

① Let $a_n = \frac{1}{\sqrt{n}} \int_0^1 \dots \int_0^1 \sqrt{x_1^2 + \dots + x_n^2} dx_1 dx_2 \dots dx_n = \int_0^1 \dots \int_0^1 \sqrt{\frac{x_1^2 + \dots + x_n^2}{n}} dx_1 dx_2 \dots dx_n$.

Then $a_n = \mathbb{E} \left[\sqrt{\frac{X_1^2 + \dots + X_n^2}{n}} \right]$ where X_1, \dots, X_n are independent $\sim \text{Uni}[0,1]$.

By the law of large numbers, $\frac{X_1^2 + \dots + X_n^2}{n} \Rightarrow \mathbb{E} X_1^2 = \int_0^1 x^2 dx = \frac{1}{3}$.

The function $f: [0,1] \rightarrow \mathbb{R}$ $f(x) = \sqrt{x}$ is bounded and continuous,

so $a_n = \mathbb{E} f\left(\frac{1}{n}\right) \rightarrow \mathbb{E} f\left(\frac{1}{3}\right) = \sqrt{\frac{1}{3}}$.

② i) Thm a) If X_n has char. fn Ψ_n , X has char. fn Ψ and $X_n \Rightarrow X$, then $\Psi_n(t) \rightarrow \Psi(t)$ for every $t \in \mathbb{R}$.

b) If X_n has char. fn Ψ_n and $\Psi(t) := \lim_{n \rightarrow \infty} \Psi_n(t)$ exists for every $t \in \mathbb{R}$, AND $\Psi = \Psi(t)$ is continuous at 0, ~~then~~ then \exists a random variable X s.t. $X_n \Rightarrow X$ and X has char. fn Ψ .

2) Thm (CLT): If X_1, \dots, X_n, \dots are iid with $\mathbb{E} X_i = m \in \mathbb{R}$, $\text{Var} X_i = \sigma^2 < \infty$, then $\frac{X_1 + \dots + X_n - nm}{\sqrt{n}\sigma} \Rightarrow N(0,1)$.

Proof: Assume w.l.o.g that $m=0$. Let $S_n = X_1 + \dots + X_n$.

Then $\Psi_{X_1}(t) = 1 + i \mathbb{E} X_1 t + \frac{1}{2} i^2 \mathbb{E}(X_1^2) t^2 + o(t^2) = 1 - \frac{\sigma^2}{2} t^2 + o(t^2)$

so $\Psi_{S_n}(t) = (\Psi_{X_1}(t))^n = \left(1 - \frac{\sigma^2}{2} t^2 + o(t^2)\right)^n$

so $\Psi_{\frac{S_n}{\sqrt{n}\sigma}}(t) = \Psi_{S_n}\left(\frac{t}{\sqrt{n}\sigma}\right) = \left[1 + \frac{-t^2/2}{n} + o\left(\frac{1}{n}\right)\right]^n \xrightarrow{n \rightarrow \infty} e^{-\frac{t^2}{2}}$, which

is exactly the char. fn of $N(0,1)$, so the continuity thm gives the statement. \square

③ a.) $E X_n^2 < \infty$ by assumption.

~~1)~~ $\sigma(X_n^2) \subset \sigma(X_n)$, so $X_n^2 \in \mathcal{F}_n$ because $X_n \in \mathcal{F}_n$

$$E(X_{n+1}^2 | \mathcal{F}_n) \underset{\text{Fensen's inequality}}{\geq} \left[E(X_{n+1} | \mathcal{F}_n) \right]^2 \underset{\substack{X_n \text{ is a} \\ \text{martingale}}}{=} X_n^2 \quad \square$$

b.) Let $X_n \equiv -\frac{1}{n}$. Then clearly $E(X_{n+1} | X_n) = \frac{1}{n+1} > -\frac{1}{n} = X_n$.

But $X_n^2 = \frac{1}{n^2}$ is decreasing.

④ Let X_n be the position of the frog after n steps. almost surely
Then X_n is a nonnegative martingale \implies it is convergent

by the martingale convergence theorem.

Since it is integer valued, it can only be convergent by being eventually constant.

However, the frog can not get stuck at any $k \neq 0$, because

$$\sum_{n \geq n_0} P(\text{it jumps away}) = \sum_{n \geq n_0} \frac{k}{k+1} \sum_{n \geq n_0} \frac{1}{n} = \infty \quad (\text{for } k \neq 0),$$

so the Borel Cantelli Lemma ensures that it would jump away.

Summary - The frog does get stuck, but it can only get stuck at 0 so $\boxed{P(\text{reaching } 0) = 1.}$

5) Let S_1, S_2, S_3, \dots be independent, $P(S_i = -1) = \frac{2}{3} =: q$, $P(S_i = +1) = \frac{1}{3} =: p$.

Let $S_0 = 0$; $S_n = S_1 + \dots + S_n$. (S_n is the simple asymmetric random walk, not trapped).

Then $M_n := \left(\frac{q}{p}\right)^{S_n} = 2^{S_n}$ is a martingale adapted to the natural filtration $\mathcal{F}_n := \sigma(S_1, \dots, S_n)$:

- Clearly M_n is bounded for every n (takes only finitely many values) so $E|M_n| < \infty$
- adapted by construction

$$\begin{aligned} E(M_{n+1} | \mathcal{F}_n) &= E\left(\left(\frac{q}{p}\right)^{S_n + S_{n+1}} | \mathcal{F}_n\right) = E\left(M_n \left(\frac{q}{p}\right)^{S_{n+1}} | \mathcal{F}_n\right) \\ &= \underbrace{M_n}_{M_n \in \mathcal{F}_n} E\left(\left(\frac{q}{p}\right)^{S_{n+1}} | \mathcal{F}_n\right) \stackrel{S_{n+1} \text{ is independent of } \mathcal{F}_n}{=} M_n E\left(\left(\frac{q}{p}\right)^{S_{n+1}}\right) \\ &= M_n \left[q \left(\frac{q}{p}\right)^{-1} + p \left(\frac{q}{p}\right)^{+1} \right] = M_n \left[q \frac{p}{q} + p \frac{q}{p} \right] = M_n \quad \checkmark \end{aligned}$$

Let τ be the 1st hitting time of $\{-10, 10\}$ by S_n :

$\tau = \inf\{n | S_n \in \{-10, 10\}\}$. Then τ is a stopping time,

so $M_{n \wedge \tau}$ is also a martingale, which is also bounded:

$M_{n \wedge \tau} = 2^{S_{n \wedge \tau}}$, so $2^{-10} \leq M_{n \wedge \tau} < 2^{10}$. The martingale

convergence thm says that $M_{n \wedge \tau}$ is convergent, so $P(\tau < \infty) = 1$.

Let $A = \{S_\tau = -10\}$; $B = \{S_\tau = +10\}$. We saw that $P(A) + P(B) = 1$,

so ~~$E S_\tau = -10P(A) + 10P(B)$~~ . $E M_\tau = P(A) 2^{-10} + P(B) 2^{10}$.

The optional stopping thm says that $E M_\tau = E M_0 = 1$, so

~~$P(A) + P(B) = 1$
 $2^{-10}P(A) + 2^{10}P(B) = 1$
 $P(A) = \frac{1 - 2^{20}P(B)}{2^{10} - 2^{-10}}$
 $P(B) = \frac{1 - 2^{-20}P(A)}{2^{10} - 2^{-10}}$~~

(5) continued

$$\text{So } \begin{cases} P(A) + P(B) = 1 & \times 2^{-10} \\ 2^{-10} P(A) + 2^{10} P(B) = 1 & \text{copy} \end{cases} \Rightarrow \begin{cases} 2^{-10} P(A) + 2^{-10} P(B) = 2^{-10} \\ 2^{-10} P(A) + 2^{10} P(B) = 1 \end{cases}$$

$$\text{Subtract } P(B)[2^{-10} - 2^{10}] = 2^{-10} - 1$$

$$\Rightarrow \underline{\underline{P(B) = \frac{1 - 2^{-10}}{2^{10} - 2^{-10}} = \frac{2^{10} - 1}{2^{20} - 1} = \frac{2^{10} - 1}{(2^{10} - 1)(2^{10} + 1)} = \frac{1}{1025}}}$$