STA2001 Probability and Statistics (I)

Lecture 12

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Key concepts and/or techniques:

Mathematical Expectation:

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

Covariance and correlation coefficient:

To study the relation between 2 RVs

$$Cov(X, Y) = E[(X - E[X])(Y - E[Y])] = E[XY] - E[X]E[Y]$$
$$\rho(X, Y) = \frac{Cov(X, Y)}{\sqrt{Var(X)}\sqrt{Var(Y)}}, \ \ Var(X) > 0, \ Var(Y) > 0.$$

 Interpretation and properties of covariance and correlation coefficient

- Cov(X, Y) = E(XY) − E(X)E(Y)
 When Cov(X, Y) = 0, X and Y are uncorrelated.
 When Cov(X, Y) > 0, X and Y are positively correlated.
 When Cov(X, Y) < 0, X and Y are negatively correlated.
- Interpretation: Roughly speaking, a positive or negative covariance indicate that the values of X E(X) and Y E(Y) obtained in a single experiment "tend" to have the same or the opposite sign respectively.
- ▶ Independence of X and Y \Rightarrow uncorrelation of X and Y, but the converse is in general not true.

Correlation coefficient

$$\rho(X,Y) = \frac{Cov(X,Y)}{\sqrt{Var(X)}\sqrt{Var(Y)}}$$

▶ It is a normalized version of Cov(X,Y) and in fact $-1 \le \rho(X,Y) \le 1$ and the size of $|\rho|$ provides a normalized measure of the extent to which this is true.

• $\rho=1$ or $(\rho=-1)$ if and only if there exists a positive. (or negative,respectively) constant c such that

$$Y - E(Y) = c(X - E(X))$$

Key concepts and/or techniques:

Conditional distribution Motivation: it is a probability distribution that describes the distribution of probability of events of a RV given the occurrence of a particular event.

For example, the conditional pmf of X given Y = y is

$$g(x|y) = \frac{f(x,y)}{f_Y(y)}, \quad x \in \overline{S_X}(y)$$

provided that $f_Y(y) > 0$.

Conditional mathematical expectations

The conditional expectation of g(Y) given X = x

$$E(g(Y)|X=x) = \sum_{y \in \overline{S_Y}(x)} g(y)h(y|x)$$

[Conditional pmf]

Conditional pmf of X given Y = y is defined by

$$g(x|y) = \frac{f(x,y)}{f_Y(y)}, \quad x \in \overline{S_X}(y)$$

provided that $f_Y(y) > 0$.

Similarly, conditional pmf of Y given that X = x is defined by

$$h(y|x) = \frac{f(x,y)}{f_X(x)}, \quad y \in \overline{S_Y}(x)$$

provided that $f_X(x) > 0$.

Conditional pmf

Conditional pmf is a well-defined pmf

- 1. h(y|x) > 0
- $2. \sum_{y \in \overline{S_Y}(x)} h(y|x) = 1$
- 3. for $A \subseteq \overline{S_Y}(x)$

$$P(Y \in A|X = x) = \sum_{y \in A} h(y|x)$$

[Conditional Mathematical Expectation]

ightharpoonup Let g(Y) be a function of Y.

Then the conditional expectation of g(Y) given X = x

$$E(g(Y)|X = x) = \sum_{y \in \overline{S_Y}(x)} g(y)h(y|x)$$

▶ When g(Y) = Y, conditional mean

$$E(Y|X=x) = \sum_{y \in \overline{S_Y}(x)} yh(y|x)$$

▶ When $g(Y) = [Y - E(Y|X = x)]^2$, conditional variance

$$Var(Y|X=x) \stackrel{\Delta}{=} E\{[Y-E(Y|X=x)]^2|X=x\}$$

Section 4.4 Bivariate Distribution of Continuous Type

Bivariate Continuous RV

Definition

Let X and Y be two continuous random variables and (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 continuous RV.

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

Then, it holds that

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

Roadmap for bivariate continuous random distributions

To study the bivariate continuous random variable

 $\mathsf{discrete}\ \mathsf{RV} \longrightarrow \mathsf{continuous}\ \mathsf{RV} \qquad \mathsf{Mathematical}\ \mathsf{expectations}$

 $pmf \longrightarrow pdf$ mean

joint pmf \longrightarrow joint pdf variance

marginal pmf — marginal pdf covariance

conditional pmf --> conditional pdf correlation coefficient

Joint pdf

Definition

The joint pdf of two continuous RVs X and Y is a function $f(x,y): \overline{S} \to (0,\infty)$ with the following properties:

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

2.
$$\iint_{\overline{S}} f(x,y) dx dy = 1$$

3.

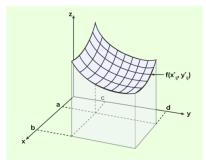
$$P((X,Y) \in A) \stackrel{\Delta}{=} P(\{(X,Y) \in A\})$$
$$= \iint_A f(x,y) dx dy, A \subseteq \overline{S}$$

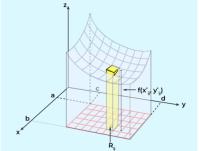
Remarks

Recall the geometric interpretation of double integral:

$$P((X,Y) \in A) = \iint_A f(x,y) dxdy$$

calculates the volume of the solid under the surface z = f(x, y) over the region A in the the xy-plane.

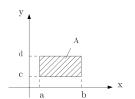




Remarks

- Very often, we extend the definition domain of f(x,y) from \overline{S} to $R \times R$ by letting f(x,y) = 0, for $(x,y) \notin \overline{S}$ and $\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x,y) dx dy = 1$
- Bottom line:

If the set A is rectangular with its line segments parallel to the coordinate axes, i.e.,



$$A = \{(x, y) | a \le x \le b, c \le y \le d\},\$$

then the double integral becomes

$$P((X,Y) \in A) = \int_{a}^{b} \int_{c}^{d} f(x,y) dy dx$$
$$= \int_{c}^{d} \int_{a}^{b} f(x,y) dx dy$$

Remarks

General case:

Let

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

Then

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

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

Then

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

Marginal pdf

Definition

The marginal pdf of X, $f_X(x): \overline{S_X} \to (0, \infty)$

$$f_X(x) = \int_{\overline{S_Y}(x)} f(x, y) dy$$
$$\overline{S_Y}(x) = \{ y | (x, y) \in \overline{S} \} \text{ for } x \in \overline{S_X}$$

The marginal pdf of Y, $f_Y(y): \overline{S_Y} \to (0, \infty)$

$$f_Y(y) = \int_{\overline{S_X}(y)} f(x, y) dx$$
$$\overline{S_X}(y) = \{x | (x, y) \in \overline{S}\} \text{ for } y \in \overline{S_Y}$$

Question

Let X and Y have the joint pdf

$$f(x,y) = \frac{3}{2}x^{2}(1 - |y|), \quad -1 < x < 1, \quad -1 < y < 1$$

$$\implies \begin{cases} \overline{S_{X}} = \{x | & -1 < x < 1\} \\ \overline{S_{Y}} = \{y | & -1 < y < 1\} \\ \overline{S} = \{(x,y) | -1 < x < 1, -1 < y < 1\} \end{cases}$$

- Q1: Let $A = \{(x, y) | 0 < x < 1, 0 < y < x\}$. What is the probability of A?
- Q2: What is the marginal pdf of X and Y?
- Q3: What is the expectation of X?

Q1:

$$P(A) = \int_0^1 \int_0^x \frac{3}{2} x^2 (1 - |y|) dy dx$$

$$= \int_0^1 \frac{3}{2} x^2 (x - \frac{1}{2} x^2) dx = \frac{3}{2} \cdot \frac{1}{4} x^4 \Big|_0^1 - \frac{3}{4} \cdot \frac{1}{5} x^5 \Big|_0^1$$

$$x = \int_0^1 \int_y^1 \frac{3}{2} x^2 (1 - |y|) dx dy$$

$$= \int_0^1 \frac{3}{2} \cdot \frac{1}{3} x^3 \Big|_y^1 (1 - |y|) dy$$

$$= \int_0^1 \frac{3}{2} \cdot \frac{1}{3} x^3 \Big|_y^1 (1 - |y|) dy$$

Q2:

For
$$x \in \overline{S_X}$$
, $f_X(x) = \int_{\overline{S_Y}(x)} f(x, y) dy$

$$= \int_{-1}^{1} \frac{3}{2} x^2 (1 - |y|) dy = \frac{3}{2} x^2 (2 + (-1)) = \frac{3}{2} x^2$$
For $y \in \overline{S_Y}$, $f_Y(y) = \int_{\overline{S_X}(y)} f(x, y) dx$

$$= \int_{-1}^{1} \frac{3}{2} x^2 (1 - |y|) dx = \frac{3}{2} (1 - |y|) \frac{1}{3} x^3 \Big|_{-1}^{1}$$

$$= 1 - |y|$$

Q3:

$$E(X) = \int_{\overline{S_X}} x f_X(x) dx = \int_{-1}^1 x \frac{3}{2} x^2 dx = 0$$

$$E(X) = \int \int_{\overline{S}} x f(x, y) dx dy = \int_{-1}^{1} \int_{-1}^{1} \frac{3}{2} x^{3} (1 - |y|) dx dy$$
$$= \int_{-1}^{1} \frac{3}{2} x^{3} dx \int_{-1}^{1} (1 - |y|) dy = 0$$

Mathematical Expectation

Definition

Let g(X, Y) be a function of X and Y, whose joint pdf $f(x, y) : \overline{S} \to (0, \infty)$. Then

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

▶
$$g(X, Y) = X \rightarrow \text{mean of } X$$

$$E[X] = \iint_{\overline{S}} x f(x, y) dx dy$$
$$= \int_{\overline{S_X}} x \int_{\overline{S_Y}(x)} f(x, y) dy dx$$
$$= \int_{\overline{S_Y}} x f_X(x) dx$$

Mathematical Expectation

Definition

$$g(X, Y) = (X - E[X])^{2} \rightarrow \text{variance of } X$$

$$Var[X] = \iint_{\overline{S}} (x - E[X])^{2} f(x, y) dx dy$$

$$= \int_{\overline{S_{X}}} (x - E[X])^{2} \int_{\overline{S_{Y}}(x)} f(x, y) dy dx$$

$$= \int_{\overline{S_{X}}} (x - E[X])^{2} f_{X}(x) dx$$

Question

Let X and Y have the joint pdf

$$f(x,y) = (\frac{4}{3})(1-xy)$$
 with $0 \le x \le 1$, $0 \le y \le 1$

- Q1 Find the marginal pdfs of X and Y?
- \mathbb{Q}^2 Find the expectation of X?
- Q3 Find the variance of X

Question

Let X and Y have the joint pdf

$$f(x,y) = (\frac{4}{3})(1-xy)$$
 with $0 \le x \le 1$, $0 \le y \le 1$

- Q1 Find the marginal pdfs of X and Y?
- Q2 Find the expectation of X?
- Q3 Find the variance of X

Q1:

$$f_X(x) = \int_{\overline{S_Y}(x)} f(x, y) dy = \int_0^1 \frac{4}{3} (1 - xy) dy$$
$$= \frac{4}{3} - \frac{4}{3} x \frac{1}{2} y^2 \Big|_0^1 = \frac{4}{3} (1 - \frac{1}{2} x)$$

$$f_Y(y) = \int_{\overline{S_X}(y)} f(x, y) dx$$
$$= \int_0^1 \frac{4}{3} (1 - xy) dx = \frac{4}{3} (1 - \frac{1}{2}y)$$

Q2:

$$E[X] = \int_{\overline{S_X}} x f_X(x) dx = \int_0^1 x \frac{4}{3} (1 - \frac{1}{2}x) dx$$
$$= \frac{4}{3} \cdot \frac{1}{2} x^2 \Big|_0^1 - \frac{4}{6} \cdot \frac{1}{3} x^3 \Big|_0^1 = \frac{4}{9}$$

Q3:

$$Var[X] = \int_{\overline{S_X}} (x - E[X])^2 f_X(x) dx = \int_0^1 (x - \frac{4}{9})^2 \frac{4}{3} (1 - \frac{1}{2}x) dx$$

Independent Continuous RVs

Definition

Two continuous RVs X and Y are independent if

$$f(x,y) = f_X(x)f_Y(y), \qquad x \in \overline{S_X}, \quad y \in \overline{S_Y}$$

If X and Y are not independent, then we say X and Y are dependent.

When X and Y are independent,

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

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

Example 2 — revisited

Note that

$$f(x,y) = (\frac{4}{3})(1-xy) \quad \text{with} \quad 0 \le x \le 1, \quad 0 \le y \le 1$$

$$f_X(x) = \frac{4}{3}(1-\frac{1}{2}x)$$

$$f_Y(y) = \frac{4}{3}(1-\frac{1}{2}y)$$

Since $f(x, y) \neq f_X(x)f_Y(y)$, X and Y are NOT independent.

Covariance and Correlation Coefficient

Definition

$$Cov(X, Y) = E[(X - E[X])(Y - E[Y])]$$
$$= E(XY) - E(X)E(Y)$$

$$E(XY) = \iint_{\overline{S}} xyf(x,y)dxdy.$$

$$\rho(X,Y) = \frac{Cov(X,Y)}{\sqrt{Var(X)}\sqrt{Var(Y)}}, \ \ Var(X) > 0, \ \ Var(Y) > 0.$$

Conditional pdf

Definition

Let X and Y have a joint pdf $f(x,y): \overline{S} \to (0,\infty)$ and marginal pdf $f_X(x): \overline{S_X} \to (0,\infty)$ and $f_Y(y): \overline{S_Y} \to (0,\infty)$.

The conditional pdf of Y, given that X = x are

$$h(y|x) = \frac{f(x,y)}{f_X(x)}$$
 for $f_X(x) > 0, y \in \overline{S_Y}(x)$

For
$$A \subseteq \overline{S_Y}(x)$$
, $P(Y \in A|X = x) = \int_{y \in A} h(y|x)dy$.

Conditional pdf

Definition

Let X and Y have a joint pdf $f(x,y): \overline{S} \to (0,\infty)$ and marginal pdf $f_X(x): \overline{S_X} \to (0,\infty)$ and $f_Y(y): \overline{S_Y} \to (0,\infty)$.

The conditional pdf of Y, given that X = x are

$$h(y|x) = \frac{f(x,y)}{f_X(x)}$$
 for $f_X(x) > 0, y \in \overline{S_Y}(x)$

For
$$A \subseteq \overline{S_Y}(x)$$
, $P(Y \in A|X = x) = \int_{y \in A} h(y|x)dy$.

The conditional pdf of X, given that Y = y are

$$g(x|y) = \frac{f(x,y)}{f_Y(y)}$$
 for $f_Y(y) > 0, x \in \overline{S_X}(y)$

For
$$A \subseteq \overline{S_X}(y)$$
, $P(X \in A|Y = y) = \int_{x \in A} g(x|y)dx$.

Conditional mathematical expectation

Definition

The conditional mathematical expectation of a function of Y, g(Y), given that X = x is

$$E(g(Y)|X=x) = \int_{\overline{S_Y}(x)} g(y)h(y|x)dy$$

The conditional mean and variance of Y, given that X = x are

$$E(Y|X=x) = \int_{\overline{S_Y}(x)} yh(y|x)dy$$

$$Var(Y|X=x) = E\{[Y - E(Y|X=x)]^2 | X = x\}$$

$$= \int_{\overline{S_Y}(x)} [y - E(Y|X=x)]^2 h(y|x)dy$$

$$= E[Y^2 | X = x] - [E(Y|X=x)]^2$$

Question

Let X and Y be two continuous RVs with

$$f(x,y) = 2, \quad 0 \le x \le y \le 1$$

$$\overline{S} = \{(x,y) | 0 \le x \le y \le 1\}, \overline{S_X} = \overline{S_Y} = [0,1]$$

Q1:
$$f_X(x), f_Y(y), E(X), E(Y)$$
 ?

Q2:
$$h(y|x), E(Y|X = x), Var(Y|X = x)$$
?

Q3:
$$P(\frac{3}{4} \le Y \le \frac{7}{8} | X = \frac{1}{4})$$
 ?

Q1:

$$f_{X}(x) = \int_{\overline{S_{Y}}(x)} f(x, y) dy = \int_{x}^{1} 2dy = 2(1 - x), 0 \le x \le 1$$

$$f_{Y}(y) = \int_{\overline{S_{X}}(y)} f(x, y) dx = \int_{0}^{y} 2dy = 2y, \quad 0 \le y \le 1$$

$$E(X) = \int_{\overline{S_{X}}} x f_{X}(x) dx = \int_{0}^{1} x 2(1 - x) = \frac{1}{3}$$

$$x E(Y) = \int_{\overline{S}} y f_{Y}(y) dy = \int_{0}^{1} 2y^{2} dy = \frac{2}{3}$$

Q2:

$$h(y|x) = \frac{f(x,y)}{f_X(x)} = \frac{1}{1-x}, 0 \le x \le y \le 1.$$

$$E(Y|X = x) = \int_{\overline{S_Y}(x)} yh(y|x)dy = \int_x^1 y \frac{1}{1-x} dy$$

$$= \frac{1}{2(1-x)} (1-x^2) = \frac{1}{2} (1+x)$$

$$Var(Y|X = x) = \int_{\overline{S_Y}(x)} [y - \frac{1}{2} (1+x)]^2 h(y|x) dy$$

$$= \int_x^1 \frac{1}{1-x} [y - \frac{1}{2} (1+x)]^2 dy$$

$$= \frac{1}{3} \frac{1}{1-x} [y - \frac{1}{2} (1+x)]^3 \Big|_x^1 = \frac{1}{12} (1-x)^2$$

Q3:

$$P(\frac{3}{4} \le Y \le \frac{7}{8}|X = \frac{1}{4}) = \int_{\frac{3}{4}}^{\frac{7}{8}} h(y|\frac{1}{4})dy = \frac{4}{3}(\frac{7}{8} - \frac{3}{4}) = \frac{1}{6}$$