

Def. X is a martingale, if

(1) X_n is adapted.

(2) $\mathbb{E}_n X_{n+1} = X_n$.

Ex. ξ_1, \dots, ξ_N i.i.d. $\mathbb{E} \xi_i = 0$.

$X_n = \xi_1 + \dots + \xi_n$ - adapted

$$\mathbb{E}_n X_{n+1} = \mathbb{E}_n (X_n + \xi_{n+1}) = \mathbb{E}_n X_n + \mathbb{E}_n \xi_{n+1} = X_n + \mathbb{E} \xi_{n+1} = X_n \quad \checkmark$$

Ex. X_1, \dots, X_N independent, $X_i \geq 0$, $\mathbb{E} X_i = 1$.

$M_0 = 1$, $M_n = X_1 \cdot \dots \cdot X_n$ - adapted

$$\mathbb{E}_n M_{n+1} = \mathbb{E}_n (M_n \cdot X_{n+1}) = M_n \mathbb{E}_n X_{n+1} = M_n \mathbb{E} X_{n+1} = M_n \cdot 1 = M_n \quad \checkmark$$

Ex. X is a r.v. $X_n = \mathbb{E}_n X$ - adapted.

$$\mathbb{E}_n X_{n+1} = \mathbb{E}_n (\mathbb{E}_{n+1} X) = \mathbb{E}_n X = X_n \quad \checkmark$$

Thm 4.57 X_n is the wealth of a portfolio at $t=n$. The portfolio is self-financing iff. $D_n X_n$ is a MRT under $\tilde{\mathbb{P}}$.

Recall: self-financing, if \exists adapted Δ_n st. $X_{n+1} = \Delta_n S_{n+1} + (1+r)(X_n - \Delta_n S_n)$

pf. (\Rightarrow)

$$\begin{aligned} \tilde{\mathbb{E}}_n(D_{n+1} X_{n+1}) &= \tilde{\mathbb{E}}_n(D_{n+1} (\Delta_n S_{n+1} + (1+r)(X_n - \Delta_n S_n))) & D_{n+1} &= (1+r)^{-(n+1)} \\ &= \tilde{\mathbb{E}}_n(\Delta_n D_{n+1} S_{n+1}) + \tilde{\mathbb{E}}_n(D_n (X_n - \Delta_n S_n)) \\ &= \Delta_n \tilde{\mathbb{E}}_n(D_{n+1} S_{n+1}) + D_n X_n - \Delta_n D_n S_n \\ &= \cancel{\Delta_n D_n S_n} + D_n X_n - \cancel{\Delta_n D_n S_n} \end{aligned}$$

(\Leftarrow)

Define $\Delta_n = \frac{X_{n+1} - (1+r)X_n}{S_{n+1} - (1+r)S_n}$. Δ_n is \mathcal{F}_{n+1} -measurable.

Goal: show Δ_n is \mathcal{F}_n -measurable.

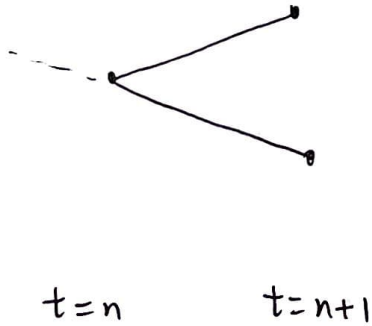
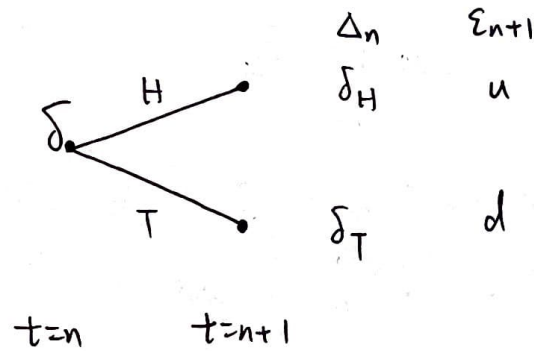
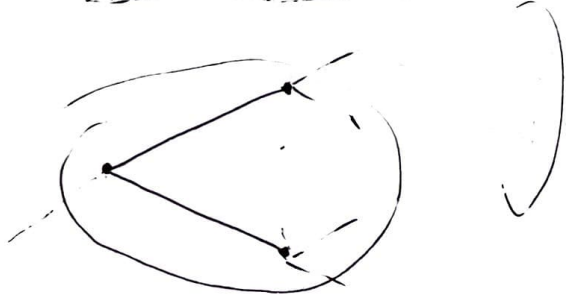
$$\Delta_n (S_{n+1} - (1+r)S_n) = X_{n+1} - (1+r)X_n.$$

$$\begin{aligned} \tilde{\mathbb{E}}_n(\Delta_n (S_{n+1} - (1+r)S_n)) &= \tilde{\mathbb{E}}_n(X_{n+1} - (1+r)X_n) = \tilde{\mathbb{E}}_n X_{n+1} - (1+r)\tilde{\mathbb{E}}_n X_n \\ &= \underbrace{(1+r)^{n+1}}_{\substack{\parallel \\ 1+r}} \tilde{\mathbb{E}}_n(D_{n+1} X_{n+1}) - (1+r)X_n = \underbrace{(1+r)^{n+1}}_{\substack{\parallel \\ 1+r}} D_n X_n - (1+r)X_n = 0. \end{aligned}$$

$$\text{Let } S_{n+1} = S_n \varepsilon_{n+1}, \quad \varepsilon_{n+1} = \begin{cases} u, & \text{if } W_{n+1} = H \\ d, & \text{if } W_{n+1} = T \end{cases}$$

$$0 = \mathbb{E}_n (\Delta_n (S_{n+1} - (1+r)S_n)) = \mathbb{E}_n (\Delta_n (S_n \varepsilon_{n+1} - (1+r)S_n)) = S_n \mathbb{E}_n (\Delta_n (\varepsilon_{n+1} - (1+r))).$$

$$\Rightarrow \mathbb{E}_n (\Delta_n (\varepsilon_{n+1} - (1+r))) = 0.$$



$$0 = \tilde{p} \cdot \delta_H (u - (1+r)) + \tilde{q} \cdot \delta_T (d - (1+r))$$

$$\Rightarrow \tilde{p} \delta_H \frac{(u - (1+r))}{u-d} = \tilde{q} \delta_T \frac{((1+r) - d)}{u-d}$$

$$\Rightarrow \tilde{p} \delta_H \tilde{q} = \tilde{q} \delta_T \tilde{p} \Rightarrow \delta_H = \delta_T =: \delta$$

$\Rightarrow \Delta_n$ is \mathcal{F}_n -measurable. □

• Intuition of Lebesgue Integral.

Discrete case: $\mathbb{E}X = \sum_{\omega \in \Omega} X(\omega) p(\omega) = \sum_{x \in \text{Range}(X)} x \mathbb{P}(X=x)$

General case: $\mathbb{E}X = \int_{\Omega} X(\omega) d\mathbb{P}(\omega) \stackrel{\text{if } X \text{ has pdf } p}{=} \int_{\mathbb{R}} x p(x) dx$

Let Δx be sufficiently small.

$$\mathbb{R} = \bigcup_{k=-\infty}^{\infty} (k\Delta x, (k+1)\Delta x]$$

$$\Omega_k = \{k\Delta x < X \leq (k+1)\Delta x\} = \{\omega \in \Omega \mid k\Delta x < X(\omega) \leq (k+1)\Delta x\}. \quad k \in \mathbb{Z}$$

$$\int_{\Omega} X(\omega) d\mathbb{P}(\omega) = \sum_{k \in \mathbb{Z}} \int_{\Omega_k} X(\omega) d\mathbb{P}(\omega) \approx \sum_{k \in \mathbb{Z}} \int_{\Omega_k} k\Delta x d\mathbb{P}(\omega) = \sum_{k \in \mathbb{Z}} k\Delta x \cdot \mathbb{P}(\Omega_k)$$

$$= \sum_{k \in \mathbb{Z}} k\Delta x \mathbb{P}(k\Delta x < X \leq (k+1)\Delta x) = \sum_{k \in \mathbb{Z}} k\Delta x \int_{k\Delta x}^{(k+1)\Delta x} p(x) dx$$

$$\approx \sum_{k \in \mathbb{Z}} \int_{k\Delta x}^{(k+1)\Delta x} x p(x) dx = \int_{\mathbb{R}} x p(x) dx.$$

• Independence lemma. (Discrete)

X is independent of \mathcal{F}_n , Y is \mathcal{F}_n -measurable, $f: \mathbb{R}^2 \rightarrow \mathbb{R}$

$$\mathbb{E}_n f(X, Y) = \sum_{x \in \text{Range}} f(x, Y) P(X=x) = g(Y)$$

$$g(y) = \mathbb{E} f(X, y) = \sum_{x \in \text{Range}} f(x, y) P(X=x).$$

• Independence lemma. (Continuous)

X is independent of \mathcal{F}_t , Y is \mathcal{F}_t -measurable, $f: \mathbb{R}^2 \rightarrow \mathbb{R}$ "nice"

$$\mathbb{E}_t f(X, Y) = g(Y), \quad g(y) = \mathbb{E} f(X, y)$$

$$\text{If } X \text{ has pdf } p_X(x), \quad g(y) = \int_{\mathbb{R}} f(x, y) p_X(x) dx$$

$$\mathbb{E}_t f(X, Y) = \int_{\mathbb{R}} f(x, Y) \tilde{p}_X(x) dx.$$

• Moment Generating Function (MGF).

X is r.v. $M_X: \mathbb{R} \rightarrow [0, \infty]$, $M_X(t) = \mathbb{E} e^{tX}$.

$X \sim N(0, 1)$, pdf: $p(x) = \frac{1}{\sqrt{2\pi}} e^{-x^2/2}$.

$$M_X(t) = \int_{\mathbb{R}} e^{tx} \cdot \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx = \frac{1}{\sqrt{2\pi}} \int_{\mathbb{R}} e^{\frac{-x^2/2 + tx - t^2/2 + t^2/2}{1}} dx = \frac{e^{t^2/2}}{\sqrt{2\pi}} \int_{\mathbb{R}} e^{-\frac{(x-t)^2}{2}} dx$$

$$f(z=x-t) = \frac{e^{-t^2/2}}{\sqrt{2\pi}} \int_{\mathbb{R}} e^{-\frac{z^2}{2}} dz = e^{-t^2/2}.$$

$$X \sim N(0,1), M_X(t) = e^{t^2/2}.$$

$$\mathbb{E} e^{tX} = M_X(t)$$

$$M_X'(t) = \mathbb{E}(X e^{tX})$$

$$M_X''(t) = \mathbb{E}(X^2 e^{tX})$$

⋮

$$M_X^{(n)}(t) = \mathbb{E}(X^n e^{tX})$$

$$t=0, M_X^{(n)}(0) = \mathbb{E} X^n.$$

Ex. $X, Y \sim N(0,1)$, independent. Compute $\mathbb{E} e^{\alpha XY}$, $\alpha \in \mathbb{R}$.

$$\sigma(X) = \{ \{X \in A\} \mid A \subseteq \mathbb{R}, (A \text{ "nice"}) \}$$

$$\mathbb{E} e^{\alpha XY} = \mathbb{E} \left(\mathbb{E}(e^{\alpha XY} \mid \sigma(X)) \right)$$

$$\mathbb{E}(e^{\alpha XY} \mid \sigma(X)) = g(X), \quad g(x) = \mathbb{E} e^{\alpha x Y} = M_Y(\alpha x) = e^{(\alpha x)^2/2}$$

$$\begin{aligned}\mathbb{E} e^{\alpha XY} &= \mathbb{E} g(X) = \mathbb{E} e^{\frac{1}{2}\alpha^2 X^2} = \int_{\mathbb{R}} e^{\frac{1}{2}\alpha^2 x^2} \cdot \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx \\ &= \frac{1}{\sqrt{2\pi}} \int_{\mathbb{R}} e^{-\frac{1}{2}(1-\alpha^2)x^2} dx.\end{aligned}$$

$$1^\circ |\alpha| \geq 1, \quad 1-\alpha^2 \leq 0 \Rightarrow \mathbb{E} e^{\alpha XY} = \infty.$$

$$2^\circ |\alpha| < 1, \quad 1-\alpha^2 > 0, \quad z = \sqrt{1-\alpha^2} x$$

$$\mathbb{E} e^{\alpha XY} = \frac{1}{\sqrt{2\pi}} \int_{\mathbb{R}} e^{-\frac{1}{2}z^2} \frac{dz}{\sqrt{1-\alpha^2}} = \frac{1}{\sqrt{1-\alpha^2}}.$$

Ex. Best Estimate in Mean Square Error.

X is r.v., for every \mathcal{F}_t -measurable r.v. Y ,

$$\mathbb{E} (X - \mathbb{E}_t X)^2 \leq \mathbb{E} (X - Y)^2.$$

$$\text{pf. } \mathbb{E} (X - Y)^2 = \mathbb{E} (X - \mathbb{E}_t X + \mathbb{E}_t X - Y)^2$$

$$= \underbrace{\mathbb{E} (X - \mathbb{E}_t X)^2}_I + 2 \underbrace{\mathbb{E} ((X - \mathbb{E}_t X)(\mathbb{E}_t X - Y))}_{II} + \underbrace{\mathbb{E} (\mathbb{E}_t X - Y)^2}_{III}.$$

$$III \geq 0, \quad II = \mathbb{E} ((X - \mathbb{E}_t X)(\mathbb{E}_t X - Y))$$

$$= \mathbb{E} \left(\mathbb{E} ((X - \mathbb{E}_t X)(\mathbb{E}_t X - Y) \mid \mathcal{F}_t) \right)$$

$$\begin{aligned}
\text{II} &= \mathbb{E} \left((\mathbb{E}_t X - Y) \mathbb{E} (X - \mathbb{E}_t X \mid \mathcal{F}_t) \right) \\
&= \mathbb{E} \left((\mathbb{E}_t X - Y) (\mathbb{E}_t X - \underline{\mathbb{E}_t(\mathbb{E}_t X)}) \right) \\
&= \mathbb{E} \left((\mathbb{E}_t X - Y) (\mathbb{E}_t X - \mathbb{E}_t X) \right) = 0.
\end{aligned}$$

$$\mathbb{E}(X - Y)^2 = \text{I} + 2\text{II} + \text{III} = \text{I} + \text{III} \Rightarrow \text{I} = \mathbb{E}(X - \mathbb{E}_t X)^2. \quad \square$$

Ex. X_1, \dots, X_n i.i.d. r.v., $S_n = X_1 + \dots + X_n$.

Find $\mathbb{E}(X_1 | S_n) = \mathbb{E}(X_1 | \sigma(S_n)) = g(S_n)$.

$$\mathbb{E}(X_i | S_n) = \mathbb{E}(X_j | S_n) \quad , 1 \leq i < j \leq n.$$

$$\mathbb{E}(X_1 | S_n) = \mathbb{E}(X_2 | S_n) = \dots = \mathbb{E}(X_n | S_n)$$

$$\begin{aligned}
n \mathbb{E}(X_1 | S_n) &= \mathbb{E}(X_1 | S_n) + \dots + \mathbb{E}(X_n | S_n) \\
&= \mathbb{E}(X_1 + \dots + X_n | S_n) = \mathbb{E}(S_n | S_n) = S_n
\end{aligned}$$

$$\Rightarrow \mathbb{E}(X_1 | S_n) = \frac{S_n}{n}.$$