

Def. Let $\mathcal{A} \subset \mathcal{B}$ - σ -algebras in Ω .

f - \mathcal{B} -measurable, $E(|f|) < \infty$. ($f \in L^1$)

Then $g = E(f|\mathcal{A})$ is the unique \mathcal{A} -measurable function
conditional expectation

Such that
$$\int_A g dP = \int_A f dP \quad \forall A \in \mathcal{A}.$$

Exists because $A \rightarrow \int_A f dP$ is an absolutely continuous measure on \mathcal{A} , so we can use Radon-Nykodim

Example For dyadic filtration,

$$E(f|\mathcal{B}_n) = \sum_{k=0}^{2^n-1} 2^n \left(\int_{k2^{-n}}^{(k+1)2^{-n}} f dx \right) \mathbb{1}_{[k2^{-n}, (k+1)2^{-n})}.$$

Some easy properties:

0) Linearity. $E(\alpha f + \beta g | \mathcal{A}) = \alpha E(f|\mathcal{A}) + \beta E(g|\mathcal{A})$

1) $\mathcal{A} \subset \mathcal{B} \Rightarrow E(E(f|\mathcal{B})|\mathcal{A}) = E(f|\mathcal{A})$
(compare \int_A of two sides).

2) If h is \mathcal{A} measurable, then

$E(hf|\mathcal{A}) = h E(f|\mathcal{A})$. In particular, $E(h|\mathcal{A}) = h$.

Proof. True for $h = \mathbb{1}_{A_0}$, $A_0 \in \mathcal{A}$, since $\forall A \in \mathcal{A}$, $A_0 \cap A \in \mathcal{A}$.
So true for step functions. Pass to the limit \equiv

3) If f is independent of \mathcal{A} then
 $E(f|\mathcal{A}) = E(f)$.

(Independent $\Rightarrow \forall A \in \mathcal{F} \int_A f = E(f)P(A)$)

Def Let $(X_t)_{t \in I}$ be a real-valued (\mathcal{F}_t) -adapted process.

$$\forall t \in I \quad E(|X_t|) < \infty.$$

X_t is called

$S < t$	<u>martingale</u>	$E(X_t \mathcal{F}_S) = X_S$
	<u>sub martingale</u>	if $\forall S \leq t \in I: E(X_t \mathcal{F}_S) \geq X_S$
	<u>super martingale</u>	$E(X_t \mathcal{F}_S) \leq X_S.$

martingale = sub + super.

Examples. 1) $f - \mathcal{F}_\infty$ -measurable, $E(|f|) < \infty$ (L^1).

Let $f_s := E(f | \mathcal{F}_s)$ - martingale.

$$E(f_t | \mathcal{F}_s) = E(E(f | \mathcal{F}_t) | \mathcal{F}_s) = E(f | \mathcal{F}_s) = f_s.$$

For dyadic:

$$f_n(x) = 2^n \int_{k2^{-n}}^{(k+1)2^{-n}} f(t) dt \quad \text{if } k2^{-n} \leq x < (k+1)2^{-n}.$$

2) Brownian Motion.

$$E(B_t | \mathcal{F}_s) = E(\underbrace{B_t - B_s}_{\substack{\text{independent} \\ \text{of } \mathcal{F}_s}} + B_s | \mathcal{F}_s) = E(B_s | \mathcal{F}_s) = B_s.$$

3) $B_t^2 - t$.

$$E(B_t^2 - t | \mathcal{F}_s) = E(\underbrace{(B_t - B_s)^2}_{\text{indep.}} + 2B_t B_s - t - B_s^2 | \mathcal{F}_s) =$$

$$t - s - t + B_s E(2B_t - B_s | \mathcal{F}_s) = 2B_s^2 - B_s^2 - s = B_s^2 - s.$$

$$4) Y_{t,s} = e^{(\alpha B_t - \frac{\alpha^2 t}{2})}$$

Markov

$$E(Y_{t,\lambda} | \mathcal{F}_s) = E\left(e^{\lambda(B_t - B_s) - \frac{\lambda^2}{2}(t-s)} \cdot e^{\lambda B_s - \frac{\lambda^2}{2}s} \mid \mathcal{F}_s\right) = e^{\lambda B_s - \frac{\lambda^2}{2}s} E\left(e^{\lambda(B_{t-s} - B_s) - \frac{\lambda^2}{2}(t-s)}\right) = Y_{s,\lambda}.$$

5) Super-example - first instance of Itô integral
Discrete time.

H_n -bounded predictable process (i.e. H_n is \mathcal{F}_{n-1} measurable)
 X_n -martingale.

Let

$$Y_0 = X_0, \quad Y_n = Y_{n-1} + H_n(X_n - X_{n-1}).$$

Then Y_n is also martingale.

Indeed: $E(Y_n | \mathcal{F}_{n-1}) = E(Y_{n-1} + H_n(X_n - X_{n-1}) | \mathcal{F}_{n-1}) =$

$$Y_{n-1} + H_n E((X_n - X_{n-1}) | \mathcal{F}_{n-1}) = Y_{n-1} \Rightarrow$$

Notation: $Y_n = (H \cdot X)_n.$

Corollary T -stopping time $X_n^T := X_{\min(n, T)}$ is a martingale

Proof. Let $H_n := \mathbb{1}_{T \geq n} = 1 - \mathbb{1}_{T \leq n-1}$ - predictable.

$$X_n^T = (H \cdot X)_n \quad \text{since} \quad X_n^T - X_{n-1}^T = \mathbb{1}_{T \geq n} (X_n - X_{n-1})$$

Thm (Discrete Optional Stopping Time Thm)

Let X_n be a discrete-time martingale,

$S \leq T$ - two bounded stopping times

Let $\widehat{\mathcal{F}}_S := \{A \in \mathcal{F}_\infty : \forall n \ A \cap \{S \leq n\} \in \mathcal{F}_n\}$.

Then $X_S = E(X_t | \widehat{\mathcal{F}}_S)$.

Proof. Let $M \geq T \geq S$, $M \in \mathbb{N}$. WLOG $X_0 = 0$ (subtract)

Let $H_n := \mathbb{1}_{T \geq n} - \mathbb{1}_{S \geq n}$.

Take $n > M$, then $(H \cdot X)_n = X_T - X_S$.

so $E(X_T) - E(X_S) = E((H \cdot X)_n) = E(X_0) = 0$.

Fix $B \in \widehat{\mathcal{F}}_S$. Consider $\begin{cases} \widetilde{S} := S \mathbb{1}_B \vee M \mathbb{1}_{B^c} \\ \widetilde{T} := T \mathbb{1}_B \vee M \mathbb{1}_{B^c} \end{cases}$ stopping times by definition of $\widehat{\mathcal{F}}_S$.

$\widetilde{S} \leq \widetilde{T} \leq M$ - stopping times.

so $E(X_T \mathbb{1}_B) = E(X_{\widetilde{T}}) - M P(B^c) = E(X_{\widetilde{S}}) - M P(B^c) = E(X_S \mathbb{1}_B)$.

Remark. Same way: if (X_t) - submartingale, (super-)

$$E(X_T | \widehat{\mathcal{F}}_S) \geq X_S$$

(\leq)