

18.175: Lecture 19

Even more on martingales

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Recollections

More martingale theorems

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More martingale theorems

Recall: conditional expectation

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- ▶ **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.
- ▶ **Theorem:** $E(X|\mathcal{F}_i)$ is a martingale if \mathcal{F}_i is an increasing sequence of σ -algebras and $E(|X|) < \infty$.

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- ▶ 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).

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- ▶ **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. Basically, you buy every time price gets below the interval, sell each time it gets above.

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- ▶ Compute probability of having a martingale price reach a before b if martingale prices vary continuously.
- ▶ Polya's urn: r red and g green balls. Repeatedly sample randomly and add extra ball of sampled color. Ratio of red to green is martingale, hence a.s. converges to limit.

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Outline

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Theorems (proofs discussed on subsequent slides)

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- ▶ **Orthogonal increment theorem:** Let X_n be a martingale with $EX_n^2 < \infty$ for all n . If $m \leq n$ and $Y \in \mathcal{F}_m$ with $EY^2 < \infty$, then $E((X_n - X_m)Y) = 0$.

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- ▶ **Cond. variance theorem:** If X_n is martingale, $EX_n^2 < \infty$ for all n , then $E((X_n - X_m)^2 | \mathcal{F}_m) = E(X_n^2 | \mathcal{F}_m) - X_m^2$.

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- ▶ **“Accumulated variance” theorems:** Consider martingale X_n with $EX_n^2 < \infty$ for all n . By Doob, can write $X_n^2 = M_n + A_n$ where M_n is a martingale, and

$$A_n = \sum_{m=1}^n E(X_m^2 | \mathcal{F}_{m-1}) - X_{m-1}^2 = \sum_{m=1}^n E((X_m - X_{m-1})^2 | \mathcal{F}_{m-1}).$$

Then $E(\sup_m |X_m|^2) \leq 4EA_\infty$. And $\lim_{n \rightarrow \infty} X_n$ exists and is finite a.s. on $\{A_\infty < \infty\}$.

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- ▶ **Proof idea:** Have $(EX_n^+)^p \leq (E|X_n|)^p \leq E|X_n|^p$ for martingale convergence theorem $X_n \rightarrow X$ a.s. Use L^p maximal inequality to get L^p convergence.

Orthogonality of martingale increments

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- ▶ **Conditional variance theorem:** If X_n is a martingale with $EX_n^2 < \infty$ for all n then $E((X_n - X_m)^2|\mathcal{F}_m) = E(X_n^2|\mathcal{F}_m) - X_m^2$.

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- ▶ We know X_n^2 is a submartingale. By Doob's decomposition, we can write $X_n^2 = M_n + A_n$ where M_n is a martingale, and

$$A_n = \sum_{m=1}^n E(X_m^2 | \mathcal{F}_{m-1}) - X_{m-1}^2 = \sum_{m=1}^n E((X_m - X_{m-1})^2 | \mathcal{F}_{m-1}).$$

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- ▶ **Proof idea:** L^2 maximal equality gives $E(\sup_{0 \leq m \leq n} |X_m|^2) \leq 4EX_n^2 = 4EA_n$. Use monotone convergence.

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- ▶ **Proof idea:** Try fixing a and truncating at time $N = \inf\{n : A_{n+1} > a^2\}$, use L^2 convergence theorem.

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 - ▶ There is an integrable random variable X so that $X_n = E(X|\mathcal{F}_n)$.
- ▶ This implies that every uniformly integrable martingale can be interpreted as a “revised expectation given latest information” sequence.

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- ▶ **Theorem:** $X_{-\infty} = \lim_{n \rightarrow -\infty} X_n$ exists a.s. and in L^1 .
- ▶ **Proof idea:** Use upcrossing inequality to show expected number of upcrossings of any interval is finite. Since $X_n = E(X_0|\mathcal{F}_n)$ the X_n are uniformly integrable, and we can deduce convergence in L^1 .

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- ▶ **Theorem:** For any stopping time $N \leq \infty$, we have $EX_0 \leq EX_N \leq EX_\infty$ where $X_\infty = \lim X_n$.
- ▶ **Fairly general form of optional stopping theorem:** If $L \leq M$ are stopping times and $Y_{M \wedge n}$ is a uniformly integrable submartingale, then $EY_L \leq EY_M$ and $Y_L \leq E(Y_M | \mathcal{F}_L)$.

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- ▶ **Classic question:** Is this also true of the stock market?

Martingales as real-time subjective probability updates

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