

# STATS310B – Lecture 8

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## 1 Polya's Urn

Consider an urn of infinite capacity. The urn initially has one white ball and one black ball inside it. At each time step, a ball is picked uniformly at random from the urn and replaced back into the urn with another ball of the same color. Equivalently we choose a color with probability proportional to the number of balls of the same color and then put in an additional ball of the chosen color.

Let  $W_n$  be the proportion of white ball at time  $n$  with  $W_0 = \frac{1}{2}$ . We would like to understand the limiting behavior of  $W_n$  as  $n \rightarrow \infty$ .

**Proposition 1.** *Let  $\mathcal{F}_n = \sigma(W_1, \dots, W_n)$ . Then the sequence  $\{W_n\}_{n \geq 0}$  is a martingale with respect to  $\{\mathcal{F}_n\}_{n \geq 0}$ .*

*Proof.* Note that the total number of balls at time  $n$  is  $n + 2$ . Let  $N_n$  be the number of white ball in the urn at time  $n$ . Thus,  $W_n = \frac{1}{n+2}N_n$ . It follows that,

$$\begin{aligned}\mathbb{E}(W_{n+1}|\mathcal{F}_n) &= \frac{1}{n+3}\mathbb{E}(N_{n+1}|\mathcal{F}_n) \\ &= \frac{1}{n+3}\left((N_n+1)\frac{N_n}{n+2} + N_n\frac{n+2-N_n}{n+2}\right) \\ &= \frac{1}{n+3}\left(\frac{1}{n+2}N_n^2 + \frac{N_n}{n+2} + N_n - \frac{1}{n+2}N_n^2\right) \\ &= \frac{1}{n+3}\left(\frac{n+3}{n+2}N_n\right) \\ &= \frac{1}{n+2}N_n \\ &= W_n.\end{aligned}\quad \square$$

Note that  $W_n \in [0, 1]$  for every  $n$  and thus  $\sup_n \mathbb{E}[W_n^+] \leq 1 < \infty$ . It follows that there exists an integrable random variable  $W$  such that  $W_n \rightarrow W$  almost surely. We will in fact prove that  $W$  is uniformly distributed on  $[0, 1]$ .

*Proof.* We will show by induction that for all  $n$ ,  $N_n$  is uniformly distributed on  $\{1, 2, \dots, n+1\}$ , where  $N_n$  is the number of white balls at time  $n$ . This is true when  $n = 0$  since  $N_0 = 1$ . Now suppose that the result is true for some  $n$ . Then, for  $k = 1, \dots, n+2$ ,

$$\begin{aligned}\mathbb{P}(M_{n+1} = k) &= \sum_{j=1}^{n+1} \mathbb{P}(M_{n+1} = k | N_n = j) \mathbb{P}(N_n = j) \\ &= \frac{1}{n+1} \sum_{j=1}^{n+1} \mathbb{P}(M_{n+1} = k | N_n = j).\end{aligned}$$

Note that  $\mathbb{P}(M_{n+1} = k | N_n = j) \neq 0$  only if  $j = k-1$  or  $j = k$ . This is even true if  $k = 1$  or  $k = n+2$  although in these cases one  $\mathbb{P}(M_{n+1} = k | N_n = k-1) = 0$  or  $\mathbb{P}(M_{n+1} = k | N_n = k) = 0$  respectively which agrees with the calculations below. Thus,

$$\begin{aligned}\mathbb{P}(M_{n+1} = k) &= \frac{1}{n+1} (\mathbb{P}(M_{n+1} = k | N_n = k-1) + \mathbb{P}(M_{n+1} = k | N_n = k)) \\ &= \frac{1}{n+1} \left( \frac{k-1}{n+2} + \frac{n+2-k}{n+2} \right) \\ &= \frac{1}{n+1} \left( \frac{n+1}{n+2} \right) \\ &= \frac{1}{n+2}.\end{aligned}$$

Thus,  $N_{n+1}$  is uniformly distributed on  $\{1, \dots, n+2\}$ . Hence,  $W_n$  is uniformly distributed on  $\left\{\frac{1}{n+2}, \dots, \frac{n+1}{n+2}\right\}$  which implies  $W_n$  converges in distribution to  $U[0, 1]$  but  $W_n$  also converges almost surely (and this  $W_n \rightarrow W$  in distribution). Thus,  $W$  must be uniformly distributed on  $[0, 1]$ .  $\square$

## 2 Lévy's downwards convergence theorem

Our next convergence theorem is Lévy's downwards convergence theorem which is also called the backwards martingale theorem.

**Theorem 1** (Lévy's downwards convergence theorem). *Let  $X$  be an integrable random variable defined on  $(\Omega, \mathcal{F}, \mathbb{P})$ . Let  $\mathcal{F}_0 \supseteq \mathcal{F}_1 \supseteq \mathcal{F}_2 \supseteq \dots$  be a decreasing sequence of sub- $\sigma$ -algebras of  $\mathcal{F}$ . Define  $\mathcal{F}^* = \bigcap_{n=0}^{\infty} \mathcal{F}_n$ , then*

$$\mathbb{E}(X | \mathcal{F}_n) \rightarrow \mathbb{E}(X | \mathcal{F}^*),$$

*almost surely and in  $L^1$ .*

*Proof.* Let  $X_n = \mathbb{E}(X | \mathcal{F}_n)$ . We will first show that  $\{X_n\}_{n \geq 0}$  has an almost sure limit  $X^*$ . We will then prove that  $X_n$  converges to  $X^*$  in  $L^1$  and then finally we will show that  $X^* = \mathbb{E}(X | \mathcal{F}^*)$ . Fix  $n \in \mathbb{N}$  and consider the time reversed finite sequence,

$$X_n, X_{n-1}, X_{n-2}, \dots, X_0.$$

The above sequence is a martingale with respect to  $\mathcal{F}_n, \mathcal{F}_{n-1}, \dots, \mathcal{F}_0$ . This is because  $\mathcal{F}_k \subseteq \mathcal{F}_{k-1}$  and thus

$$\mathbb{E}(X_{k-1} | \mathcal{F}_k) = \mathbb{E}(\mathbb{E}(X | \mathcal{F}_{k-1}) | \mathcal{F}_k) = \mathbb{E}(X | \mathcal{F}_k).$$

Fix an interval  $[a, b]$  and let  $U_n$  be the number of complete upcrossings of  $[a, b]$  by  $X_n, X_{n-1}, \dots, X_0$ .

By the upcrossing lemma,

$$\begin{aligned}
 \mathbb{E}[U_n] &\leq \frac{\mathbb{E}[(X_0 - a)^+] - \mathbb{E}[(X_n - a)^+]}{b - a} \\
 &\leq \frac{\mathbb{E}[(X_0 - a)^+]}{b - a} \\
 &= \frac{\mathbb{E}[(\mathbb{E}(X|\mathcal{F}_0) - a)^+]}{b - a} \\
 &\leq \frac{\mathbb{E}[\mathbb{E}((X - a)^+|\mathcal{F}_0)]}{b - a} \\
 &\leq \frac{\mathbb{E}[(X - a)^+]}{b - a}.
 \end{aligned}$$

Note that  $U_n \leq U_{n+1}$  and thus  $U_n \nearrow U$  for some random variable  $U$ . By the monotone convergence theorem,  $\mathbb{E}[U] \leq \frac{\mathbb{E}[(X-a)^+]}{b-a} < \infty$ . As with Doob's sub-martingale convergence theorem, this implies that  $X^* = \lim_{n \rightarrow \infty} X_n$  exists almost surely. We will now show that  $X^*$  is integrable. Note that

$$\begin{aligned}
 \mathbb{E}[|X^*|] &= \mathbb{E}\left[\lim_{n \rightarrow \infty} |X_n|\right] \\
 &\leq \liminf_{n \rightarrow \infty} \mathbb{E}[|X_n|] \\
 &= \liminf_{n \rightarrow \infty} \mathbb{E}[\mathbb{E}(|X_n||\mathcal{F}_n)] \\
 &\leq \liminf_{n \rightarrow \infty} \mathbb{E}[\mathbb{E}(|X||\mathcal{F}_n)] \\
 &= \liminf_{n \rightarrow \infty} \mathbb{E}[|X|] \\
 &= \mathbb{E}[|X|] < \infty.
 \end{aligned}$$

To show that  $X^* = \mathbb{E}(X|\mathcal{F}^*)$  and that  $X_n \rightarrow X^*$  in  $L^1$  we need to first review the concept of *uniform integrability*.  $\square$

### 3 Uniform integrability

**Definition 1.** A sequence of random variables  $\{X_n\}_{n \geq 1}$  is *uniformly integrable* if for  $\epsilon > 0$ , there exists  $K > 0$  such that,

$$\sup_n \mathbb{E}[|X_n| \mathbf{1}_{\{|X_n| > k\}}] < \epsilon.$$

Uniform integrability allows one to calculate the expectation of a limit.

**Lemma 1.** Suppose that  $\{X_n\}_{n \geq 0}$  is a uniformly integrable sequence and  $X_n \rightarrow X$  almost surely. Then  $X$  is integrable and  $X_n \rightarrow X$  in  $L^1$ .

*Proof.* Take any  $\epsilon > 0$  and take  $k$  such that  $\mathbb{E}[|X_n| \mathbf{1}_{\{|X_n| > k\}}] < \epsilon$ . Then,

$$\mathbb{E}[|X_n|] \leq \mathbb{E}[|X_n| \mathbf{1}_{\{|X_n| > k\}}] + \mathbb{E}[|X_n| \mathbf{1}_{\{|X_n| \leq k\}}] \leq \epsilon + k.$$

Thus,  $\mathbb{E}[|X|] \leq \liminf_n \mathbb{E}[|X_n|] \leq \epsilon + k < \infty$ . So  $X$  is integrable. Note that for all  $L > 0$ ,  $|X| \mathbf{1}_{\{|X| > L\}} \leq |X|$  and, almost surely

$$\lim_{L \rightarrow \infty} |X| \mathbf{1}_{\{|X| > L\}} = 0.$$

Thus, by the dominated convergence theorem,

$$\lim_{L \rightarrow \infty} \mathbb{E}[|X| \mathbf{1}_{\{|X| > L\}}] = 0.$$

This shows that given  $\varepsilon > 0$ , we can choose  $k > 0$  so that  $\mathbb{E}[|X|\mathbf{1}_{\{|X|>k\}}] < \varepsilon$  and

$$\sup_n \mathbb{E}[|X_n|\mathbf{1}_{\{|X_n|>k\}}] < \varepsilon.$$

Let  $\varepsilon > 0$  be arbitrary and fix such a corresponding  $k > 0$ . Let  $\phi : \mathbb{R} \rightarrow \mathbb{R}$  be given by

$$\phi(x) = \begin{cases} -k & \text{if } x \leq -k, \\ x & \text{if } x \in (-k, k), \\ k & \text{if } x \geq k. \end{cases}$$

The function  $\phi$  is bounded and continuous and  $|\phi(x) - x| \leq |x|\mathbf{1}_{\{|x|>k\}}$  for all  $x \in \mathbb{R}$ . Thus,

$$\begin{aligned} \mathbb{E}[|X_n - X|] &\leq \mathbb{E}[|X_n - \phi(X_n)|] + \mathbb{E}[|\phi(X_n) - \phi(X)|] + \mathbb{E}[|\phi(X) - X|] \\ &\leq \mathbb{E}[|X_n|\mathbf{1}_{\{|X_n|>k\}}] + \mathbb{E}[|\phi(X_n) - \phi(X)|] + \mathbb{E}[|X|\mathbf{1}_{\{|X|>k\}}] \\ &< 2\varepsilon + \mathbb{E}[|\phi(X_n) - \phi(X)|]. \end{aligned}$$

The random variable  $|\phi(X_n) - \phi(X)|$  is bounded above by  $2k$  and goes to 0 almost surely. Thus, by the dominated convergence theorem,

$$\limsup_n \mathbb{E}[|X_n - X|] \leq 2\varepsilon + \limsup_n \mathbb{E}[|\phi(X_n) - \phi(X)|] = 2\varepsilon.$$

Thus,  $X_n \rightarrow X$  in  $L^1$ . □

We will next state a characterization of uniform integrability that we will need in proving Lévy's downwards convergence theorem.

**Proposition 2.** *Let  $\{X_n\}_{n \geq 1}$  be a sequence of random variables. The sequence  $\{X_n\}_{n \geq 0}$  is uniformly integrable if and only if the following both hold,*

1.  $\sup_n \mathbb{E}[|X_n|] < \infty$ .
2. For all  $\varepsilon > 0$  there exists  $\delta > 0$  such that for all  $A$  and  $n$ , if  $\mathbb{P}(A) < \delta$ , then  $\mathbb{E}[|X_n|\mathbf{1}_A] < \varepsilon$ .

We will prove this proposition in the next lecture.