STSCI 5080 Probability Models and Inference

Lecture 4: Discrete Random Variables

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Notation

For a random variable X (a function on Ω taking values in the real numbers) and a subset S of the real numbers (e.g. $S = \{0, 1, 2\}$ or S = [0, 1]), we use the shorthand notation

$${X \in S} = {\omega \mid X(\omega) \in S}.$$

In addition, we also write

$$P(X \in S) = P(\{X \in S\}).$$

Discrete random variable

Definition (Discrete random variable)

A random variable X is discrete if X takes values in a finite or countably infinite set.

Probability mass function

Definition

For a discrete random variable X, the probability mass function (pmf) p(x) is a function defined by

$$p(x) = P(X = x)$$

for any real number x.

Properties of pmf

• If X that takes values in $\{x_1, x_2, \dots\}$ (called the support of X), then

$$p(x) = 0$$
 for any $x \notin \{x_1, x_2, \dots\}$.

• The pmf p(x) satisfies that $p(x) \ge 0$ for any x and $\sum_{x} p(x) = 1$ (why?). In addition,

$$P(X \in S) = \sum_{x \in S} p(x).$$

Example 4.1

Example

Suppose that we toss a coin three times and all the outcomes occur equally likely. Consider

$$X(\omega) = \text{total number of heads in } \omega.$$

Then the support is $\{0, 1, 2, 3\}$ and the pmf is

$$p(x) = P(X = x) = \begin{cases} \frac{1}{8} & x = 0\\ \frac{3}{8} & x = 1\\ \frac{3}{8} & x = 2\\ \frac{1}{8} & x = 3\\ 0 & x \notin \{0, 1, 2, 3\} \end{cases}.$$

Cumulative distribution function

Definition

For a discrete random variable X with pmf p(x), the cumulative distribution function (cdf) F(x) is defined as

$$F(x) = P(X \le x) = \sum_{y \le x} p(y)$$

for any real number x.

What is the cdf of the random variable in Example 4.1?

Example 4.2

Example (CDF as a measure of risk)

Suppose you consider to invest on Portfolio α or β .

- X: gain of Portfolio α with cdf F_X ;
- Y: gain of Portfolio β with cdf F_Y .

Now, we know that

$$F_X(-100) = 0.05$$
 and $F_Y(-1000) = 0.05$.

Which portfolio is more risky?

Bernoulli random variable

• If a random variable X takes values in 0 or 1 with

$$P(X = 1) = p$$
 and $P(X = 0) = 1 - p$,

then X is called a Bernoulli random variable with success probability p.

• What is the pmf?

$$p(x) = \begin{cases} p & x = 1\\ 1 - p & x = 0\\ 0 & x \notin \{0, 1\} \end{cases} = \begin{cases} p^x (1 - p)^{1 - x} & x = 0, 1\\ 0 & x \notin \{0, 1\} \end{cases}.$$

What is the cdf?

Independent Bernoulli trials

Definition

Let X_1, \ldots, X_n be Bernoulli random variables with the same success probability p that are independent in the sense that n events

$${X_1 = x_1}, \dots, {X_n = x_n}$$

are independent for any $x_1, \ldots, x_n \in \{0, 1\}$. In this case, we call X_1, \ldots, X_n independent Bernoulli trials with success probability p.

Binomial random variable

Definition

Let X_1, \ldots, X_n be independent Bernoulli trials with success probability p. Then the random variable

$$Y = X_1 + \cdots + X_n$$

is called a binomial random variable with parameters n and p.

"Y follows the binomial distribution with parameters n and p"

$$Y \sim Bin(n,p)$$
.

Binomial coefficients

For a positive integer n and $k = 0, 1, \dots, n$,

$$\binom{n}{k}$$
 = number of k -element subsets of $\{1, \dots, n\}$.

For example,

$$\binom{n}{0} = 1, \ \binom{n}{1} = n, \ \binom{n}{n} = 1.$$

In general,

$$\binom{n}{k} = \frac{n(n-1)\cdots(n-k+1)}{k(k-1)\cdots1} = \frac{n!}{(n-k)!k!},$$

where

$$n! = n(n-1)\cdots 1.$$

What are $\binom{n}{2}$ and $\binom{n}{3}$?

PMF of Bin(n, p)

Theorem

The pmf of $Y \sim Bin(n, p)$ is

$$p(k) = P(Y = k) = \binom{n}{k} p^k (1 - p)^{n-k}, \ k = 0, 1, \dots, n.$$

Poisson random variable

Definition

Let $\lambda>0$. X is a Possion random variable with parameter λ if its takes values in $\{0,1,2,\dots\}$ and its pmf is

$$p(k) = \frac{\lambda^k}{k!} e^{-\lambda}, \ k = 0, 1, 2, \dots$$

"X follows the Poisson distribution with parameter λ "

$$X \sim Po(\lambda)$$
.