

eg Toss a fair coin 100 times
 Estimate the probability that the number
 of heads obtained is not strictly between 40 and 60

Solution (3 ways)

X_1, X_2, \dots, X_{100} indep

$$P(X_i=1) = \frac{1}{2} = P(X_i=0)$$

\uparrow
heads

\uparrow
tails

$$S_n = X_1 + \dots + X_n$$

$$E(S_n) = \frac{n}{2}. \quad \text{Var}(S_n) = \frac{n}{4}, \quad \sigma_{S_n} = \frac{\sqrt{n}}{2}.$$

Specialize to $n=100$.

$$P(|S_{100} - 50| \geq 10) = P((S_{100} - 50)^2 \geq 100)$$

$$\begin{aligned} & \leq \frac{1}{100} E((S_{100} - 50)^2) \\ & \stackrel{\substack{\text{markov's} \\ \text{inequality}}}{\curvearrowright} = \frac{1}{100} \text{Var}(S_{100}) \\ & = \frac{1}{100} \cdot \frac{100}{4} = \frac{1}{4}. \end{aligned}$$

We used Chebyshev's inequality:

$$P(|Z - E(Z)| \geq \varepsilon) \leq \frac{1}{\varepsilon^2} \text{Var}(Z).$$

(special case of Markov's inequality).

What if we try Bernstein's inequality?

$$\begin{aligned} P(|S_{100} - 50| \geq 10) &= P\left(\left|\frac{1}{100} S_{100} - \frac{1}{2}\right| \geq \frac{1}{10}\right) \\ &\leq 2e^{-n\delta^2/4} && \begin{array}{l} n=100 \\ \delta=10 \end{array} \\ &= 2e^{-1/4} \\ &= 1.557... > 1 \end{aligned}$$

(useless!!)

Let's try the Central Limit Theorem

$$P(|S_{100} - 50| \geq 10) = P(|S_{100}^*| \geq 2)$$

$$\begin{aligned} &\approx 1 - \int_{-2}^2 \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx \\ &= 2 \int_{-\infty}^{-2} \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx \end{aligned}$$

$$\approx 2(0.0228) = 0.0456$$

The Poisson Distribution

$$\lambda \in [0, \infty) \quad (\lambda = \text{"intensity"})$$

$$\pi_k(\lambda) = \frac{\lambda^k e^{-\lambda}}{k!} \quad \text{for } k=0, 1, 2, \dots$$

$\pi(\lambda)$ is a prob. dist on the non-negative integers because $\pi_k(\lambda) \geq 0$ and

$$\sum_{k=0}^{\infty} \frac{\lambda^k e^{-\lambda}}{k!} = e^{-\lambda} \sum_{k=0}^{\infty} \frac{\lambda^k}{k!} = e^{-\lambda} e^{\lambda} = 1.$$

(suppose
 $P(X=k) = \pi_k(\lambda)$
 $\forall k \in \mathbb{N} \cup \{0\}$)

$\pi(\lambda)$ has mean λ and variance λ because

$$\hookrightarrow E(X) = \sum_{k=0}^{\infty} k \frac{\lambda^k e^{-\lambda}}{k!} = e^{-\lambda} \lambda \sum_{k=1}^{\infty} \frac{\lambda^{k-1}}{(k-1)!} = e^{-\lambda} \lambda e^{\lambda} = \lambda$$

$$\text{And } \sum_{k=0}^{\infty} k^2 \frac{\lambda^k e^{-\lambda}}{k!} = e^{-\lambda} \lambda \sum_{k=1}^{\infty} k \frac{\lambda^{k-1}}{(k-1)!} = e^{-\lambda} \lambda \left(\frac{d}{dx} x e^x \right) \Big|_{x=\lambda}$$

$$= e^{-\lambda} \lambda (e^{\lambda} + \lambda e^{\lambda}) = \lambda + \lambda^2$$

$$\text{So } \text{Var}(X) = E(X^2) - E(X)^2 = \lambda + \lambda^2 - \lambda^2 = \lambda.$$

$$\begin{aligned} \text{Also, } E(X(X-1)) &= \sum_{k=0}^{\infty} k(k-1) \pi_k(\lambda) \\ &= \sum_{k=2}^{\infty} \frac{\lambda^k e^{-\lambda}}{(k-2)!} \\ &= \lambda^2 e^{-\lambda} e^{\lambda} \\ &= \lambda^2 \end{aligned}$$

$$\text{So } E(X^2) - E(X) = \lambda^2$$

$$\text{So } E(X^2) = \lambda^2 + \lambda$$

$$\text{So } \text{Var}(X) = E(X^2) - E(X)^2 = \lambda^2 + \lambda - \lambda^2 = \lambda.$$

Recall $B_k(n, p) = \binom{n}{k} p^k q^{n-k}$ for $k=0, 1, 2, \dots, n$,

where $n \in \{0, 1, 2, \dots\}$, $p \in [0, 1]$, and $q = 1 - p$.

Theorem Let (p_n) be a sequence in $[0, 1]$
 such that $np_n \rightarrow \lambda$, where $\lambda \in [0, \infty)$.

Then for each k , $B_k(n, p_n) \xrightarrow[n \rightarrow \infty]{\omega} \pi_k(\lambda)$

$$\begin{aligned}
 \text{pf } B_k(n, p_n) &= \binom{n}{k} p_n^k q_n^{n-k} \\
 &= \frac{n!}{k!(n-k)!} p_n^k (1-p_n)^{n-k} \\
 &= \frac{1}{k!} n(n-1)\cdots(n-k+1) p_n^k (1-p_n)^{n-k} \\
 &= \frac{(np_n)^k}{k!} \frac{\frac{n}{n} \left(\frac{n-1}{n}\right) \left(\frac{n-2}{n}\right) \cdots \left(\frac{n-k+1}{n}\right)}{(1-p_n)^k} (1-p_n)^n.
 \end{aligned}$$

$$\begin{aligned}
 \text{Now } (1-p_n)^n &= \left(1 - \frac{np_n}{n}\right)^n = e^{n \log(1-p_n)} \\
 &= \exp\left(n \left(-p_n + \frac{p_n^2}{2} - \frac{p_n^3}{3} + \dots\right)\right) \\
 &= \exp\left(\underbrace{-np_n}_{-\lambda} \left(\underbrace{1 - \frac{p_n}{2} + \frac{p_n^2}{3} - \dots}_1\right)\right) \\
 &\rightarrow e^{-\lambda}
 \end{aligned}$$

$$\text{Thus } B_k(n, p_n) \rightarrow \frac{\lambda^k}{k!} \cdot 1 \cdot e^{-\lambda} = \pi_k(\lambda).$$

eg Cosmic Rays

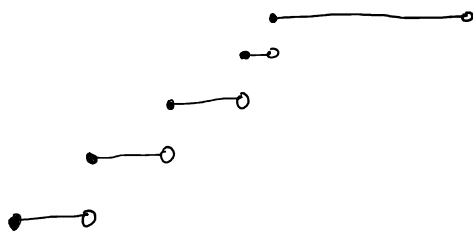


P_n = probability that a cosmic ray arrives
in an interval $[\frac{k}{n}, \frac{k+1}{n})$.

of intervals (of length $\frac{1}{n}$) in which
a cosmic ray arrives $\sim B(n, P_n)$.

The Poisson Process

Defn A simple counting process is a family $(N_t)_{0 \leq t < \infty}$
of RVs such that for each $\omega \in \Omega$,
There is a set $J(\omega) \subseteq (0, \infty)$ such that
 $\forall t \in [0, \infty), N_t(\omega) = |J(\omega) \cap [0, t]| < \infty$.



The paths of a simple counting process
are increasing and right cts, with jumps

of size 1, and they only change by jumping.

Defn A poisson process is a simple counting process $(N_t)_{0 \leq t < \infty}$ such that

(A) (N_t) has independent increments;

i.e. for each n , for all $0 \leq t_0 < t_1 < \dots < t_n < \infty$,

$$N_{t_1} - N_{t_0}, N_{t_2} - N_{t_1}, \dots, N_{t_n} - N_{t_{n-1}}$$

are independent.

(B) (N_t) has stationary increments;

i.e. for all $s, t, u, v \in [0, \infty)$, if $t - s = v - u$

$$\text{Then } \text{law}(N_t - N_s) = \text{law}(N_v - N_u).$$

Remark let X be an RV in $(\mathbb{X}, \mathcal{A})$.

Define μ on \mathcal{A} by $\mu(A) = P(X \in A) = P(X^{-1}[A])$.

Then $\text{law}(X) := \mu$.

a.k.a. \uparrow the distribution of X
or the law of X . \uparrow a probability measure on \mathcal{A} .

let $f: \mathbb{X} \rightarrow [0, \infty]$ be mble.

$\int f \, d\mu = \int f(x) \, dP_x$

$$\text{Then } E[f(X)] = \int_{\mathbb{X}} f \, d\text{law}(X)$$

Let's check this.

$$\begin{aligned} \textcircled{1} \text{ If } f = 1_A, \text{ where } A \in \mathcal{A}, \text{ then } f(X)(\omega) &= f(X(\omega)) \\ &= \begin{cases} 1 & \text{if } X(\omega) \in A \\ 0 & \text{if } X(\omega) \notin A \end{cases} \\ &= 1_{X^{-1}(A)}(\omega). \end{aligned}$$

$$\text{So } E[f(X)] = E[1_{X^{-1}(A)}] = P(X^{-1}(A)) = \text{law}(X)(A) = \int_{\mathbb{X}} 1_A \, d\text{law}(X).$$

② If f is simple, use linearity of the integral...

③ In the general case, let (f_n) be an increasing sequence of nonnegative simple functions on \mathbb{X} s.t. $f_n \rightarrow f$ pointwise. Use MCT.