

Probability Theory and Simulation Methods

March 2nd, 2018

Lecture 11: Expectation

Countdown to midterm (March 21st): 26 days

Week 1 Chapter 1: Axioms of probability

Week 2 Chapter 3: Conditional probability and independence

Week 4 **Chapters 4, 5, 6, 7: Random variables**

Week 9 Chapters 8, 9: Bivariate and multivariate distributions

Week 10 Chapter 10: Expectations and variances

Week 11 Chapter 11: Limit theorems

Week 12 Chapters 12, 13: Selected topics

Chapter 4: Discrete random variables

4.1 Random variables

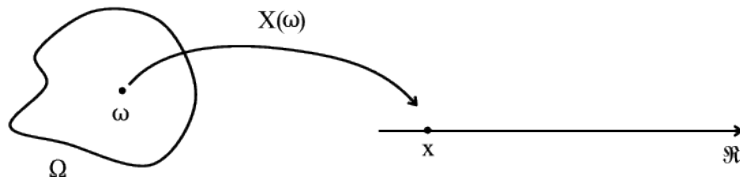
4.3 Discrete random variables

4.4 Expectations of discrete random variables

4.5 Variances and moments of discrete random variables

4.2 Distribution functions

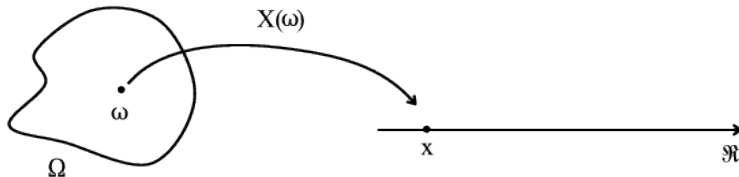
Random variable



Definition

Let S be the sample space of an experiment. A real-valued function $X : S \rightarrow \mathbb{R}$ is called a random variable of the experiment.

Notation



When we write $\{X \in (0, 1)\}$, we are (implicitly) referring to the event

$$\{s \in S : X(s) \in (0, 1)\}.$$

Definition

A random variables X is discrete if the set of all possible values of X

- is finite
- is countably infinite

Note: A set A is countably infinite if its elements can be put in one-to-one correspondence with the set of natural numbers, i.e, we can index the element of A as a sequence

$$A = \{x_1, x_2, \dots, x_n, \dots\}$$

A random variable X is described by its probability mass function

Definition *The probability mass function p of a random variable X whose set of possible values is $\{x_1, x_2, x_3, \dots\}$ is a function from \mathbf{R} to \mathbf{R} that satisfies the following properties.*

- (a) $p(x) = 0$ if $x \notin \{x_1, x_2, x_3, \dots\}$.
- (b) $p(x_i) = P(X = x_i)$ and hence $p(x_i) \geq 0$ ($i = 1, 2, 3, \dots$).
- (c) $\sum_{i=1}^{\infty} p(x_i) = 1$.

Read the pmf table

x	1	2	3	4	5	6	7
$p(x)$.01	.03	.13	.25	.39	.17	.02

Definition *The **expected value** of a discrete random variable X with the set of possible values A and probability mass function $p(x)$ is defined by*

$$E(X) = \sum_{x \in A} xp(x).$$

We say that $E(X)$ exists if this sum converges absolutely.

The expected value of a random variable X is also called the **mean**, or the **mathematical expectation**, or simply the **expectation** of X . It is also occasionally denoted by $E[X]$, $E(X)$, EX , μ_X , or μ .

Problem

A random variable X has the following pmf table

x	1	2	3	4	5	6	7
$p(x)$.01	.03	.13	.25	.39	.17	.02

What is the expected value of X ?

Problem

We write the numbers $a_1 = 2, a_2 = 4, \dots, a_n = 2n$ on n identical balls and mix them in a box. What is the expected value of the number on a ball selected at random?

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X	2	4	6	...	2n
probability	1/n	1/n	1/n	...	1/n

Problem

Suppose that we flip a coin until a head first appears, and if the number of tosses equals n , then we are paid 2^n dollars.

- *Write down the probability mass function of the payment*
- *Compute its expected value*

St. Petersburg Paradox

Problem

Suppose that we flip a coin until a head first appears, and if the number of tosses equals n , then we are paid 2^n dollars.

- *Write down the probability mass function of the payment*
- *Compute its expected value*

X	2	2^2	2^3	...	2^n	...
probability	$\frac{1}{2}$	$\frac{1}{2^2}$	$\frac{1}{2^3}$...	$\frac{1}{2^n}$...

Function of a random variable

- A random variable X has the following pmf table

x	1	2	3	4	5	6	7
$p(x)$.01	.03	.13	.25	.39	.17	.02

- Define $Y = X^2$, then Y is a random variable of the same experiment as X
- Write down the pmf table of Y
- What is the expected value of Y ?

Since $Y = X^2$, the two expressions

$$\{s \in S : X(s) = 2\}$$

and

$$\{s \in S : Y(s) = 4\}$$

describes the same events.

Thus

$$P(X = 2) = P(X^2 = 4) = P(Y = 4)$$

Function of a random variable

- A random variable X has the following pmf table

x	1	2	3	4	5	6	7
$p(x)$.01	.03	.13	.25	.39	.17	.02

- Define $Y = g(X)$, then Y is a random variable of the same experiment as X
- What is the expected value of Y ?

Theorem 4.2 *Let X be a discrete random variable with set of possible values A and probability mass function $p(x)$, and let g be a real-valued function. Then $g(X)$ is a random variable with*

$$E[g(X)] = \sum_{x \in A} g(x)p(x).$$

Corollary *Let X be a discrete random variable; g_1, g_2, \dots, g_n be real-valued functions, and let $\alpha_1, \alpha_2, \dots, \alpha_n$ be real numbers. Then*

$$\begin{aligned} E[\alpha_1 g_1(X) + \alpha_2 g_2(X) + \dots + \alpha_n g_n(X)] \\ = \alpha_1 E[g_1(X)] + \alpha_2 E[g_2(X)] + \dots + \alpha_n E[g_n(X)]. \end{aligned}$$

Example 4.23 The probability mass function of a discrete random variable X is given by

$$p(x) = \begin{cases} x/15 & x = 1, 2, 3, 4, 5 \\ 0 & \text{otherwise.} \end{cases}$$

What is the expected value of $X(6 - X)$?

Problem

Show that if X is a constant random variable, that is, if $P(X = c) = 1$ for a constant c , then $E(X) = c$.

The probability mass function:

x	c
$p(x)$	1

What we need to know so far

- What is a discrete random variable?
- What is a pmf?
- How to compute expected value of a random variable?
- How to compute the expectation of $g(X)$?

Generalization

- Inputs: X_1, X_2, \dots, X_n
- Model: Function h
- Output $T = h(X_1, X_2, \dots, X_n)$

Question: If we know the pmf of the inputs, can we estimate the pmf of the output?

Worldwide nuclear testing, 1945 - 2013

