

① LECTURE 7

End of pf. of cond. expectation existence: we did case $X \geq 0$

In general, $X^+ := X \mathbb{1}_{\{X \geq 0\}}$, $X^- := -X \mathbb{1}_{\{X < 0\}}$, so that $X = X^+ - X^-$. Then, can check correct definition is:

$$E[X | \mathcal{G}] := E[X^+ | \mathcal{G}] - E[X^- | \mathcal{G}]$$

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Main properties of conditional expectation:

• Linearity: $E[aX + bY | \mathcal{G}] = a E[X | \mathcal{G}] + b E[Y | \mathcal{G}]$

• Monotonicity: $X \leq Y$ a.s. $\implies E[X | \mathcal{G}] \leq E[Y | \mathcal{G}]$ a.s.
PF: WLOG $X = 0$. $E := E[Y | \mathcal{G}] \rightarrow E[E \mathbb{1}_{\{E < 0\}}] = E[Y \mathbb{1}_{\{E < 0\}}] \geq 0$.

• Factorization: $E|X|, E|Y|, E|XY| < \infty$ and X \mathcal{G} -measurable $\implies E[XY | \mathcal{G}] = X E[Y | \mathcal{G}]$.

PF: Say $X = \mathbb{1}_A$ for $A \in \mathcal{G}$: for $B \in \mathcal{G}$,

$$E[X E[Y | \mathcal{G}] \mathbb{1}_B] = E[E[Y | \mathcal{G}] \mathbb{1}_{A \cap B}] = E[Y \mathbb{1}_{A \cap B}] = E[XY \mathbb{1}_B] \implies \text{satisfies defining property.}$$

Then, standard argument: indicators \rightarrow step functions (linearity) \rightarrow measurable functions (monotone conv.)

• Independence: If X ind. of \mathcal{G} (i.e., any $A \in \sigma(X), B \in \mathcal{G}$ ind.) then $E[X | \mathcal{G}] = E[X]$. In particular, X, Y ind. $\implies E[XY] = E[X]E[Y]$.
PF: $\hookrightarrow \mathcal{G} := \mathcal{E}$. Then, $E[E \mathbb{1}_B] = E[X \mathbb{1}_B] = E[X] E \mathbb{1}_B = E[E[X] \cdot \mathbb{1}_B]$.

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- Tower rule: $\mathcal{G}_1 \subseteq \mathcal{G}_2 \subseteq \mathcal{F} \Rightarrow \mathbb{E}[\mathbb{E}[X|\mathcal{G}_2]|\mathcal{G}_1] = \mathbb{E}[X|\mathcal{G}_1]$
 In particular, w/ $\mathcal{G}_1 = \{\emptyset, \Omega\}$, $\mathbb{E}[\mathbb{E}[X|\mathcal{G}_2]] = \mathbb{E}[X]$.
 $\mathbb{E}[\mathbb{E}[X|\mathcal{G}_1] \mathbb{1}_A] = \mathbb{E}[X \mathbb{1}_A] = \mathbb{E}[\mathbb{E}[X|\mathcal{G}_2] \mathbb{1}_A]$
 \uparrow
 $\mathcal{G}_1, \text{ also } \mathcal{G}_2$

- Jensen inequality: f convex, $\mathbb{E}[X], \mathbb{E}[f(X)] < \infty \Rightarrow$
 $\mathbb{E}[f(X)|\mathcal{G}] \geq f(\mathbb{E}[X|\mathcal{G}])$ a.s. \rightarrow both sides random!

Martingales (discrete time):

- Def: $\mathcal{F}_0 \subseteq \mathcal{F}_1 \subseteq \dots \subseteq \mathcal{F}_{n-1} \subseteq \mathcal{F}_n \subseteq \mathcal{F}$ is filtration.
 (Intuition: $\mathcal{F}_i =$ "info available at time i ."
 $(X_n)_{n \geq 0}$ r.v.'s are adapted to filtration if X_i is \mathcal{F}_i m'sble, $\forall i$.
 $(X_n)_{n \geq 1}$ are predictable/previsible if X_i is \mathcal{F}_{i-1} m'sble, $\forall i \geq 1$.

- Def: $(M_n)_{n \geq 0}$ is $\begin{cases} \text{martingale} \\ \text{submartingale} \\ \text{supermartingale} \end{cases}$ if (1) $\mathbb{E}|M_n| < \infty$, (2) adapted,
 and (3) $\mathbb{E}[M_{n+1} | \mathcal{F}_n] \begin{cases} = \\ \geq \\ \leq \end{cases} M_n$ a.s., $\forall n \geq 0$

Rk: Some properties hold w/o/t filtration $\tilde{\mathcal{F}}_n = \sigma(M_0, M_1, \dots, M_n)$.

Prop: For $\begin{cases} \text{mgl} \\ \text{submgl} \\ \text{super mgl} \end{cases}$, $\mathbb{E}M_n \begin{cases} = \\ \geq \\ \leq \end{cases} \mathbb{E}M_0 \quad \forall n \geq 1$. PF: Tower property + induction.

Prop: $\mathbb{E}[M_{n+k} | \mathcal{F}_n] = M_n \quad \forall n \geq 0, k \geq 1$. PF: Same.

Examples:

↙ or just independent.

(1) Random walk: X_i iid, $E X_i = 0$, $S_0 := 0$, $S_n := \sum_{i=1}^n X_i$.
Then (S_n) is martingale (w.r.t $\mathcal{F}_n = \sigma(X_1, \dots, X_n) = \sigma(S_1, \dots, S_n)$):

$$E[S_{n+1} | \mathcal{F}_n] = E[S_n + X_{n+1} | \mathcal{F}_n] = E[S_n | \mathcal{F}_n] + E[X_{n+1} | \mathcal{F}_n]$$

Think: LLN, CLT, etc. for mgd's?

\downarrow
= S_n since $E X_i = 0$ by \mathcal{F}_n m.s.b.l. ind. property.

$\Delta_n := M_n - M_{n-1}$ martingale differences / increments
 $M_n = M_0 + \sum_{i=1}^n \Delta_i$ mgd iff $E[\Delta_{n+1} | \mathcal{F}_n] = 0 \forall n$.
(sub-mgd / super-mgd) $\begin{pmatrix} \geq \\ \leq \end{pmatrix}$

(2) Random product (geometric RW: X_i iid, $X_i \geq 0$ a.s., $E X_i = 1$)

$\iff M_0 = 1$, $M_n = \prod_{i=1}^n X_i$ is mgd.

Ex: $X_i = \begin{cases} 0 & \text{w/prob } 1/2 \\ 2 & \text{w/prob } 1/2 \end{cases}$. Exc: $M_n \rightarrow 0$ a.s.
(Note $0 \neq E M_n = 1$!)

Very non-RW behavior!

(3) Doob martingale: X any r.v. w/ $E|X| < \infty$, (\mathcal{F}_n) any filtration

$\implies M_n := E[X | \mathcal{F}_n]$ is mgd. (tower rule)

~~Say~~ Say $\mathcal{F}_0 = \{\emptyset, \Omega\}$, $\mathcal{F}_N = \mathcal{F}$, then $M_0 = EX$, $M_N = X$, in between "gradually reveal info", "forecast X ".

(4) Martingale transform: (H_n) predictable, (M_n) ~~sub~~ martingale \implies

$(H \cdot M)_n = \begin{cases} 0 & \text{if } n=0 \\ \sum_{i=1}^n H_i (M_i - M_{i-1}) & \text{if } n \geq 1 \end{cases}$ is mgd.

Also, if $H \geq 0$ and (M_n) sub/super-mgd, then $H \cdot M$ sub/super-mgd.

(4)

(Increments of H.M are $\tilde{\Delta}_n = H_n \Delta_n$, $E[\tilde{\Delta}_{n+1} | \mathcal{F}_n] = H_n E[\Delta_{n+1} | \mathcal{F}_n] = 0$.)

Rk: Precursor to stochastic integral, $\int_0^t H dM$ (cf. Stieltjes integrals.)

Ex: Mgl transform of simple RW: $X_i = \begin{cases} +1 & \text{w/prob } 1/2 \\ -1 & \text{w/prob } 1/2 \end{cases}$, $S_n = \sum_{i=1}^n X_i$, $\mathcal{F}_n = \sigma(X_1, \dots, X_n)$, (H_n) previsible, i.e. $H_n \in \mathcal{F}_{n-1}$, i.e. $H_n = h_n(X_1, \dots, X_{n-1})$.

$$(H \cdot S)_n = \sum_{i=1}^n H_i X_i = \sum_{i=1}^n h_i(X_1, \dots, X_{i-1}) X_i.$$

Gambling interpretation: $X_i =$ outcome of i th bet,
 $H_i =$ size of i th bet, (can depend on history!)
 $(H \cdot S)_n =$ profit after n bets

Martingale: Name for specific strategy

$$H_1 = 1, \quad H_n = \begin{cases} 2H_{n-1} & \text{if } X_{n-1} = -1 \\ 0 & \text{if } X_{n-1} = +1 \end{cases}$$

I.e., double my bet until I win, then stop.

Analysis: a.s. some $X_i = 1$, let $n =$ first such time.

$$H_1 = 1, H_2 = 2, H_3 = 2^2, \dots, H_n = 2^{n-1}, H_{n+1} = 0 = H_{n+k}$$

$$X_1 = -1, X_2 = -1, X_3 = -1, \dots, X_n = +1$$

$$\text{profit at all times } (\geq n) = 2^{n-1} - 2^{n-2} - \dots - 2 - 1 = 1.$$

Always win?!

- Rk:
- RW: martingale not converging
 - GRW: mgl converging below mean ($E M_n = 1, M_n \rightarrow 0$)
 - (H.S): mgl converging above mean ($E M_n = 0, M_n \rightarrow 1$)