

Multivariate distributions

Introduction

- Generally interested in more than one random variable at a time

1. n observations of a single characteristic from some population

X_1, \dots, X_n $X_i \equiv$ weight of the i th Unit

2. k different characteristics from a single individual

One person Y_1 Y_2 Y_3 \dots
 \downarrow \downarrow \downarrow
 age height gender

- Notation: $\tilde{X} = \mathbf{X} = (X_1, \dots, X_n)$ is an n -dimensional random vector (r.v.)

In other words, \mathbf{X} is a function from sample space $S \rightarrow \mathbb{R}^n$

$$\begin{aligned} \tilde{X} : \Omega &\longrightarrow \mathbb{R}^n \\ S &\longrightarrow \mathbb{R}^n \\ \tilde{X}(\omega) &= (X_1(\omega), \dots, X_n(\omega)) \in \mathbb{R}^n \end{aligned}$$

- General plan:

use the bivariate case (X_1, X_2) ($n = 2$) to derive results and then extend to any n

For notational simplicity, in the bivariate case, we'll denote a pair of random variables as (X, Y) instead of (X_1, X_2)

Multivariate distributions

Discrete random vectors & probability mass functions

Definition:

~~Definition: If there exists a countable set of points $\{(x_i, y_i)\}$ such that~~

$$\sum_{i=1}^{\infty} P[(X, Y) = (x_i, y_i)] = 1$$

$X \in \{x_1, x_2, \dots, x_n\}$

$Y \in \{y_1, y_2, \dots, y_m\}$

$P(X, Y)$ is joint pmf

~~we say that the distribution of X and Y is **discrete** and the function~~

$$f(x, y) = P(X = x, Y = y) \geq 0 \quad \text{for any } (x, y) \in \mathbb{R}^2$$

If $\sum_{j=1}^m \sum_{i=1}^n P(x_i, y_j) = 1$

is called the joint probability mass function (joint pmf) for X and Y .

Example (artificial): (X, Y) with joint pmf given in tabular form as

		x		
		1	2	3
y	3	1/12	1/12	1/6
	2	1/12	1/6	1/12
	1	1/6	1/12	1/12

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

$$f(1, 2) = P(X = 1, Y = 2) = 1/12$$

$$f(3, 3) = 1/6$$

$$f(x, y) = 0 \text{ for all other } x, y \in \mathbb{R}$$

For (X, Y) jointly discrete with pmf $f(x, y)$,

$$P[(X, Y) \in A] = \sum_{(x, y) \in A} f(x, y) \quad \text{for } A \subset \mathbb{R}^2$$

$$P(Y \geq X) = \sum_{(x, y): y \geq x} f(x, y) = f(1, 3) + f(2, 3) + f(3, 3) + f(1, 2) + f(2, 2) + f(1, 1) = 3/4$$

Multivariate distributions

Another joint pmf example (i.e., discrete case again)

- Suppose there are n independent trials, where each trial has three possible outcomes: outcome a with prob. p_1 , outcome b with prob. p_2 and outcome c with prob. $1 - p_1 - p_2$

- Let $X = \#$ of trials having outcome a & $Y = \#$ of trials having outcome b

- Then the joint pmf of (X, Y) is

$$P(X=x, Y=y) = f(x, y) = \binom{n}{x} \binom{n-x}{y} p_1^x p_2^y (1-p_1-p_2)^{n-x-y}$$

Out of n trials choose x trials to "a"

$$= \frac{n!}{x!y!(n-x-y)!} p_1^x p_2^y (1-p_1-p_2)^{n-x-y}$$

X has the range from $0, 1, 2, \dots, n$
 Y has the range from $0, 1, 2, \dots, n$
 $0 \leq X+Y \leq n$

Defined for $0 < p_1, p_2, p_1 + p_2 < 1$, and the support/range is given by all integer pairs (x, y) where $0 \leq x, y, x + y \leq n$

otherwise

- If $n = 1$, then $f(1, 0) = p_1$ $f(0, 1) = p_2$ $f(0, 0) = 1 - p_1 - p_2$

$f(x, y)$			
		0	1
y	0	$1 - p_1 - p_2$	p_1
	1	p_2	0

- If $n = 2$, then $f(2, 0) = p_1^2$, $f(1, 1) = 2p_1p_2$, etc.

		0	1	2
y	0	$(1 - p_1 - p_2)^2$	$2p_1(1 - p_1 - p_2)$	p_1^2
	1	$2p_2(1 - p_1 - p_2)$	$2p_1p_2$	0
	2	p_2^2	0	0

- multinomial distribution with three outcomes: $(X, Y) \sim \text{Multinomial}(n, p_1, p_2)$
a binominal distribution is a multinomial distribution with two outcomes

Multivariate distributions

Continuous random vectors & probability density functions

Definition: If there exists a function $f(x, y) \geq 0$ for $(x, y) \in \mathbb{R}^2$ such that

$$\int_{\mathbb{R}^2} f(x, y) \underline{dx dy} = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) \underline{dx dy} = 1$$

and

$$P((X, Y) \in A) = \int_A \int f(x, y) dx dy \quad \text{for } A \subset \mathbb{R}^2$$

then we say that the distribution of X and Y is **jointly (absolutely) continuous**. The function $f(x, y)$ is called the **joint probability density function** (joint pdf) for X and Y .



Note: As in the univariate case, any function $f(x, y)$ with

✓ 1. $f(x, y) \geq 0$ for all $(x, y) \in \mathbb{R}^2$

✓ 2. $\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) dx dy = \underline{1}$ (continuous case)



or $\sum_{(x, y) \in \mathbb{R}^2} f(x, y) = 1$ (discrete case)

$$\sum_x \sum_y P(X=x, Y=y)$$

specifies the joint pdf or pmf of some bivariate random vector (X, Y)

Calculus we need for the Course

$$\int x^m dx = \frac{1}{m+1} x^{m+1}$$

$$\int e^x dx = e^x$$

$$\textcircled{*} \rightarrow \int U dV = UV - \int V dU$$

$$\rightarrow \int e^{-x} x^{m-1} dx = \Gamma(m)$$

$$(f(g(x)))' = g'(x) f'(g(x))$$

$$[e^{f(x)}]' = f'(x) e^{f(x)}$$

$$(\log f(x))' = \frac{f'(x)}{f(x)}$$

$$\left(\frac{f(x)}{g(x)}\right)' = \frac{f'g - g'f}{g^2}$$

$$[x^m]' = m x^{m-1}$$

$$[f(x)^m]' = m f'(x) [f(x)]^{m-1}$$

$$\int_{-a}^a f(x) dx \begin{matrix} \text{even} \\ f \text{ is} \end{matrix} 2 \int_0^a f(x) dx$$

$$\int_{-a}^a f(x) dx \begin{matrix} \text{odd} \\ f \text{ is} \end{matrix} 0$$

$$\left\{ \begin{array}{l} \text{we say } f \text{ is even If } f(x) = f(-x) \text{ e.g. } f(x) = x^2 \\ \text{we say } f \text{ is odd If } f(-x) = -f(x) \text{ e.g. } f(x) = x^3 \\ f(x) = \sin x \end{array} \right.$$

$$(\sin u)' = u' \cos u$$

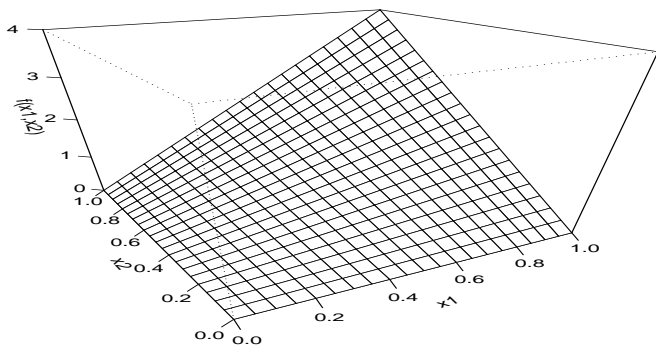
$$(\cos u)' = -u' \sin u$$

$$\left\{ \begin{array}{l} B(a,b) = \frac{\Gamma(b)\Gamma(a)}{\Gamma(a+b)} \\ B(a,b) = \int_0^1 x^{a-1} (1-x)^{b-1} dx = \frac{\Gamma(b)\Gamma(a)}{\Gamma(a+b)} \end{array} \right.$$

Multivariate distributions

Joint pdf example (i.e., continuous case)

Suppose (X, Y) have joint pdf $\underline{\underline{f(x, y) = 4xy}}$ for $0 < x, y < 1$



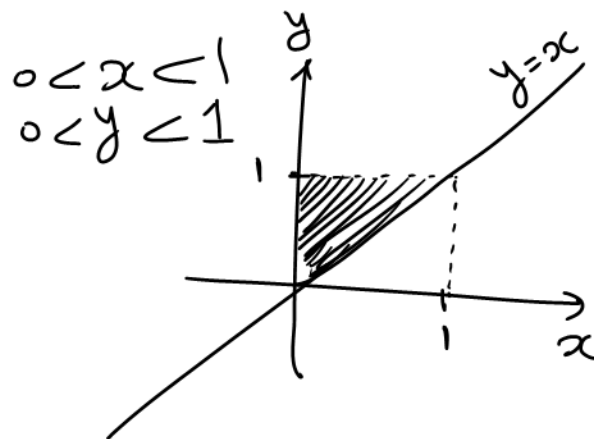
Find $P(X < Y)$

$$P(X < Y) = \iint_A f(x, y) dx dy$$

$$A = \{(x, y) : y - x > 0\}$$

$$= \int_0^1 \left[\int_0^y 4xy dx \right] dy$$

$$= \int_0^1 \left[y(2x^2) \Big|_0^y \right] dy = \int_0^1 y(2y^2) dy = 2 \int_0^1 y^3 dy = 2 \left(\frac{1}{4} y^4 \right) \Big|_0^1 = \frac{1}{2}$$



Find $P(X + Y < 1)$

$$\int_0^1 \int_0^{1-y} 4xy dx dy$$

$$x + y < 1 \Rightarrow x < 1 - y$$

