

## Martingale representation theorem

$\Omega = C[0, T]$ ,  $\mathcal{F}_T$  = smallest  $\sigma$ -field with respect to which  $B_s$  are all measurable,  $s \leq T$ ,  $P$  the Wiener measure,  $B_t$  = Brownian motion  
 $M_t$  square integrable martingale with respect to  $\mathcal{F}_t$

Then there exists  $\sigma(t, \omega)$  which is

- 1 progressively measurable
- 2 square integrable
- 3  $\mathcal{B}([0, \infty)) \times \mathcal{F}$  mble

such that

$$M_t = M_0 + \int_0^t \sigma(s) dB_s$$

## Lemma

$\mathcal{A}$  = set of all linear combinations of random variables of the form

$$e^{\int_0^T h dB - \frac{1}{2} \int_0^T h^2 dt}, \quad h \in L^2([0, T])$$

$\mathcal{A}$  is dense in  $L^2(\Omega, \mathcal{F}_T, P)$

## Proof

Suppose  $g \in L^2(\Omega, \mathcal{F}_T, P)$  is orthogonal to all such functions

We want to show that  $g = 0$

By an easy choice of simple functions  $h$  we find that for any  $\lambda_1, \dots, \lambda_n \in \mathbb{R}$  and  $t_1, \dots, t_n \in [0, T]$ ,

$$E^P[ge^{\lambda_1 B_{t_1} + \dots + \lambda_n B_{t_n}}] = 0$$

lhs real analytic in  $\lambda$  and hence has an analytic extension to  $\lambda \in \mathbb{C}^n$

Since  $E^P[ge^{\lambda_1 B_{t_1} + \dots + \lambda_n B_{t_n}}]$  is analytic and vanishes on the real axis, it is zero everywhere. In particular

$$E^P[ge^{i(y_1 B_{t_1} + \dots + y_n B_{t_n})}] = 0$$

Suppose  $\phi \in C_0^\infty(\mathbb{R}^n)$

$$\hat{\phi}(y) = (2\pi)^{-n/2} \int_{\mathbb{R}^n} \phi(x) e^{-ix \cdot y} dx$$

Fourier inversion:

$$\phi(x) = (2\pi)^{-n/2} \int_{\mathbb{R}^n} \hat{\phi}(y) e^{ix \cdot y} dy$$

$$E^P[g\phi(B_{t_1}, \dots, B_{t_n})] = (2\pi)^{-n/2} \int_{\mathbb{R}^n} \hat{\phi}(y) E^P[e^{iy_1 B_{t_1} + \dots + y_n B_{t_n}}] dy = 0$$

Hence  $g$  is orthogonal to fns of form  $\phi(B_{t_1}, \dots, B_{t_n})$  where  $\phi \in C_0^\infty(\mathbb{R}^n)$   
Dense in  $L^2(\Omega, \mathcal{F}_T, P) \Rightarrow g = 0$

## Lemma

$F \in L^2(\Omega, \mathcal{F}_T, P)$  There exists a unique  $f(t, \omega)$  which is

- 1 progressively measurable
- 2 square integrable
- 3  $\mathcal{B}([0, \infty)) \times \mathcal{F}$  measurable

such that

$$F(\omega) = E[F] + \int_0^T f dB.$$

## Proof of Uniqueness

suppose

$$F = E[F] + \int_0^T f_1 dB = E[F] + \int_0^T f_2 dB$$

$$\Rightarrow \int_0^T (f_2 - f_1) dB = 0 \Rightarrow \int_0^T E[(f_2 - f_1)^2] dt = 0 \Rightarrow f_2 = f_1$$

## Proof of existence

First we prove it if  $F$  is of the form  $F = e^{\int_0^T h dB - \frac{1}{2} \int_0^T h^2 ds}$

Defining  $F_t = e^{\int_0^t h dB - \frac{1}{2} \int_0^t h^2 ds}$  gives

$$dF = hF dB, \quad F_0 = 1,$$

so

$$F_t = 1 + \int_0^t F_s h dB.$$

Plugging in  $t = T$  gives the result.

If  $F$  is a linear combination of such functions the result follows by linearity

## Proof of existence for $F \in L^2(\Omega, \mathcal{F}_T, P)$

$F_n \in L^2(\Omega, \mathcal{F}_T, P)$  with  $F_n \rightarrow F$  and

$$F_n = E[F_n] + \int_0^T f_n dB.$$

$E[F_n] \rightarrow E[F]$ , so wlog  $E[F_n] = E[F] = 0$

$$E[(F_n - F_m)^2] = \int_0^T E[(f_n - f_m)^2] dt \rightarrow 0 \quad \text{as } n, m \rightarrow \infty$$

$\Rightarrow f_n$  Cauchy in  $L^2([0, T] \times \Omega, dx \times dP)$ .

Let  $f$  be the limit. Taking limits we have

$$F = E[F] + \int_0^T f dB.$$

## Proof of the martingale representation theorem

By previous lemma, for each  $t$  we have  $\sigma_t(s, \omega)$  such that

$$M_t = E[M_t] + \int_0^t \sigma_t(s) dB_s$$

Let  $t_2 > t_1$

$$M_{t_1} = E[M_{t_2} \mid \mathcal{F}_{t_1}]$$

$$\int_0^{t_1} \sigma_{t_2}(s) dB_s = \int_0^{t_1} \sigma_{t_1}(s) dB_s$$

Uniqueness  $\Rightarrow \sigma_{t_1} = \sigma_{t_2}$

## Quadratic variation of $X_t = \int_0^t \sigma(s) dB_s$

$$e^{\lambda \int_0^t \sigma(s) dB_s - \frac{\lambda^2}{2} \int_0^t \sigma^2(s) ds} = \text{martingale}$$

$$E[e^{\lambda \int_{t_i}^{t_{i+1}} \sigma(s) dB_s - \frac{\lambda^2}{2} \int_{t_i}^{t_{i+1}} \sigma^2(s) ds} \mid \mathcal{F}_{t_i}] = 0$$

$$E[Z(t_i, t_{i+1}) \mid \mathcal{F}_{t_i}] = 0, \quad Z(t_i, t_{i+1}) = \left( \int_{t_i}^{t_{i+1}} \sigma(s) dB_s \right)^2 - \int_{t_i}^{t_{i+1}} \sigma^2(s) ds$$

$$E[\{Z(t_i, t_{i+1})\}^2 - 4 \left( \int_{t_i}^{t_{i+1}} \sigma^2(s) ds \right)^2 \mid \mathcal{F}_{t_i}] = 0$$

$$E\left[\left(\sum_i Z(t_i, t_{i+1})\right)^2\right] \leq 4E\left[\left(\sum_i \int_{t_i}^{t_{i+1}} \sigma^2(s) ds\right)^2\right] \rightarrow 0$$

$$\langle X_t, X_t \rangle = \int_0^t \sigma^2(s, \omega) ds$$



## Levy's Theorem

Let  $X_t$  be a process adapted to a filtration  $\mathcal{F}_t$  which

- 1 has continuous sample paths
- 2 is a martingale
- 3 has quadratic variation  $t$

Then  $X_t$  is a Brownian motion

## Proof of Levy's theorem

Enough to show that for each  $\lambda$ ,

$$E[e^{i\lambda(X_t - X_s)} \mid \mathcal{F}_s] = e^{-\frac{1}{2}\lambda^2(t-s)}$$

Call  $M_t = e^{i\lambda X_t + \frac{1}{2}\lambda^2 t}$ ,  $t_j = s + \frac{j}{2^n}(t - s)$

$$\begin{aligned} M_t - M_s &= \sum_{j=1}^{2^n} M_{t_j} - M_{t_{j-1}} \\ &= \sum_{j=1}^{2^n} i\lambda M_{t_{j-1}}(X_{t_j} - X_{t_{j-1}}) - \frac{1}{2}\lambda^2 M_{\xi_j}[(X_{t_j} - X_{t_{j-1}})^2 - (t_j - t_{j-1})] \end{aligned}$$

$$\begin{aligned} E[M_{t_{j-1}}(X_{t_j} - X_{t_{j-1}}) \mid \mathcal{F}_s] &= E[E[M_{t_{j-1}}(X_{t_j} - X_{t_{j-1}}) \mid \mathcal{F}_{t_{j-1}}] \mid \mathcal{F}_s] \\ &= E[M_{t_{j-1}} E[(X_{t_j} - X_{t_{j-1}}) \mid \mathcal{F}_{t_{j-1}}] \mid \mathcal{F}_s] = 0 \end{aligned}$$

## Proof of Levy's theorem

Fix  $m$ . Let  $\xi^m = \max\{\frac{i}{2^m} : \frac{i}{2^m} \leq \xi\}$

$$\lim_{n \rightarrow \infty} \sum_{j=1}^{2^n} M_{\xi_j^m} [(X_{t_j} - X_{t_{j-1}})^2 - (t_j - t_{j-1})] = 0$$

So we only have to show

$$\lim_{n \rightarrow \infty} \sum_{j=1}^{2^n} [M_{\xi_j^m} - M_{\xi_j}] (X_{t_j} - X_{t_{j-1}})^2 = 0$$

Would follow from

$$\lim_{n \rightarrow \infty} \sum_{j=1}^{2^n} (X_{\xi_j^m} - X_{\xi_j}) (X_{t_j} - X_{t_{j-1}})^2 = 0$$

Left hand side =  $t \lim_{n \rightarrow \infty} \max_{1 \leq j \leq 2^n} |X_{\xi_j^m} - X_{\xi_j}| = 0$  a.s.

Note the same proof gives

## Itô formula for semimartingales

Let  $M_t^1, \dots, M_t^d$  be martingales with respect to a filtration  $\mathcal{F}_t$ ,  $t \geq 0$ ,  $A_t^1, \dots, A_t^d$  adapted processes of bounded variation,  $X_t = x_0 + A_t + M_t$  where  $x_0 \in \mathcal{F}_0$ , and  $f(t, x) \in C^{1,2}$ . Then

$$\begin{aligned} f(t, X_t) &= f(0, X_0) + \int_0^t \frac{\partial f}{\partial t}(s, X_s) ds + \sum_{i=1}^d \int_0^t \frac{\partial f}{\partial x_i}(s, X_s) dA_s^i + dM_s^i \\ &\quad + \sum_{i,j=1}^d \int_0^t \frac{\partial^2 f}{\partial x_i \partial x_j}(s, X_s) d\langle M^i, M^j \rangle_s \end{aligned}$$

## Multidimensional Levy's theorem

Let  $M_t^1, \dots, M_t^d$  be continuous martingales with respect to a filtration  $\mathcal{F}_t$ ,  $t \geq 0$ , with

$$\langle M^i, M^j \rangle_t = \delta_{ij} t$$

Then  $M_t^1, \dots, M_t^d$  is a Brownian motion in  $\mathbb{R}^d$

## Time change

Let  $Y_t$  be a stochastic integral

$$Y_t = \int_0^t g ds + \int_0^t f dB$$

where  $f$  and  $g$  are adapted square integrable processes

Let  $c_t > 0$  be another adapted process and define

$$\beta_t = \int_0^t c_s ds.$$

Then  $\beta_t$  is adapted and strictly increasing. We call  $\alpha_t$  its inverse. We can check that

$$Y_{\alpha_t} = \int_0^t \frac{f}{c} ds + \int_0^t \frac{g}{\sqrt{c}} d\tilde{B}$$

for some Brownian motion  $\tilde{B}$ . In particular, if we are given a stochastic integral  $\int_0^t f dB$  we can choose  $f^2 = c$  as the rate of our time change and the resulting  $Y_{\alpha_t}$  is a Brownian motion

# Time change

## Theorem

Let  $B_t$  be Brownian motion and  $\mathcal{F}_t$  its canonical  $\sigma$ -field

Suppose that  $M_t$  is a square integrable martingale with respect to  $\mathcal{F}_t$

Let

$$M_t = M_0 + \int_0^t f(s)dB_s$$

be its representation in terms of Brownian motion. Suppose that  $f^2 > 0$  (i.e. its quadratic variation is strictly increasing)

Let  $c = f^2$  and define  $\alpha_t$  as above

Then  $M_{\alpha_t}$  is a Brownian motion

## Example. Stochastic growth model

$$dX = rXdt + \sigma\sqrt{X}dB$$

Solution is  $X_t = r\tau_t + B(\tau_t)$  where  $\tau'_t = X_t$

Because if

$$dY = rdt + \sigma dB$$

then by time change

$$X_t = Y_{\tau_t}$$

satisfies

$$dX = r\tau'_t dt + \sigma\sqrt{\tau'_t} dB$$