

Orthogonal Martingales

1. Let \mathcal{M}^2 be the set of L^2 martingales adapted to the filtration (\mathcal{F}_t) on the stochastic base,

$$(\Omega, \mathcal{F}_T, \mathbb{P}, (\mathcal{F}_t), [0, \infty)) \text{ "usual conditions"}$$

and such that $\sup_t \|X_t\|_2 < \infty$ for $X \in \mathcal{M}^2$. This means $X_t = M_t(X_\infty)$ for some X_∞ . By \mathcal{M}_0^2 we mean the subset of \mathcal{M}^2 comprising martingales, (X_t) , for which $X_0 = 0$.

There is a one to one correspondance between elements of \mathcal{M}_0^2 and elements, U , of $L^2(\Omega, \mathcal{F}_T, \mathbb{P})$ given by

$$L^2(\Omega, \mathcal{F}_T, \mathbb{P}) \ni U \longmapsto (M_t(U)) \in \mathcal{M}_0^2.$$

This correspondance, restricted to \mathcal{M}_0^2 , identifies \mathcal{M}_0^2 with the closed subspace of $L^2(\Omega, \mathcal{F}_T, \mathbb{P})$ comprising random variables with zero expectation.

In $L^2(\Omega, \mathcal{F}_T, \mathbb{P})$ there is a natural notion of orthogonality furnished by the inner product; for $U, V \in L^2$,

$$\langle U, V \rangle \equiv \int_{\Omega} UV \, d\mathbb{P} = 0.$$

Since each of U and V correspond to martingales in \mathcal{M}_0^2 it is natural to ask if this idea of orthogonality might be transferred to \mathcal{M}_0^2 and become a useful tool? It turns out that this idea has a limited use. However, there is a notion of orthogonality.

for elements of \mathcal{M}^2 , which is "far stronger" than the one discussed above, and this is a useful tool in Stochastic Calculus. Here it is,

1.1 Definition. Let $(X_t), (Y_t)$ belong to \mathcal{M}^2_0 . We say (X_t) and (Y_t) are (strongly) orthogonal if their 'product' is in \mathcal{M}^1_0 , that is,

$(X_t Y_t)$ is an L^1 -martingale adapted to $(\Omega, \mathcal{F}_T, \mathbb{P}, (\mathcal{F}_t), [0, T])$, and $X_0 Y_0 = 0$.

We write $(X_t) \perp (Y_t)$, for this relationship.

Let X be an L^2 -martingale and τ a stopping time. Suppose τ takes the values $0 \leq t_0 < t_1 < \dots < t_n \leq \infty$, so $\tau = \sum_{i=0}^n t_i I_{E_i}$, $E_i = \{\tau = t_i\} \in \mathcal{F}_{t_i}$.

$X_\tau = \sum_{i=0}^n X_{t_i} I_{E_i}$ and if $t_j \leq t < t_{j+1}$ then

$X_{\tau \wedge t} = \sum_{i=0}^j X_{t_i} I_{E_i} + X_t I_{\{\tau \geq t_{j+1}\}}$. Now let $t_k \leq s < t_{k+1}$ with $s < t$. Consider $M_s(X_{\tau \wedge t})$.

$$M_s(X_{\tau \wedge t}) = \sum_{i=0}^k X_{t_i} I_{E_i} + M_s\left(\sum_{i=k+1}^j X_{t_i} I_{E_i}\right) + M_s\left(X_t I_{\{\tau \geq t_{j+1}\}}\right)$$

Recall that $X_{t_i} = M_{t_i}(X_\infty)$ and since $E_i \in \mathcal{F}_{t_i}$ then, $X_{t_i} I_{E_i} = M_{t_i}(X_\infty I_{E_i})$

$$M_s\left(\sum_{i=k+1}^j X_{t_i} I_{E_i}\right) = \sum_{i=k+1}^j M_s(M_{t_i}(X_\infty I_{E_i})) = M_s\left(X_\infty \sum_{i=k+1}^j I_{E_i}\right)$$

$$\text{while } M_s(X_t I_{\{\tau \geq t_{j+1}\}}) = M_s(M_t(X_\infty) I_{\{\tau \geq t_{j+1}\}})$$

$$= M_s (M_t (X_\infty I_{\{\tau \geq t_{j+1}\}})) = M_s (X_\infty I_{\{\tau \geq t_{j+1}\}})$$

since $\{\tau \geq t_{j+1}\} = \Omega \setminus \{\tau \leq t_j\} \in \mathcal{F}_t$. So

$$M_s (X_{\tau \wedge t}) = \sum_{i=0}^k X_{t_i} I_{E_i} + M_s (X_\infty I_{\{\tau \geq t_{k+1}\}})$$

$$= \sum_{i=0}^k X_{t_i} I_{E_i} + M_s (X_\infty) I_{\{\tau \geq t_{k+1}\}}$$

as $\{\tau \geq t_{k+1}\} = \Omega \setminus \{\tau \leq t_k\} \in \mathcal{F}_s$.

$$= \sum_{i=0}^k X_{t_i} I_{E_i} + X_s I_{\{\tau \geq t_{k+1}\}} = X_{\tau \wedge s}$$

CB So $(X_{\tau \wedge t})$ is a martingale. This works for all τ .
Do an approximation proof. Note also $(X_{\tau \wedge t}) \approx (M_t(X_\tau))$

1.3 Lemma Suppose that $X, Y \in \mathcal{M}_0^2$ and for every stopping time of the filtration, typically τ , we have X_τ is orthogonal to Y_τ in $L^2(\Omega, \mathcal{F}_\infty, \mathbb{P})$. Then $X \perp Y$.

Proof

We have $\mathbb{E}(X_\tau Y_\tau) = \mathbb{E}((XY)_\tau) = 0$ for all stopping times. Let $t \in [0, \infty)^c$ and $E \in \mathcal{F}_t$. Let $\tau = t I_E + \infty I_{\Omega \setminus E}$. This

is a stopping time of the filtration:

$$\{\tau \leq s\} = \begin{cases} \emptyset & \text{if } s < t \text{ and } \emptyset \in \mathcal{F}_s \\ E & \text{if } t \leq s < \infty \text{ and } E \in \mathcal{F}_s \\ \Omega & \text{if } t = \infty \text{ and } \Omega \in \mathcal{F}_\infty \end{cases}$$

$$\text{Now } (XY)_\tau = X_t Y_t I_E + X_\infty Y_\infty I_{\Omega \setminus E}$$

$$\text{and } \mathbb{E}((XY)_\tau) = \mathbb{E}(X_\tau Y_\tau I_E) + \mathbb{E}(X_\infty Y_\infty I_{\Omega \setminus E}).$$

We also have,

$$\mathbb{E}(X_\infty Y_\infty) = \mathbb{E}(X_\infty Y_\infty I_E) + \mathbb{E}(X_\infty Y_\infty I_{\Omega \setminus E})$$

Now, by hypothesis,

$$0 = \mathbb{E}(X_\infty Y_\infty) = \mathbb{E}((XY)_\tau), \text{ which leads to}$$

$$\mathbb{E}(X_t Y_t I_E) = \mathbb{E}(X_\infty Y_\infty I_E) \quad \forall t \in \mathcal{F}_t.$$

Recalling the Radon-Nikodym theorem this amounts to the statement that

$$M_t(X_\infty Y_\infty) = X_t Y_t, \quad 0 \leq t \leq T.$$

$$\text{i.e. } XY \in \mathcal{M}^1.$$

The converse is also true, the proof I have uses optional stopping. See later!

Using the correspondence between $L^2(\Omega, \mathcal{F}_T, \mathbb{P})$ and \mathcal{M}^2 we can define a norm and inner product on \mathcal{M}^2 - we just pull them back from $L^2(\Omega, \mathcal{F}_T, \mathbb{P})$ onto \mathcal{M}^2 :

$$L^2(\Omega, \mathcal{F}_T, \mathbb{P}) \ni U \mapsto (M_t(U)) \in \mathcal{M}^2$$

$$\text{so } \|(M_t(U))\| = \|U\|_2 \text{ and}$$

$$\langle (M_t(U)), (M_t(V)) \rangle = \langle U, V \rangle.$$

Clearly, M^2 is a vector space and, with the definitions above, it is a Hilbert space.

Here is a key idea:

1.4 Definition A closed subspace, F , of M^2 is called stable if it is stable under stopping, that is,

If $(X_t) \in F$ and τ is a stopping time then $X^\tau \equiv (X_{\tau \wedge t}) \in F$.

Note that since $\mathcal{F}_0 = \{\Omega, \emptyset\}$ then for each $E \in \mathcal{F}_0$ and $(x_t) \in F$, $(x_t I_E)$ is also in F . In a more general setting this property would be part of the definition.

* Stopping a martingale gives a martingale.

There is a deep relationship between stopping and stochastic integration. One can take stopping as a departure point for developing stochastic integration. Because of this a subspace of M^2 which is stable under stopping is also stable under stochastic integration for suitable integrands. In fact there are a number of characterisations of stable subspaces:

1.5 Lemma.

The following are equivalent for a closed subspace, F , of M^2 .

* Borrowed from Philip Protter's book
(+) Random Times, Predictable.... Proc, London Math Soc (3) 67 (1993)

(i) F is a (closed) stable subspace of \mathcal{M}^2 .

(ii) For $t \in [0, T]$ and $E \in \mathcal{F}_t$ and $X \in F$ the process $(X - X^t)I_E$ belongs to F too.

(iii) If $X \in F$ and (h_s) is a bounded predictable process then

$\left(\int_0^t h_s dX_s \right)$
belongs to F too.

(iv) If $X \in F$ and (h_s) is a predictable process such that,

$$\mathbb{E} \left(\int_0^T h_s^2 d\langle X, X \rangle_s \right) < \infty$$

then

$\int_0^T h_s dX_s$
belongs to F too.

Proof Let $\tau = tI_E + \infty I_{\Omega \setminus E}$ where $E \in \mathcal{F}_t$.

For $X \in F$, $X^\tau \equiv (X_{\tau \wedge s})$ where

$$X_{\tau \wedge s} = \begin{cases} X_s & \text{if } s < t \\ X_t I_E + X_s I_{\Omega \setminus E} & \text{for } t \leq s \leq \infty \end{cases}$$

So $X_s - X_{\tau \wedge s} = \begin{cases} 0 & \text{if } s < t \\ (X_s - X_t) I_E & \text{for } t \leq s \leq \infty \end{cases}$

and this is $(X - X^\tau)I_E$, so it's in F . (i) \Rightarrow (ii)

A moment's thought shows that $(X - X^T)I_E$ is simply the process obtained from E a stochastic integral,

$$(X - X^T)I_E = \left(\int_0^s f(r) dX_r \right)$$

where f is the elementary process

$$f(s) = I_E I_{[t, \infty)}(s).$$

Because F is a subspace, linear combinations of such f 's can be assembled and we conclude

$$\left(\int_0^s f(r) dX_r \right) \in F$$

Whenever $f(s) = h I_{[t, \infty)}(s)$ and h

is a simple \mathcal{F}_t random variable. Now because F is closed the stochastic integrals of (appropriate) limit of these processes will be in F , that is,

$$\left(\int_0^s \hat{f}(r) dX_r \right) \in F$$

Whenever $f(s) = k I_{[t, \infty)}(s)$ and

k is a bounded \mathcal{F}_t random variable.

Now we let t vary and using linearity and the fact that F is a subspace we conclude that

$$\left(\int_0^s f(r) dX_r \right) \in F$$

Whenever f is a simple predictable process. Now the integral for a bounded predictable process is a limit at each time t , of the integrals of a sequence of simple predictable processes and the convergence is in L^2 norm. As a consequence

$$\left(\int_0^{\cdot} f \, dX_r \right) \in F$$

For every bounded predictable process, f So (ii) \Rightarrow (iii). A similar argument shows that (iii) \Rightarrow (iv). Now consider a time T , taking just two values, $0 \leq u < v < \infty$, on sets E and E^c . For any martingale $X \in F$,

$$X_T \equiv (X_{T \wedge u}) = \begin{cases} X_0 & \text{if } 0 \leq u \\ X_u I_E + X_u I_{E^c}, & u < v \\ X_u I_E + X_v I_{E^c} & v < \infty \end{cases}$$

So let f be the simple predictable process

$$f(t) = \begin{cases} I_{[0, u)} & \text{if } 0 \leq u \\ I_{[u, v)} - I_E & \text{if } u < v \\ 0 & \text{if } v < \infty \end{cases}$$

Then, as $X \in M^2$,

$$\int_0^s f(x) dx = \begin{cases} X_0 - X_s = X_s & \text{if } s \leq u \\ X_u + (X_0 - X_u) - (X_s - X_u) I_E & \text{if } u < s \leq v \\ X_u + (X_v - X_u) - (X_v - X_u) I_E & \text{if } v < s \end{cases}$$

$$= \begin{cases} X_s & \text{if } s \leq u \\ X_s - (X_0 - X_u) I_E = X_u I_E + X_s I_{s \in A} & \text{if } u < s \leq v \\ X_u I_E + X_v I_{s \in A} & \text{if } v < s \end{cases}$$

$$= X^\tau.$$

So (vi) \Rightarrow (i) in this case.

One can extend this result to times τ which take a finite number of distinct values. Moreover, any time τ is the limit of a decreasing sequence of such times. (Exercise). One can prove that for $X \in M^2$,

$$X_{\tau_n} \rightarrow X_\tau \text{ in } L^2 \text{ norm.}$$

All of which shows, (iv) \Rightarrow (i).

Let $\mathcal{A} \subseteq \mathcal{M}_0^2$. Then as \mathcal{M}_0^2 is stable under stopping and a closed subspace it follows that $\langle \mathcal{A} \rangle = \bigcap \{F : F \text{ is closed + stable, } F \supseteq \mathcal{A}\}$ exists. We call it the stable subspace, generate \mathcal{A} by \mathcal{A} . Let $\mathcal{A} = \{X\}$ where $X \in \mathcal{M}_0^2$. Consider $\langle X \rangle \equiv \langle \{X\} \rangle$. From the result above, part (iv) we know that for every predictable F such that $\mathbb{E}(\int_0^\infty f_{\tau}^2 d\langle X \rangle_\tau) < \infty$ the martin-

$$\text{gale, } (Z_t) = \left(\int_0^t f_{\tau} dX_\tau \right)$$

lies in $\langle X \rangle$. But as we saw in the proof of (iv) \Rightarrow (i), stopping can be expressed as stochastic integration. So stopping the martingale (Z_t) amounts to integrating $\mathbb{I}_{\tau \leq T} \mathbb{I}_{\tau \leq t}$ with respect to (Z_t) , that is

$$Z_{t \wedge T} = \int_0^{t \wedge T} (\mathbb{I}_{\tau \leq T} \mathbb{I}_{\tau \leq s}) dZ_s = \int_0^{t \wedge T} (\mathbb{I}_{\tau \leq T}) f_{\tau} dX_\tau$$

So, if you stop a martingale obtained by stochastic integration, it is itself a stochastic integral. This says that these stochastic integrals are stable under stopping! Therefore

$$\langle X \rangle = \left\{ \left(\int_0^t h_\tau dX_\tau \right) : \mathbb{E} \left(\int_0^\infty h_\tau^2 d\langle X \rangle_\tau \right) < \infty \right\}.$$

Let F be a stable subspace of M^2 . Write $F^\perp = \{N : N \perp K \forall K \in F\}$. So F^\perp is the strong orthogonal of F . We are going to look at the case where F is $\langle W \rangle$ (W Brownian Motion, and $(\Omega, \mathcal{F}_\infty, \mathbb{P}, (\mathcal{F}_t), [0, \infty)$) is the (completed) filtration given by Brownian Motion. Since W is not in M^2 we need to localize "everything" to $[0, T]$ for $T < \infty$. So our filtration is indexed by $[0, T]$, the definitions and results above carry over to this "smaller" filtration with appropriate modifications. We will write $M^2(T)$ to indicate the restriction to $[0, T]$. Before we proceed, a theorem about $\langle W \rangle$.

Theorem

Let F be a stable subspace of \mathcal{M}_0^2 and F^\perp its strong orthogonal. Then F^\perp is a closed stable subspace of \mathcal{M}_0^2 .

Prf Let $X \in F$ and $M, N \in F^\perp$ and $\alpha, \beta \in \mathbb{R}$.

(i) $X(\alpha M + \beta N) = \alpha XM + \beta XN$ a sum of martingales so $\alpha M + \beta N \in F^\perp$.

(ii) If $(X_n) \subset F^\perp$ and $X_n \rightarrow X$ in \mathcal{M}_0^2 then $X_\infty \rightarrow X_\infty$ in $\|\cdot\|_2$. So $X_n \rightarrow X_\infty$ in $\|\cdot\|_2$ for each t . Let $N_t^2 \in F$. Then $X_t N_t \rightarrow X_\infty N_t^2$.

$X_t N_t \rightarrow X_t N_t$ in L^1 norm for $t \in [0, \infty]$

Since $(X_t N_t)$ is an L^1 -martingale $(X_t N_t)$ is an L^1 -mart, so $X \in F^\perp$. So F^\perp is a closed subspace.

(iii) Let τ be a stopping time taking the distinct values t_0, t_1, \dots, t_n . We have already seen that for a martingale $N \in F^\perp$

$$N_{\tau \wedge t} = \sum_{i=0}^{j-1} N_{t_i} I_{E_i} + N_t I_{\{\tau \geq t_{j+1}\}}$$

where $t_j \leq t < t_{j+1}$ and $E_i = \{\tau = t_i\} \in \mathcal{F}_{t_i}$,

let $X \in F$. Then XN is a L^1 martingale and it is exactly $(M_t(X_{\infty} N_{\infty}))$. Let $t_k \leq \Delta < t_{k+1}$, $\Delta < t_j$,

$$M_{\Delta}(X_{\infty} N_{\infty}) = M_{\Delta}(X_{\infty} \sum_{t_i \leq \Delta} N_{t_i} I_{E_i}) + M_{\Delta}(X_{\infty} \sum_{t_{k+1} \leq \Delta} N_{t_i} I_{E_i}) + M_{\Delta}(X_{\infty} N_{\Delta} I_{\{\tau \geq t_{j+1}\}})$$

$$\text{Now, } M_{\Delta}(X_{\infty} \sum_{t_{k+1} \leq \Delta} N_{t_i} I_{E_i}) = \sum_{k+1}^j M_{\Delta}(X_{\infty} N_{t_i} I_{E_i})$$

$$= \sum_{k+1}^j M_{\Delta}(M_{t_i}(X_{\infty} N_{t_i} I_{E_i})) = \sum_{k+1}^j M_{\Delta}(M_{t_i}(X_{\infty}) N_{t_i} I_{E_i})$$

$$= \sum_{k+1}^j M_{\Delta}(X_{t_i} N_{t_i} I_{E_i}) = \sum_{k+1}^j M_{\Delta}(M_{t_i}(X_{\infty} N_{\infty}) I_{E_i})$$

$$= \sum_{k+1}^j M_{\Delta}(M_{t_i}(X_{\infty} N_{\infty} I_{E_i})) = \sum_{k+1}^j M_{\Delta}(X_{\infty} N_{\infty} I_{E_i}) .$$

Also

$$M_{\Delta}(X_{\infty} N_{\Delta} I_{\{\tau \geq t_{j+1}\}}) = M_{\Delta}(M_{\Delta}(X_{\infty} N_{\infty}) I_{\{\tau \geq t_{j+1}\}})$$

Since $\{\tau \geq t_{j+1}\} = \Omega \setminus \{\tau \leq t_j\} \in \mathcal{F}_{t_j}$ and $t \geq t_j$

$$= M_{\Delta}(M_{t_j}(X_{\infty} N_{\infty} I_{\{\tau \geq t_{j+1}\}})) = M_{\Delta}(X_{\infty} N_{\infty} I_{\{\tau \geq t_{j+1}\}})$$

$$\begin{aligned}
& \text{So, } M_\Delta \left(X_{\tau} \cdot \sum_{t_k \leq \tau} N_{t_k} I_{E_{t_k}} \right) + M_\Delta \left(X_{\tau} N_{\tau} I_{\{\tau \geq t_{j+1}\}} \right) \\
&= M_\Delta \left(X_{\infty} N_{\infty} \left(\sum_{k \geq j} I_{E_{t_k}} + I_{\{\tau \geq t_{j+1}\}} \right) \right) \\
&= M_\Delta \left(X_{\infty} N_{\infty} I_{\{\tau \geq t_{k+1}\}} \right) \text{ but (again)} \\
&\quad \{\tau \geq t_{k+1}\} \in \mathcal{F}_{t_k} \text{ and } t_k \leq \Delta, \text{ therefore} \\
&= M_\Delta \left(X_{\infty} N_{\infty} \right) I_{\{\tau \geq t_{k+1}\}} = X_\Delta N_\Delta I_{\{\tau \geq t_{k+1}\}}
\end{aligned}$$

While,

$$M_\Delta \left(X_{\tau} \sum_{t_k \leq \tau} N_{t_k} I_{E_{t_k}} \right) = M_\Delta \left(X_{\tau} \right) \sum_{t_k \leq \tau} N_{t_k} I_{E_{t_k}} = X_\Delta \sum_{t_k \leq \tau} N_{t_k} I_{E_{t_k}}$$

So $M_\Delta \left(X_{\tau} N_{\tau} \right) = X_\Delta N_{\tau}$ i.e. F^1 is stable under stopping at τ by times τ_n which take a finite number of values. Once again, any I is the limit of a decreasing sequence of times τ_n , taking only finitely many values, and

$$N_{\tau_n} \rightarrow N_{\tau} \text{ in } L^2$$

(and for each t , $N_{\tau_n t} \rightarrow N_{\tau t}$ in L^2) This shows that $(N_{\tau t})$ is the limit of a sequence in F^1 , which,

Since F^{-1} is closed, means $N_{\mathcal{F}} \in F^{-1}$, ie. F^{-1} is stable.

