

## The Doob-Meyer Decomposition

Let  $X = (X_n)$  be a stochastic process adapted to the stochastic base,  $(\Omega, \mathcal{F}, \mathbb{P}, (\mathcal{F}_n), n \in \mathbb{N})$ . We will assume that  $X$  is an  $L^1$ -process, that is,  $\forall n, X_n \in L^1(\Omega, \mathcal{F}_n, \mathbb{P}) \subseteq L^2(\Omega, \mathcal{F}_n, \mathbb{P})$ . As usual, we will take  $\mathcal{F} = \sigma(\bigcup_{n \in \mathbb{N}} \mathcal{F}_n)$ . As you know, there is a conditional expectation map,  $M_n$ , for each  $n \in \mathbb{N}$ .

$$M_n : L^2(\Omega, \mathcal{F}, \mathbb{P}) \rightarrow L^2(\Omega, \mathcal{F}_n, \mathbb{P})$$

With the usual properties. At this point it is worth remarking that the restriction of  $M_n$  to  $L^2(\Omega, \mathcal{F}, \mathbb{P})$  is simply the orthogonal projection of  $L^2(\Omega, \mathcal{F}, \mathbb{P})$  onto the closed subspace  $L^2(\Omega, \mathcal{F}_n, \mathbb{P})$ . In this case the conditional expectation has many wonderful properties. One of these, that is retained by

the expectation on  $L^2(\Omega, \mathcal{F}, \mathbb{P})$ , albeit in a weaker form, is the orthogonality of  $M_n(X_n)$  and  $M_{n-1}(X_n)$  for  $X_n \in L^2(\Omega, \mathcal{F}_n, \mathbb{P})$ . This orthogonality remains for  $X_n$  in  $L^1(\Omega, \mathcal{F}_n, \mathbb{P})$  in the form,

$$M_{n-1}(M_n(X_n)) = M_{n-1}(M_{n-1}(X_n)) = 0.$$

Of course we must have  $n \geq 2$  above, although by defining  $\mathcal{F}_0 = \{\Omega, \emptyset\}$  and setting  $M_0(x) = \mathbb{E}(x) \mathbb{I}_\Omega$  we can extend this to  $n \geq 1$ .

(†) See the notes; "Families of  $\sigma$ -fields, Theorem M1".  
 (\*)  $M_n^\perp = \mathbb{I} - M_n$

One consequence of this is that

$$M_1 ( M_1^T(X_2) + X_1 ) = X_1$$

indeed

$$M_2 ( M_2^T(X_3) + X_2 ) = X_2$$

and

$$M_3 ( M_3^T(X_4) + X_3 ) = X_3$$

and so on! Now it is easy to see that

$$M_1 ( M_1^T(X_2) + X_1 ) = X_1,$$

$$M_2 ( M_2^T(X_3) + M_1^T(X_2) + X_1 ) = M_1^T(X_2) + X_1,$$

and

$$M_n ( M_n^T(X_{n+1}) + M_{n-1}^T(X_n) + \dots + X_1 ) = M_{n-1}^T(X_n) + M_{n-2}^T(X_{n-1}) + \dots + M_1^T(X_2) + X_1.$$

indeed

What this is showing is that the process  $U$ , defined by,

$$U_1 = X_1$$

$$U_2 = M_1^T(X_2) + X_1$$

$$U_3 = M_2^T(X_3) + M_1^T(X_2) + X_1$$

$$\vdots$$

$$U_n = M_{n-1}^T(X_n) + M_{n-2}^T(X_{n-1}) + \dots + M_1^T(X_2) + X_1$$

is a martingale.

Notice that  $U$  is adapted, i.e.  $U_n \in \mathcal{L}(\Omega, \mathcal{F}_n, \mathbb{P})$ .

This observation also allows us to write the original process  $X$  in a particular way, starting slowly:

$$U_1 = X_1 \text{ so}$$

$$A_1 \triangleq X_1 - U_1 = 0,$$

hear with me! Now,

$$U_2 = M_T^1(X_2) + X_1 = X_2 - M_1(X_2) + X_1$$

so that

$$A_2 \triangleq X_2 - U_2 = M_1(X_2) - X_1.$$

More generally,

$$U_n = M_T^{n-1}(X_n) + M_T^{n-2}(X_{n-1}) + \dots + M_T^1(X_2) + X_1$$

$$= X_n - M_{n-1}(X_n) + M_T^{n-2}(X_{n-1}) + \dots + M_T^1(X_2) + X_1.$$

We are now going to rearrange the terms on the right side of the last equation as a prelude to writing  $X$  in the 'particular way' described above. What we do is to gather  $(M_T^{n-2}(X_{n-1}))$  together, this leaves  $-M_{n-2}(X_{n-1})$  'spare'. We put this together with  $X_{n-2}$  (from  $M_T^{n-3}(X_{n-2})$ ) leaving  $M_{n-3}(X_{n-2})$  'spare' and so on. Writing this out formally:

$$X_n - M_{n-1}(X_n) + M_T^{n-2}(X_{n-1}) + \dots + M_T^1(X_2) + X_1$$

$$= X_n - \{M_{n-1}(X_n) - X_{n-1}\} - \{M_{n-2}(X_{n-1}) - X_{n-2}\} - \dots - \{M_1(X_2) - X_1\} + X_1$$

So that eventually we get

$$U_n = X_n - M_{n-1}^+(X_n) + M_{n-2}^+(X_{n-1}) + M_{n-3}^+(X_{n-2}) + \dots + M_1^+(X_2) + X_1$$

$$= X_n - \{M_{n-1}^+(X_n) - X_{n-1}\} - \{M_{n-2}^+(X_{n-1}) - X_{n-2}\} - \{M_{n-3}^+(X_{n-2}) - X_{n-3}\} - \dots$$

$$\dots - \{M_2^+(X_3) - X_2\} - \{M_1^+(X_2) - X_1\}.$$

Hence,

$$X_n = U_n + \sum_{k=1}^{n-1} \{M_k^+(X_{k+1}) - X_k\}$$

$$= U_n + A_n \quad \text{--- Equation 1}$$

where  $A_n = \sum_{k=1}^{n-1} \{M_k^+(X_{k+1}) - X_k\}$ . Notice that  $A_n \in \mathcal{F}(A_n)$

is a predictable process, that is,  $A_n \in \mathcal{L}(\Omega, \mathcal{F}_{n-1}, \mathbb{P})$ , it is adapted "one time step earlier" than  $(X_n)$ .

Equation 1 is the Doob-Meyer decomposition of the process  $X$ . Should  $X$  be such that

$$\sup_n \|A_n\|_1 = \sup_n \sum_{k=1}^{n-1} \|M_k^+(X_{k+1}) - X_k\|_1 < \infty$$

We can describe  $X$  as the sum of an  $L^1$ -martingale and an  $L^1$ -bounded process. This is a matter we may take further.....

### The Doob-Meyer Decomposition

Let  $(\Omega, \mathcal{F}, \mathbb{P}, (\mathcal{F}_n), n \in \mathbb{N}_0)$  be a discrete (time-) parameter stochastic base. An adapted sequence,  $(A_n)$ , is increasing if  $\mathbb{P}$ -a.s. we have  $0 = A_0 \leq A_1 \leq A_2 \leq \dots$  and  $(A_n)$  is called integrable if  $\mathbb{E}(\sup A_n) < \infty$ . We will require, always, that  $\forall n, \mathbb{E}(A_n) < \infty$ . An adapted sequence is called predictable if  $\mathbb{E}_n \in \mathcal{M}(\mathcal{F}_{n-1})$  for  $n \geq 1$ . (as Define super as well)

Now let  $(X_n)$  be a submartingale adapted to this stochastic base. Recall that this means that

$$M_n(X_{n+1}) \geq X_n \text{ for } n = 0, 1, 2, \dots$$

By defining;

$$A_0 = 0, \quad A_{n+1} = A_n + M_n(X_{n+1}) - X_n, \quad n \geq 0$$

We see that,

(i)  $A_n \geq 0, n \geq 0,$

(ii)  $A_n \leq A_{n+1},$

(iii)  $(A_n)$  is predictable.

Moreover, defining  $U_n = X_n - A_n$ , we see that

$$M_n(U_{n+1}) = M_n(X_{n+1} - A_{n+1}) = M_n(X_{n+1}) - A_{n+1}$$

(A<sub>n</sub> predictable)

$$= M_n(X_{n+1}) - (A_n + M_n(X_{n+1}) - X_n)$$

$$= -A_n + X_n$$

$$= U_n.$$

$$+ N_0 \equiv \mathbb{N} \cup \{0\}$$

So  $(U^n)$  is a martingale. We can write

$$X = U + A$$

where  $U$  is a martingale and  $A$  is a predictable increasing process.

Predictability has an equivalent formulation:

An increasing process,  $A$ , is called natural if for every bounded martingale,  $Y$ , we have

$$\mathbb{E}(Y_n \Delta A_n) = \mathbb{E}\left(\sum_{k=1}^n Y_k \Delta A_k\right), \quad n \geq 1.$$

### Proposition 1

An increasing process,  $A$ , is predictable iff it is natural.

**PF** If  $A$  is natural and  $Y$  a bounded martingale. Then

$$\mathbb{E}\left((A_n - M_{n-1})(A_n) Y_n\right) = \mathbb{E}(A_n Y_n) - \mathbb{E}(A_{n-1} Y_{n-1})$$

$$= \mathbb{E}(A_n Y_n) - \mathbb{E}(A_n Y_{n-1}) + \mathbb{E}(A_{n-1} Y_{n-1})$$

$$- \mathbb{E}(A_{n-1} Y_{n-1})$$

$$= \mathbb{E}(A_n Y_n) - \mathbb{E}(A_n Y_{n-1}) - \mathbb{E}(A_{n-1} Y_{n-1})$$

because  $A$  is natural

$$= \mathbb{E}\left(\sum_{k=1}^n Y_k \Delta A_k\right) - \mathbb{E}\left(\sum_{k=1}^{n-1} Y_k \Delta A_k\right) - \mathbb{E}\left(\sum_{k=1}^{n-1} Y_k \Delta A_k\right)$$

$$= \mathbb{E}\left(\sum_{k=1}^n Y_k \Delta A_k\right) - \mathbb{E}\left(\sum_{k=1}^{n-1} Y_k \Delta A_k\right)$$

$$= 0$$

Since  $Y$  is arbitrary,  $\oplus A_n = M_{n-1}(A_n)$ .

$\oplus$  The dual of  $L^2$  is  $L^\infty$  and for  $z \in L^\infty$   $\|z\|_1 = \sup_{y \in L^2} |\int xy dP|$

So  $A$  is predictable. Conversely, suppose that  $A$  is predictable and consider for  $n \geq 1$ , and bounded mart  $Y$ ,  
is predictable and consider for  $n \geq 1$ , and bounded mart  $Y$ ,

$$E \left( \sum_{k=1}^n Y_{k-1} \Delta A_k \right) = \sum_{k=1}^n \left\{ E(Y_{k-1} A_k) - E(Y_{k-1} A_{k-1}) \right\}$$

$$= \sum_{k=1}^n \left\{ E(M_{k-1}(Y) A_k) - E(Y A_{k-1}) \right\}$$

$$= \sum_{k=1}^n \left\{ E(Y_{k-1} A_k) - E(Y_{k-1} A_{k-1}) \right\}$$

This is a telescoping sum,  $= E(Y_n A_n) - E(Y_0 A_0)$

$$= E(Y_n A_n) \text{ since } A_0 = 0$$

So  $A$  is natural. [CS 1999 Now discuss Elementary Stochastic]

We move now to the question of Doob-Meyer decomposition for continuous parameter processes. So, ...

### STOCHASTIC BASE OR

Definition 1.1 An adapted process,  $A$ , is called increasing if  $E A_0 = 0$ .

$$(i) A_0(\omega) = 0$$

$$(ii) t \mapsto A_t(\omega) \text{ is monotone increasing right continuous}$$

If, further,  $E(\text{Sup } A_t) < \infty$  then  $A$  is called integrable. We require, always,  $E(A_t) < \infty, t \in [0, \infty)$ .

(b) An increasing process is called natural if for every bounded martingale,  $(Y_t) \equiv Y$ , we have,  $E \left( \int_0^t Y_{s-} dA_s \right) = E \left( \int_0^t Y_{s-} dA_s \right)_{(0,t]}$  for every  $t \in (0, \infty)$ .

The integrals used in this definition are Lebesgue-Stieltjes integrals. For a bounded martingale  $Y$ , the paths  $t \mapsto Y_t(\omega)$  are B.V. bounded measurable functions, so these integrals are with respect to  $dA_t(\omega)$ .

Proposition 2

An increasing process  $A_t$  is natural iff

$$E(Y_t A_t) = E\left(\int_0^t Y_s dA_s\right),$$

for each bounded martingale  $Y$ .

PF

Let  $\pi = \{t_i\}$  partition  $[0, t]$ . For a bounded martingale  $Y$ , set

$$Y_\pi = \sum_{k=1}^n Y_{t_k} I_{(t_{k-1}, t_k]}.$$

Then,

$$E\left(\int_0^t Y_\pi dA_s\right) = E\left(\sum_{k=1}^n Y_{t_k} (A_{t_k} - A_{t_{k-1}})\right)$$

$$= E\left(\sum_{k=1}^n \left\{E(Y_{t_k} A_{t_k}) - E(Y_{t_k} A_{t_{k-1}})\right\}\right)$$

$$= E\left(\sum_{k=1}^n \left\{E(Y_{t_k} A_{t_k}) - E(M_{t_{k-1}}(Y_{t_k}) A_{t_{k-1}})\right\}\right) \quad (\#)$$

$$= E\left(\sum_{k=1}^n \left\{E(Y_{t_k} A_{t_k}) - E(Y_{t_k} A_{t_{k-1}})\right\}\right)$$

$$= E\left(\sum_{k=1}^n \left\{E(Y_{t_k} A_{t_k}) - E(Y_{t_{k-1}} A_{t_{k-1}})\right\}\right)$$



The last sum is a telescoping sum with value  $E(Y_n A_n)$ . Now as  $\text{mesh } \pi \rightarrow 0$  then  $Y_n \rightarrow Y(\omega)$  for each  $\omega \in [0, t]$  and  $\mathbb{F}$  almost every  $\omega \in \Omega$ . Using Bounded Convergence,

$$\int Y_n dA_n \xrightarrow{\|\pi\| \rightarrow 0} \int Y dA_n \quad \mathbb{F} \text{ a.s.}$$

and  $\text{Sup } \pi \left| \int Y_n dA_n \right| \leq \text{Sup}_{\omega \in [0, t]} |Y(\omega)| A_t(\omega)$

$$\leq K A_t(\omega) \quad \mathbb{F} \text{ a.s.}$$

where  $K$  is a bound for  $Y$ . Again bounded convergence shows,

$$E \left( \int Y_n dA_n \right) \xrightarrow{\|\pi\| \rightarrow 0} E \left( \int Y dA_n \right)$$

So,  $E \left( \int Y dA_n \right) = E(Y A_t)$ . Now, if  $A$  is natural then  $E \left( \int Y dA_n \right) = E \left( \int Y dA_0 \right) = E(Y A_t)$ . On the other hand if  $E \left( \int Y dA_n \right) = E(Y A_t)$  then, as we have seen, we must have

$$E \left( \int Y dA_n \right) = E \left( \int Y dA_0 \right)$$

So  $A$  is natural.

(\*) Since  $A^{t_{k-1}} \in L^2(\mathcal{F}_{t_{k-1}})$  and  $Y^k \in L^\infty(\mathcal{F}_{t_k})$  we cannot simply apply the theorem ( $M_2$ ) on expectations. Let  $A^{t_{k-1}} = A^{t_{k-1}}$  and in  $L^2$ . Then  $A^{t_k} \rightarrow A^{t_{k-1}}$   $\mathbb{F}$ -almost surely. So  $Y^k A^{t_{k-1}} \rightarrow Y^k A^{t_{k-1}}$  in  $L^1$ . ( $M_2$  in an  $L^1$  contradiction) But by  $M_2$ ,  $M^{t_{k-1}}(Y^k A^{t_{k-1}}) = M^{t_{k-1}}(Y^k A^{t_{k-1}}) = M^{t_{k-1}}(Y^k A^{t_{k-1}}) \rightarrow Y^k A^{t_{k-1}}$  in  $L^1$ . The two limits must agree, so,  $Y^k A^{t_{k-1}} = M^{t_{k-1}}(Y^k A^{t_{k-1}})$ .  $\mathbb{F}$  a.s. and in  $L^1$ .

Consider now the class  $\mathcal{T}$  of stopping times of the filtration  $(\mathcal{F}_t)_{t \in [0, \infty)}$  for which  $P\{T < \infty\} = 1$ . A right continuous process  $X$  is said to be of class  $\mathcal{D}$  if the family

$$\{X_T : T \in \mathcal{T}\}$$

is uniformly integrable. One variant this definition is: " $\mathcal{T}$  of stopping times,  $\sigma$ , of  $(\mathcal{F}_t)_{t \in [0, \infty)}$  if  $\{X_\sigma : \sigma \in \mathcal{T}\}$  is a right continuous process  $X$  of class  $\mathcal{D}$  if  $\{X_\sigma : \sigma \in \mathcal{T}\}$  is uniformly integrable, for every  $0 < \alpha < \infty$ ."

Theorem 1 With  $(\Omega, \mathcal{F}, P, (\mathcal{F}_t), [0, \infty))$  satisfying the usual conditions and  $X$  of class  $\mathcal{D}$ . Then  $X$  has a decomposition,

$$X_t = U_t + A_t$$

where  $(U_t)$  is a right continuous martingale and  $(A_t)$  an increasing process. The process  $(A_t)$  can be taken to be predictable and in this case it is unique (up to indistinguishability). Moreover, if  $X$  is of class  $\mathcal{D}$ , then  $(U_t)$  is a uniformly integrable martingale and  $(A_t)$  is integrable.

Pf

Uniqueness

Suppose that  $X_t = U_t + A_t = U'_t + A'_t$  are two decompositions, with  $(A_t)$  and  $(A'_t)$  both natural. The process  $B_t = A_t - A'_t = U'_t - U_t$  is of course of bounded variation and a martingale. Now,  $B$  is also a natural process in the sense of Proposition 2.

To see this let  $\pi^n$  be an increasing sequence of partitions of  $[0, t]$  with  $|\pi^n| \rightarrow 0$  as  $n \rightarrow \infty$ . Then, for a bounded martingale,  $(S_t^n)$ ,

$$\mathbb{E} \int_0^t S_{s-} dB_s = \lim_n \int_0^t S_{s-}^n dB_s \triangleq \lim_n \sum_{\pi^n} S_{t_i}^n \Delta B_{t_i} \quad (\text{BOUNDED (GMV?)})$$

So, by bounded convergence,

$$\mathbb{E} \left( \int_0^t S_{s-} dB_s \right) = \lim_n \mathbb{E} \left( \sum_{\pi^n} S_{t_i}^n \Delta B_{t_i} \right) = \mathbb{E} (S_t^B)$$

as  $(A_t)$  and  $(A_t')$  are natural.

Since  $\Delta B_{t_i}^n$  is a martingale difference;

$$\mathbb{E} ( \sum_{\pi^n} \Delta B_{t_i}^n ) = \mathbb{E} ( M_{t_i}^n ) = 0$$

$$= \mathbb{E} ( \sum_{\pi^n} M_{t_i}^n \Delta B_{t_i}^n ) = 0$$

So  $\int_0^t E(B_t S_t) = 0$ , for every bounded martingale  $(S_t)$ .

This shows that  $B_t = 0$  as for every  $t \in (0, \infty)$ .

Finally, as  $(A_t)$  and  $(A_t')$  are right continuous

then for each  $q \in \mathbb{Q}^+$  there is a null set  $N_q$ , with

$A_q = A_q'$  on  $\Omega \setminus E_q$ . Let  $E = \cup_{q \in \mathbb{Q}^+} E_q$ . It is a  $\mathbb{P}$ -null set.

On  $\Omega \setminus E$  we have  $A_q = A_q'$   $\forall q \in \mathbb{Q}^+$ . By

right continuity,  $A_t = A_t'$   $\forall t \in [0, \infty)$  on  $\Omega \setminus E$ . So they

are indistinguishable.

$\mathbb{Q}^+ = \mathbb{Q} \cap [0, \infty)$  They are already equal  $\mathbb{P}$ -a.s. for  $t=0$ , for  $t>0$ , look above.