

18.175: Lecture 26

More on martingales

Scott Sheffield

MIT

Conditional expectation

Regular conditional probabilities

Martingales

Arcsin law, other SRW stories

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Recall: conditional expectation

- ▶ Say we're given a probability space $(\Omega, \mathcal{F}_0, P)$ and a σ -field $\mathcal{F} \subset \mathcal{F}_0$ and a random variable X measurable w.r.t. \mathcal{F}_0 , with $E|X| < \infty$. The **conditional expectation of X given \mathcal{F}** is a new random variable, which we can denote by $Y = E(X|\mathcal{F})$.
- ▶ We require that Y is \mathcal{F} measurable and that for all A in \mathcal{F} , we have $\int_A X dP = \int_A Y dP$.
- ▶ Any Y satisfying these properties is called a **version** of $E(X|\mathcal{F})$.
- ▶ **Theorem:** Up to redefinition on a measure zero set, the random variable $E(X|\mathcal{F})$ exists and is unique.
- ▶ This follows from Radon-Nikodym theorem.

Conditional expectation observations

- ▶ Linearity: $E(aX + Y|\mathcal{F}) = aE(X|\mathcal{F}) + E(Y|\mathcal{F})$.
- ▶ If $X \leq Y$ then $E(X|\mathcal{F}) \leq E(Y|\mathcal{F})$.
- ▶ If $X_n \geq 0$ and $X_n \uparrow X$ with $EX < \infty$, then $E(X_n|\mathcal{F}) \uparrow E(X|\mathcal{F})$ (by dominated convergence).
- ▶ If $\mathcal{F}_1 \subset \mathcal{F}_2$ then
 - ▶ $E(E(X|\mathcal{F}_1)|\mathcal{F}_2) = E(X|\mathcal{F}_1)$.
 - ▶ $E(E(X|\mathcal{F}_2)|\mathcal{F}_1) = E(X|\mathcal{F}_1)$.
- ▶ Second is kind of interesting: says, after I learn \mathcal{F}_1 , my best guess of what my best guess for X will be after learning \mathcal{F}_2 is simply my current best guess for X .
- ▶ Deduce that $E(X|\mathcal{F}_i)$ is a martingale if \mathcal{F}_i is an increasing sequence of σ -algebras and $E(|X|) < \infty$.

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- ▶ Consider probability space (Ω, \mathcal{F}, P) , a measurable map $X : (\Omega, \mathcal{F}) \rightarrow (S, \mathcal{S})$ and $\mathcal{G} \subset \mathcal{F}$ a σ -field. Then $\mu : \Omega \times \mathcal{S} \rightarrow [0, 1]$ is a **regular conditional distribution for X given \mathcal{G}** if
 - ▶ For each A , $\omega \rightarrow \mu(\omega, A)$ is a version of $P(X \in A | \mathcal{G})$.
 - ▶ For a.e. ω , $A \rightarrow \mu(\omega, A)$ is a probability measure on (S, \mathcal{S}) .
- ▶ **Theorem:** Regular conditional probabilities exist if (S, \mathcal{S}) is nice.

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- ▶ Let \mathcal{F}_n be increasing sequence of σ -fields (called a **filtration**).
- ▶ A sequence X_n is **adapted** to \mathcal{F}_n if $X_n \in \mathcal{F}_n$ for all n . If X_n is an adapted sequence (with $E|X_n| < \infty$) then it is called a **martingale** if

$$E(X_{n+1}|\mathcal{F}_n) = X_n$$

for all n . It's a **supermartingale** (resp., **submartingale**) if same thing holds with $=$ replaced by \leq (resp., \geq).

Martingale observations

- ▶ **Claim:** If X_n is a supermartingale then for $n > m$ we have $E(X_n|\mathcal{F}_m) \leq X_m$.
- ▶ **Proof idea:** Follows if $n = m + 1$ by definition; take $n = m + k$ and use induction on k .
- ▶ Similar result holds for submartingales. Also, if X_n is a martingale and $n > m$ then $E(X_n|\mathcal{F}_m) = X_m$.
- ▶ **Claim:** if X_n is a martingale w.r.t. \mathcal{F}_n and ϕ is convex with $E|\phi(X_n)| < \infty$ then $\phi(X_n)$ is a submartingale.
- ▶ **Proof idea:** Immediate from Jensen's inequality and martingale definition.
- ▶ Example: take $\phi(x) = \max\{x, 0\}$.

- ▶ Call H_n **predictable** if each H_{n+1} is \mathcal{F}_n measurable.
- ▶ Maybe H_n represents amount of shares of asset investor has at n th stage.
- ▶ Write $(H \cdot X)_n = \sum_{m=1}^n H_m(X_m - X_{m-1})$.
- ▶ **Observe:** If X_n is a supermartingale and the $H_n \geq 0$ are bounded, then $(H \cdot X)_n$ is a supermartingale.
- ▶ Example: take $H_n = 1_{N \geq n}$ for stopping time N .

Two big results

- ▶ **Optional stopping theorem:** Can't make money in expectation by timing sale of asset whose price is non-negative martingale.
- ▶ **Proof:** Just a special case of statement about $(H \cdot X)$.
- ▶ **Martingale convergence:** A non-negative martingale almost surely has a limit.
- ▶ **Idea of proof:** Count upcrossings (times martingale crosses a fixed interval) and devise gambling strategy that makes lots of money if the number of these is not a.s. finite.

- ▶ How many primary candidates ever get above twenty percent in expected probability of victory? (Asked by Aldous.)
- ▶ Compute probability of having conditional probability reach a before b .

- ▶ **Wald's equation:** Let X_j be i.i.d. with $E|X_j| < \infty$. If N is a stopping time with $EN < \infty$ then $ES_N = EX_1EN$.
- ▶ **Wald's second equation:** Let X_j be i.i.d. with $E|X_j| = 0$ and $EX_j^2 = \sigma^2 < \infty$. If N is a stopping time with $EN < \infty$ then $ES_N = \sigma^2EN$.

- ▶ $S_0 = a \in \mathbb{Z}$ and at each time step S_j independently changes by ± 1 according to a fair coin toss. Fix $A \in \mathbb{Z}$ and let $N = \inf\{k : S_k \in \{0, A\}\}$. What is $\mathbb{E}S_N$?
- ▶ What is $\mathbb{E}N$?

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- ▶ How many walks from $(0, x)$ to (n, y) that don't cross the horizontal axis?
- ▶ Try counting walks that *do* cross by giving bijection to walks from $(0, -x)$ to (n, y) .

Ballot Theorem

- ▶ Suppose that in election candidate A gets α votes and B gets $\beta < \alpha$ votes. What's probability that A is ahead throughout the counting?
- ▶ Answer: $(\alpha - \beta)/(\alpha + \beta)$. Can be proved using reflection principle.

- ▶ Theorem for last hitting time.
- ▶ Theorem for amount of positive time.

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