

Lecture #2: Random Variables

Definition - A random variable is a function X from Ω to \mathbb{R} , i.e., $X: \Omega \rightarrow \mathbb{R}$.

Example:

Flip a coin twice. $X(\omega) = \text{sum of heads that appear}$.

Notation:

$$\begin{aligned} P(X=2) &= P(\{\omega: X(\omega)=2\}) \\ &= P(X^{-1}(2)) \\ &= \frac{1}{4} \end{aligned}$$

$$\begin{aligned} P(X=1) &= P(\{\omega: X(\omega)=1\}) \\ &= P(X^{-1}(1)) \\ &= P(\{(0,1) \cup (1,0)\}) \\ &= \frac{1}{2} \end{aligned}$$

$$\begin{aligned} P(X=0) &= P(\{\omega: X(\omega)=0\}) \\ &= P(\{(0,0)\}) \\ &= \frac{1}{4} \end{aligned}$$

$$\begin{aligned} P(X=-1) &= P(\{\omega: X(\omega)=-1\}) \\ &= P(\emptyset) \\ &= 0 \end{aligned}$$

$$\begin{aligned} P(X \geq 1) &= P(\{\omega: X(\omega) \geq 1\}) \\ &= \frac{3}{4} \end{aligned}$$

defines new
probability on
 $\Omega' = \{0, 1, 2\}$

Example:

$$1_A(\omega) = \begin{cases} 1 & \text{if } \omega \in A \\ 0 & \text{if } \omega \notin A \end{cases}$$

is a random variable.

Definition - The distribution of a random variable X , denoted

S_X is given by

$$S_X((a, b]) = P(X \in (a, b]) \\ = P(\{\omega \in \Omega : X(\omega) \in (a, b]\}).$$

The cumulative distribution is a function $F_X: \mathbb{R} \rightarrow [0, 1]$ defined by

↑ abbreviated CDF

$$F_X(x) = P(X \leq x) \\ \Rightarrow S_X((a, b]) = F_X(b) - F_X(a)$$

If F_X is differentiable, there exists a probability density function $f(x)$ such that

$$S_X((a, b]) = F_X(b) - F_X(a) \\ = \int_a^b \frac{dF}{dx} dx \\ = \int_a^b f(x) dx$$

* The key point is that a distribution defines a probability on \mathbb{R} !

Example:

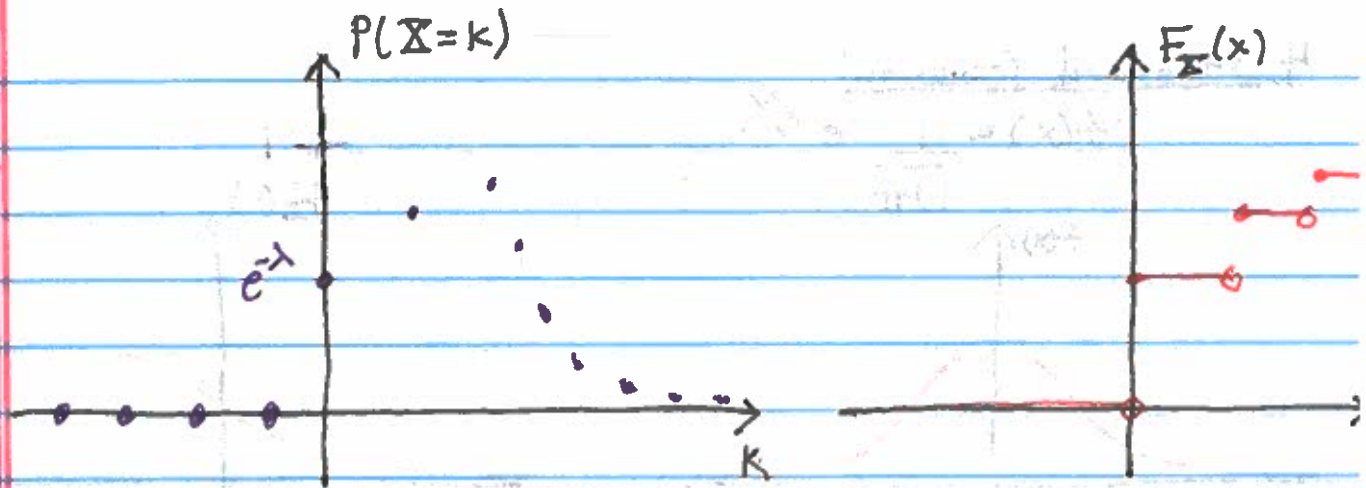
1. Poisson Distribution -

$$P(X=k) = \frac{\lambda^k}{k!} e^{-\lambda}, \quad k \in \{0, 1, 2, \dots\}$$

$$\sum_{i=0}^{\infty} \frac{\lambda^i}{i!} e^{-\lambda} = e^{-\lambda} \sum_{i=0}^{\infty} \frac{\lambda^i}{i!} = e^{-\lambda} e^{\lambda} = 1.$$

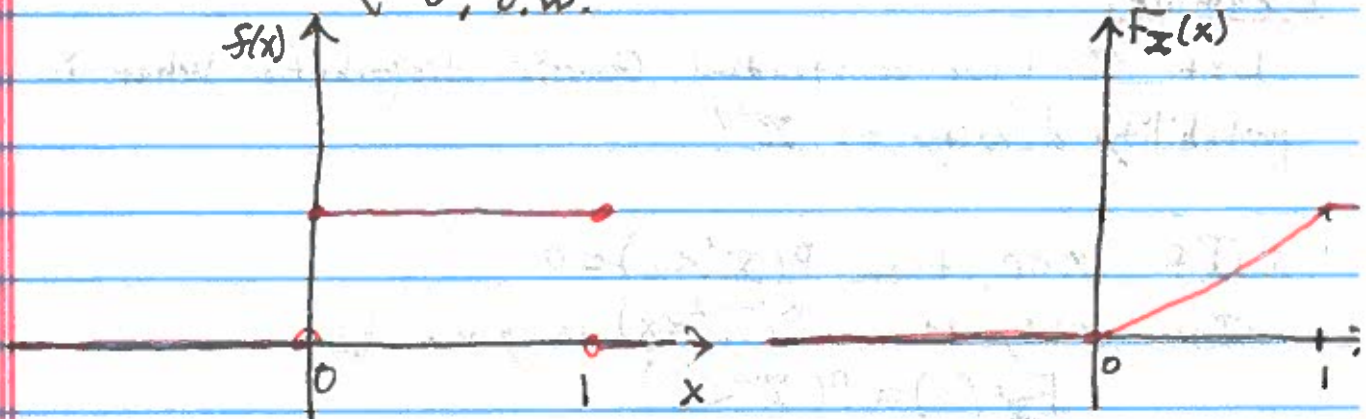
The CDF is

$$F_X(x) = \begin{cases} 0, & x < 0 \\ \sum_{j=0}^k \frac{\lambda^j}{j!} e^{-\lambda} & \text{if } k \leq x < k+1 \end{cases}$$



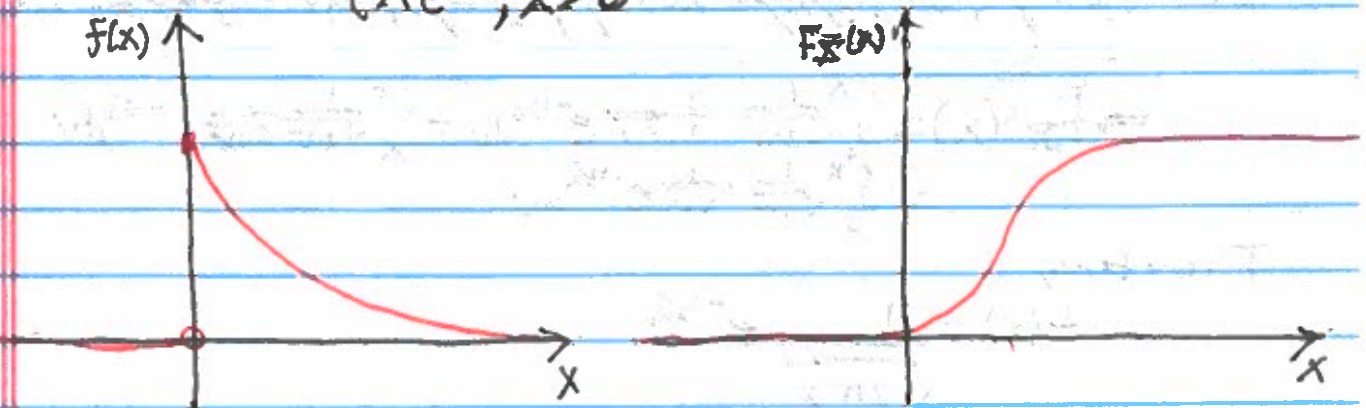
2. Uniform Distribution -

$$f(x) = \begin{cases} 1, & 0 \leq x < 1 \\ 0, & \text{o.w.} \end{cases}$$



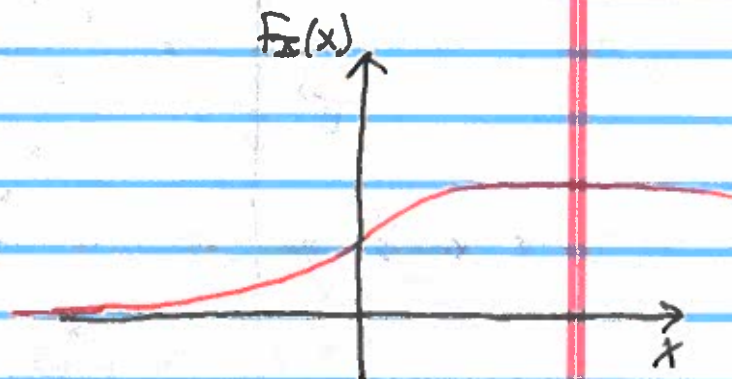
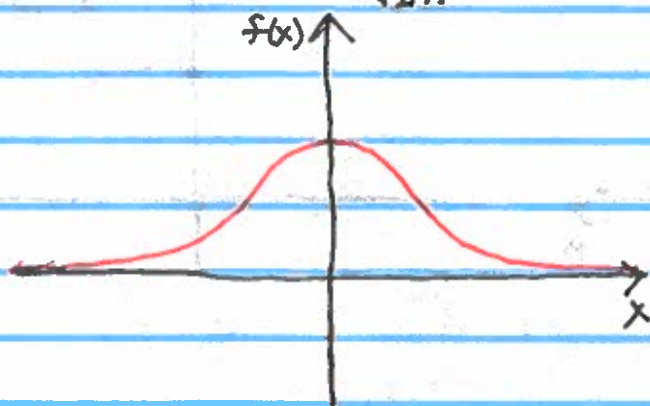
3. Exponential Distribution -

$$f(x) = \begin{cases} 0, & x \leq 0 \\ \lambda e^{-\lambda x}, & x > 0 \end{cases}$$



4. Standard Gaussian:

$$f(x) = \frac{1}{\sqrt{2\pi}} e^{-x^2/2}$$



Example:

Let X have a standard Gaussian distribution. What is the probability density of X^2 ?

- If $x < 0$, then $P(X^2 < x) = 0$

- If $x > 0$, then $P(X^2 < x)$ is given by

$$\begin{aligned} F_{X^2}(x) &= P(X^2 < x) \\ &= P(0 < X^2 < x) \\ &= P(0 < X < \sqrt{x} \text{ or } -\sqrt{x} < X < 0) \\ &= \int_0^{\sqrt{x}} \frac{1}{\sqrt{2\pi}} e^{-y^2/2} dy + \int_{-\sqrt{x}}^0 \frac{1}{\sqrt{2\pi}} e^{-y^2/2} dy \end{aligned}$$

$$\text{Let } v = y^2, \quad dv = 2y dy = 2\sqrt{v} dy, \quad v = y^2, \quad dv = 2y dy = -2\sqrt{v} dv$$

$$\begin{aligned} \Rightarrow F_{X^2}(x) &= \int_0^x \frac{1}{\sqrt{2\pi}} \cdot \frac{1}{2\sqrt{v}} e^{-v/2} dv - \int_x^0 \frac{1}{\sqrt{2\pi}} \frac{1}{2\sqrt{v}} e^{-v/2} dv \\ &= \int_0^x \frac{1}{\sqrt{2\pi}} \cdot \frac{1}{\sqrt{v}} e^{-v/2} dv. \end{aligned}$$

Therefore,

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