PROBABILITY THEORY AND RANDOM PROCESSES (MA225)

LECTURE SLIDES Lecture 06 (August 05, 2019)

Discrete Random Variable

Def: [Discrete RV] A random variable is said to have discrete distribution if there exists an atmost countable set $S_X \subset \mathbb{R}$ such that P(X = x) > 0 for all $x \in S_X$ and $\sum_{x \in S_X} P(X = x) = 1$. S_X is called

the support of X.

Def: Define a function $f_X : \mathbb{R} \to [0,1]$ by

$$f_X(x) = \begin{cases} P(X = x) & \text{if } x \in S_X \\ 0 & \text{otherwise} \end{cases}$$

The function f_X is called the probability mass function of X.

Example 1, X is discrete, Example 2, X is discrete, Example 3, X_1 is not discrete, but X_2 is discrete.

Remarks

 \blacktriangleright For a discrete random variable X,

$$F_X(x) = \sum_{\substack{y \in S_X \\ y \le x}} f_X(y).$$

 \blacktriangleright For a discrete random variable X,

$$f_X(x) = F_X(x) - F_X(x-).$$

Properties of PMF

①
$$f_X(x) \ge 0$$
 for all $x \in \mathbb{R}$.

Theorem: Suppose a real valued function $h : \mathbb{R} \to \mathbb{R}$ satisfies the following conditions:

- ① $h(x) \ge 0$ for all $x \in \mathbb{R}$. $D = \{x : h(x) > 0\}$ is atmost countable.

Then $h(\cdot)$ is a probability mass function of some discrete random variable.

Example

- ① (Bernoulli Distribution) $X \sim Bernoulli(p)$: $S_X = \{0, 1\}$, $f_X(0) = 1 p$, $f_X(1) = p$.
- ② (Binomial Distribution) $X \sim Bin(n, p)$: $S_X = \{0, 1, ..., n\}$, $f_X(k) = \binom{n}{k} p^k (1-p)^{n-k}$.
- ③ (Geometric Distribution) $X \sim Geo(p)$: $S_X = \{0, 1, ...\}$, $f_X(k) = p(1-p)^k$.
- ① (Poisson Distribution) $X \sim Poi(\lambda)$ $(\lambda > 0)$: $S_X = \{0, 1, ...\}$, $f_X(k) = \frac{e^{-\lambda}\lambda^k}{k!}$.

Application

Example 1: Suppose that an airplane engine will fail, when in flight, with probability 1-p independently from engine to engine. The airplane will make a successful flight if at least 50 percent of its engines remain operating. For what values of p is a four engine plane preferable to a two engine plane? (Ans: p > 2/3)

Continuous Random Variable

Def: A random variable is said to have a continuous distribution if there exists a non-negative integrable function f_X : $\mathbb{R} \to [0, \infty)$ such that

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

for all $x \in \mathbb{R}$. The function f_X is called the probability density function. The set $S_X = \{x \in \mathbb{R} : f_X(x) > 0\}$ is called support of X.

Example 1: (Exponential Distribution: $Exp(\lambda)$)

$$F_X(x) = egin{cases} 1 - e^{-\lambda x} & \text{if } x > 0 \\ 0 & \text{otherwise.} \end{cases}$$

Example 2: (Uniform Distribution: U(a, b))

$$F_X(x) = \begin{cases} 0 & \text{if } x < a \\ \frac{x-a}{b-a} & \text{if } a \le x < b \\ 1 & \text{if } x \ge b. \end{cases}$$

Example 3: (Normal Distribution: $N(\mu, \sigma^2)$)

$$F_X(x) = \int_{-\infty}^{x} \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{(t-\mu)^2}{2\sigma^2}} dt \quad \text{if } -\infty < x < \infty.$$

Remarks on CRV

- ▶ For a continuous random variable X, P(X = a) = 0 for all $a \in \mathbb{R}$.
- ► CDF of a continuous random variable is continuous.
- ▶ PDF is not unique.
- ► Support of a continuous random variable is not unique.
- $P(a \le X \le b) = \int_a^b f_X(t) dt.$
- ▶ $f_X(x)$ is not P(X = x).