

Lecture 3

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Scribe: Aya Vituri

In this lecture we prove the law of iterated logarithm; First we prove the following lemma:

- (i) (LCLT - local CLT) If $k = o(n^{3/4})$ and $n + k$ is even, then $P(S_n = k) \underset{n \rightarrow \infty}{\sim} \sqrt{\frac{2}{\pi n}} e^{-\frac{k^2}{2n}}$.
- (ii) For all $n, k \geq 0$, $P(S_n \geq k) \leq e^{-\frac{k^2}{2n}}$ (not tight, by a polynomial factor).

Then by using the Borel-Cantelli lemma we show that $\limsup_{n \rightarrow \infty} \frac{S_n}{\sqrt{2n \log \log n}} = 1$ a.s.,

and by symmetry, $\liminf_{n \rightarrow \infty} \frac{S_n}{\sqrt{2n \log \log n}} = -1$ a.s.

We discuss Higher dimensional RW and general 1D RW;

We prove the Hewitt-Savage 0-1 law: If $A \in \varepsilon$ (for a RW) then $P(A) \in \{0, 1\}$,

and its following application: For a RW in \mathbb{R} , exactly one of the following has probability 1:

- (i) $S_n = 0$ for all n (trivial RW)
- (ii) $S_n \rightarrow \infty$
- (iii) $S_n \rightarrow -\infty$
- (iv) $\limsup S_n = \infty$ and $\liminf S_n = -\infty$

1 Law of the iterated logarithm

For SRW, we know by CLT that $S_n \approx N(0, n)$.

Is it true that $P(S_n = k) \approx \frac{1}{\sqrt{2\pi n}} e^{-\frac{k^2}{2n}}$ (density at k of $N(0, n)$)?

For parity reasons, this is false; $P(S_n = k) = 0$ if $k + n$ is odd.

Is it true that $P(S_n = k) \approx \frac{2}{\sqrt{2\pi n}} e^{-\frac{k^2}{2n}}$ when $k + n$ is even (and n is large)?

Yes, this is true.

$$P(S_n = k) \sim \sqrt{\frac{2}{\pi n \left(1 - \frac{k^2}{n^2}\right)}} e^{-\frac{k^2}{2n} + O\left(\frac{k^4}{n^3}\right)} \sim \sqrt{\frac{2}{\pi n}} e^{-\frac{k^2}{2n}} \quad \left(\begin{array}{l} n \rightarrow \infty \\ k = o(n^{3/4}) \end{array} \right)$$

(ii) We will use that for any X ,

$$P(X \geq t) \underset{\text{assuming } \theta > 0}{=} P\left(e^{\theta X} \geq e^{\theta t}\right) \underset{\text{Markov}}{\leq} \frac{E e^{\theta X}}{e^{\theta t}}$$

Taking $X = S_n$, $t = k$ and by comparing Taylor coefficients,

$$\begin{aligned} E e^{\theta X_1} &= \left(\frac{1}{2}e^{\theta} + \frac{1}{2}e^{-\theta}\right) \leq e^{\theta^2/2} \\ P(S_n \geq k) &\underset{\theta > 0}{\leq} \frac{E e^{\theta S_n}}{e^{\theta k}} = \frac{\left(\frac{1}{2}e^{\theta} + \frac{1}{2}e^{-\theta}\right)^n}{e^{\theta k}} \leq \frac{e^{\theta^2 \frac{n}{2}}}{e^{\theta k}} \underset{\text{take } \theta = \frac{k}{n}}{=} e^{-\frac{k^2}{2n}} \quad \square \end{aligned}$$

Lemma. For $k > \sqrt{n}$, $k = o(n^{3/4})$ we have $P(S_n \geq k) \geq c \frac{\sqrt{n}}{k} e^{-k^2/2n}$ for some $c > 0$.

Remark: If $|k| \leq \sqrt{n}$, then $P(S_n \geq k) \geq c$ for some $c > 0$ by the CLT.

Proof.

$$P(S_n \geq k) \geq P\left(k \leq S_n \leq k + \frac{n}{k}\right) \underset{(i)}{\geq} \sqrt{\frac{c}{n}} \sum_{k \leq j \leq \lfloor k + \frac{n}{k} \rfloor} e^{-j^2/2n}$$

For these j we have

$$e^{-j^2/2n} \geq e^{-(k + \frac{n}{k})^2/2n} \geq e^{-\frac{k^2}{2n} - 1 - \frac{n}{2k^2}} \underset{k \geq \sqrt{n}}{\geq} c' e^{-k^2/2n}$$

Thus,

$$P(S_n \geq k) \geq \frac{c''}{\sqrt{n}} e^{-k^2/2n} \cdot \frac{n}{k} \quad \square$$

Reminder: Define $M_n = \max_{0 \leq j \leq n} S_j$. Then $P(M_n \geq k) = P(S_n = k) + 2P(S_n > k)$.

S_n is typically of order \sqrt{n} . By (ii) and Borel-Cantelli, a.s. $S_n \leq \sqrt{2n \log n} (1 + \epsilon)$ eventually.

Lemma. (Borel-Cantelli)

If $\{A_n\}$ is a sequence of events, then:

1) If $\sum P(A_n) < \infty$, then a.s. only finitely many occur.

2) If $\sum P(A_n) = \infty$ and A_n are independent, then a.s. infinitely many of them occur.

Application: Define $A_n = \{S_n \geq (1 + \epsilon) \sqrt{2n \log n}\}$. By (ii), $P(A_n) \leq \frac{1}{n^{1+\epsilon}}$.

So $\sum P(A_n) < \infty$, and therefore a.s. $S_n \leq (1 + \epsilon) \sqrt{2n \log n}$ eventually.

Theorem. (LIL)

$$\limsup_{n \rightarrow \infty} \frac{S_n}{\sqrt{2n \log \log n}} = 1 \text{ a.s.}$$

Remark: By symmetry, $\liminf_{n \rightarrow \infty} \frac{S_n}{\sqrt{2n \log \log n}} = -1$ a.s.

Notation: $u(n) = \sqrt{2n \log \log n}$

Proof. *Remark:* For ease of readability, we will not worry about integrality of indices in the proof. One may round the indices appropriately everywhere and the argument will still go through.

For $\gamma > 0$, $a > 1$, $k \in \mathbb{N}$

$$\begin{aligned} P\left(\max_{0 \leq j \leq a^k} S_j \geq (1 + \gamma) u(a^k)\right) &\leq 2P\left(S_{a^k} \geq (1 + \gamma) u(a^k)\right) \leq \\ &\stackrel{(ii)}{\leq} 2e^{-\frac{(1+\gamma)^2 u(a^k)^2}{2a^k}} = 2e^{-(1+\gamma)^2 \log \log(a^k)} = \frac{2}{(k \log a)^{(1+\gamma)^2}} \end{aligned}$$

These probability estimates are summable in k . Thus by B-C,

$$P\left(\max_{0 \leq j \leq a^k} S_j \leq (1 + \gamma) u(a^k) \text{ from some } k \text{ on}\right) = 1$$

Now, for large n , write $a^{k-1} \leq n \leq a^k$. Then

$$\frac{S_n}{u(n)} = \frac{S_n}{u(a^k)} \cdot \frac{u(a^k)}{a^k} \cdot \frac{a^k}{n} \cdot \frac{n}{u(n)} \leq a(1 + \gamma)$$

since $\frac{S_n}{u(a^k)} \leq 1 + \gamma$, $\frac{a^k}{n} \leq a$, and $\frac{u(a^k)}{a^k} \cdot \frac{n}{u(n)} \leq 1$ because $\frac{u(t)}{t}$ is eventually decreasing.

We deduce that

$$P\left(\max_{0 \leq j \leq a^k} S_j \leq (1 + \gamma) u(a^k) \text{ from some } k \text{ on}\right) = 1$$

$$P\left(\limsup_n \frac{S_n}{u(n)} \leq a(1 + \gamma)\right) = 1$$

Since $\gamma > 0$ and $a > 1$ are arbitrary, we have

$$P\left(\limsup_n \frac{S_n}{u(n)} \leq 1\right)$$

For the lower bound, fix $0 < \gamma < 1$, $a > 1$.

Denote $A_k = \{S_{a^k} - S_{a^{k-1}} \geq (1 - \gamma) u(a^k - a^{k-1})\}$.

We need to show that $\sum P(A_k) = \infty$.

Denote: $n = a^k - a^{k-1}$.

For large k ,

$$P(A_k) \underset{\text{lemma}}{\geq} c \frac{\sqrt{n}}{(1 - \gamma) u(n)} e^{-\frac{(1 - \gamma)^2 u^2(n)}{2n}} = \frac{c}{1 - \gamma} \cdot \frac{1}{\sqrt{2 \log \log n}} \cdot \frac{1}{(\log n)^{(1 - \gamma)^2}}$$

Noting $\log n \approx k$. So this expression is not summable in k .

Therefore, by B-C,

$$P(\text{infinitely many of } A_k \text{ occur}) = 1$$

By the upper bound,

$$P\left(\liminf_n \frac{S_n}{u(n)} \geq -1\right) = 1$$

So we have, for large k ,

\textrm{is\:\ permutable}

$$\begin{aligned} \frac{S_{a^k}}{u(a^k)} &\geq (1-\gamma) \frac{u(a^k - a^{k-1})}{u(a^k)} + \frac{S_{a^k}}{u(a^k)} \geq \\ &\geq (1-\gamma) \frac{u(a^k - a^{k-1})}{u(a^k)} - \frac{(1+\epsilon)u(a^{k-1})}{u(a^k)} \xrightarrow{k \rightarrow \infty} (1-\gamma) \sqrt{1 - \frac{1}{a}} - \frac{1+\epsilon}{\sqrt{a}} \end{aligned}$$

Thus

$$P \left(\limsup_n \frac{S_n}{u(n)} \geq (1-\gamma) \sqrt{1 - \frac{1}{a}} - \frac{1+\epsilon}{\sqrt{a}} \right) = 1$$

and we take $\gamma \rightarrow 0$ and $a \rightarrow \infty$ to obtain the lower bound. \square

2 Higher dimensional RW and general 1D RW

In 1D and in 2D SRW is recurrent. In 3D (and more D) it is transient.

For a general RW in \mathbb{R}^d ($S_n = X_1 + \dots + X_n$, I.I.D., $X_1 \in \mathbb{R}^d$), what can we say about $P(S_n = 0 \text{ infinitely often})$? It turns out that this probability is 0 or 1.

Say that an event A is *permutable* if the occurrence of A is unaffected by applying a *finite permutation* (a function $\pi : \mathbb{N} \rightarrow \mathbb{N}$ $\begin{matrix} 1-1 \\ \text{onto} \end{matrix}$ s.t. $\pi(n) = n$ for all $n \geq n_0$) to the increments of S_n .

More formally, if the increments take values in a state space S , let our probability space Ω be $S^{\mathbb{N}}$, with the product probability over increments.

An event $A \in \mathcal{F}$ (σ -field) is *permutable* if $A = \{\omega \in \Omega \mid \pi(\omega) \in A\}$ for all finite permutations π .

The collection of all permutable events forms σ -field ε , the *exchangable σ -field*.

Theorem. (*Hewitt-Savage 0-1 law*)

If $A \in \varepsilon$ (for a RW) then $P(A) \in \{0, 1\}$.

More Examples:

1. For any B , $\{S_n \in B \text{ infinitely often}\} \in \varepsilon$.

2. For any C_n , $\left\{ \limsup_{n \rightarrow \infty} \frac{S_n}{C_n} \geq 1 \right\} \in \varepsilon$.
3. Any tail event is permutable (tail σ -field $\subseteq \varepsilon$).

A *tail event* is an event which, for any n , is a function only of X_n, X_{n+1}, \dots .

Proof. Fix $A \in \varepsilon$. We will show that $P(A) = P(A)^2$ (in other words, A , is independent of itself).

Take a sequence of events A_n s.t. A_n is a function only of X_1, \dots, X_n and:

$$(1) \quad P(A_n \Delta A) \rightarrow 0$$

(Xor: $B \Delta C = (B \cup C) \setminus (B \cap C)$)

Let $\pi = \pi_n$ be the permutation which exchanges $1, \dots, n$ with $n+1, \dots, 2n$.

Write $A'_n = \pi(A_n)$. Notice that A_n and A'_n are independent.

Since π preserves probability (since X_1, \dots are I.I.D.),

$$(2) \quad P(A_n \Delta A) = P(\pi(A_n \Delta A)) \underset{A \text{ is permutable}}{=} P(A'_n \Delta A)$$

Noticing that $|P(B) - P(C)| \leq P(B \Delta C)$, then (1) and (2) imply that $P(A_n) \rightarrow P(A)$ and

$$(3) \quad P(A'_n) \rightarrow P(A)$$

However, we also obtain that

$$P(A_n \Delta A'_n) \leq P(A_n \Delta A) + P(A'_n \Delta A) \rightarrow 0$$

from which

$$0 \leq P(A_n) - P(A_n \cap A'_n) \leq P(A_n \cup A'_n) - P(A_n \cap A'_n) = P(A_n \Delta A'_n) \rightarrow 0$$

But $P(A_n) \rightarrow P(A)$, and by independence $P(A_n)P(A'_n) \underset{(3)}{\rightarrow} P(A)^2$. Thus,

$$P(A) = P(A)^2$$

as we wanted. \square

Application: For a RW in \mathbb{R} , exactly one of the following has probability 1:

(i) $S_n = 0$ for all n (trivial RW)

(ii) $S_n \rightarrow \infty$

(iii) $S_n \rightarrow -\infty$

(iv) $\limsup S_n = \infty$ and $\liminf S_n = -\infty$

Proof. By the 0-1 law, $\limsup_{n \rightarrow \infty} S_n$ is a constant $c \in [-\infty, \infty]$ (since it is a permutable RV).

Notice that by considering the first increment X_1 , $c = c - X_1$.

Thus, either $X_1 \equiv 0$ (this is case (i)), or $c \in \{-\infty, \infty\}$.

Similarly, $\liminf S_n \in \{-\infty, \infty\}$. \square

Exercise: If $X_1 \in \mathbb{R}$ is non-degenerate (not always 0: $P(X_1 = 0) < 1$), then we are in case (iv) if either:

1. X_1 is symmetric.
2. $E(X_1) = 0$ and $E(X_1^2) < \infty$. (It is actually true that $E(X_1) = 0$ suffices, as we will see next time.)