inline=symbol

Probability and Random Processes

Keivan Mallahi-Karai 30 September 2022

Jacobs University

Distribution function of a random variable (for confinuous Canlon Umille)

Definition

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

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

Know
$$F_{X}(t) = P(X \le t)$$
 $P(X > t) = 1 - F_{X}(t)$

$$P(X < t) = \lim_{S \to t^{-}} F_{X}(s)$$

$$P(X \le t) = F_{X}(t) = \lim_{S \to t^{+}} F_{X}(s)$$

$$\mathbb{P}(X=t) = F_X(t) - \lim_{s \to t^-} F_X(s).$$

 $P(X=t) = F_X(t) - \lim_{S \to t-} F_X(s).$ Conlinor X
with lensit finite $P(a \le X \le b) = \int_{a}^{b} f_X(x) dx.$

Continuous random variables and the density function

Definition

Let $X : \Omega \to \mathbb{R}$ be a *continuous random variable* with the probability density function f_X . The for all real values of $a \le b$ we have

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

Definition (version 2)

Definition

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

$$\mathbb{P}\left[s \leq X \leq t\right] = \int_{s}^{t} f_{X}(x) dx.$$

Relation between the density and the distribution functions

Theorem

If F_X is differentiable, then

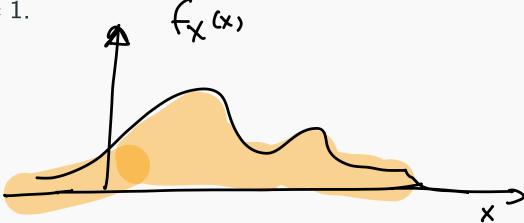
$$F_{X}(t) = \frac{d}{dt}F_{X}(t).$$

$$F_{X}(t) = \int_{-\infty}^{t} f_{X}(x) dx$$

Properties of the density function and comparison with discrete case

• (Non-negativity: $f_X \ge 0$.

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



Uniform random variables

Definition

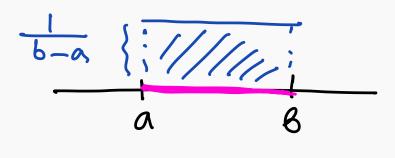
A random variable X has uniform distribution over the interval [a, b], if its probability density function is given by

$$f_X(t) = egin{cases} rac{1}{b-a} & ext{if} \ a \leq t \leq b \ 0 & ext{otherwise} \end{cases}$$

warm -
$$\varphi$$

X uniform [-1,1]

$$f_{\chi}(t) = \begin{cases} \frac{1}{2} & -1 \leq t \leq 1 \\ 0 & \text{otherwise,} \end{cases}$$



Distribution function of a uniform random variable

The distribution function of a random variable X with uniform distribution over the interval [a, b] is given by

$$X$$
 random vaniable

Uniform distribute over $[0,1]$
 $Y = X^2$

Y is a random vaniable.

P(CYS)= 1

Y is not uniform

continos

X random vaniable my I=f(X) a what i

The densit

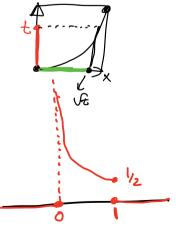
function

Question Find the desit function of 7?

$$F_{Y}(t) = P(Y \leq t) = \begin{cases} 0 \\ ? \\ 1 \end{cases}$$

 $F(t) = P(Y \le t) = P(X \le \sqrt{\epsilon})$

$$\left(\mathbf{y}(t) = \frac{\mathbf{d}}{\mathbf{d}t} \sqrt{t} = \frac{\mathbf{d}}{\mathbf{d}t} (t^{V_2}) = \frac{1}{2\sqrt{t}}$$



Density function of f(X)

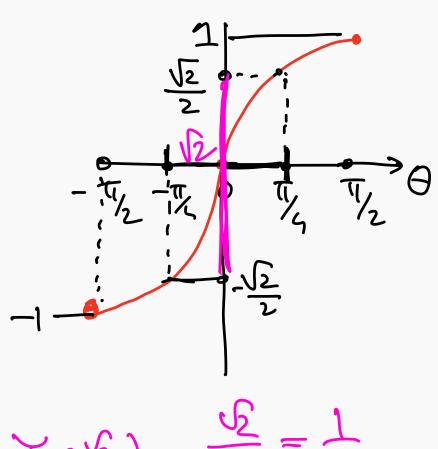
Example

Suppose θ has uniform distribution over the interval $[-\pi/2, \pi/2]$ and $X = \sin \theta$. Compute the probability density function of X.

Y=Sin
$$\Theta$$

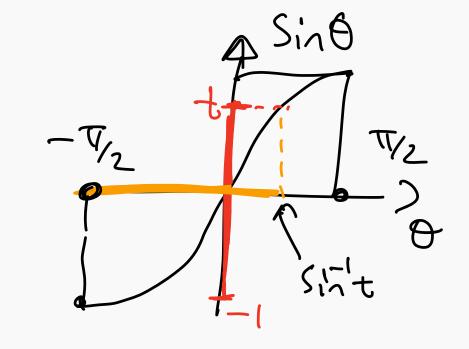
O uniform $[-\sqrt{2}, \sqrt{2}]$
 $-1 \le Y \le 1$
 $P(-\sqrt{2} \le Y \le \sqrt{2}) = \frac{1}{2}$

If Y were uniform P



density functai

$$F_{y}(t) = P(Y \leq t)$$



$$P(Y \leq t) = P(Sin \theta \leq t)$$

$$= P(\theta \leq \sin t) = \frac{\sin t + \frac{\pi}{2}}{\pi} = \frac{1}{2} + \frac{1}{\pi} \sin t$$

$$F_{\gamma}(t) = \frac{1}{2} + \frac{1}{\pi} \sin^{2} t$$

$$F_{\gamma}(t) = \frac{1}{2} + \frac{1}{\pi} \sin^{2} t$$

$$= \frac{1}{\pi} \cdot \frac{1}{1 + 2}$$

Example

Example

Let X be uniformly distributed in [-1,1], and $Y=X^2$. Compute the distribution function of Y.

Exponential random variables / Exponential distribution

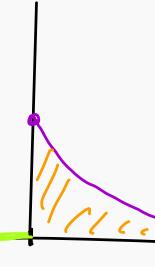
Definition

A continuous random variable X has an exponential distribution with parameter λ if

$$f_X(t) = \begin{cases} \lambda e^{-\lambda t} & \text{if } t \geq 0 \\ 0 & \text{otherwise} \end{cases}$$

$$\int_{-\infty}^{+\infty} f_{X}(t)dt = \int_{0}^{\infty} \lambda e^{-\lambda t} dt$$

$$\lambda \cdot \frac{e^{\lambda t}}{-\lambda} \Big|_{0}^{\infty} - e^{\lambda t} \Big|_{0}^{\infty} = 0 - (-1) = 1$$



Distribution function of an exponential random variable

Theorem

The distribution function of a random variable X with geometric distribution with parameter λ is given by

$$F_X(t) = egin{cases} 1 - e^{-\lambda t} & \textit{if } t \geq 0 \\ 0 & \textit{otherwise} \end{cases}$$

$$F_{X}(t) = P(X \le t) = 0 \quad \text{if } t \le 0$$
if the
$$P(X \le t) = \int_{0}^{t} f_{X}(x) dx = \int_{0}^{t} \lambda e^{\lambda x} dx$$

$$= -e^{\lambda x} \Big|_{0}^{t} = -e^{\lambda t} + 1 = 1 - e^{\lambda t}$$

$$P(X > t) = P(X > t) = 1 - F_{X}(t) = e^{\lambda t}$$

Condition Probly

X random vaniable with exp. distribute with parameter 2:

$$P(x>t) = e^{\lambda t} P(x>t) = e^{\lambda}$$

$$t_2 > t_1$$

$$\frac{e}{P(X>5|X>1)} = \frac{P((X>1))}{P(X>1)}$$

$$= \frac{e}{e^{\lambda}} = e^{\lambda}$$

$$= \frac{P((X>5))(X>1)}{P(X>1)}$$

$$P(X > t_1 + t_2 | X > t_1) = \frac{P(X > t_1 + t_2 \cap X > t_1)}{P(X > t_1 + t_2 \cap X > t_1)}$$

$$= \frac{\mathbb{P}(X > t_1 + t_2)}{\mathbb{P}(X > t_1 + t_2)} = \frac{e^{\lambda(t_1 + t_2)}}{e^{\lambda t_1}} = \frac{e^{\lambda t_2}}{e^{\lambda t_1}} = e^{\lambda t_2}$$

$$= \mathbb{P}(X > t_2)$$

Exponential random variables are Memoryless

Gaussian (Normal) random variables

Gaussian (Normal) random variables

Definition

A continuous random variable X is said to have standard Gaussian or standard normal distribution if the probability density function of X is given by

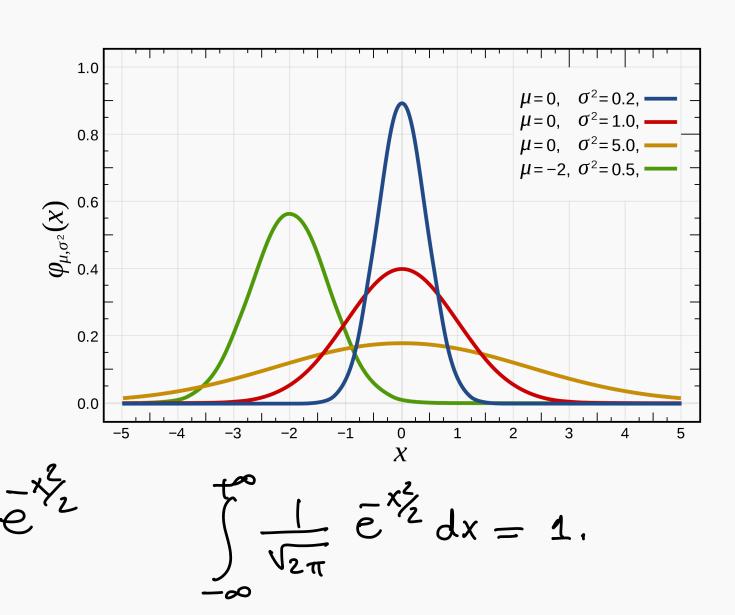
with possible
$$\mu,\sigma$$
)
$$f_X(t) = \frac{1}{\sqrt{2\pi}\sigma}e^{-\frac{\left(\frac{x-\mu}{\sigma}\right)^2}{2}}.$$

A random variable with normal distribution with parameters $\mu = 0$ and $\sigma = 1$ is called a *standard normal distribution*.

The density function of a standard normal random variable is given by

$$f_X(t) = \frac{1}{\sqrt{2\pi}}e^{-\frac{x^2}{2}}.$$

Density functions



The first of the exprending a cloud firm

F(x) cannot be exprend in a cloud firm

Using plyonil, trig, exp. log.

F(1)

 $\int_{-\infty}^{\infty} e^{x^{2}/2} dx = \sqrt{2\pi}$