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Probability and Random Processes

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Distribution function of a random variable

Definition

Let $X : \Omega \rightarrow \mathbb{R}$ be a random variable. The *probability distribution function* of X is the function $F_X : \mathbb{R} \rightarrow [0, 1]$ defined by

$$F_X(t) = \mathbb{P}[X \leq t].$$

$X : \Omega \rightarrow \mathbb{R}$ values of X are real numbers

define events in terms of

$$A_t = \{X \leq t\}$$

$$B = \{s \leq X \leq t\}$$

$$\mathbb{P}(A) = \mathbb{P}(X \leq t)$$

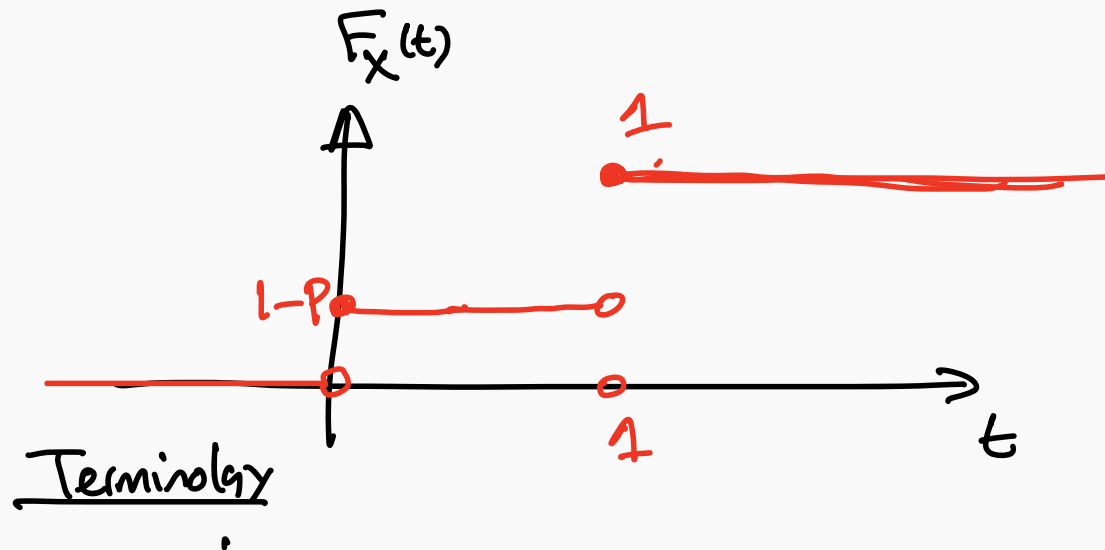
$$\mathbb{P}(B) = \mathbb{P}(s \leq X \leq t)$$

\mathbb{R}

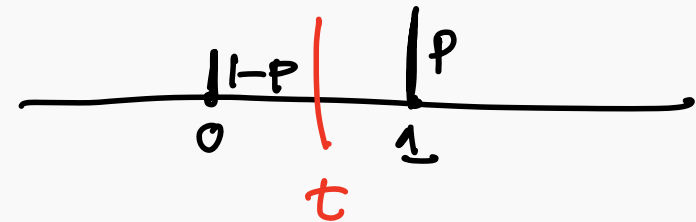


Examples: the distribution function of a Bernoulli random variable

Plot $F_X(t)$



x	0	1
$P(X=x)$	$1-p$	p

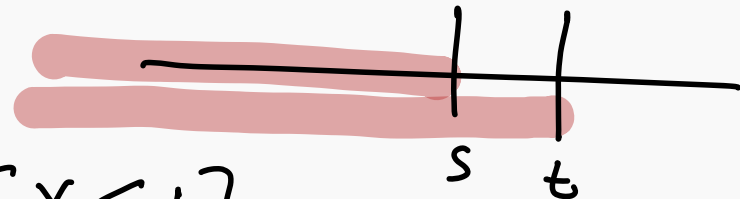


increasing if $x < y \Rightarrow f(x) \leq f(y)$

$$s \leq t \quad F_X(s) \leq F_X(t)$$

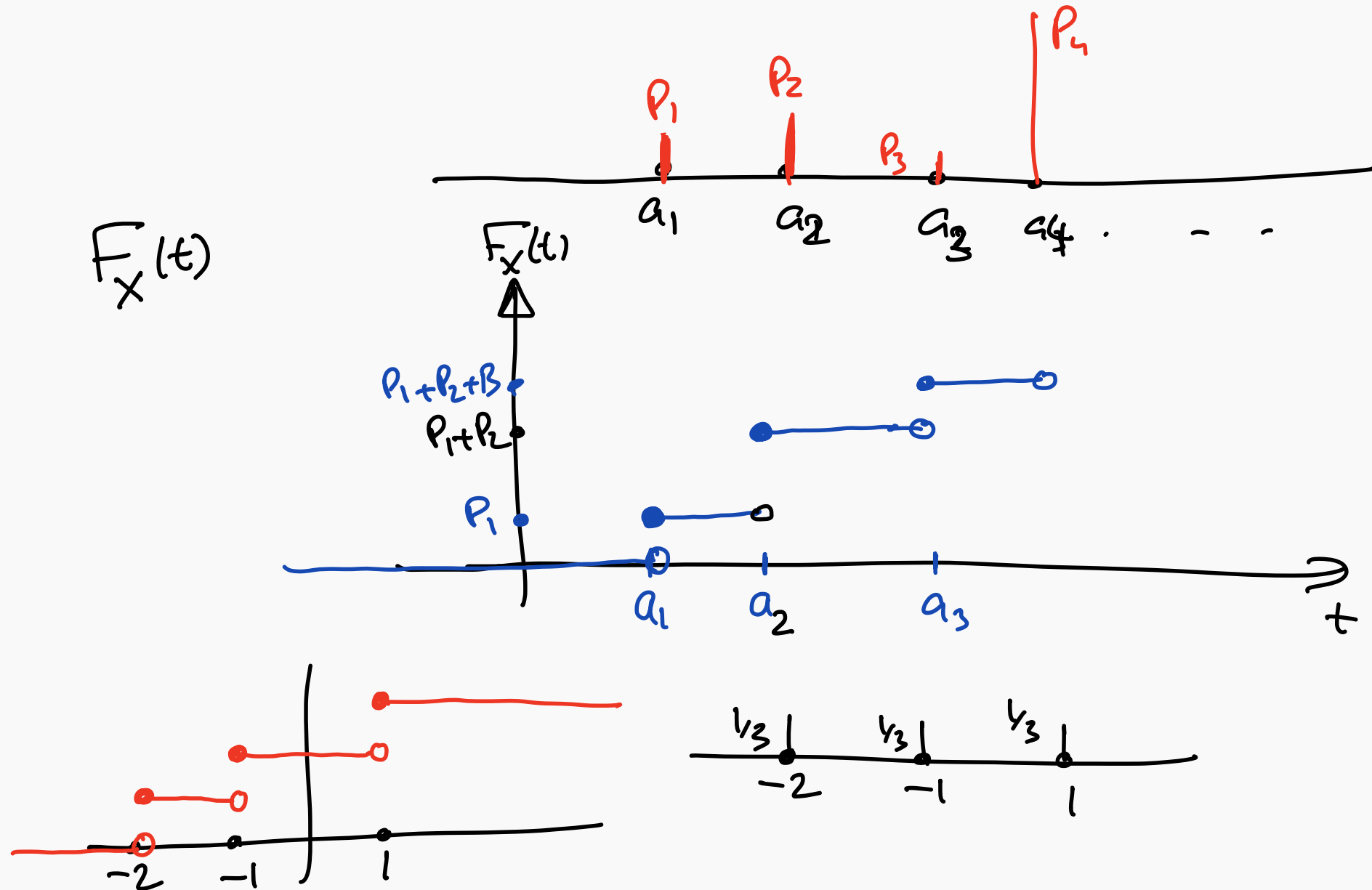
$$P[X \leq s] \leq P[X \leq t]$$

$$A_s = \{X \leq s\} \subseteq \{X \leq t\} = A_t$$



The distribution function of a general discrete random variable

X



Properties of the distribution function

Theorem

The probability distribution function enjoys the following properties:

1. F_X is (non-strictly) increasing: if $t_1 \leq t_2$, then $F_X(t_1) \leq F_X(t_2)$. ✓
2. F_X is right-continuous, that is, for every $t \in \mathbb{R}$:

$$\lim_{s \rightarrow t+} F_X(s) = F_X(t).$$

3. F_X has limits at $\pm\infty$, namely, $F_X(-\infty) = 0$ and $F_X(\infty) = 1$.



Computing probabilities using the distribution function

$$X \rightsquigarrow F_X(t)$$

Theorem

For a random variables X and $t \in \mathbb{R}$, we have

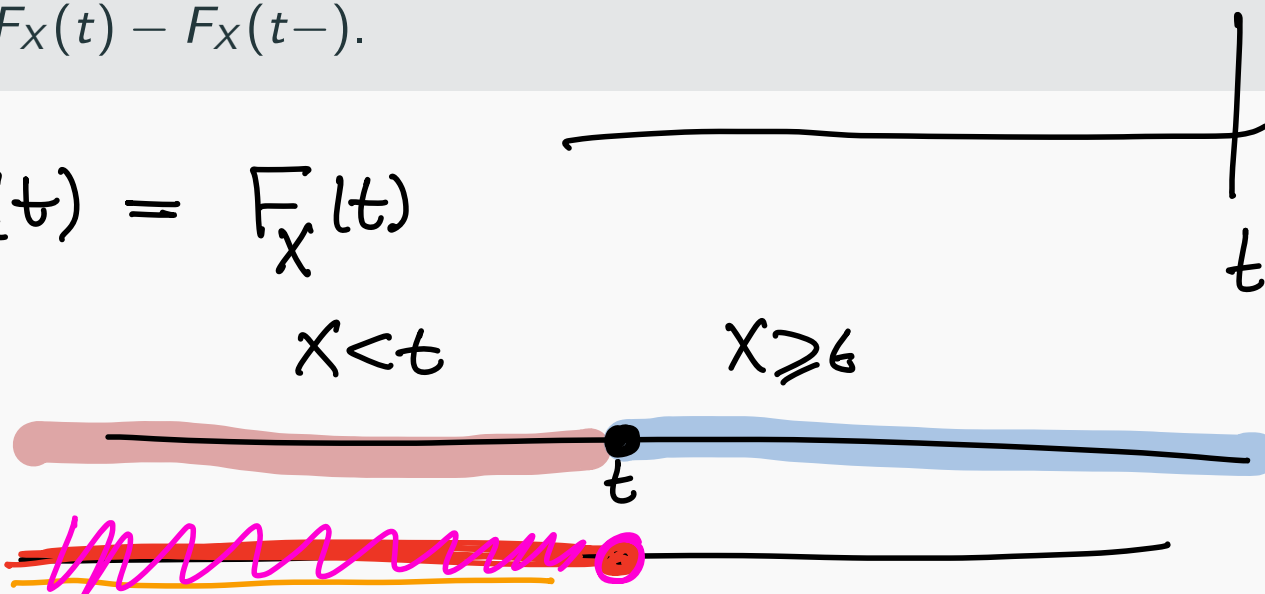
1. $\mathbb{P}[X < t] = F_X(t-) := \lim_{s \rightarrow t-} F_X(s).$

2. $\mathbb{P}[X \geq t] = 1 - F_X(t-).$

$$\mathbb{P}(X \geq t) = 1 - \mathbb{P}(X < t)$$

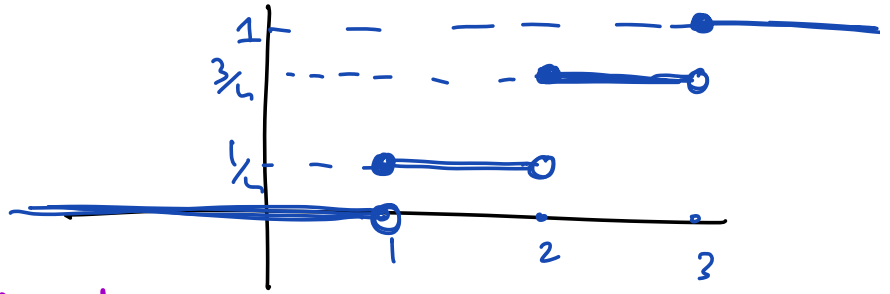
3. $\mathbb{P}[X = t] = F_X(t) - F_X(t-).$

$$\mathbb{P}(X \leq t) = F_X(t)$$



$$\mathbb{P}(X = t) = \mathbb{P}(X \leq t) - \mathbb{P}(X < t)$$

X RV



$$P(X \leq 1.5) = \frac{1}{4}$$

$$P(X > 2) = 1 - P(X \leq 2) = 1 - \frac{3}{4} = \frac{1}{4}$$

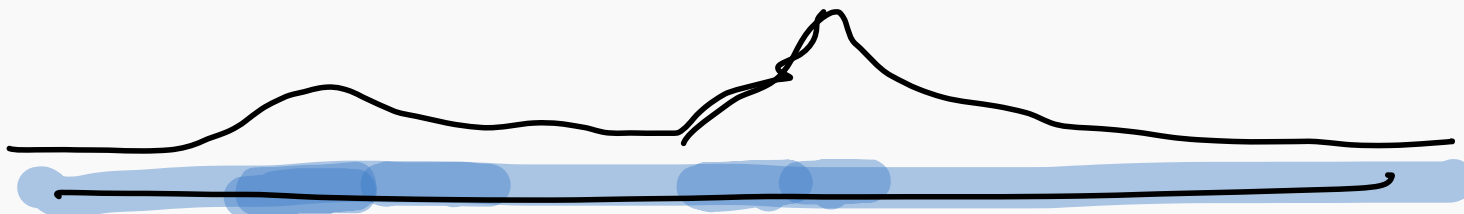
$$\begin{aligned} P(1 \leq X \leq 2) &= P(X \leq 2) - P(X < 1) \\ &= \frac{3}{4} - 0 = \frac{3}{4} \end{aligned}$$

The idea of continuous random variables

Point mass



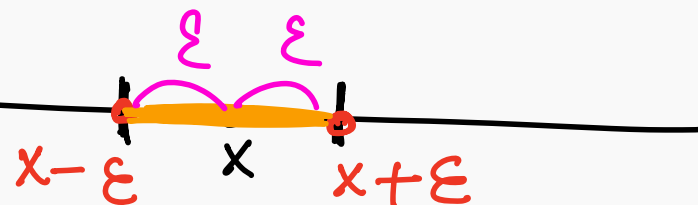
density



density function

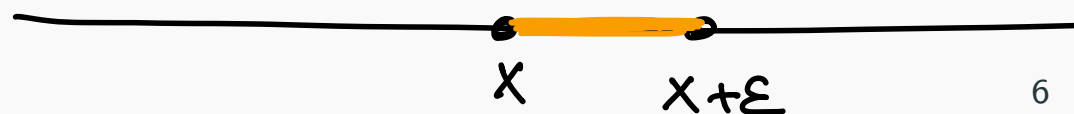
$$\delta(x) = \lim_{\epsilon \rightarrow 0} \frac{\text{mass between } (x-\epsilon) \text{ and } (x+\epsilon)}{2\epsilon}$$

length of little rod



ε)

$$\delta(x) = \lim_{\epsilon \rightarrow 0} \frac{\text{mass between } x \text{ and } x+\epsilon}{\epsilon}$$



Third version



$$F(x) = \text{mass up to } x$$

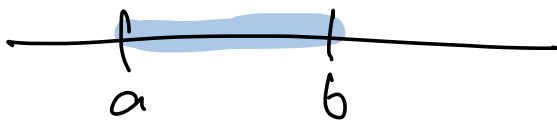
$$F(x+\epsilon) = \text{mass up to } x+\epsilon$$

$$\begin{array}{l} \text{mass between } x \\ \text{and } x+\epsilon \end{array} = F(x+\epsilon) - F(x)$$

$$\delta(x) = \lim_{\epsilon \rightarrow 0} \frac{F(x+\epsilon) - F(x)}{\epsilon} = F'(x).$$

If $\delta(x)$ is given

$$F(t) = \int_{-\infty}^t \delta(x) dx$$



$$\begin{array}{l} \text{mass between} \\ a \text{ \& } b \end{array} = \int_a^b \delta(x) dx$$

Definition (version 1)

Definition

A random variable $X : \Omega \rightarrow \mathbb{R}$ is called *continuous* if there exists a non-negative function $f_X : \mathbb{R} \rightarrow \mathbb{R}$, called the probability density function of X , such that for all $t \in \mathbb{R}$, we have

$$\mathbb{P}(X \leq t) = F_X(t) = \int_{-\infty}^t f_X(x) dx.$$

_____t

Definition (version 2)

Definition

A random variable $X : \Omega \rightarrow \mathbb{R}$ is called *continuous* if there exists a non-negative function $f_X : \mathbb{R} \rightarrow \mathbb{R}$, called the *probability density function* of X , such that for all $s \leq t$, we have

$$\mathbb{P}[s \leq X \leq t] = \int_s^t f_X(x) dx.$$

Relation between the density and the distribution functions

Theorem

If F_X is differentiable, then

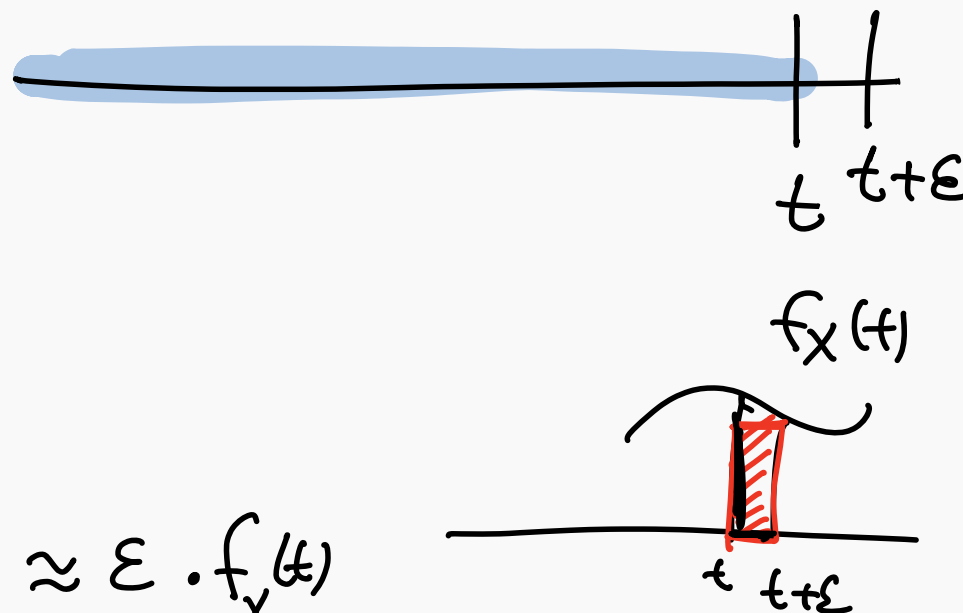
$$f_X(t) = \frac{d}{dt} F_X(t).$$

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

$$F_X(t+\varepsilon) = \int_{-\infty}^{t+\varepsilon} f_X(x) dx$$

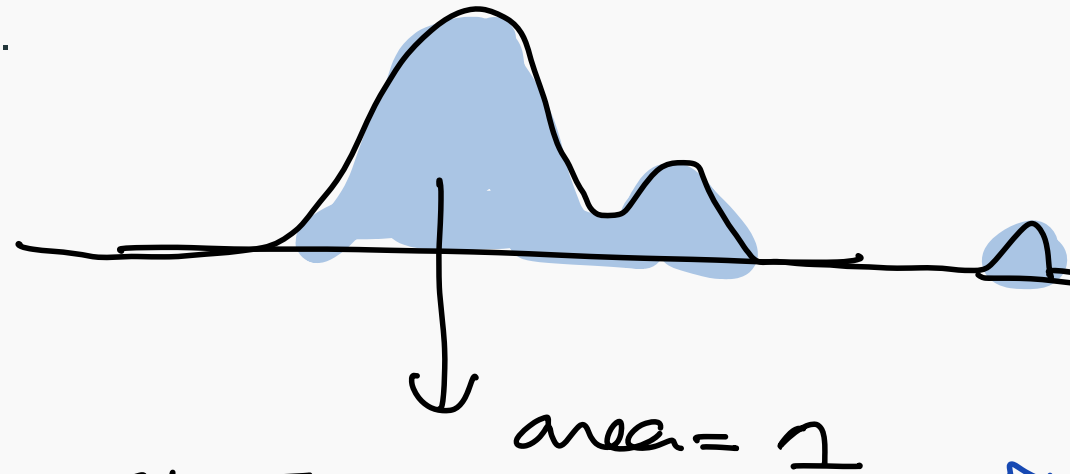
$$F_X(t+\varepsilon) - F_X(t) = \int_t^{t+\varepsilon} f_X(x) dx \approx \varepsilon \cdot f_X(t)$$

$$\lim_{\varepsilon \rightarrow 0} \frac{F_X(t+\varepsilon) - F_X(t)}{\varepsilon} = f_X(t).$$



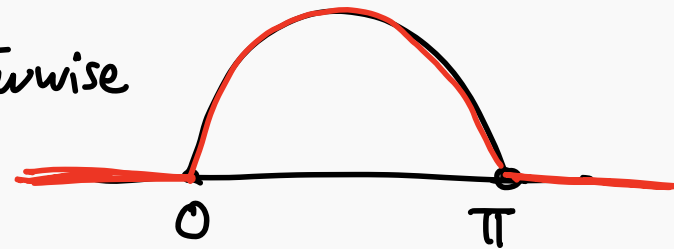
Properties of the density function and comparison with discrete case

- (Non-negativity: $f_X \geq 0$.)
- Total mass one: $\int_{-\infty}^{\infty} f_X(x) dx = 1$.



example

$$f_X(t) = \begin{cases} \frac{1}{2} \sin t & 0 \leq t \leq \pi \\ 0 & \text{otherwise} \end{cases}$$



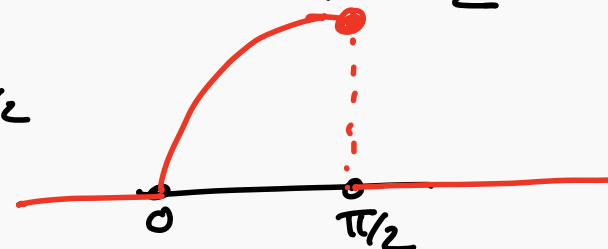
$$\int_{-\infty}^{\infty} f_X(x) dx = 1$$

$$\int_{-\infty}^{\infty} f_X(x) dx = \int_0^{\pi} \sin x dx$$

$$= -\cos x \Big|_0^{\pi} = -\cos \pi + \cos 0 = 1 + 1 = 2$$

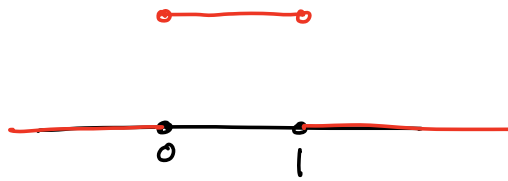
$$\int_0^{\pi/2} \sin x dx = 1$$

$$f_X(t) = \begin{cases} \sin t & 0 \leq t \leq \pi/2 \\ 0 & \text{otherwise} \end{cases}$$



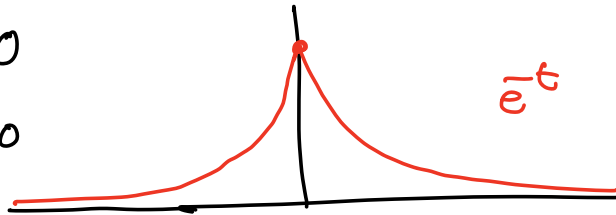
disturb.

$$f_X(t) = \begin{cases} 1 & 0 \leq t \leq 1 \\ 0 & \text{otherwise.} \end{cases}$$



$$f_X(t) = \begin{cases} \bar{e}^{-t/2} & t \geq 0 \\ e^{t/2} & t \leq 0 \end{cases}$$

$$= \bar{e}^{-|t|}$$



$$\int_{-\infty}^{+\infty} f_X(x) dx = 2 \int_0^{\infty} \bar{e}^x dx = -2 \bar{e}^x \Big|_0^{\infty}$$

$$= -2(0 - 1) = 2$$

}

Uniform random variables

Definition

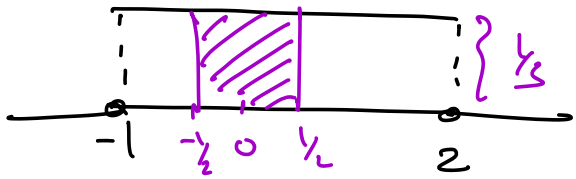
A random variable X has uniform distribution over the interval $[a, b]$, if its probability density function is given by

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



X uniform RV on the interval $[-1, 2]$

compute $P(-\frac{1}{2} \leq X \leq \frac{1}{2})$



$$P\left(\underbrace{-\frac{1}{2} \leq X \leq \frac{1}{2}}\right) = \int_{\text{interval}} f_X(x) dx$$

$$= \int_{-\frac{1}{2}}^{\frac{1}{2}} \frac{1}{3} dx = \frac{1}{3} x \Big|_{-\frac{1}{2}}^{\frac{1}{2}} = \frac{1}{3}$$

$$P\left(-2 \leq X \leq \frac{1}{2}\right) = \int_{\text{interval}} f_X(x) dx$$

$$= \int_{\substack{-2 \\ -5}}^{\frac{1}{2}} \frac{1}{3} dx = \frac{1}{3} x \Big|_{-2}^{\frac{1}{2}} = \frac{1}{3} \left(\frac{1}{2} + 2\right)$$

$$\frac{2.5}{3}$$

$$\int_{-1}^{\frac{1}{2}} \frac{1}{3} dx = \frac{1}{3} \left(\frac{3}{2}\right) = \frac{1}{2}$$

$$\frac{1}{3} \left(5 + \frac{1}{2}\right) = \frac{5.5}{3}$$