

Defn

A random variable $X : \Omega \rightarrow \mathbb{R}$ is called continuous iff there exists a function

$$f_X : \mathbb{R} \rightarrow \mathbb{R}$$

called the Probability Density Function (PDF) such that

$$P(X \in I) = \int_I f_X(x) dx$$

for 'any' $I \subseteq \mathbb{R}$. In Particular
if $I = [a, b]$, then

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

Note : integrating over $[a, b]$

$(a, b]$, $[a, b)$ or (a, b) are the same:

$$P(X=a) = \int_a^a f_X(x) dx = 0$$

Note

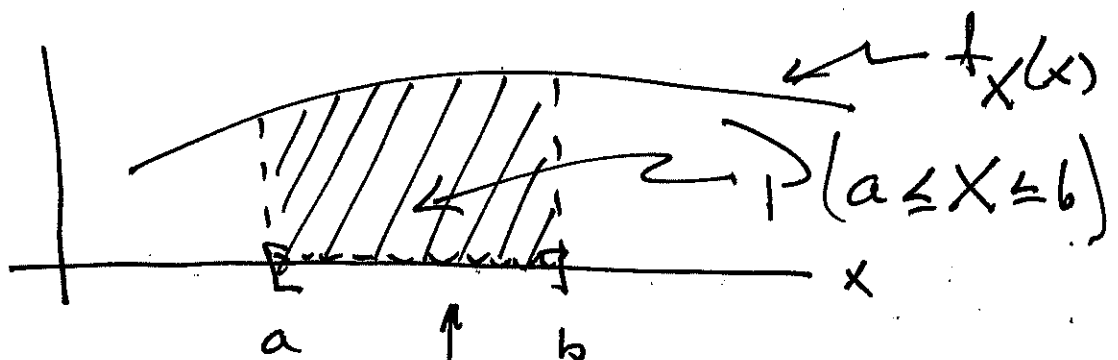
To be a valid PDF we must have

- $f_X(x) \geq 0$ for all $x \in \mathbb{R}$

- $\int_{-\infty}^{\infty} f_X(x) dx = 1 \quad (X \in \mathbb{R})$

$$= P(\Omega) = 1$$

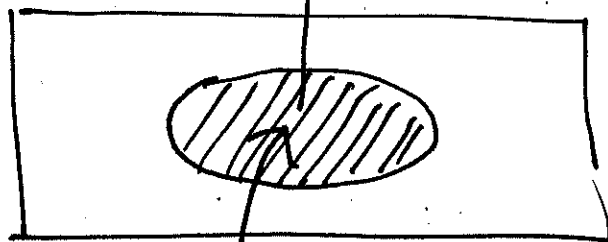
Picture



$\mathbb{R}$

$X \uparrow$

Ω



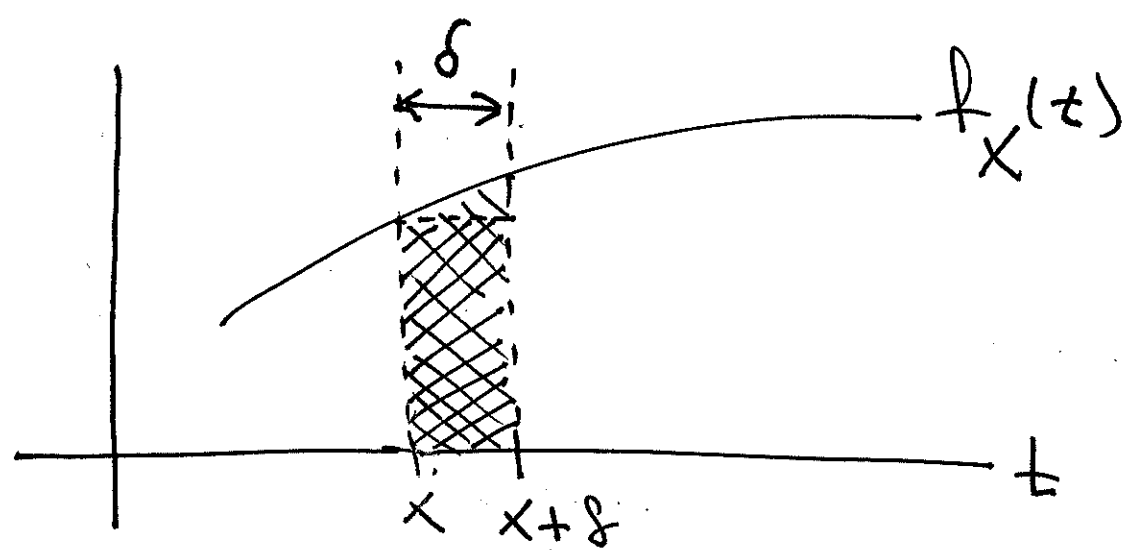
$\{a \leq X \leq b\}$

Note

Unlike the PMF, the PDF does not give Probability (or 'mass') at a point.

Instead, $f_X(x)$ gives the 'mass density' at x , i.e. mass per unit length.

Let $\delta > 0$ be a ^{small} real number.
 Then $f_X(t)$ is approx. constant
 on the interval $[x, x+\delta]$
 (assuming f_X is continuous).



$$\underbrace{P(x \leq X \leq x+\delta)}_{\substack{\text{Probability i.e.} \\ \text{'mass'}}} = \int_x^{x+\delta} f_X(t) dt \approx \underbrace{f_X(x)}_X \cdot \underbrace{\delta}_{\text{length}}$$

Thus $f_X(x)$ has units

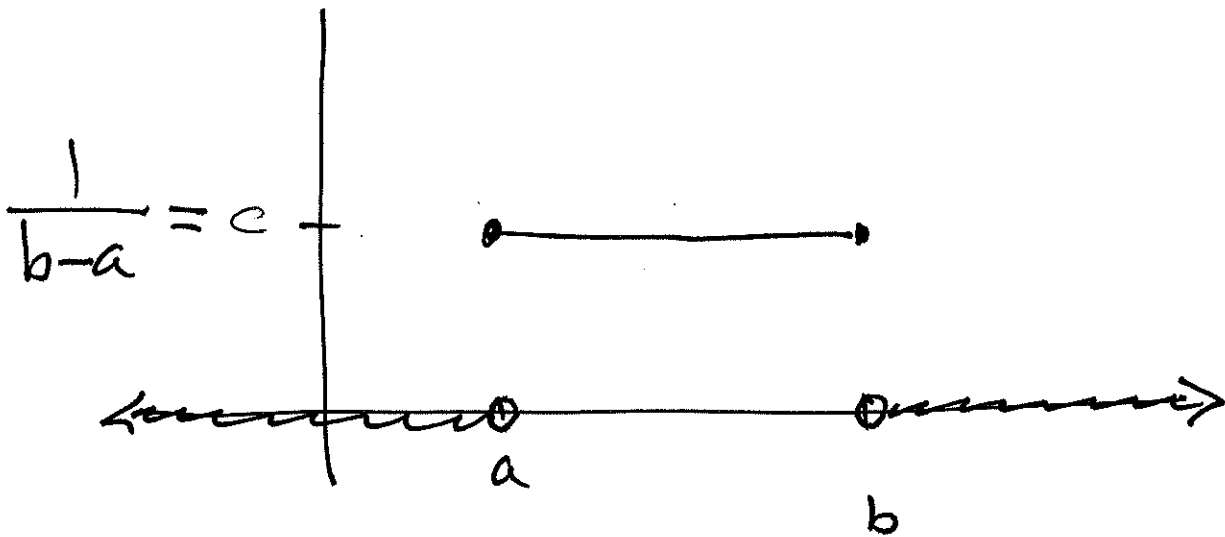
$$\frac{\text{'mass'}}{\text{length}} = \text{density}.$$

note: f_X may have some jump discontinuities.

Ex. Continuous uniform r.v. on $[a, b]$.

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

□



The const c is determined by

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

so

$$1 = \int_{-\infty}^{\infty} f_X(x) dx = \int_a^b c dx = cx \Big|_a^b = c(b-a)$$

∴ $c = \frac{1}{b-a}$

Ex Piecewise Constant $\rightarrow$ PDF.

Alice's driving time to work is 40-44 min. in good weather and 44-50 min in bad, uniformly distributed in each case.

Suppose $P(\text{good weather}) = \frac{2}{3}$.

Determine the PDF of X , Alice's driving time.

Solution We are given

$$f_X(x) = \begin{cases} c_1 & \text{if } 40 \leq x \leq 44 \text{ (good)} \\ c_2 & \text{if } 44 < x \leq 50 \text{ (bad)} \\ 0 & \text{otherwise} \end{cases}$$

note

$$\frac{2}{3} = P(\text{good}) = \int_{40}^{44} c_1 dx = c_1 \cdot x \Big|_{40}^{44} = 4c_1$$

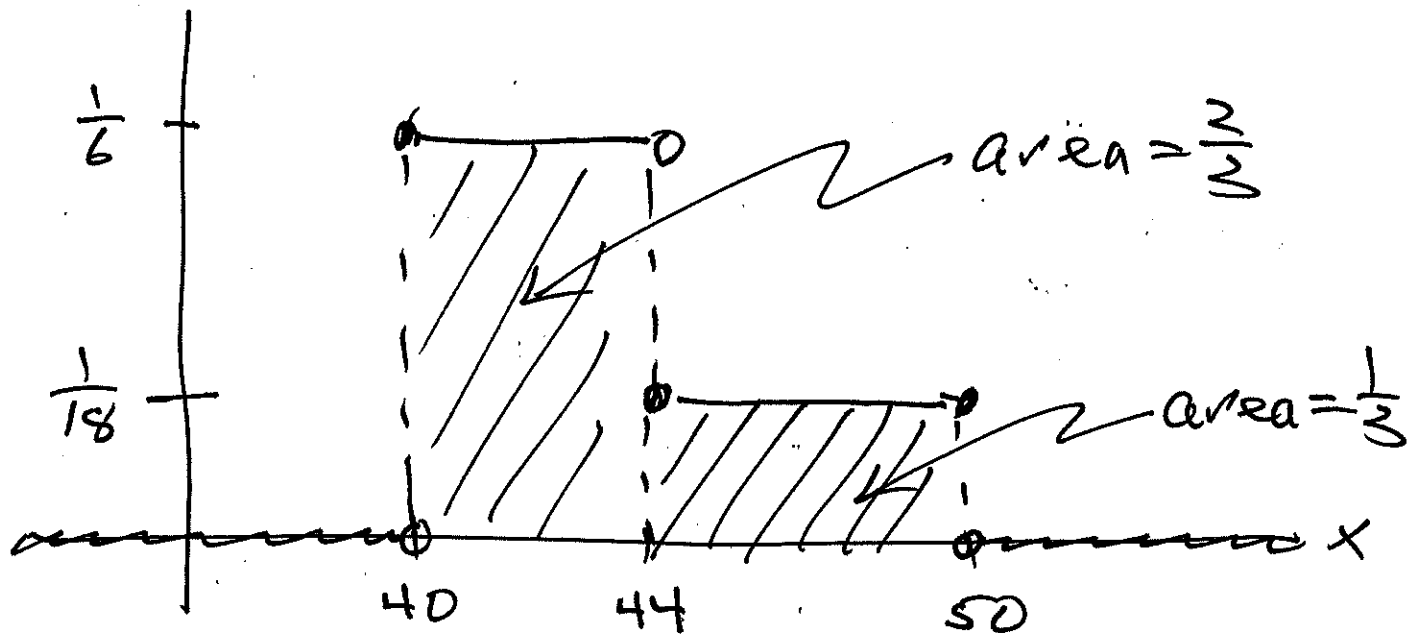
$$\therefore c_1 = \frac{1}{4} \cdot \frac{2}{3} = \boxed{\frac{1}{6}}$$

also

$$\frac{1}{3} = P(\text{bad}) = \int_{44}^{50} c_2 dx = 6 \cdot c_2$$

$$\therefore c_2 = \frac{1}{6} \cdot \frac{1}{3} = \frac{1}{18}$$

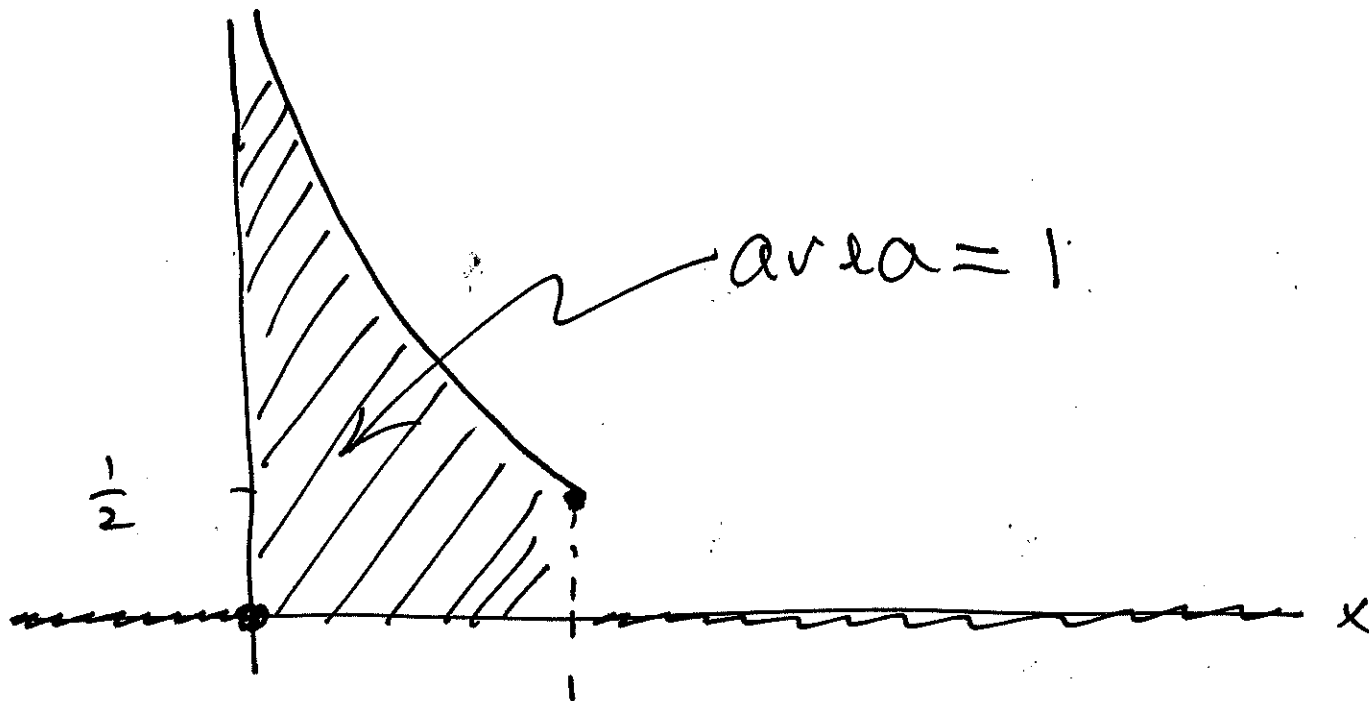
$$\therefore f_X(x) = \begin{cases} \frac{1}{6} & \text{if } 40 \leq x \leq 44 \\ \frac{1}{18} & \text{if } 44 < x \leq 50 \\ 0 & \text{otherwise} \end{cases}$$



A PDF need not be bounded above.

Ex.

$$f_X(x) = \begin{cases} \frac{1}{2\sqrt{x}} & \text{if } 0 < x \leq 1 \\ 0 & \text{otherwise} \end{cases}$$



$$\int_{-\infty}^{\infty} f_X(x) dx = \int_0^1 \frac{1}{2} x^{-\frac{1}{2}} dx$$

$$= \frac{1}{2} \frac{x^{\frac{1}{2}}}{\frac{1}{2}} \bigg|_0^1 = \sqrt{1} - \sqrt{0} = \boxed{1}$$

actually

$$\int_0^1 \frac{1}{2} x^{-\frac{1}{2}} dx = \lim_{a \rightarrow 0^+} \int_a^1 \frac{1}{2} x^{-\frac{1}{2}} dx$$

$$= \lim_{a \rightarrow 0^+} (x^{\frac{1}{2}}) \bigg|_a^1 = \lim_{a \rightarrow 0^+} (\sqrt{1} - \sqrt{a})$$

$$= \boxed{1}$$

Defn

The expected value of a continuous r.v. X is given by

$$E[X] = \int_{-\infty}^{\infty} x f_X(x) dx$$

Expected value rule

Let X be a cont. r.v., $g: \mathbb{R} \rightarrow \mathbb{R}$,

then

$$E[g(X)] = \int_{-\infty}^{\infty} g(x) \cdot f_X(x) dx$$

The n^{th} moment of X is

$$E[X^n]$$

Defn

The variance of X is

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

Theorem (var in terms of moments)

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

Proof. Let $\mu = E[X]$.

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

$$= \int_{-\infty}^{\infty} (x - \mu)^2 \cdot f_X(x) dx$$

$$= \int_{-\infty}^{\infty} (x^2 - 2\mu x + \mu^2) f_X(x) dx$$

$$= \underbrace{\int_{-\infty}^{\infty} x^2 f_X(x) dx}_{E[X^2]} - 2\mu \underbrace{\int_{-\infty}^{\infty} x f_X(x) dx}_{E[X]} + \mu^2 \underbrace{\int_{-\infty}^{\infty} f_X(x) dx}_1$$

μ

$$= E[X^2] - 2\mu^2 + \mu^2 = E[X^2] - \mu^2$$

$$= E[X^2] - E[X]^2$$



Let $Y = aX + b$ where

$a, b \in \mathbb{R}$ and X is a cont. r.v.

Then

$$\bullet E[Y] = aE[X] + b$$

$$\bullet \text{Var}(Y) = a^2 \text{Var}(X)$$

Exercise

Prove these.

Ex. cont. unif. on $[a, b]$

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

so

$$E[X] = \int_{-\infty}^{\infty} x f_X(x) dx$$

$$= \int_a^b x \cdot \left(\frac{1}{b-a} \right) dx$$

$$= \frac{1}{b-a} \cdot \frac{1}{2} x^2 \Big|_a^b = \frac{1}{b-a} \cdot \frac{1}{2} (b^2 - a^2)$$

$$= \frac{1}{2} \cdot \frac{\cancel{(b-a)}(b+a)}{\cancel{b-a}} = \boxed{\frac{a+b}{2}}$$

$$E[X^2] = \int_a^b x^2 \left(\frac{1}{b-a} \right) dx$$

$$= \frac{1}{b-a} \cdot \frac{1}{3} x^3 = \frac{1}{b-a} \cdot \frac{1}{3} (b^3 - a^3)$$

$$= \frac{1}{3} \frac{\cancel{(b-a)}(b^2 + ab + a^2)}{\cancel{(b-a)}}$$

$$= \frac{b^2 + ab + a^2}{3}$$

Thus

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

$$= \frac{b^2 + ab + a^2}{3} - \left(\frac{a+b}{2} \right)^2$$

$$= \dots = \boxed{\frac{(b-a)^2}{12}}$$