

2.1 Discrete Random Variables

Terms:

- Random Variable
- Support
- Discrete Random Variable
- Probability Mass Function
- Cumulative Distribution Function
- Expected Value (Expectation)

Random Variables

Definition 2.1-1

Given a random experiment with an outcome space S , a function X that assigns one and only one real number $X(s) = x$ to each element s in S is called a **random variable**. The **space** of X is the set of real numbers $\{x : X(s) = x, s \in S\}$, where $s \in S$ means that the element s belongs to the set S .

A random variable associates a numerical value to each outcome of a random experiment.

Used to describe the elements of S numerically.

Random variables – Coin Example

A fair coin is flipped. The set of possible outcomes is Heads and Tails. $S = \{H, T\}$

Suppose we are interested in whether the outcome was Tails or not.

Let X be a real valued function defined on S such that $X(H) = 0$ and $X(T) = 1$.

- Domain of X : Outcome space, S
- Range of X : real numbers $\{x: x = 0, 1\}$

X is a random variable, and the space of X is the set of numbers $\{0, 1\}$.

Types of Random Variables

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- A random variable is **discrete** if it has a countable number of values. (can be countably infinite)

连续

- A random variable is **continuous** if it is uncountable. (represents something on a continuous scale and can take any values in an interval)

Example: Defining a Random Variable

Consider the sample space for rolling two die: 36种可能

$S =$

Support (?)

Say we are interested only in the sum of the number of spots.

We can define a random variable, X , by assigning a real number to each outcome in S .

$$X = \{2, 3, 4, \dots, 12\}$$

$$\begin{aligned} X(\{\text{red 1, green 1}\}) &= 2, & X(\{\text{red 1, green 2}\}) &= 3, \dots \\ X(\{\text{red 6, green 6}\}) &= 12 \end{aligned}$$

What is the space of X ?

Example: Defining a Random Variable (continued)

The space of X is $\mathcal{X} = \{2,3,4,5,6,7,8,9,10,11,12\}$

For convenience, we don't need the original sample space anymore. We can use the space of X instead.

Discrete random variable

If a sample space, S , contains a **countable** number of points, we call S a **discrete sample space**.

Any random variable X (with space, \mathcal{X}) arising from a discrete sample space, S , is called a **discrete random variable**.



Discrete random variable (examples)

- ✓ • Number of “heads” in 3 flips of a coin
 $S = \{HHH, HHT, \dots, TTH, TTT\} \rightarrow \mathcal{X} = \{0, 1, 2, 3\}$
- ✓ • Number of diaper changes I need to make per night:
 $S = \{0, 1, 2, \dots\}$
- ✗ • Hulk start with 15mL saliva for Covid testing but accidentally spill random amount. Want to know **exactly** how much saliva have left in tube?!?! $S = [0, 15)$
- ✓ • “Hulk start with 15mL saliva... **rounded to nearest mL?**”
 $S = \{0, 1, 2, \dots, 14, 15\}$

Probability Mass Function

For a random variable X , the probability that the random variable takes a value, x , is denoted $P[X = x]$.

For a discrete random variable, this is also denoted by $f(x)$.

$f(x)$ is called the **probability mass function**.

Properties of a pmf

Definition 2.1-2

The pmf $f(x)$ of a discrete random variable X is a function that satisfies the following properties:

(a) $f(x) > 0, \quad x \in S;$

(b) $\sum_{x \in S} f(x) = 1;$

(c) $P(X \in A) = \sum_{x \in A} f(x), \quad \text{where } A \subset S.$

If a pmf satisfies all 3 properties, then it is a valid discrete probability distribution.

Probability mass function

x	$f(x)$
x_1	$f(x_1)$
x_2	$f(x_2)$
x_3	$f(x_3)$
\vdots	\vdots
\vdots	\vdots
\vdots	\vdots
x_n	$f(x_n)$

If the pmf is simple, it may be written as a table or list.

- Example: Flip a fair coin twice. let X be defined as the number of heads observed. What is the pmf of X ?

Most of the time, it is written as a formula.

Examples:

$$f(x) = \frac{(|x| + 1)^2}{9}, \quad x = -1, 0, 1.$$

$$f(x) = \frac{5 - x}{10}, \quad x = 1, 2, 3, 4.$$

$$f(x) = \begin{cases} 0.9, & x = 0, \\ \frac{c}{x}, & x = 1, 2, 3, 4, 5, 6, \end{cases}$$

Cumulative Distribution Function

The function defined by

$$F(x) = P(X \leq x), -\infty < x < \infty$$

is called the **cumulative distribution function**.

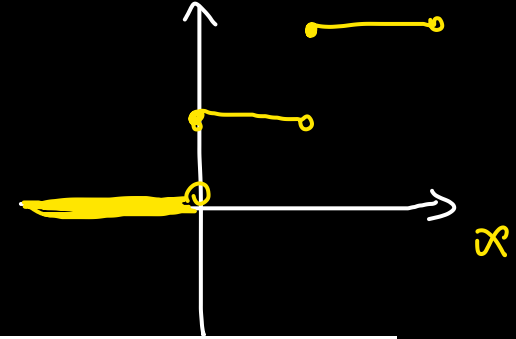
You may also see the cdf referred to as the **distribution function** of a random variable, X .

CDF examples

1. Flip a fair coin twice. let X be defined as the number of heads observed. What is the cdf of X ?

$$F(x) = P[X \leq x]$$

$$F(3) = 1$$



2. Given the following pmf, what is $F(x)$?

$$f(x) = \frac{5-x}{10}, \quad x = 1, 2, 3, 4.$$

$$F(1) = f(1)$$

$$F(2) = f(1) + f(2)$$

$$F(3) = f(1) + f(2) + f(3)$$

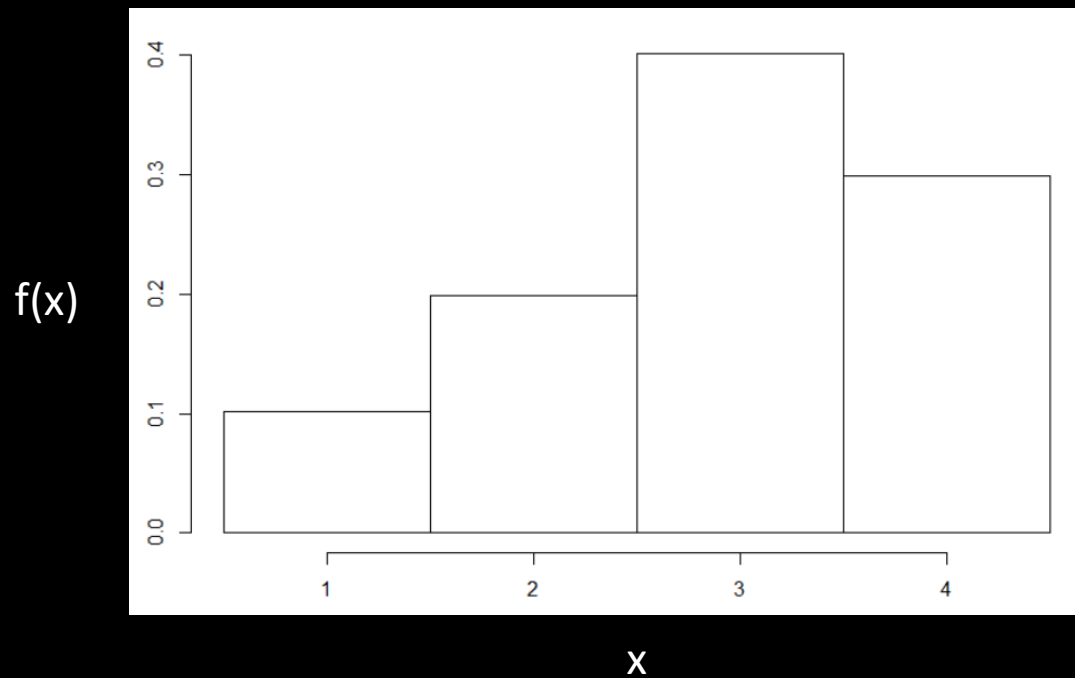
2.2 Expected Value

Expected Value of X (or ‘expectation’ of X)

Let X be a discrete random variable with probability mass function $f(x)$. The **expected value** of X can be denoted $E[X]$, μ , or μ_x , and is given by:

$$E[X] = \mu = \mu_x = \sum_{\{all\ x\}} x \cdot f(x)$$

Expected Value of X



x	$f(x)$	$F(x)$
1	0.1	
2	0.2	
3	0.4	
4	0.3	

$$E[X] =$$

Expectation of a function of X

Definition 2.2-1

If $f(x)$ is the pmf of the random variable X of the discrete type with space S , and if the summation

$$\sum_{x \in S} u(x)f(x), \quad \text{which is sometimes written} \quad \sum_S u(x)f(x),$$

exists, then the sum is called the **mathematical expectation** or the **expected value** of $u(X)$, and it is denoted by $E[u(X)]$. That is,

$$E[u(X)] = \sum_{x \in S} u(x)f(x).$$

Some properties of expectation

Theorem 2.2-1

When it exists, the mathematical expectation E satisfies the following properties:

- (a) If c is a constant, then $E(c) = c$.
- (b) If c is a constant and u is a function, then

$$E[c u(X)] = cE[u(X)].$$

- (c) If c_1 and c_2 are constants and u_1 and u_2 are functions, then

$$E[c_1 u_1(X) + c_2 u_2(X)] = c_1 E[u_1(X)] + c_2 E[u_2(X)].$$

Examples:

(a) $E[5] = 5,$ $E[2] = 2$

(b) $E[2X] = 2E[X],$ $E[7X^2] = 7E[X^2]$

(c) $E[2X + 5Y] = 2E[X] + 5E[Y],$ $E[3X + 6X^2 - X^3] = 3E[X] + 6E[X^2] - E[X^3]$

Examples



1. Thanos has some flavored JUULs at home with the following proportion of flavors:

1/2 berry, 1/4 orange, 1/4 lemon.

If he picks up the following JUUL flavor, it will vaporize the following number of people:

Berry – 50, Orange – 30, Lemon – 40. $X = \{30, 40, 50\}$

Let the random variable, X , represent the number of vaporized people if Thanos randomly selects a JUUL.

a) What is the pmf of X ?

$$E = \sum x \cdot f(x)$$



x	$f(x)$
30	$\frac{1}{4}$
40	$\frac{1}{4}$
50	$\frac{1}{2}$

b) Does the expected value need to be an element of the sample space?

b) Let X be a random variable that represents the number of people vaporized if Thanos eats a random gummy. Find the expected value of X .

$$E[X] = (30)(\frac{1}{4}) + (40)(\frac{1}{4}) + (50)(\frac{1}{2})$$

c) If Thanos selects a random JUUL, find the expected value of the number of eyeballs vaporized. $E[2X] = 2E[X]$

e) Find $E[X^2]$.

x	x^2	$f(x)$
30	900	$\frac{1}{4}$
40	1600	$\frac{1}{4}$
50	2500	$\frac{1}{2}$

$$E[X^2] = 900 \times \frac{1}{4} + 1600 \times \frac{1}{4} + 2500 \times \frac{1}{2}$$

Example 2

Suppose a discrete random variable X has the following probability distribution:

$$P(X=0) = 2 - \sqrt{e}, \quad P(X=k) = \frac{1}{2^k \cdot k!}, \quad k=1, 2, 3, \dots$$

a) Find $E[X]$. $E[X] = 1 \cdot \frac{1}{2^1 \cdot 1!} + 2 \cdot \frac{1}{2^2 \cdot 2!} + 3 \cdot \frac{1}{2^3 \cdot 3!} + \dots$

知识点 $e^x = \sum_{n=0}^{\infty} \frac{x^n}{n!}$ ☆

$$\begin{aligned} E(X) &= \sum_{\text{all } x} x \cdot f(x) = 0 \cdot (2 - e^{1/2}) + \sum_{k=1}^{\infty} k \cdot \frac{1}{2^k \cdot k!} = \sum_{k=1}^{\infty} \frac{1}{2^k \cdot (k-1)!} \\ &= \frac{1}{2} \cdot \sum_{k=1}^{\infty} \frac{1}{2^{k-1} \cdot (k-1)!} = \frac{1}{2} \cdot \sum_{n=0}^{\infty} \frac{1}{2^n \cdot n!} = \frac{e^{1/2}}{2}. \end{aligned}$$

Example

Suppose a discrete random variable X has the following probability distribution:

$$P(X=0) = 2 - \sqrt{e}, \quad P(X=k) = \frac{1}{2^k \cdot k!}, \quad k=1, 2, 3, \dots$$

b) Find $E[X^2]$.

$$\begin{aligned}
 E[X^2] &= \sum_{x=1}^{\infty} x^2 \cdot \frac{1}{2^x \cdot x!} = \sum_{x=1}^{\infty} x^2 \frac{\left(\frac{1}{2}\right)^x}{x!} = \sum_{x=1}^{\infty} \frac{\left(\frac{1}{2}\right)^{x-1}}{(x-1)!} \cdot \frac{1}{2} \cdot (x-1+1) = \sum_{x=1}^{\infty} \left[\frac{1}{2} \cdot \frac{1}{2} \cdot \frac{\left(\frac{1}{2}\right)^{x-2}}{(x-2)!} + \frac{1}{2} \cdot \frac{\left(\frac{1}{2}\right)^{x-1}}{(x-1)!} \right] \\
 &= \frac{1}{4} \cdot e^{\frac{1}{2}} + \frac{1}{2} \cdot e^{\frac{1}{2}} = \frac{3}{4} \sqrt{e}
 \end{aligned}$$

$$\begin{aligned}
 E(X(X-1)) &= \sum_{k=2}^{\infty} k \cdot (k-1) \cdot \frac{1}{2^k \cdot k!} = \sum_{k=2}^{\infty} \frac{1}{2^k \cdot (k-2)!} \\
 &= \frac{1}{4} \cdot \sum_{k=2}^{\infty} \frac{1}{2^{k-2} \cdot (k-2)!} = \frac{1}{4} \cdot \sum_{n=0}^{\infty} \frac{1}{2^n \cdot n!} = \frac{e^{1/2}}{4}.
 \end{aligned}$$