

Multivariate distributions

Introduction

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

1. n observations of a single characteristic from some population

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

2. k different characteristics from a single individual



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

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

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

- General plan:

use the bivariate case $(\underline{X_1, X_2})$ ($n = 2$) 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 $(\underline{X}, \underline{Y})$ instead of $(\underline{X_1, X_2})$

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

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

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

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
		1	1/12 1/12 1/6
y	3	1/12 1/12 1/6	1/6
	2	1/12 1/6 1/12	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 $f(1, 0) = p_1$ $f(0, 1) = p_2$ $f(0, 0) = 1 - p_1 - p_2$

		x	y
		0	1
x	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.

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

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

a binomial 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

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

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. $\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) dx dy = 1$ (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

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

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

\rightarrow $\int v dv = uv - \int u dv$

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

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

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

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

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

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

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

$$\int_{-a}^a f(x) dx \stackrel{\text{even}}{\underline{f \text{ is}}} 2 \int_0^a f(x) dx$$

$$\int_{-a}^a f(x) dx \stackrel{\text{odd}}{\underline{f \text{ is}}} 0$$

} We say f is even If $f(x) = f(-x)$ e.g. $f(x) = x^2$
 } We say f is odd If $f(-x) = -f(x)$ e.g. $f(x) = x^3$
 } $f(x) = \sin x$

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

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

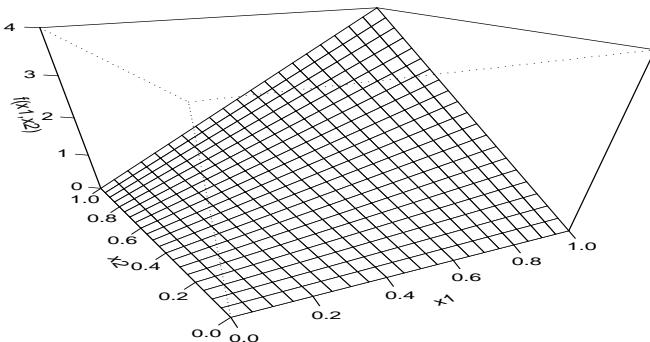
$$\left\{ B(a,b) = \frac{\Gamma(b)\Gamma(a)}{\Gamma(a+b)} \right.$$

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

Multivariate distributions

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

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



Find $P(X < Y)$

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

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

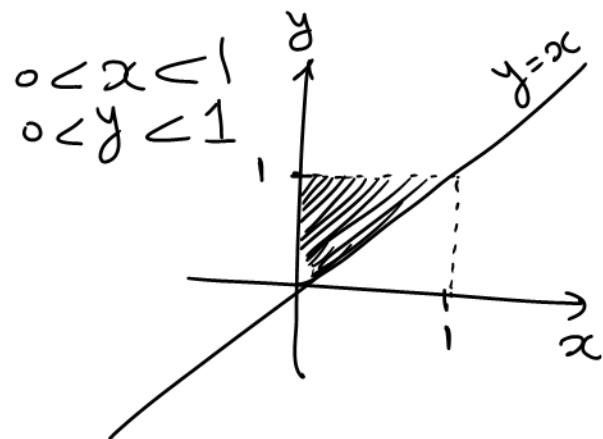
$$= \int_0^1 \left[\int_0^y 4xy dx \right] dy$$

$$= \int_0^1 \left[y(2x^2) \Big|_0^y \right] dy$$

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

Find $P(X + Y < 1)$

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



$$dy = \int_0^1 y(2y^2) dy = 2 \int_0^1 y^3 dy = 2/4 y^2 \Big|_0^1 = 1/2$$

