

Notes 15 : UI Martingales

Math 733 - Fall 2013

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References: [Wil91, Chapter 13, 14], [Dur10, Section 5.5, 5.6, 5.7].

1 Uniform Integrability

We give a characterization of L^1 convergence. First note:

LEM 15.1 *Let $Y \in L^1$. $\forall \varepsilon > 0$, $\exists K > 0$ s.t.*

$$\mathbb{E}[|Y|; |Y| > K] < \varepsilon.$$

Proof: Immediate by (MON) to $\mathbb{E}[|Y|; |Y| \leq K]$. ■

What we need is for this condition to hold uniformly over the sequence:

DEF 15.2 (Uniform Integrability) *A collection $\mathcal{C}$ of RVs on $(\Omega, \mathcal{F}, \mathbb{P})$ is uniformly integrable (UI) if: $\forall \varepsilon > 0$, $\exists K > +\infty$ s.t.*

$$\mathbb{E}[|X|; |X| > K] < \varepsilon, \quad \forall X \in \mathcal{C}.$$

THM 15.3 (Necessary and Sufficient Condition for L^1 Convergence) *Let $\{X_n\} \in L^1$ and $X \in L^1$. Then $X_n \rightarrow X$ in L^1 if and only if:*

- $X_n \rightarrow X$ in prob
- $\{X_n\}$ is UI.

Before giving the proof, we look at a few examples.

EX 15.4 (L^1 -bddness is not sufficient) *Let $\mathcal{C}$ is UI and $X \in \mathcal{C}$. Note that*

$$\mathbb{E}|X| \leq \mathbb{E}[|X|; |X| \geq K] + \mathbb{E}[|X|; |X| < K] \leq \varepsilon + K < +\infty,$$

so UI implies L^1 -bddness. But the opposite is not true by our last example (take $n^2 > K$).

EX 15.5 (L^p -bdd RVs) But L^p -bddness works for $p > 1$. Let $\mathcal{C}$ be L^p -bdd and $X \in \mathcal{C}$. Then

$$\mathbb{E}[|X|; |X| > K] \leq \mathbb{E}[K^{1-p}|X|^p; |X| > K] \leq K^{1-p}A \rightarrow 0,$$

as $K \rightarrow +\infty$.

EX 15.6 (Dominated RVs) Assume $\exists Y \in L^1$ s.t. $|X| \leq Y \forall X \in \mathcal{C}$. Then

$$\mathbb{E}[|X|; |X| > K] \leq \mathbb{E}[Y; |X| > K] \leq \mathbb{E}[Y; Y > K],$$

and apply lemma above.

2 Proof of main theorem

Proof: We start with the if part. Fix $\varepsilon > 0$. We want to show that for n large enough:

$$\mathbb{E}|X_n - X| \leq \varepsilon.$$

It is natural to truncate at K to apply the UI property. Let $\phi_K(x) = \text{sgn}(x)[|x| \wedge K]$. Then,

$$\begin{aligned} \mathbb{E}|X_n - X| &\leq \mathbb{E}|\phi_K(X_n) - X_n| + \mathbb{E}|\phi_K(X) - X| + \mathbb{E}|\phi_K(X_n) - \phi_K(X)| \\ &\leq \mathbb{E}[|X_n|; |X_n| > K] + \mathbb{E}[|X|; |X| > K] + \mathbb{E}|\phi_K(X_n) - \phi_K(X)|. \end{aligned}$$

The 1st term $\leq \varepsilon/3$ by UI and the 2nd term $\leq \varepsilon/3$ by lemma above. Check, by case analysis, that

$$|\phi_K(x) - \phi_K(y)| \leq |x - y|,$$

so $\phi_K(X_n) \rightarrow_P \phi_K(X)$. By bounded convergence for convergence in probability, the claim is proved.

LEM 15.7 (Bounded convergence theorem (convergence in probability version))

Let $X_n \leq K < +\infty \forall n$ and $X_n \rightarrow_P X$. Then

$$\mathbb{E}|X_n - X| \rightarrow 0.$$

Proof:(Sketch) By

$$\mathbb{P}[|X| \geq K + m^{-1}] \leq \mathbb{P}[|X_n - X| \geq m^{-1}],$$

it follows that $\mathbb{P}[|X| \leq K] = 1$. Fix $\varepsilon > 0$

$$\begin{aligned} \mathbb{E}|X_n - X| &= \mathbb{E}[|X_n - X|; |X_n - X| > \varepsilon/2] + \mathbb{E}[|X_n - X|; |X_n - X| \leq \varepsilon/2] \\ &\leq 2K\mathbb{P}[|X_n - X| > \varepsilon/2] + \varepsilon/2 < \varepsilon, \end{aligned}$$

for n large enough. ■

Proof of only if part. Suppose $X_n \rightarrow X$ in L^1 . We know that L^1 implies convergence in probability. So the first claim follows.

For the second claim, if $n \geq N$ (large enough),

$$\mathbb{E}|X_n - X| \leq \varepsilon.$$

We can choose K large enough so that

$$\mathbb{E}[|X_n|; |X_n| > K] < \varepsilon,$$

$\forall n \leq N$. (Because there is only a finite number.) So only need to worry about $n > N$. To use L^1 convergence, natural to write

$$\mathbb{E}[|X_n|; |X_n| > K] \leq \mathbb{E}[|X_n - X|; |X_n| > K] + \mathbb{E}[|X|; |X_n| > K].$$

First term $\leq \varepsilon$. The issue with the second term is that we cannot apply the lemma because the event involves X_n rather than X . In fact, a stronger version exists:

LEM 15.8 (Absolute continuity) Let $X \in L^1$. $\forall \varepsilon > 0, \exists \delta > 0$, s.t. $\mathbb{P}[F] < \delta$ implies

$$\mathbb{E}[|X|; F] < \varepsilon.$$

Proof: Argue by contradiction. Suppose there is ε and F_n s.t. $\mathbb{P}[F_n] \leq 2^{-n}$ and

$$\mathbb{E}[|X|; F_n] \geq \varepsilon.$$

By BC,

$$\mathbb{P}[H] \equiv \mathbb{P}[F_n \text{ i.o.}] = 0.$$

By reverse Fatou (applied to $|X|\mathbb{1}_H = \limsup |X|\mathbb{1}_{F_n}$),

$$\mathbb{E}[|X|; H] \geq \varepsilon,$$

a contradiction. ■

To conclude note that

$$\mathbb{P}[|X_n| > K] \leq \frac{\mathbb{E}|X_n|}{K} \leq \frac{\sup_{n \geq N} \mathbb{E}|X_n|}{K} \leq \frac{\sup_{n \geq N} \mathbb{E}|X| + \mathbb{E}|X_n - X|}{K} < \delta,$$

uniformly in n for K large enough. We are done. ■

3 UI MGs

THM 15.9 (Convergence of UI MGs) *Let M be UI MG. Then*

$$M_n \rightarrow M_\infty,$$

a.s. and in L^1 . Moreover,

$$M_n = \mathbb{E}[M_\infty | \mathcal{F}_n], \quad \forall n.$$

Proof: UI implies L^1 -bddness so we have $M_n \rightarrow M_\infty$ a.s. By necessary and sufficient condition, we also have L^1 convergence.

Now note that for all $r \geq n$ and $F \in \mathcal{F}_n$, we know $\mathbb{E}[M_r | \mathcal{F}_n] = M_n$ or

$$\mathbb{E}[M_r; F] = \mathbb{E}[M_n; F],$$

by definition of CE. We can take a limit by L^1 convergence. More precisely

$$|\mathbb{E}[M_r; F] - \mathbb{E}[M_\infty; F]| \leq \mathbb{E}[|M_r - M_\infty|; F] \leq \mathbb{E}[|M_r - M_\infty|] \rightarrow 0,$$

as $r \rightarrow \infty$. So plugging above

$$\mathbb{E}[M_\infty; F] = \mathbb{E}[M_n; F],$$

and $\mathbb{E}[M_\infty | \mathcal{F}_n] = M_n$. ■

4 Applications I

THM 15.10 (Levy's upward thm) *Let $Z \in L^1$ and define $M_n = \mathbb{E}[Z | \mathcal{F}_n]$. Then M is a UI MG and*

$$M_n \rightarrow M_\infty = \mathbb{E}[Z | \mathcal{F}_\infty],$$

a.s. and in L^1 .

Proof: M is a MG by (TOWER). We first show it is UI:

LEM 15.11 *Let $X \in L^1(\Omega, \mathcal{F}, \mathbb{P})$. Then*

$$\{\mathbb{E}[X | \mathcal{G}] : \mathcal{G} \text{ is a sub-}\sigma\text{-field of } \mathcal{F}\},$$

is UI.

Proof: We use the absolute continuity lemma again. Let $Y = \mathbb{E}[X | \mathcal{G}] \in \mathcal{G}$. Since $\{|Y| > K\} \in \mathcal{G}$,

$$\begin{aligned} \mathbb{E}[|Y|; |Y| > K] &= \mathbb{E}[|\mathbb{E}[X | \mathcal{G}]|; |Y| > K] \\ &\leq \mathbb{E}[\mathbb{E}[|X| | \mathcal{G}]; |Y| > K] \\ &= \mathbb{E}[|X|; |Y| > K]. \end{aligned}$$

By Markov and (JENSEN)

$$\mathbb{P}[|Y| > K] \leq \frac{\mathbb{E}[|Y|]}{K} \leq \frac{\mathbb{E}[|X|]}{K} \leq \delta,$$

for K large enough (uniformly in $\mathcal{G}$). And we are done. ■

In particular, we have convergence a.s. and in L^1 to $M_\infty \in \mathcal{F}_\infty$.

Let $Y = \mathbb{E}[Z | \mathcal{F}_\infty] \in \mathcal{F}_\infty$. By dividing into negative and positive parts, we assume $Z \geq 0$. We want to show, for $F \in \mathcal{F}_\infty$,

$$\mathbb{E}[Z; F] = \mathbb{E}[M_\infty; F].$$

By Uniqueness Lemma, it suffices to prove equality for all $\mathcal{F}_n$. If $F \in \mathcal{F}_n \subseteq \mathcal{F}_\infty$, then by (TOWER)

$$\mathbb{E}[Z; F] = \mathbb{E}[Y; F] = \mathbb{E}[M_n; F] = \mathbb{E}[M_\infty; F].$$

The first equality is by definition of Y ; the second equality is by definition of M_n ; the third equality is from our main theorem. ■

THM 15.12 (Levy's 0 – 1 law) Let $A \in \mathcal{F}_\infty$. Then

$$\mathbb{P}[A | \mathcal{F}_n] \rightarrow \mathbb{1}_A.$$

Proof: Immediate. ■

COR 15.13 (Kolmogorov's 0 – 1 law) Let $X_1, X_2, \dots$ be iid RVs. Recall that the tail σ -field is

$$\mathcal{T} = \cap_n \mathcal{T}_n = \cap_n \sigma(X_{n+1}, X_{n+2}, \dots).$$

If $A \in \mathcal{T}$ then $\mathbb{P}[A] \in \{0, 1\}$.

Proof: Since $A \in \mathcal{T}_n$ is independent of $\mathcal{F}_n$,

$$\mathbb{P}[A | \mathcal{F}_n] = \mathbb{P}[A],$$

$\forall n$. By Levy's law,

$$\mathbb{P}[A] = \mathbb{1}_A \in \{0, 1\}. \quad \blacksquare$$

5 Applications II

THM 15.14 (Levy's Downward Thm) Let $Z \in L^1(\Omega, \mathcal{F}, \mathbb{P})$ and $\{\mathcal{G}_{-n}\}_{n \geq 0}$ a collection of σ -fields s.t.

$$\mathcal{G}_{-\infty} = \cap_k \mathcal{G}_{-k} \subseteq \cdots \subseteq \mathcal{G}_{-n} \subseteq \cdots \subseteq \mathcal{G}_{-1} \subseteq \mathcal{F}.$$

Define

$$M_{-n} = \mathbb{E}[Z | \mathcal{G}_{-n}].$$

Then

$$M_{-n} \rightarrow M_{-\infty} = \mathbb{E}[Z | \mathcal{G}_{-\infty}]$$

a.s. and in L^1 .

Proof: We apply the same argument as in the Martingale Convergence Thm. Let $\alpha < \beta \in \mathbb{Q}$ and

$$\Lambda_{\alpha, \beta} = \{\omega : \liminf X_{-n} < \alpha < \beta < \limsup X_{-n}\}.$$

Note that

$$\begin{aligned} \Lambda &\equiv \{\omega : X_n \text{ does not converge}\} \\ &= \{\omega : \liminf X_{-n} < \limsup X_{-n}\} \\ &= \cup_{\alpha < \beta \in \mathbb{Q}} \Lambda_{\alpha, \beta}. \end{aligned}$$

Let $U_N[\alpha, \beta]$ be the number of upcrossings of $[\alpha, \beta]$ between time $-N$ and -1 . Then by the Upcrossing Lemma applied to the MG $M_{-N}, \dots, M_{-1}$

$$(\beta - \alpha)\mathbb{E}U_N[\alpha, \beta] \leq |\alpha| + \mathbb{E}|M_{-1}| \leq |\alpha| + \mathbb{E}|Z|.$$

By (MON)

$$U_N[\alpha, \beta] \uparrow U_\infty[\alpha, \beta],$$

and

$$(\beta - \alpha)\mathbb{E}U_\infty[\alpha, \beta] \leq |\alpha| + \mathbb{E}|Z| < +\infty,$$

so that

$$\mathbb{P}[U_\infty[\alpha, \beta] = \infty] = 0.$$

Since

$$\Lambda_{\alpha, \beta} \subseteq \{U_\infty[\alpha, \beta] = \infty\},$$

we have $\mathbb{P}[\Lambda_{\alpha, \beta}] = 0$. By countability, $\mathbb{P}[\Lambda] = 0$. Therefore we have convergence a.s.

By lemma in previous class, M is UI and hence we have L^1 convergence as well.

Finally, for all $G \in \mathcal{G}_{-\infty} \subseteq \mathcal{G}_{-n}$,

$$\mathbb{E}[Z; G] = \mathbb{E}[M_{-n}; G].$$

Take the limit $n \rightarrow +\infty$ and use L^1 convergence. ■

5.1 Law of large numbers

An application:

THM 15.15 (Strong Law; Martingale Proof) *Let $X_1, X_2, \dots$ be iid RVs with $\mathbb{E}[X_1] = \mu$ and $\mathbb{E}|X_1| < +\infty$. Let $S_n = \sum_{i \leq n} X_i$. Then*

$$n^{-1}S_n \rightarrow \mu,$$

a.s. and in L^1 .

Proof: Let

$$\mathcal{G}_{-n} = \sigma(S_n, S_{n+1}, S_{n+2}, \dots) = \sigma(S_n, X_{n+1}, X_{n+2}, \dots),$$

and note that, for $1 \leq i \leq n$,

$$\mathbb{E}[X_1 | \mathcal{G}_{-n}] = \mathbb{E}[X_1 | S_n] = \mathbb{E}[X_i | S_n] = \mathbb{E}[n^{-1}S_n | S_n] = n^{-1}S_n,$$

by symmetry. By Levy's Downward Thm

$$n^{-1}S_n \rightarrow \mathbb{E}[X_1 | \mathcal{G}_{-\infty}],$$

a.s. and in L^1 . But the limit must be trivial by Kolmogorov's 0-1 law and we must have $\mathbb{E}[X_1 | \mathcal{G}_{-\infty}] = \mu$. ■

5.2 Hewitt-Savage

DEF 15.16 *Let $X_1, X_2, \dots$ be iid RVs. Let $\mathcal{E}_n$ be the σ -field generated by events invariant under permutations of the X s that leave $X_{n+1}, X_{n+2}, \dots$ unchanged. The exchangeable σ -field is $\mathcal{E} = \bigcap_m \mathcal{E}_m$.*

THM 15.17 (Hewitt-Savage 0-1 law) *Let $X_1, X_2, \dots$ be iid RVs. If $A \in \mathcal{E}$ then $\mathbb{P}[A] \in \{0, 1\}$.*

Proof: The idea of the proof is to show that A is independent of itself. Indeed, we then have

$$0 = \mathbb{P}[A] - \mathbb{P}[A \cap A] = \mathbb{P}[A] - \mathbb{P}[A]\mathbb{P}[A] = \mathbb{P}[A](1 - \mathbb{P}[A]).$$

Since $A \in \mathcal{E}$ and $A \in \mathcal{F}_\infty$, it suffices to show that $\mathcal{E}$ is independent of $\mathcal{F}_n$ for every n (by the π - λ theorem).

WTS: for every bounded ϕ , $B \in \mathcal{E}$,

$$\mathbb{E}[\phi(X_1, \dots, X_k); B] = \mathbb{E}[\phi(X_1, \dots, X_k)]\mathbb{E}[B] = \mathbb{E}[\mathbb{E}[\phi(X_1, \dots, X_k)]; B],$$

or equivalently

$$Y = \mathbb{E}[\phi(X_1, \dots, X_k) | \mathcal{E}] = \mathbb{E}[\phi(X_1, \dots, X_k)].$$

It suffices to show that Y is independent of $\mathcal{F}_k$. Indeed, by the L^2 characterization of conditional expectation and independence,

$$0 = \mathbb{E}[(\phi(X_1, \dots, X_k) - Y)Y] = \mathbb{E}[\phi(X_1, \dots, X_k)]\mathbb{E}[Y] - \mathbb{E}[Y^2] = -\text{Var}[Y],$$

and Y is constant.

1. Since ϕ is bounded, it is integrable and Levy's Downward Thm implies

$$\mathbb{E}[\phi(X_1, \dots, X_k) | \mathcal{E}_n] \rightarrow \mathbb{E}[\phi(X_1, \dots, X_k) | \mathcal{E}].$$

2. We make ϕ “exchangeable” by averaging over all configurations and taking a limit as $n \rightarrow +\infty$. Define

$$A_n(\phi) = \frac{1}{(n)_k} \sum_{1 \leq i_1 \neq \dots \neq i_k \leq n} \phi(X_{i_1}, \dots, X_{i_k}),$$

where $(n)_k = n(n-1) \cdots (n-k+1)$. Note by symmetry

$$A_n(\phi) = \mathbb{E}[A_n(\phi) | \mathcal{E}_n] = \mathbb{E}[\phi(X_1, \dots, X_k) | \mathcal{E}_n] \rightarrow \mathbb{E}[\phi(X_1, \dots, X_k) | \mathcal{E}].$$

3. The reason we did this is that now the first k X s have little influence on this quantity and therefore the limit is independent of them. However, note that

$$\frac{1}{(n)_k} \sum_{1 \leq i} \phi(X_{i_1}, \dots, X_{i_k}) \leq \frac{k(n-1)_{k-1}}{(n)_k} \sup \phi = \frac{k}{n} \sup \phi \rightarrow 0,$$

so that the limit of $A_n(\phi)$ is independent of X_1 and

$$\mathbb{E}[\phi(X_1, \dots, X_k) | \mathcal{E}] \in \sigma(X_2, \dots),$$

and by induction

$$Y = \mathbb{E}[\phi(X_1, \dots, X_k) | \mathcal{E}] \in \sigma(X_{k+1}, \dots).$$

■

6 Optional Sampling

6.1 Review: Stopping times

Recall:

DEF 15.18 A random variable $T : \Omega \rightarrow \overline{\mathbb{Z}}_+ \equiv \{0, 1, \dots, +\infty\}$ is called a stopping time if

$$\{T = n\} \in \mathcal{F}_n, \forall n \in \overline{\mathbb{Z}}_+.$$

EX 15.19 Let $\{A_n\}$ be an adapted process and $B \in \mathcal{B}$. Then

$$T = \inf\{n \geq 0 : A_n \in B\},$$

is a stopping time.

THM 15.20 Let $\{M_n\}$ be a MG and T be a stopping time. Then M_T is integrable and

$$\mathbb{E}[M_T] = \mathbb{E}[X_0].$$

if one of the following holds:

1. T is bounded.
2. M is bounded and T is a.s. finite.
3. $\mathbb{E}[T] < +\infty$ and M has bounded increments.
4. M is UI. (This one is new. The proof follows from the Optional Sampling Theorem below.)

DEF 15.21 ($\mathcal{F}_T$) Let T be a stopping time. Denote by $\mathcal{F}_T$ the set of all events F such that $\forall n \in \overline{\mathbb{Z}}_+$

$$F \cap \{T = n\} \in \mathcal{F}_n.$$

6.2 More on the σ -field $\mathcal{F}_T$

The following two lemmas help clarify the definition of $\mathcal{F}_T$:

LEM 15.22 $\mathcal{F}_T = \mathcal{F}_n$ if $T \equiv n$, $\mathcal{F}_T = \mathcal{F}_\infty$ if $T \equiv \infty$ and $\mathcal{F}_T \subseteq \mathcal{F}_\infty$ for any T .

Proof: In the first case, note $F \cap \{T = k\}$ is empty if $k \neq n$ and is F if $k = n$. So if $F \in \mathcal{F}_T$ then $F = F \cap \{T = n\} \in \mathcal{F}_n$ and if $F \in \mathcal{F}_n$ then $F = F \cap \{T = n\} \in \mathcal{F}_T$. Moreover $\emptyset \in \mathcal{F}_n$ so we have proved both inclusions. This works also for $n = \infty$. For the third claim note

$$F = \cup_{k \in \mathbb{Z}_+} F \cap \{T = k\} \in \mathcal{F}_\infty.$$

■

LEM 15.23 *If X is adapted and T is a stopping time then $X_T \in \mathcal{F}_T$ (where we assume that $X_\infty \in \mathcal{F}_\infty$, e.g., $X_\infty = \liminf X_n$).*

Proof: For $B \in \mathcal{B}$

$$\{X_T \in B\} \cap \{T = n\} = \{X_n \in B\} \cap \{T = n\} \in \mathcal{F}_n.$$

■

LEM 15.24 *If S, T are stopping times then $\mathcal{F}_{S \wedge T} \subseteq \mathcal{F}_T$.*

Proof: Let $F \in \mathcal{F}_{S \wedge T}$. Note that

$$F \cap \{T = n\} = \cup_{k \leq n} [(F \cap \{S \wedge T = k\}) \cap \{T = n\}] \in \mathcal{F}_n.$$

Indeed, the expression in parenthesis is in $\mathcal{F}_k \subseteq \mathcal{F}_n$ and $\{T = n\} \in \mathcal{F}_n$. ■

6.3 Optional Sampling Theorem (OST)

THM 15.25 (Optional Sampling Theorem) *If M is a UI MG and S, T are stopping times with $S \leq T$ a.s. then $\mathbb{E}[M_T] < +\infty$ and*

$$\mathbb{E}[M_T | \mathcal{F}_S] = M_S.$$

Proof: Since M is UI, $\exists M_\infty \in \mathcal{L}^1$ s.t. $M_n \rightarrow M_\infty$ a.s. and in $\mathcal{L}^1$. We prove a more general claim:

LEM 15.26

$$\mathbb{E}[M_\infty | \mathcal{F}_T] = M_T.$$

Indeed, we then get the theorem by (TOWER) and (JENSEN).

Proof:(Lemma) Wlog we assume $M_\infty \geq 0$ so that $M_n = \mathbb{E}[M_\infty | \mathcal{F}_n] \geq 0 \forall n$. Let $F \in \mathcal{F}_T$. Then (trivially)

$$\mathbb{E}[M_\infty; F \cap \{T = \infty\}] = \mathbb{E}[M_T; F \cap \{T = \infty\}]$$

so STS

$$\mathbb{E}[M_\infty; F \cap \{T < +\infty\}] = \mathbb{E}[M_T; F \cap \{T < +\infty\}].$$

In fact, by (MON), STS

$$\mathbb{E}[M_\infty; F \cap \{T \leq k\}] = \mathbb{E}[M_T; F \cap \{T \leq k\}] = \mathbb{E}[M_{T \wedge k}; F \cap \{T \leq k\}],$$

$\forall k$. To conclude we make two observations:

1. $F \cap \{T \leq k\} \in \mathcal{F}_{T \wedge k}$. Indeed if $n \leq k$

$$F \cap \{T \leq k\} \cap \{T \wedge k = n\} = F \cap \{T = n\} \in \mathcal{F}_n,$$

and if $n > k$

$$= \emptyset \in \mathcal{F}_n.$$

2. $\mathbb{E}[M_\infty | \mathcal{F}_{T \wedge k}] = M_{T \wedge k}$. Since $\mathbb{E}[M_\infty | \mathcal{F}_k] = M_k$, STS $\mathbb{E}[M_k | \mathcal{F}_{T \wedge k}] = M_{T \wedge k}$. But note that if $G \in \mathcal{F}_{T \wedge k}$

$$\mathbb{E}[M_k; G] = \sum_{l \leq k} \mathbb{E}[M_k; G \cap \{T \wedge k = l\}] = \sum_{l \leq k} \mathbb{E}[M_l; G \cap \{T \wedge k = l\}] = \mathbb{E}[M_{T \wedge k}; G]$$

since $G \cap \{T \wedge k = l\} \in \mathcal{F}_l$.

■

■

6.4 Example: Biased RW

DEF 15.27 The asymmetric simple RW with parameter $1/2 < p < 1$ is the process $\{S_n\}_{n \geq 0}$ with $S_0 = 0$ and $S_n = \sum_{k \leq n} X_k$ where the X_k s are iid in $\{-1, +1\}$ s.t. $\mathbb{P}[X_1 = 1] = p$. Let $q = 1 - p$. Let $\phi(x) = (q/p)^x$ and $\psi_n(x) = x - (p - q)n$.

THM 15.28 Let $\{S_n\}$ as above. Let $a < 0 < b$. Define $T_x = \inf\{n \geq 0 : S_n = x\}$. Then

1. We have

$$\mathbb{P}[T_a < T_b] = \frac{\phi(b) - \phi(0)}{\phi(b) - \phi(a)}.$$

In particular, $\mathbb{P}[T_a < +\infty] = 1/\phi(a)$ and $\mathbb{P}[T_b < \infty] = 1$.

2. We have

$$\mathbb{E}[T_b] = \frac{b}{2p - 1}.$$

Proof: There are two MGs here:

$$\mathbb{E}[\phi(S_n) | \mathcal{F}_{n-1}] = p(q/p)^{S_{n-1}+1} + q(q/p)^{S_{n-1}-1} = \phi(S_{n-1}),$$

and

$$\mathbb{E}[\psi_n(S_n) | \mathcal{F}_{n-1}] = p[S_{n-1}+1-(p-q)(n)] + q[S_{n-1}-1-(p-q)(n)] = \psi_{n-1}(S_{n-1}).$$

Let $N = T_a \wedge T_b$. Now note that $\phi(S_{N \wedge n})$ is a bounded MG and therefore applying the MG property at time n and taking limits as $n \rightarrow \infty$ (using (DOM))

$$\phi(0) = \mathbb{E}[\phi(S_N)] = \mathbb{P}[T_a < T_b]\phi(a) + \mathbb{P}[T_a > T_b]\phi(b),$$

where we need to prove that $N < +\infty$ a.s. Indeed, since $(b-a) + 1$ -steps always take us out of (a, b) ,

$$\mathbb{P}[T_b > n(b-a)] \leq (1 - q^{b-a})^n,$$

so that

$$\mathbb{E}[T_b] = \sum_{k \geq 0} \mathbb{P}[T_b > k] \leq \sum_n (b-a)(1 - q^{b-a})^n < +\infty.$$

In particular $T_b < +\infty$ a.s. and $N < +\infty$ a.s. Rearranging the formula above gives the first result. (For the second part of the first result, take $b \rightarrow +\infty$ and use monotonicity.)

For the third one, note that $T_b \wedge n$ is bounded so that

$$0 = \mathbb{E}[S_{T_b \wedge n} - (p-q)(T_b \wedge n)].$$

By (MON), $\mathbb{E}[T_b \wedge n] \uparrow \mathbb{E}[T_b]$. Finally, using

$$\mathbb{P}[-\inf_n S_n \geq -a] = \mathbb{P}[T_a < +\infty],$$

and the fact that $-\inf_n S_n \geq 0$ shows that $\mathbb{E}[-\inf_n S_n] < +\infty$. Hence, we can use (DOM) with $|S_{T_b \wedge n}| \leq \max\{b, -\inf_n S_n\}$. ■

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