Performance Evaluation

- One in-class exams: 10%
 - -4/13
- One midterm exams: 20%
 - $-\frac{5}{4}$
- One final exam: 40%
 - -6/8
- Homework assignments: 25%
- In-class performance: 5%

Chapter 2 Discrete Distributions

- Random Variables of the Discrete Type
 - Uniform Distribution
 - Hypergeometric Distribution
- Mathematical Expectation
- Moment Generating Function
- Bernoulli Trials and the Binomial Distribution
- Geometric and Negative Binomial Distribution
- The Poisson Distribution

2.1 Random Variables of the Discrete Type

- Def 2.1-1: Random Variable X (abbreviated by r.v.)
 - a **function** that maps the possible outcomes of an experiment to real numbers. (把實驗結果轉成實數)
 - i.e., assign to each element s in S one and only one real number
 - Notice that: a function assigns one and only one number in the range (output) to each number in the domain (input)
 - $-X: S \rightarrow R$, where S is the set of all outcomes of an experiment, and R is the set of real numbers.
- The space of X is the set of real numbers S_X , where

$$S_X = \{x: X(s) = x, s \in S\}$$

From now on, S_X is replaced by S in this class

An Example of Random Variable

- If we toss a coin one time, then there are two possible outcomes
 - namely "head up" and "tail up".
- We can define a random variable *X* that maps "head up" to 1 and "tail up" to 0.
- We also can define a random variable *Y* that maps "head up" to 0 and "tail up" to 1.
- The spaces of both random variables X and Y are $S = \{0,1\}$.

Further Illustration of Random Variables

- A random variable corresponds to a quantitative interpretation of the outcomes of an experiment.
- For example, a company offers its employees a drawing in its yearend party.
- A computer will randomly select an employee for the <u>first prize of \$100,000</u> based on the employees' ID number, which ranges from 1 to 100.

- In addition, the computer will randomly select two more employees for the second and third prizes of \$50,000 and \$10,000, respectively.
- Assume that each employee can receive only one award and the drawing starts with the third prize and ends with the first prize.
- Number of possible outcomes
 - totally $100 \times 99 \times 98 = 970200$

• To Edward, whose employee ID number is 10, the random variable of his interest is as follows:

$$X(<10, *, *>) = 10,000$$

 $X(<*, 10, *>) = 50,000$
 $X(<*, *, 10>) = 100,000$
 $X($ all other outcomes $) = 0$

• To Grace, whose employee ID number is 30, the random variable of her interest is as follows:

$$Y(<30, *, *>) = 10,000$$

 $Y(<*, 30, *>) = 50,000$ space?
 $Y(<*, *, 30>) = 100,000$
 $Y(\text{all other outcomes}) = 0$

- X and Y map some outcomes to different real numbers.
- However, the spaces of *X* and *Y*
 - are identical
 - both are {0, 10000, 50000, 100000}.
- The probability functions of X and Y are also equal.

$$Prob(X=10,000) = Prob(Y=10,000) = 0.01$$

 $Prob(X=50,000) = Prob(Y=50,000) = 0.01$
 $Prob(X=100,000) = Prob(Y=100,000) = 0.01$
 $Prob(X=0) = Prob(Y=0) = 0.97$

- The expected values of *X* and *Y*
 - are equal to

$$E[X] = E[Y]$$

- = 10,000 * 0.01 + 50,000 * 0.01
 - + 100,000 * 0.01
- = 1600.

Discrete Random Variables

- Given a random variable X,
 - let S denote the space of X.
- If *S* is a finite or countable infinite set, then *X* is said to be a discrete random variable.

Countable Infinite

- A set is said to be countable infinite if
 - it contains infinite number of elements
 - there exists a <u>one-to-one mapping</u>
 between each element of the set and the <u>positive integers</u>.

Examples of Countable / Uncountable Infinite

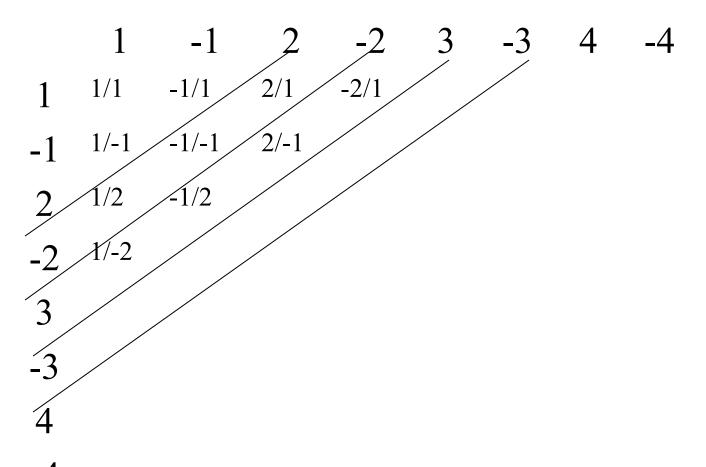
- The set of integer numbers is?
 - countable.
- The set of fractional numbers is?
 - countable.
- The set of real numbers is?
 - uncountable.

Mapping of Integer Numbers to Positive Integer Numbers

- 0 1 -1 2 -2 3 -3
- 0 1 2 3 4 5 6

Mapping of Fractional Numbers to Positive Integer Numbers?

Mapping of Fractional Numbers to Positive Integer Numbers



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Probability Mass Function

• The probability mass function (p.m.f.) of a discrete random variable *X* is defined to be

$$P_X(k) \equiv Prob(X = k) = \sum_{q \in Q_k} Prob(q),$$

where Q_k contains all outcomes that are mapped to k by random variable X.

課本的定義?

In the previous example of drawing,

$$P_{X}(10,000) = Prob(X = 10,000)$$

$$= \sum_{\substack{<10,i,j>\\i\neq 10,j\neq 10\\i\neq j}} Prob(<10,i,j>)$$

$$= \sum_{\substack{<10,i,j>\\i\neq 10,j\neq 10\\i\neq j}} \frac{1}{100\times 99\times 98} = 0.01.$$

$$P_X(k) \equiv Prob(X = k) = \sum_{q \in Q_k} Prob(q),$$

where Q_k contains all outcomes that are mapped to k by random variable X.

- In fact, the p.m.f. of a random variable is defined on a set of events of the experiment conducted.
 - $f(x) \equiv P(X=x)$ as shown in the Hogg's textbook
- In the previous drawing example, the set of outcomes that are mapped to 10,000 by X is an event.

• To Edward, whose employee ID number is 10, the random variable of his interest is as follows:

$$X(<10, *, *>) = 10,000$$

 $X(<*, 10, *>) = 50,000$
 $X(<*, *, 10>) = 100,000$
 $X($ all other outcomes $) = 0$

• To Grace, whose employee ID number is 30, the random variable of her interest is as follows:

$$Y(<30, *, *>) = 10,000$$

 $Y(<*, 30, *>) = 50,000$
 $Y(<*, *, 30>) = 100,000$
 $Y($ all other outcomes $) = 0$

• Furthermore, in the previous drawing example, random variables X and Y map some outcomes to different real numbers. However, X and Y have the same distribution, i.e. the p.m.f. of X and the p.m.f. of Y are equal. More precisely,

$$P_X(k) = P_Y(k)$$

for every $k \in \{0,10000,50000,100000\}$.

- The distribution of the random variables X
 - The distribution of probability associated with space S of the r.v. X

Properties of Probability Mass Function

$$P_X(k) \equiv Prob(X = k) = \sum_{q \in Q_k} Prob(q),$$

• The p.m.f. of a random variable X satisfies the following three properties:

(1)
$$P_X(x) > 0$$
 , $x \in S$: the space of X .

$$(2) \sum_{x_i \in S} P_X(x_i) = 1.$$

(3)
$$\operatorname{Prob}(A) = \sum_{x_j \in A} P_X(x_j)$$
, where $A \subseteq S$.

課本將以上的Properties視為定義,如下一頁:

For a random variable X of the discrete type, the probability P(X = x) is frequently denoted by f(x), and this function f(x) is called the **probability mass** function. Note that some authors refer to f(x) as the probability function, the frequency function, or the probability density function. In the discrete case, we shall use "probability mass function," and it is hereafter abbreviated pmf.

Let f(x) be the pmf of the random variable X of the discrete type, and let S be the space of X. Since f(x) = P(X = x) for $x \in S$, f(x) must be nonnegative for $x \in S$, and we want all these probabilities to add to 1 because each P(X = x) represents the fraction of times x can be expected to occur. Moreover, to determine the probability associated with the event $A \in S$, we would sum the probabilities of the x values in A. This leads us to the following definition.

Definition 2.1-2

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

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

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

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

課本第29-30頁

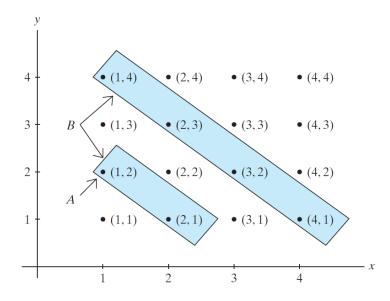


Figure 1.3-2 Dice example

Example 1.3-4

A pair of fair four-sided dice is rolled and the sum is determined. Let A be the event that a sum of 3 is rolled, and let B be the event that a sum of 3 or a sum of 5 is rolled. In a sequence of rolls, the probability that a sum of 3 is rolled before a sum of 5 is rolled can be thought of as the conditional probability of a sum of 3 given that a sum of 3 or 5 has occurred; that is, the conditional probability of A given B is

$$P(A \mid B) = \frac{P(A \cap B)}{P(B)} = \frac{P(A)}{P(B)} = \frac{2/16}{6/16} = \frac{2}{6}.$$

Note that for this example, the only outcomes of interest are those having a sum of 3 or a sum of 5, and of these six equally likely outcomes, two have a sum of 3. (See Figure 1.3-2 and Exercise 1.3-13.)