STA2001 Probability and Statistics (I)

Lecture 10

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Chapter 4. Bivariate Distribution

Section 4.1 Bivariate Distribution of Discrete Type

Motivation

Very often, we are interested to study two random experiments jointly, each of whose outcome is a scalar, or a random experiment whose outcome is a pair of two scalars.

- 1. observe college students to obtain information such as height x and weight y.
- 2. observe high school students to obtain information such as rank x and score of college entrance examination y.

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Very often, we are interested to study two random experiments jointly, each of whose outcome is a scalar, or a random experiment whose outcome is a pair of two scalars.

- 1. observe college students to obtain information such as height *x* and weight *y*.
- 2. observe high school students to obtain information such as rank x and score of college entrance examination y.
- → univariate RV
- two random experiments jointly each of whose outcome is a scalar, or a random experiment whose outcome is a pair of two scalars, \rightarrow bivariate RV

Bivariate RV

Definition

Let (X, Y) be a pair of RVs with their range denoted by $\overline{S} \subseteq R^2$. Then (X, Y) or X and Y is said to be a bivariate RV. If \overline{S} is finite or countably infinite, then (X, Y) is said to be a discrete bivariate RV.

Moreover, let $\overline{S_X} \subseteq R$ and $\overline{S_Y} \subseteq R$ denote the range of X and Y, respectively.

$$\overline{S} = \{\text{all possible values of } (X, Y)\}$$

$$\overline{S_X} = \{ \text{all possible values of } X \} = \{ x | (x, y) \in \overline{S} \}$$

$$\overline{S_Y} = \{ \text{all possible values of } Y \} = \{ y | (x, y) \in \overline{S} \}$$

Then, it holds that

$$\overline{S} \subseteq \overline{S_X} \times \overline{S_Y} = \{(x,y)|x \in \overline{S_X}, y \in \overline{S_Y}\}$$

Example 1, Page 134

Roll a pair of 4-sided fair dice. Then the original sample space

$$S = \begin{cases} (1,1), & (1,2), & (1,3), & (1,4), \\ (2,1), & (2,2), & (2,3), & (2,4), \\ (3,1), & (3,2), & (3,3), & (3,4), \\ (4,1), & (4,2), & (4,3), & (4,4) \end{cases},$$

where the two numbers in each pair represent the outcome of the first die and the second die, respectively.

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Sample space $\overline{S} \subseteq \overline{S_X} \times \overline{S_Y}$: $\overline{S} \overline{\Lambda} - \dot{\overline{\mathcal{L}}} = \overline{S_X} \times \overline{S_Y}$ 但一定 $\underline{C} \overline{S_X} \times \overline{S_Y}$

$$\overline{S}_X = \overline{S}_Y = \{1, 2, 3, 4\}, \overline{S} = \left\{ egin{array}{ll} (1, 1), & (1, 2), & (1, 3), & (1, 4), \\ & (2, 2), & (2, 3), & (2, 4), \\ & & (3, 3), & (3, 4), \\ & & & (4, 4) \end{array}
ight\},$$

where the two numbers in each pair represent the possible values of X and Y, respectively.

Joint pmf

Joint PMF

Definition

The function $f(x,y): \overline{S} \to (0,1]$ is called the joint probability mass function (joint pmf) of X and Y or (X,Y), if

1.
$$f(x,y) > 0$$
 for $(x,y) \in \overline{S}$, $f(x,y) \neq \overline{S}$

- 2. $\sum_{(x,y)\in\overline{S}} f(x,y) = 1, \qquad \sum_{(x,y)\in\overline{S}} f(x,y) = 1$
- 3. For $A \subseteq \overline{S}$,

$$P[(X,Y) \in A] \stackrel{\Delta}{=} P(\{(X,Y) \in A\}) = \sum_{(x,y) \in A} f(x,y)$$

which defines the probability function for a set A. In particular, taking $A = \{(x, y)\}$ yields the probability of X = x and Y = y, i.e.,

$$P(X = x, Y = y) = f(x, y)$$

Question:

$$P(X = 2, Y = 3) = ?, P(X = 2, Y = 2) = ?$$

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$$P(X = 2, Y = 3) = \frac{1}{16} + \frac{1}{16} = \frac{2}{16}$$

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$$\overline{S} = \left\{ egin{array}{cccc} (1,1) & (1,2) & (1,3) & (1,4) \ & (2,2) & (2,3) & (2,4) \ & & (3,3) & (3,4) \ & & & (4,4) \end{array}
ight\}$$

$$f(x,y) = \begin{cases} \frac{2}{16}, & 1 \le x < y \le 4\\ \frac{1}{16}, & 1 \le x = y \le 4. \end{cases}$$

A Remark on Computation of The Probability

For
$$A \subseteq \overline{S}$$
, $P[(X, Y) \in A] \stackrel{\triangle}{=} P(\{(X, Y) \in A\}) = \sum_{(x,y) \in A} f(x,y)$,

where the double summation can be split into 2 single summation.

Let

$$A_X = \{x | (x, y) \in A\}, A_Y(x) = \{y | (x, y) \in A\}, \text{ for } x \in A_X$$

Then

$$P((X,Y) \in A) = \sum_{x \in A_X} \sum_{y \in A_Y(x)} f(x,y)$$

Let

$$A_Y = \{y | (x, y) \in A\}, A_X(y) = \{x | (x, y) \in A\}, \text{ for } y \in A_Y$$

Then

$$P((X,Y) \in A) = \sum_{y \in A_Y} \sum_{x \in A_X(y)} f(x,y)$$

Marginal pmf

Definition

Let (X, Y) be a bivariate RV or X and Y be two RVs and have the joint pmf $f(x, y) : \overline{S} \to (0, 1]$. Sometimes, we are interested in the pmf of X or Y alone, which is called the marginal pmf of X or Y and described by

For
$$x \in \overline{S_X}$$
, Marginal PMF

$$f_X(x) = P_X(X = x) \stackrel{\triangle}{=} P\left(\left\{X = x, Y \in \overline{S_Y}(x)\right\}\right)$$

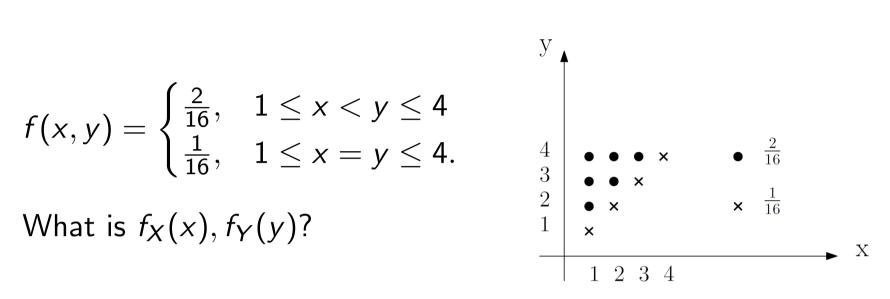
$$= \sum_{y \in \overline{S_Y}(x)} f(x, y)$$

$$y \in \overline{S_Y}(x)$$

where

$$\overline{S_Y}(x) = \{y | (x, y) \in \overline{S}\}$$
 for the given $x \in \overline{S_X}$.

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First,
$$\overline{S_X} = \overline{S_Y} = \{1, 2, 3, 4\}.$$

$$f_X(x) = \sum_{y \in \overline{S_Y}(x)} f(x,y), x \in \overline{S_X} = \{1,2,3,4\}$$

$$\Longrightarrow f_X(1) = \frac{7}{16}, \quad f_X(2) = \frac{5}{16}, \quad f_X(3) = \frac{3}{16}, \quad f_X(4) = \frac{1}{16}$$

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For
$$y \in \overline{S_Y}$$
,

$$f_Y(y) = P_Y(Y = y) \stackrel{\Delta}{=} P\left(\left\{X \in \overline{S_X}(y), Y = y\right\}\right)$$
$$= \sum_{x \in \overline{S_X}(y)} f(x, y)$$

where

$$\overline{S_X}(y) = \{x | (x,y) \in \overline{S}\}$$
 for the given $y \in \overline{S_Y}$.

$$f(x,y) = \begin{cases} \frac{2}{16}, & 1 \le x < y \le 4\\ \frac{1}{16}, & 1 \le x = y \le 4. \end{cases}$$

What is $f_X(x), f_Y(y)$?

First,
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$$f_{Y}(y) = \sum_{x \in \overline{S_{X}}(y)} f(x,y), y \in \overline{S_{Y}} = \{1,2,3,4\}$$

$$x \in \overline{S_{X}}(y) \quad \text{for } f_{Y}(y) = \text{f$$

Remarks on Marginal pmf

It is crucial to understand the following definitions

$$\overline{S}, \overline{S_X}, \overline{S_Y}, \overline{S_X}(y), \overline{S_Y}(x)$$

$$\overline{S} = \{ \text{all possible values of } (X, Y) \}$$

$$\overline{S_X} = \{ \text{all possible values of } X \} = \{ x | (x, y) \in \overline{S} \}$$

$$\overline{S_Y} = \{ \text{all possible values of } Y \} = \{ y | (x, y) \in \overline{S} \}$$

$$\overline{S_X}(y) = \{x | (x, y) \in \overline{S}\} \text{ for a given } y \in \overline{S_Y}$$
$$\overline{S_Y}(x) = \{y | (x, y) \in \overline{S}\} \text{ for a given } x \in \overline{S_X}$$

Description: The random experiment has three mutually exclusive and exhaustive outcomes:

- "perfect",
- "second"
- "defective"



We repeat the experiment n independent times, and moreover, the probabilities

- $\triangleright p_X$: the probability of "perfect",
- $\triangleright p_Y$: the probability of "second"
- \triangleright p_Z : the probability of "defective"

<u>remain the same</u> for each repetition. Such *n* repetitions can be called a trinomial experiment.

For the trinomial experiment, we are interested in the number of perfects, the number of seconds and the number of defectives.

For the n trinomial trials, we let

- X be number of perfects,
- Y be number of seconds,
- Z = n X Y be the number of defectives

We are interested in the joint pmf of (X, Y), $f(x, y) : \overline{S} \to \mathbb{R}^2$

- $\overline{S} = \{(x,y)|x+y \le n, x = 0, 1, \dots, n, y = 0, 1, \dots, n\}$
- f(x,y) = P(X = x, Y = y) which is the probability of having x perfects, y seconds, and n x y defectives

Joint pmf: to calculate f(x, y) = P(X = x, Y = y),

the probability for each way of having x perfects, y seconds, and n-x-y defectives is

$$p_X^x p_Y^y (1 - p_X - p_Y)^{n-x-y}$$

the total number of ways of having x perfects, y seconds, and n-x-y defectives is $\binom{n}{x_1, x_2, x_3} P_x^x P_y^y I - P_x - P_y^y$

$$\binom{n}{x,y,n-x-y} = \frac{n!}{x!y!(n-x-y)!}$$

Therefore, the joint pmf for trinomial distribution is

$$f(x,y) = \frac{n!}{x!y!(n-x-y)!} p_X^x p_Y^y (1-p_X-p_Y)^{n-x-y}, (x,y) \in \overline{S}$$

It's called trinomial distribution because of the trinomial expansion.

$$(a+b+c)^{n} = \sum_{x=0}^{n} \binom{n}{x} a^{x} (b+c)^{n-x}$$

$$= \sum_{x=0}^{n} \binom{n}{x} a^{x} \sum_{y=0}^{n-x} \binom{n-x}{y} b^{y} c^{n-x-y}$$

$$= \sum_{x=0}^{n} \sum_{y=0}^{n-x} \frac{n!}{x! y! (n-x-y)!} a^{x} b^{y} c^{n-x-y}$$

Marginal pmf: to calculate $f_X(x)$ or $f_Y(y)$

$$(a+b+c)^{n} = \sum_{x=0}^{n} \binom{n}{x} a^{x} (b+c)^{n-x}$$

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Marginal pmf: to calculate $f_X(x)$ or $f_Y(y)$

$$f_{X}(x) = \sum_{y \in \overline{S_{Y}}(x)} f(x,y) = \sum_{y=0}^{n-x} \binom{n}{x} \binom{n-x}{y} p_{X}^{x} p_{Y}^{y} (1-p_{X}-p_{Y})^{n-x-y}$$

$$= \binom{n}{x} p_{X}^{x} (1-p_{X})^{n-x} \int_{X} (x) = \sum_{y=0}^{n-x} \binom{n}{x} \binom{n-y}{y} p_{X}^{x} p_{Y}^{y} [1-p_{X}-p_{Y})^{n-x-y}$$

$$= \binom{n}{x} p_{X}^{x} (1-p_{X})^{n-x} \int_{X} (x) = \sum_{y=0}^{n-x} \binom{n}{x} \binom{n-y}{y} p_{X}^{x} p_{Y}^{y} [1-p_{X}-p_{Y})^{n-x-y}$$
Without summing, we know $X \sim b(n, p_{X})$ and $Y \sim b(n, p_{Y})$

Independent Random Variables

Definition

The random variables X and Y are said to be independent if for every $x \in S_X$ and $y \in S_Y$

$$f(x,y) = f_X(x)f_Y(y)$$
 Inde.

or equivalently,

$$P(X = x, Y = y) = P_X(X = x)P_Y(Y = y).$$

X and Y are said to be dependent if otherwise.

 $\overline{S} = \overline{S_X} \times \overline{S_Y}$, \overline{S} is said to be rectangular

which is a necessary condition for independence of X and Y.

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or equivalently,

$$f(x,y)=f_X(x)f_Y(y).$$

The definition of independent RVs has root in the definition of independent events.

$$A = \{X = x, Y \in \overline{S_Y}(x)\}, B = \{X \in \overline{S_X}(y), Y = y\}$$

X and Y are independent if and only if A and B are independent.

Example 2, Page 135

Let the joint pmf of X and Y be defined by

$$f(x,y) = \frac{x+y}{21}, \quad x = 1,2,3, \quad y = 1,2.$$

$$\overline{S} = \{(x, y) | x = 1, 2, 3, y = 1, 2.\}$$

$$f: \overline{S} \longrightarrow (0,1] \text{ with } \overline{S_X} = \{1,2,3\}, \quad \overline{S_Y} = \{1,2\}.$$

Question

Are X and Y independent or dependent?

Example 2, Page 135

$$\int_{y \in \overline{S_Y}(x)} f(x,y) = \sum_{y=1}^{2} \frac{x+y}{21} = \frac{2x+3}{21}, \quad x = 1, 2, 3.$$

$$f_Y(y) = \sum_{x \in \overline{S_X}(y)} f(x,y) = \sum_{x=1}^{3} \frac{x+y}{21} = \frac{3y+6}{21}, \quad y = 1, 2$$

$$f(x,y) = \frac{x+y}{21} \neq \frac{2x+3}{21} \cdot \frac{3y+6}{21} = f_X(x)f_Y(y)$$

$$\Rightarrow X \text{ and } Y \text{ are dependent}$$

What is the implication of independent RVs?

Implication of Independent RVs

Implication of independent RVs

For any $A \subset \overline{S_X}$ and $B \subset \overline{S_Y}$, the two events $X \in A$ and $Y \in B$ are independent. A $\subset \overline{S_X}$, $B \subset \overline{S_Y}$

We only need to show $P(A \cap B) = P(A)P(B)$:

$$P(A \cap B) = P(X \in A, Y \in B) = \sum_{x \in A, y \in B} f(x, y)$$

$$= \sum_{x \in A} \sum_{y \in B} f_X(x) f_Y(y)$$

$$= \sum_{x \in A} f_X(x) \sum_{y \in B} f_Y(y)$$

$$= P(X \in A) P(Y \in B) = P(A) P(B)$$

Let X and Y be discrete RVs with their joint pmf

$$f(x,y): \overline{S} \to (0,1]$$

Consider a function g(X, Y) of X and Y.

Then the expectation of g(X, Y) is

$$E[g(X,Y)] = \sum_{(x,y)\in\overline{S}} g(x,y)f(x,y)$$

When g(X, Y) = X, E[X] is the mean of XWhen $g(X, Y) = (X - E[X])^2$, $E[(X - E[X])^2]$ is the variance of X

When
$$g(X, Y) = X$$
, $E[X]$ is the mean of X
When $g(X, Y) = (X - E[X])^2$, $E[(X - E[X])^2]$ is the variance of X

There are seemingly two ways to calculate E[X]:

$$E[X] = \sum_{x \in \overline{S_X}} x f_x(x) \quad \text{Marginal}$$

$$E[X] = \sum_{x \in \overline{S_X}} x f(x, y)$$

$$E[X] = \sum_{(x,y) \in \overline{S}} x f(x, y)$$
Joint pmf

When
$$g(X, Y) = X$$
, $E[X]$ is the mean of X
When $g(X, Y) = (X - E[X])^2$, $E[(X - E[X])^2]$ is the variance of X

There seems two ways to calculate E[X]:

Equivalent
$$\begin{cases} E(X) = \sum_{x \in \overline{S_X}} x f_x(x) \\ E(X) = \sum_{x \in \overline{S$$

Example 1, [Page 134] —— Revisited

Question

Recall that X and Y are discrete RVs with joint pmf

$$f(x,y): \overline{S} \to (0,1]$$
 with $\overline{S_X} = \overline{S_Y} = \{1,2,3,4\}$

$$f(x,y) = \begin{cases} \frac{2}{16} & 1 \le x < y \le 4 \\ \frac{1}{16} & 1 \le x = y \le 4 \end{cases}$$

What is E[X + Y]?

Example 1, [Page 134] —— Revisited

$$\sum_{(X+Y)\in\overline{S}} (X+Y)f(X,Y) = \sum_{(X+Y)\in\overline{S}} (X+Y)\frac{2}{16}$$

$$E(X+Y) = \sum_{(x,y)\in\overline{S}} (x+y)f(x,y) + \sum_{(X+Y)\in\overline{S}} (x+y)\frac{2}{16}$$

$$= \sum_{1\leq x=y\leq 4} (x+y)\frac{1}{16} + \sum_{1\leq x< y\leq 4} (x+y)\frac{2}{16}$$

$$= \sum_{x=1}^{4} (2x)\frac{1}{16} + \sum_{x=1}^{4} \sum_{y\in\overline{S_Y}(x),x< y} (x+y)\frac{2}{16}$$
Note: the expectation is w.r.t all random variable, i.e. X and Y .

X
$$SX \in R$$
 $PX(A)$ $A \subseteq SX$

Y $SY \in R$ $PY(B)$ $B \subseteq SY$

(XY) $S \subseteq R^2$ $A \subseteq SX$

P(C) $C \subseteq S \subseteq R^2$ $B \subseteq SY$

PX $(A) = P(A \times SY)$
 $\Rightarrow Inde.$