{3} < X < \frac{2}{3}\right) = F\left(\frac{2}{3}\right) - F\left(\frac{1}{3}\right) = \left(\frac{2}{3}\right)^2 - \left(\frac{1}{3}\right)^2 = \frac{1}{8},$$

and

$$P\left(\frac{1}{2} < X < 2\right) = F(2) - F\left(\frac{1}{2}\right) = 1 - \left(\frac{1}{2}\right)^2 = \frac{3}{4}.$$

Alternatively,

$$\begin{aligned} P\left(\frac{1}{2} < X < 2\right) &= \int_{\frac{1}{2}}^2 f(x) dx = \int_{\frac{1}{2}}^1 2x dx + \int_1^2 0 dx \\ &= \frac{3}{4}. \end{aligned}$$

Often, PDF is piecewise-defined.

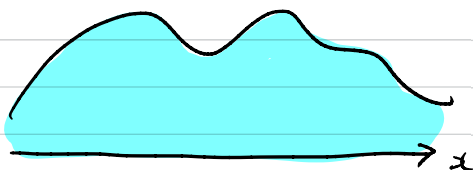
□

[2] Expectation

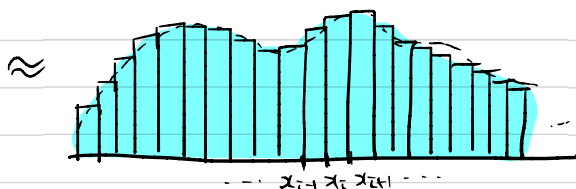
DEF If X : conti. RV with PDF $f(x)$, then its expected value (or mean) is

$$\mu = E(X) := \int_{-\infty}^{\infty} x f(x) dx.$$

Motivation



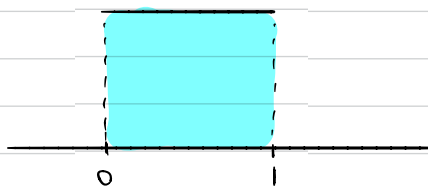
graph of $f(x)$



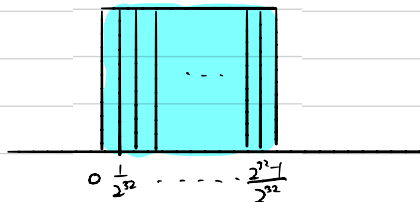
prob. histogram of
 $p(x_i) = P(x_i \leq X < x_{i+1})$
 $\approx f(x_i) \Delta x$
 with bin size Δx .

$$\Rightarrow \int_a^b u(x) f(x) dx = E[u(X)] \approx \sum_{i: a < x_i < b} u(x_i) f(x_i) \Delta x_i.$$

(For instance in software, pseudo-random number gen. uses this idea to approx the uniform dist. $U(0,1)$ by a discrete uniform dist:)

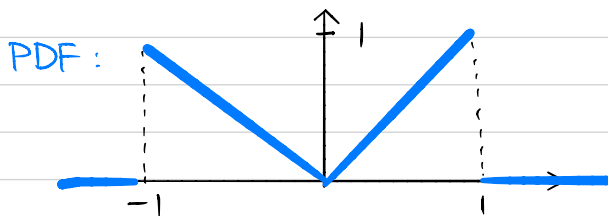


$\approx$



- DEF • Variance : $\sigma^2 = \text{Var}(X) = E[(X-\mu)^2]$
 $= \int_{-\infty}^{\infty} (x-\mu)^2 f(x) dx,$
- Std. Dev. : $\sigma = \sqrt{\text{Var}(X)},$
- MGF : $M(t) = E[e^{tx}] = \int_{-\infty}^{\infty} e^{tx} f(x) dx.$

Ex Let X has PDF $f(x) = |x|, -1 < x < 1.$

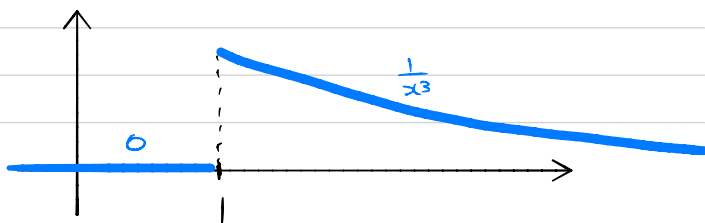


Then

$$\begin{aligned} E(X) &= \int_{-1}^1 x \cdot |x| dx = \int_{-1}^0 (-x^2) dx + \int_0^1 x^2 dx \\ &= \left[-\frac{x^3}{3} \right]_{-1}^0 + \left[\frac{x^3}{3} \right]_0^1 = -\frac{1}{3} + \frac{1}{3} = 0. \end{aligned}$$

$$\begin{aligned} \text{Var}(X) &\stackrel{(\because \mu=0)}{=} E(X^2) = \int_{-1}^1 x^2 |x| dx \\ &= \int_{-1}^0 (-x^3) dx + \int_0^1 x^3 dx = \frac{1}{4} + \frac{1}{4} = \frac{1}{2}. \quad \square \end{aligned}$$

Ex Let X has PDF $f(x) = \frac{2}{x^3}, x > 1.$



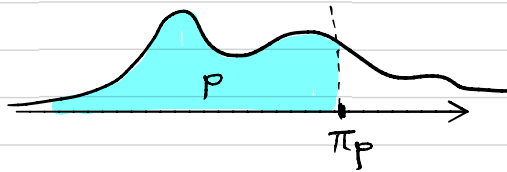
Then

$$\begin{aligned} E(X) &= \int_1^{\infty} x \cdot \frac{2}{x^3} dx = \lim_{b \rightarrow \infty} \int_1^b \frac{2}{x^2} dx \\ &= \lim_{b \rightarrow \infty} \left[-\frac{2}{x} \right]_1^b = \lim_{b \rightarrow \infty} \left(2 - \frac{2}{b} \right) \\ &= 2. \end{aligned}$$

□

DEF The $(100p)$ quantile is a number π_p s.t.

$$F(\pi_p) = p$$



• Special cases:

p	name of π_p
0.25	first quartile
0.5	median (= second quartile)
0.75	third quartile

Rmk) Median need not equal the mean.