

Def

Let X be a continuous r.v with density f . Then expected value of X is $E[X] = \int_{-\infty}^{+\infty} x f(x) dx$

Ex

$X \sim \text{Uni}[a, b]$, $f(x) = \begin{cases} \frac{1}{b-a}, & a \leq x \leq b \\ 0, & \text{otherwise} \end{cases}$

$$E[X] = \int_a^b \frac{x}{b-a} dx = \frac{x^2}{2(b-a)} \Big|_a^b = \frac{b+a}{2}$$

Def

Let X be a continuous r.v with density f . Then expected value of r.v. $H(x)$, denoted by $E[H(X)]$ is given by $E[H(X)] = \int_{-\infty}^{+\infty} H(x) f(x) dx$

Ex

$H(x) = x^2$, $X \sim \text{Uni}[a, b]$

$$E[X^2] = \int_a^b x^2 \frac{1}{b-a} dx = \frac{x^3}{3(b-a)} \Big|_a^b = \frac{a^3 + ab^2 + b^3}{3}$$

Def

Let X be a continuous r.v with density f and mean $E[X]$

$$\text{Var}[X] = E[X^2] - E[X]^2$$

Ex

$$\text{Var}[X] = E[X^2] - E[X]^2 = \frac{a^3 + ab^2 + b^3}{3} - \left(\frac{a+b}{2}\right)^2 = \frac{a^4 - 2ab^2 + b^4}{12} = \frac{(a-b)^2}{12} = \frac{(b-a)^2}{12}$$

Def

A r.v X with density $f(x) = \frac{1}{\sqrt{2\pi}\sigma^2} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$, where $-\infty < \mu < +\infty$, $\sigma^2 > 0$ have a normal distribution with parameters μ and σ^2 .

Remark

$X \sim N(\mu, \sigma^2)$

$E[X] = \mu$

$\text{Var}[X] = \sigma^2$

Pic



Expectations: I: 0
II: 0
III: $\mu_3 > 0$

Variations: $\sigma_{II}^2 > \sigma_I^2$
 $\sigma_{III}^2 = \sigma_{II}^2$

$$X_{III} = X_I + \mu_3$$

Def

A r.v with parameters $\mu=0, \sigma^2=1$ is denoted by Z and is called Standard Normal Random Variable.

Proposition

Let $X \sim N(\mu, \sigma^2)$. The variable $\frac{X-\mu}{\sqrt{\sigma^2}}$ is standard normal.

Theorem

Let $X \sim \text{Bin}(n, p)$ $P(X=k) = \binom{n}{k} p^k (1-p)^{n-k}$, $k=0,1,\dots,n$

For large values of n , X is approximately normal with mean np and variance $\underbrace{np(1-p)}_q$

Ex

Rolling a dice 108 times. X = number of ones after rolling

$$X \sim \text{Bin}(108, \frac{1}{6}), \quad E[X] = 108 \cdot \frac{1}{6} = 18, \quad \text{Var}[X] = 108 \cdot \frac{1}{6} \left(1 - \frac{1}{6}\right) = 15$$

1. Exact calculations

$$P(12 \leq X \leq 20) = \sum_{i=12}^{20} P(X=i) = \sum_{i=12}^{20} \binom{108}{i} \left(\frac{1}{6}\right)^i \left(\frac{5}{6}\right)^{108-i}$$

2. Normal approximation, for $n \geq 25$

$$P(12 \leq X \leq 20) \approx P(11.5 \leq X \leq 20.5) = P\left(\frac{11.5-18}{\sqrt{15}} \leq \frac{X-18}{\sqrt{15}} \leq \frac{20.5-18}{\sqrt{15}}\right) \stackrel{\text{approx } N(0,1)}{=} F(0.645) - F(-1.678) = 0.691$$

where F is a c.d.f. of $N(0,1)$

Reminder

$$\Phi(x) = P(Z \leq x) = \int_{-\infty}^x \frac{1}{\sqrt{2\pi}} e^{-\frac{u^2}{2}} du$$

↑ c.d.f. for Z

Kika i VLE-Tables

Def

Let X and Y be discrete r.v.s, defined on the same sample space. The ordered pair (X, Y) is called a two-dimensional (bivariate) discrete random variable.

A function $f_{X,Y}(x, y) = P(X=x, Y=y)$ is called the joint density for (X, Y) .

Properties

- $f(x, y) \geq 0$, all x, y
- $f(x, y) > 0$ only in a countable number of points.
- $\sum_{\text{all } x} \sum_{\text{all } y} f(x, y) = 1$

EX

Three coin tossings with a fair coin, i.e. $P(\text{Heads}) = P(\text{Tails}) = \frac{1}{2}$

X = {number of heads}

"run"

Y = {number of runs}

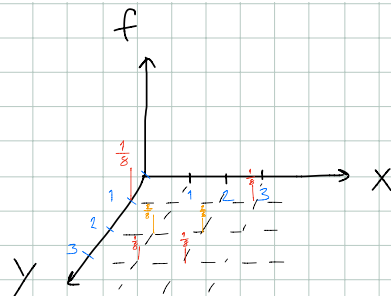
maximal number of consecutive coin flips that are the same

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	$\frac{1}{8}$
	2	0	$\frac{2}{8}$	$\frac{2}{8}$
	3	0	$\frac{1}{8}$	$\frac{1}{8}$



Def

Let X and Y be a bivariate r.v with joint density $f_{X,Y}$. The marginal density for X is given by $f_X(x) = \sum_y f(x, y)$

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