

M378K: February 13th, 2026.

Random Vectors.

Say, we are interested in two (or more) r.v.s as a
PAIR (or VECTOR), i.e.,

we look at (Y_1, Y_2)

Then, it's not sufficient to only look @ their
"individual" dist^{'ns}, but also their association.

Example. $Y_i \dots$ cointoss for $i=1,2$ of fair coins

independence.

$\{Y_1=H, Y_2=H\} \{Y_1=T, Y_2=H\}$

$\{Y_1=H, Y_2=T\} \{Y_1=T, Y_2=T\}$

complete dependence

$\{Y_1=H, Y_2=H\}$ X

X

$\{Y_1=T, Y_2=T\}$

Discrete 2D Environment.

The Joint Distribution Table.

$X \backslash Y$	y_1	y_2	\dots	y_j	\dots	y_ℓ
x_1						
x_2						
\vdots						
x_i				p_{ij}		
\vdots						
x_m						

$$p_X(x_1) = \sum_{j=1}^{\ell} p_{1j}$$

\vdots

$$p_X(x_i) = \sum_{j=1}^{\ell} p_{ij}$$

\vdots

The
MARGI-
NAL
dist'n
of
X

The Marginal Dist'n
of Y

$p_{ij} = \mathbb{P}[X=x_i, Y=y_j]$, i.e., the joint pmf
for all i, j

X and Y are independent iff

$$p_{ij} = p_X(x_i) \cdot p_Y(y_j) \text{ for } i, j$$

Example. We independently throw two fair dice and record the results as Y_1 and Y_2

joint pmf $p_{ij} = \frac{1}{36}$ for all $1 \leq i, j \leq 6$

Define $Z = Y_1 + Y_2$

Q: What is the joint dist'n table for (Y_1, Z) ?

$Y_1 \backslash Z$	2	3	4	5	6	7	8	9	10	11	12
1	0	0	0	0	0	0	0	0	0	0	0
2	0	0	0	0	0	0	0	0	0	0	0
3	0	0	0	0	0	0	0	0	0	0	0
4	0	0	0	0	0	0	0	0	0	0	0
5	0	0	0	0	0	0	0	0	0	0	0
6	0	0	0	0	0	0	0	0	0	0	0
	$\frac{1}{36}$	$\frac{2}{36}$	$\frac{3}{36}$	$\frac{4}{36}$	$\frac{5}{36}$	$\frac{6}{36}$	$\frac{5}{36}$	$\frac{4}{36}$	$\frac{3}{36}$	$\frac{2}{36}$	$\frac{1}{36}$

Joint Dist's: The Continuous Case

Recall: For a continuous r.v. Y w/ a pdf f_Y , we can calculate probabilities as

$$\begin{aligned} \mathbb{P}[Y \in [a, b]] &= \mathbb{P}[a \leq Y \leq b] \\ &= \int_a^b f_Y(y) dy \quad \text{for all } a \leq b \end{aligned}$$

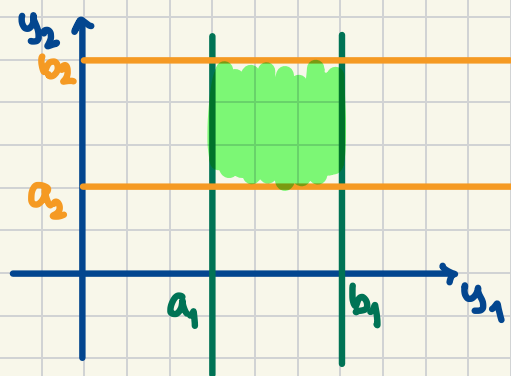
In multiple dimensions.

Say the random vector (Y_1, Y_2, \dots, Y_n) is jointly continuous w/ density

$$f_{Y_1, \dots, Y_n}$$

Then,

$$\begin{aligned} \mathbb{P}[Y_1 \in [a_1, b_1], Y_2 \in [a_2, b_2], \dots, Y_n \in [a_n, b_n]] &= \\ &= \int_{a_1}^{b_1} \int_{a_2}^{b_2} \dots \int_{a_n}^{b_n} \underline{f_{Y_1, \dots, Y_n}(y_1, \dots, y_n)} dy_n \dots dy_2 dy_1 \end{aligned}$$



For any "nice" region $A \subseteq \mathbb{R}^n$,

$$\mathbb{P}[(Y_1, \dots, Y_n) \in A] = \underbrace{\int \int \dots \int}_A f_{Y_1, \dots, Y_n}(y_1, \dots, y_n) dy_n dy_{n-1} \dots dy_1$$

Example. $(Y_1, Y_2) \dots$ represents a point chosen
a random in a unit square $[0, 1]^2$

$$f_{Y_1, Y_2}(y_1, y_2) = \underset{?}{C} \cdot \mathbb{1}_{[0, 1] \times [0, 1]}(y_1, y_2)$$

$$f_{Y_1, Y_2}(y_1, y_2) = 1 \cdot \mathbb{1}_{[0, 1] \times [0, 1]}(y_1, y_2)$$

$$\mathbb{P}[Y_1 > Y_2] = \cancel{\frac{1}{2}}$$

