

# Section 5 Discrete & Continuous Random Variable

TA: Yasi Zhang

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## 1 Random Variable

### 1.1 Discrete Random Variable

**[Prob (Mass) Function]** of a discrete R.V.  $X$  is defined as

$$f_X(x) = P(X = x), \forall x \in \mathbb{R}$$

**Properties of pmf:**

(i)  $0 \leq P(X = x) \leq 1$  for all  $x \in \mathbb{R}$ .

(ii)  $\sum_x P(X = x) = 1$ .

**Relationship between pmf and cdf:**

pmf  $\rightarrow$  cdf:  $F_X(x) = P(X \leq x) = \sum_{y \leq x} P(X = y)$ ;

cdf  $\rightarrow$  pmf:  $P(X = x_i) = \begin{cases} F(x_i) & i = 1 \\ F(x_i) - F(x_{i-1}) & i > 1. \end{cases}$

**Remark 1** *pmf and cdf are in one-to-one correspondence.*

#### 1.1.1 Practice

Derive the cdf of  $X$  if

$$P(X = x) = \begin{cases} 0.2 & X = 0 \\ 0.5 & X = 2 \\ 0.3 & X = 8 \\ 0 & otherwise \end{cases}$$

## 1.2 Discrete Distribution Examples

### 1. Discrete Uniform Distribution

$$P(X = x) = \frac{1}{k}, \quad x = 1, 2, \dots, k$$

### 2. Bernoulli Distribution

$$P(X = x) = \begin{cases} p & x = 1 \\ 1 - p & x = 0 \end{cases}$$

**Remark 2** Toss a coin and the prob of getting a head ( $X = 1$ ) is  $p$ .

$Ber(p)$ :  $X = 1$  head up;  $X = 0$  tail up.

### 3. Binomial Distribution

$$P(X = x) = C_n^x p^x (1 - p)^{n-x}, \quad x = 0, 1, \dots, n$$

**Remark 3** Toss a coin  $n$  times and the prob of getting a head each time is  $p$ .

$B(n, p)$ : Number of getting a head.  $X = 0, 1, \dots, n$

**Remark 4** If  $Y_1, Y_2, \dots, Y_n$  are independent and all follow a Bernoulli distribution  $Ber(p)$ , then  $Y_1 + Y_2 + \dots + Y_n$  follows  $B(n, p)$ .

### 4. Poisson Distribution

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

#### 1.2.1 Practice

1. Verify that  $\sum_x P(X = x) = 1$  when  $X$  follows Binomial distribution or Poisson distribution.
2. Calculate the probability of getting at least 2 heads in the experiment of tossing a fair coin 10 times.
3. Calculate  $P(2 \leq X \leq 4)$  when  $X$  follows Poisson distribution when  $\lambda = 1$ .

#### 1.2.2 Take-Home Practice

Derive the cdf of the distributions mentioned above.

### 1.3 Continuous Random Variable

**[Prob Density Function]** Suppose the (cumulative) distribution function  $F_X(x)$  of a continuous R.V.  $X$  is absolutely continuous, then there exists a prob density function  $f_X(x)$  such that

$$F_X(x) = \int_{-\infty}^x f_X(u)du, \quad \forall x \in (-\infty, \infty).$$

**Remark 5** Normally, for a C.R.V.  $X$ ,  $P(X = x) = 0$  for any  $x$ .

Therefore,  $P(a < X < b) = P(a \leq X \leq b)$  for a C.R.V.  $X$ .

**Remark 6** An easy way to understand pdf: The magnitude of  $f_X(x)$  at each  $x$  indicates the probability that  $X$  takes values in a small neighborhood of each point  $x$ , i.e.

$$P(X \in [x, x + dx]) = f_X(x)dx$$

**Properties of pdf:**

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

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

**Remark 7** Can  $f_X(x) > 1$ ? Yes.

**Relationship between pdf and cdf:**

pdf  $\rightarrow$  cdf:  $F_X(x) = P(X \leq x) = \int_{-\infty}^x f_X(u)du$ ;

cdf  $\rightarrow$  pdf:  $f_X(x) = \frac{dF_X(x)}{dx}$ .

**Remark 8** Is  $f_X(x)$  unique for a given  $F_X(x)$ ? No!

Is  $F_X(x)$  unique for a given  $f_X(x)$ ? Yes!

Again, we emphasize that the fact that  $P(X = x) = 0$  for any given  $x$  for a C.R.V. allows us to change the value of the pdf of a continuous random variable  $X$  at a single point without altering the distribution of  $X$ . For instance, the pdf

$$f_X(x) = \lambda e^{-\lambda x}, \quad x > 0$$

can be written as

$$f_X(x) = \lambda e^{-\lambda x}, \quad x \geq 0$$

without changing  $F_X(x)$  for any  $x$ . We observe these two functions differ only at  $x = 0$  and  $P(X = 0) = 0$ . **More generally, if two pdf's of continuous random variables differ only on a set having probability zero, the two corresponding probability distribution functions are exactly the same.** Unlike the continuous random variables, the pmf of a discrete random variable cannot be changed at any point, since a change in such a pmf alters the distribution of probability.

### 1.3.1 Practice

1. For each of the following, determine the value of  $c$  that makes  $f_X(x)$  a pdf.

(a)  $f_X(x) = c(x+1)^2, -1 < x < 1$

(b)  $f_X(x) = ce^{-2x}, 0 < x < \infty$

(c)  $f_X(x) = c \cos x, 0 < x < \frac{\pi}{2}$

**Remark 9** When the type of  $f_X(x)$  (like exponential, triangular, or composite) is given, then  $f_X(x)$  can be determined (because we only need to calculate the constant  $c$ ).

Think about how to derive the  $c$  if  $f_X(x) = ce^{-x^2/2}, x \in \mathbb{R}$ .

1.1 Is it possible to construct a pdf  $f_X(x)$  from any nonnegative  $g(x)$  with a finite integral, i.e.  $0 < \int_{\mathbb{R}} g(x)dx < \infty$ ?

1.2 Is it possible to construct a symmetric pdf  $f_X(x)$  from any nonnegative  $g(x)$  with a finite integral, i.e.  $0 < \int_{\mathbb{R}} g(x)dx < \infty$ ?

2.

Find  $f_Y(y)$ , the pdf of  $Y$ , where

$$F_Y(y) = P(Y \leq y) = \begin{cases} 1 - \frac{1}{y^2}, & 1 \leq y < \infty \\ 0 & \text{otherwise.} \end{cases}$$

3. Suppose a C.R.V.  $X$  has a pdf  $f_X(x) = \begin{cases} |x| & \text{for } -1 < x < 1 \\ 0 & \text{otherwise} \end{cases}$ ,

Find  $P(X < \frac{1}{2})$ .

## 1.4 Gaussian Integral

The Gaussian integral, also known as the Euler–Poisson integral, is the integral of the Gaussian function  $f(x) = e^{-x^2}$  over the entire real line. Named after the German mathematician Carl Friedrich Gauss, the integral is

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

**Proof:**

$$\begin{aligned} \iint_{\mathbb{R}^2} e^{-(x^2+y^2)} dx dy &= \int_0^{2\pi} \int_0^{\infty} e^{-r^2} r dr d\theta \\ &= 2\pi \int_0^{\infty} r e^{-r^2} dr \\ &= 2\pi \int_{-\infty}^0 \frac{1}{2} e^s ds && s = -r^2 \\ &= \pi \int_{-\infty}^0 e^s ds \\ &= \pi (e^0 - e^{-\infty}) \\ &= \pi, \end{aligned}$$

In addition,

$$\iint_{\mathbb{R}^2} e^{-(x^2+y^2)} dx dy = \left( \int_{\mathbb{R}} e^{-x^2} dx \right) \left( \int_{\mathbb{R}} e^{-y^2} dy \right) = \left( \int_{\mathbb{R}} e^{-x^2} dx \right)^2$$

### 1.4.1 Practice

Use the result of Gaussian Integral to calculate

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

$$\int_{-\infty}^{\infty} e^{-x^2/(2\sigma^2)} dx$$

$$\int_{-\infty}^{\infty} e^{-(x-\mu)^2/(2\sigma^2)} dx$$

## 1.5 Continuous Distribution Examples

### 1. Uniform Distribution

$$f_X(x) = 1, \quad 0 \leq x \leq 1$$

### 2. Generalized Uniform Distribution

$$f_X(x) = \frac{1}{b-a}, \quad a \leq x \leq b$$

### 3. Exponential Distribution

$$f_X(x) = \lambda e^{-\lambda x}, \quad x > 0$$

or

$$f_X(x) = \frac{1}{\beta} e^{-x/\beta}, \quad x > 0$$

### 4. Standard Normal Distribution

$$f_X(x) = \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}, \quad x \in \mathbb{R}$$

### 5. Normal Distribution

$$f_X(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-\mu)^2}{2\sigma^2}}, \quad x \in \mathbb{R}$$

#### 1.5.1 Practice

Derive the cdf of Uniform distribution, generalized uniform distribution, and exponential distribution.

**Remark 10** *The cdf of normal distribution does not have analytical form, which means we cannot write it down explicitly.*