

Random vectors

Stat 561
11-27-18

A random vector \vec{X} is a function from the sample space S into \mathbb{R}^n .

①

(i.e. it is a vector whose components are random variables)

A random vector is discrete if its range in \mathbb{R}^n is finite or countably infinite.

For $A \subset \mathbb{R}^n$,

$$P(A) = P[\vec{X} \in A]$$

Defn: For a discrete random vector, the joint probability mass function is

$$p(x_1, x_2, \dots, x_n) = P[X_1 = x_1 \cap X_2 = x_2 \cap \dots \cap X_n = x_n]$$

$$\text{So } P(A) = \sum_{(x_1, \dots, x_n) \in A} \sum \dots \sum p(x_1, x_2, \dots, x_n)$$

②

Example: 4 slips of paper, numbered 1, 2, 3, 4

Draw 2 w.o.r.

Let $X_1 = \min$ $X_2 = \max$

$$S = \{12, 13, 14, 21, 23, 24, 31, 32, 34, 41, 42, 43\}$$

$p(x_1, x_2)$		x_2			
		2	3	4	
x_1	1	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{3}{6}$
	2	0	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{2}{6}$
	3	0	0	$\frac{1}{6}$	$\frac{1}{6}$
		$\frac{1}{6}$	$\frac{2}{6}$	$\frac{3}{6}$	1

Theorem:
$$p_1(x_1) = \sum_{\text{all } x_2} p(x_1, x_2)$$

Proof:
$$p_1(x_1) = P[X_1 = x_1]$$

$$= P[X_1 = x_1 \cap -\infty < X_2 < \infty]$$

$$= \sum_{\text{all } x_2} P[X_1 = x_1 \cap X_2 = x_2]$$

$$= \sum_{\text{all } x_2} p(x_1, x_2)$$

(5)

Continuous Case: Assume every component of the random vector is continuous

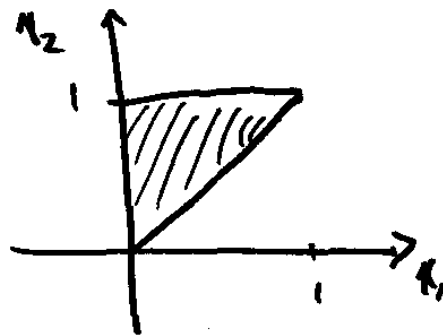
Defn: $f(x_1, \dots, x_n)$ is a joint probability density function if, $\forall A \subset \mathbb{R}^n$

$$P[\vec{X} \in A] = \int \int \dots \int_A f(x_1, \dots, x_n) dx_1 \dots dx_n$$

(6)

Example: $f(x_1, x_2) = 8x_1x_2$ $0 < x_1 < x_2 < 1$

Check to see if $f(x_1, x_2)$ integrates to 1.



$$\int_0^1 \int_0^{x_2} 8x_1x_2 dx_1 dx_2 \quad \text{OR} \quad \int_0^1 \int_{x_1}^1 8x_1x_2 dx_2 dx_1$$

$$8 \int_0^1 x_2 \left. \frac{x_1^2}{2} \right|_{x_1=0}^{x_2} dx_2 = 8 \int_0^1 \frac{x_2^3}{2} dx_2 = \left. x_2^4 \right|_0^1 = 1 \checkmark$$

The marginal prob. density fn. for X_1 is

(7)

$$\begin{aligned} f_1(x_1) &= \int_{x_1}^1 f(x_1, x_2) dx_2 \\ &= \int_{x_1}^1 8x_1 x_2 dx_2 = 8x_1 \left. \frac{x_2^2}{2} \right|_{x_2=x_1}^1 \\ &= 4x_1 - 4x_1^3 \end{aligned}$$

$$f_1(x_1) = 4x_1 - 4x_1^3, \quad 0 < x_1 < 1$$

The marginal prob. density fn. for X_2 is

(8)

$$\begin{aligned} f_2(x_2) &= \int_0^{x_2} 8x_1 x_2 dx_1 \\ &= 8x_2 \left. \frac{x_1^2}{2} \right|_{x_1=0}^{x_2} = 4x_2^3 \end{aligned}$$

$$f_2(x_2) = 4x_2^3 \quad 0 < x_2 < 1$$

(9)

Let $g(X_1, X_2, \dots, X_n)$ be a new random variable.

$$\text{Defn. } E[g(\vec{X})] = \begin{cases} \sum \dots \sum_{\text{all } \vec{x}} g(\vec{x}) p(\vec{x}) \\ \int \dots \int_{\mathbb{R}^n} g(\vec{x}) f(\vec{x}) d\vec{x} \end{cases}$$

In our example, find $E[X_1 + X_2]$

$$\begin{aligned} \text{Method 1: } E[X_1 + X_2] &= \int_0^1 \int_0^{x_2} (x_1 + x_2) \delta_{x_1, x_2} dx_1 dx_2 \quad (10) \\ &= \int_0^1 \int_0^{x_2} (\delta_{x_1} x_2 + \delta_{x_1} x_1^2) dx_1 dx_2 \\ &= \int_0^1 \left(\frac{\delta_{x_1}^3 x_2}{3} + \frac{\delta_{x_1}^2 x_2^2}{2} \right) \Big|_{x_1=0}^{x_2} dx_2 \\ &= \int_0^1 \frac{\delta_{x_2}^4}{3} + \frac{\delta_{x_2}^4}{2} dx_2 = \frac{20}{3} \int_0^1 x_2^4 dx_2 \end{aligned}$$

$$= \frac{20}{3} \frac{x_2^5}{5} \Big|_0^1 = \frac{4}{3}.$$

(11)

Method 2: $E[X_1 + X_2] = E[X_1] + E[X_2]$

$$E[X_1] = \int_0^1 x_1 (4x_1 - 4x_1^3) dx_1 = \frac{8}{15}$$

$$E[X_2] = \int_0^1 x_2 4x_2^3 dx_2 = \frac{4}{5}$$

So $E[X_1 + X_2] = \frac{8}{15} + \frac{4}{5} = \frac{4}{3}.$

Conditional distributions

(12)

$$\text{Defn: } \begin{cases} f(x_2 | x_1) = \frac{f(x_1, x_2)}{f_1(x_1)} \\ p(x_2 | x_1) = \frac{p(x_1, x_2)}{p_1(x_1)} \end{cases}$$

Note: $\int_{\text{all } x_2} f(x_2 | x_1) dx_2 = \int_{\text{all } x_2} \frac{f(x_1, x_2)}{f_1(x_1)} dx_2$

$$= \frac{1}{f_1(x_1)} \underbrace{\int_{\text{all } x_2} f(x_1, x_2) dx_2}_{f_1(x_1)} = 1 \quad \checkmark \quad (13)$$

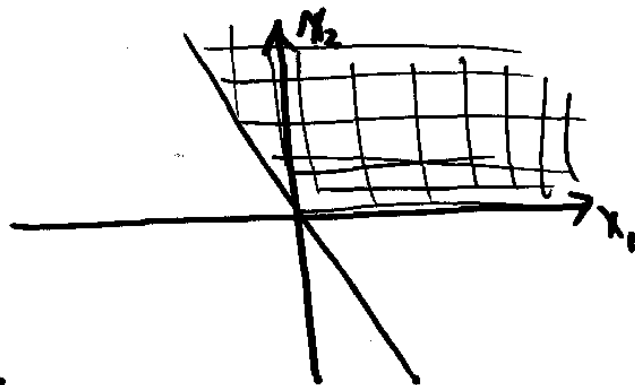
Let $g(x_2)$ be a function of x_2 .

Defn: $E[g(x_2) | x_1] = \begin{cases} \sum_{\text{all } x_2} g(x_2) p(x_2 | x_1) \\ \int_{\text{all } x_2} g(x_2) f(x_2 | x_1) dx_2 \end{cases}$

Note: $E[g(x_2)] = \text{number}$
 $E[g(x_2) | x_1] = \text{function of } x_1$

Example: $f(x_1, x_2) = \frac{1}{2} e^{-(x_1 + x_2)}$, $x_2 > 0$ (14)
 $x_2 > -2x_1$

Find $f(x_2 | x_1)$



$$f_1(x_1) = \int_{\text{all } x_2} f(x_1, x_2) dx_2$$

$$= \begin{cases} \int_0^{\infty} \frac{1}{2} e^{-(x_1 + x_2)} dx_2 & x_1 \geq 0 \\ \int_{-2x_1}^{\infty} \frac{1}{2} e^{-(x_1 + x_2)} dx_2 & x_1 < 0 \end{cases}$$

(15)

$$= \begin{cases} \frac{1}{2} e^{-x_1} \left[-e^{-x_2} \right]_{x_2=0}^{\infty} & x_1 \geq 0 \\ = 0 - \left(-\frac{1}{2} e^{-x_1} \right) = \frac{1}{2} e^{-x_1} \\ \frac{\frac{1}{2} e^{-x_1} \left[-e^{-x_2} \right]_{x_2=-2x_1}^{\infty}}{=} & x_1 < 0 \\ = 0 + \frac{1}{2} e^{-x_1} e^{2x_1} = \frac{1}{2} e^{x_1} \end{cases}$$

$$= \frac{1}{2} e^{-|x_1|} \quad -\infty < x_1 < \infty$$

(16)

$$\begin{aligned} f(x_2|x_1) &= \frac{f(x_1, x_2)}{f_1(x_1)} \\ &= \frac{\frac{1}{2} e^{-(x_1+x_2)}}{\frac{1}{2} e^{-|x_1|}} \\ &= e^{|x_1| - x_1 - x_2} \end{aligned}$$

$x_2 > 0$
 $x_2 > -2x_1$