## Hultivariate random vectors

$$X = (X_1, \dots, X_p)'$$
 $X_i \cap \alpha r. V. \text{ on } (\Omega, \mathcal{F}_t, \mathcal{P}) \text{ say}$ 
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Dist for joint dist for X is a random vector

 $F_{X}(x_{1},\ldots,x_{p}) = P(\omega: X_{1}(\omega) \leq x_{1}, X_{2}(\omega) \leq x_{2},\ldots X_{p}(\omega) \leq x_{p})$   $= P(X_{1} \leq x_{1}, X_{2} \leq x_{2},\ldots,X_{p} \leq x_{p})$ 

Note that Fx (.) as defined above satisfies

(i)  $\lim_{\substack{m \in X \\ m \neq x}} F_X(\underline{x}) = 1$ 

(ii)  $\lim_{x \to -\alpha} F_{x(x)} = 0$  for i = 1, -., b

(iii)  $F_{X}(x)$  is non decreasing in each argument (iv).  $F_{X}(x)$  is right continuous in each argument

Remark:  $\frac{1}{p=2} \lim_{X_2 \to a^*} F_X(x_1, x_2) = F_X(x_1, a^*) = F_X(x_1)$ marginal dist" f" f f

In general for any K=1, -- 11.

lim F (x1, -, xb) = F (x1, .xk-1, xk+1, -.xb)

xk >d x (x1, .xk-1, xk+1, ..xb)

marginal joint 1-f. } (X1, -. X K-1, X K+1) -. X b)

## Remark:

P(X & B) can be expressed through Joint d.f.

p-dimensional semiclosed rectangle of the form

(a, b, ] x (a2, b2] x - · · x (ax, bx)

a; < b; for i=1,...,k

Consider p=2, to have a feel

a, <b, ; a2 < b2

$$P(a, \langle X, \leq b_1, a_2 \langle X_2 \leq b_2)$$

$$= P(x_1 \le b_1, \dot{x}_2 \le b_2)$$

$$=F_{x_1,x_2}(b_1,b_2)-F_{x_1,x_2}(a_1,b_2)-F_{x_1,x_2}(b_1,a_2)$$

Remark: The four conditions that Fx(.) satisfies (i)-(iv) are

not n. s. c. for a f' to be d.f. of rondom vector.

We additionally need condition (efor \$=2) that

$$P(a_1 < X_1 \leq b_1, a_2 < X_2 \leq b_2) > 0 + a_1 < b_1$$

## Discrete random vector

Ad': A random vector  $X = (X_1, ..., X_p)'$  is said to be discrete

If there exist a countable set  $E \subset \mathbb{R}^p \supset P(X \in E) = 1$ (finite or countably infinite)

i.e.  $P_X(X) = P(X_1 = X_1, \dots, X_p = X_p)$ .

Consider a birariate setup b=2 case

(i) marginal p.m.t. of X; can be obtained by summing over possible values of the other variable

e.g  $P(X_1=x_1) = \sum_{y} P(X_1=x_1, X_2=y)$ ; i=1,2,--Sly  $P(X_2=y_1) = \sum_{x} P(X_1=x_1, X_2=y_2)$ ; j=1,2,--

(ii) Conditional dist of X, given X2 or X2 given X, conditional p.m.t of X2 given X,

$$\frac{|Y|}{|X_2|} = \frac{P(X=Y, X_1=x_i)}{P(X_1=X_i)}$$

Sty  $P_{X_1|X_2=y_i}$  =  $\frac{P(X_1=X_1,X_2=y_i)}{P(X_1=y_i)}$ 

For a p-dimensional random vector  $X = (X_1, -.. \times p)$ , We can obtain p non 1-variate marginals for  $X_1, X_2, -.. \times p$ (p) 2-variate joint marginals for  $(X_i, X_i)$ 

and so on

One can obtain conditional p.m.t. of (Xi, Xiz, ..., Xir)

Given the remaining variables on or any subset of

the remaining variables.

Independence: Discrete random variables are XII. Xp

are independent iff the joint p.m. f can be expressed

as

 $P(X_1=x_1,...,X_p=x_p) = \frac{1}{11} P(X_1=x_1)$   $(x_1,...,x_p) \in E$ 

Note that in such a case conditional p.m.f. will be identical to unconditional p.m.f. s.

## Example à a multivariate discrete dist

Consider a random experiment with 3 mutually exclusive and exhaustive outcomes,  $A_{1,1}A_{2} \& A_{3}$  with probabilities  $\theta_{1,1}, \theta_{2,1}, \theta_{3,2}$ , respectively. Repeat the trials in times along the fine.

Alefine

X,: number of times A, occurs out of n trials

X2: ---
A2 occurs out of n trials

X3: ---
A3 occurs out of n trials

Let  $(x_1, x_2, x_3)$  denote the observed count in ntrials  $x_i \ge 0$ ,  $x_i \le n$   $\forall i=1,2,3$  &  $\sum_{i=1}^3 x_i = n$   $E = \left\{ (x_1, x_2, x_3) : 0 \le x_i \le n , \sum_{i=1}^3 x_i = n \right\} - \text{finite number of points}$ 

It p.m.f.

$$P(X_1=x_1, X_2=x_2, X_3=x_3) = \frac{n!}{x_1! x_2! x_3!} P_1^{x_1} P_2^{x_2} P_3^{x_3};$$

Note that  $x_3=n-x_1-x_2$  and  $\sum x_i=n$ 

 $P(X_1 = X_1, X_2 = X_2) = \frac{n!}{x_1! x_2! (n-x_1-x_2)!} P(X_1 = X_1, X_2 = X_2)$ 

 $x_1, x_2 \ge 0$  $x_1 + x_2 \le n$ 

 $(X_1, X_2)$  in said to follow a trinomial dist<sup>n</sup>  $(n, \theta_1, \theta_2)$ Marginal dist<sup>n</sup>s:

Marginal p.m. f. of X1: n-x1

$$= \begin{pmatrix} x^{1} \\ y \end{pmatrix} \theta_{x_{1}}^{1} \left(1 - \theta^{1} - \theta^{2} + \theta^{2}\right)_{x_{1} - x_{1}}$$

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$$= \begin{pmatrix} x^{1} \\ y \end{pmatrix} \theta_{x_{1} - x_{1}} \left(1 - \theta^{1} - \theta^{2}\right)_{x_{1} - x_{1}}$$

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$$= \begin{pmatrix} x^{1} \\ y \end{pmatrix} \theta_{x_{1}}^{1} \left(1 - \theta^$$

i.e. X, ~ Bin (n, 0,) Sly X, ~ Bin (n, 0,)