

Sample Point	Number of heads	Number of runs
HHH	3	1
HHT	2	2
HTH	2	3
HTT	1	2
THH	2	2
THT	1	3
TTH	1	2
TTT	0	1

$$X \in \{0, 1, 2, 3\}$$

$$Y \in \{1, 2, 3\}$$

		Number of heads				
		*	0	1	2	3
Number of runs	1	$\frac{1}{8}$	0	0	$\frac{1}{8}$	
	2	0	$\frac{2}{8}$	$\frac{2}{8}$	0	
	3	0	$\frac{1}{8}$	$\frac{1}{8}$	0	
$P(X=*)$		$\frac{1}{8}$	$\frac{3}{8}$	$\frac{3}{8}$	$\frac{1}{8}$	

Def

Let X and Y be r.v. with joint density f_{xy} and marginal densities f_x , f_y . X and Y are independent if $f_{xy}(x,y) = f_x(x)f_y(y)$ for all x and y .

Ex

$$\begin{aligned} f_{xy}(3,3) &= 0 \\ f_x(3) &= \frac{1}{8} \\ f_y(3) &= \frac{2}{8} \end{aligned} \quad \left. \begin{aligned} f_{xy}(3,3) &\neq f_x(3)f_y(3) \end{aligned} \right.$$

Note

If we variate the probability we can make f_{xy} independent.

Def

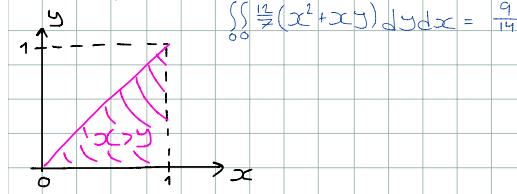
Let X and Y be continuous r.v. The ordered pair (X, Y) is called a two-dimensional, continuous r.v. A function f_{xy} such that:

1. $f_{xy}(x,y) \geq 0$, all x, y
2. $\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x,y) dx dy = 1$
3. $P(a \leq X \leq b \text{ and } c \leq Y \leq d) = \int_a^b \int_c^d f(x,y) dx dy$

EK

$$f(x,y) = \frac{12}{7}(x^2 + xy), 0 \leq x, y \leq 1$$

$$P(X > Y) ?$$



$$\int_0^1$$

$$\int_0^x \int_0^y \frac{12}{7}(x^2 + xy) dy dx = \frac{9}{14}$$

Def

Let (X, Y) be a two-dimensional r.v. with joint density f_{xy} . H is a function then

$$E[H(X, Y)] = \begin{cases} \sum_{\text{all } x, y} H(x, y) f(x, y) & \text{discrete} \\ \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} H(x, y) f(x, y) dx dy & \text{continuous} \end{cases}$$

Remark

$$1. H(x, y) = xy$$

$$2. \text{If } X, Y \text{ are independent, } E[XY] = E[X] \cdot E[Y]$$

Def: Covariance

Let X and Y be r.v with means $E[X]$ and $E[Y]$. The covariance between X and Y , denoted $\text{Cov}(X, Y)$ or Cov_{XY} , is given by:

$$\text{Cov}(X, Y) = E[(X - E[X])(Y - E[Y])]$$

Proposition

$$1. \text{Var}(X+Y) = \text{Var}(X) + \text{Var}(Y) + 2\text{Cov}(X, Y)$$

Def

Let X and Y be r.v with expectations $E[X]$ and $E[Y]$, and variances $\text{Var}[X]$ and $\text{Var}[Y]$. The correlation, p_{XY} , between X and Y is given by:

$$-1 \leq p_{XY} = \frac{\text{Cov}(X, Y)}{\sqrt{\text{Var}[X]\text{Var}[Y]}} \leq 1$$

Remark

$$\text{If } X = Y \quad p_{XY} = \frac{\text{Cov}(X, X)}{\sqrt{\text{Var}[X]\text{Var}[X]}} = \frac{\text{Var}[X]}{\text{Var}[X]} = 1$$

$$\text{If } X = -Y \quad p_{XY} = \frac{\text{Cov}(X, -X)}{\sqrt{\text{Var}[X]\text{Var}[X]}} = \frac{-\text{Var}[X]}{\text{Var}[X]} = -1$$

Note

Correlation measures only linear dependence.

Prelims

1. Systems which changes states in discrete time
2. The collection of all possible states $I = \{i_0, \dots, i_n\}$ is called state space
3. An initial distribution $\lambda = \{\lambda_{i_0}, \dots, \lambda_{i_n}\}$ defined on I , specifies the starting state.
4. The random mechanism is described by transition matrix P
Entry P_{ij} gives the probability that the system will change state from i to j in a unit of time.
5. Each entry in P is non-negative and not greater than one. The sum of entries on each row equals to one.
A matrix with these properties is called stochastic.

Def

A sequence of random variables X_n with values in a finite set I is a discrete-time Markov chain (DTMC). With initial distribution λ and transition matrix P if for all $i_0, \dots, i_n \in I$, the joint probability $P(X_0 = i_0, \dots, X_n = i_n) = \lambda_{i_0} P_{i_0 i_1} \cdots P_{i_n i_n}$.

Def

A state i of MC is called absorbing if it is impossible to leave it ($P_{ii} = 1$). A MC is called absorbing if it has at least one absorbing state and it is possible to reach that state from all other states. It is not necessary to be able to go directly there.

Ex



Theorem

$$1. P(X_n = j) = (\lambda P^n)_{j,j}, P^n - n\text{-th power of matrix } P$$

$$2. P_{ij}^{(n)} = P(X_{k+n} = j | X_k = i) \quad n\text{-step transition probability}$$

Ex

$$P(X_{n+1} = j | X_0 = i_0, \dots, X_{n-1} = i_{n-1}, X_n = i) = \frac{P(X_0 = i_0, \dots, X_n = i, X_{n+1} = j)}{P(X_0 = i_0, \dots, X_n = i)} = \frac{\lambda_{i_0} P_{i_0 i_1} \cdots P_{i_n i}}{\lambda_{i_0} P_{i_0 i_1} \cdots P_{i_n i}} = P_{ij}$$