

correction - from last time

21/10/12

FTAP - Single time period

$\inf \mathbb{E} \exp\{-\theta \cdot (S_t - S_0)\}$  attained  $\Rightarrow \exists$  EMM  
 not attained  $\Rightarrow \exists$  arbitrage

We still need to show that we can't have both  $\exists$  EMM and  $\exists$  arbitrage. Suppose the contrary, that  $\mathbb{Q}$  is an EMM, and  $\theta$  is an arbitrage. Then  $\theta \cdot (S_t - S_0) \geq 0$  and  $P(\theta \cdot (S_t - S_0) > 0) > 0$ ; so  $\mathbb{Q}(\theta \cdot (S_t - S_0) > 0) > 0$

Now we have

$$\mathbb{E}^{\mathbb{Q}}(\theta \cdot (S_t - S_0)) = 0 \quad \text{because } S_t \text{ is a } \mathbb{Q} \text{ martingale} \Rightarrow \Leftarrow$$

Remarks from earlier

- (i) B.M. exists
- (ii)  $\forall c \neq 0$   $cB(\cdot/c)$  is a Brownian motion
- (iii) the process  $\tilde{B}_t = \begin{cases} tB(1/t) & t > 0 \\ 0 & t = 0 \end{cases}$  is a B.M.
- (iv) If  $\mathcal{F}_t^\circ = \sigma(\{B_u : u \leq t\})$ , and  $\mathcal{F}_t = \bigcap_{s \geq t} \mathcal{F}_s^\circ$ , then for any  $t$  finite  $(\mathcal{F}_t)$ -stopping time  $T$ , the process

$B_{t+T} - \frac{B_t}{T}$  is a B.M., independent of  $\mathcal{F}_t$

$$(v) P\left(\sup_t \frac{B_t}{t} = +\infty, \inf_t B_t = -\infty\right) = 1$$

- So no last visit to 0...

and in fact the Brownian motion is with probability 1 nowhere differentiable

(vi) The process  $B$  is a zero mean Gaussian Process with

$$\mathbb{E}(B_s B_t) = S_{\lambda} t$$

(vii) Easy to see that  $B_t^2 - t$  is a martingale

- but the converse is also true; if  $M$  is a continuous martingale  $M^2 - t$  is a martingale, then  $M$  is a B.M. (Levy)

(viii)  $\forall a \in \mathbb{R}$ ,  $\exp(a B_t - \frac{1}{2} a^2 t)$  is a martingale

(ix) If  $H_a = \inf \{t : B_t = a\}$ , then the density of

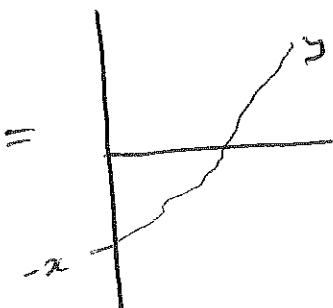
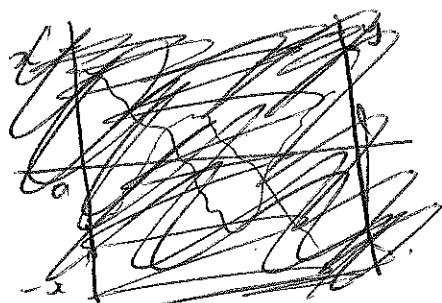
$$H_a \text{ is } \frac{P(H_a \in dt)}{dt} = \frac{ae^{-a^2/2t}}{\sqrt{2\pi t^3}}$$

(x) Write  $P_t(x, y) = \exp(-\frac{1}{2}(y-x)^2/t)/\sqrt{2\pi t}$

the transition density of the B.M. ( $P_t(x, y) dy$ )  
 $P(B_t \in dy | B_0 = x)$  ... we'll often call  $B_{\alpha t} B_t + a$  a Brownian motion when  $a \neq 0$ .)

We have by the reflection principle for  $x, y > 0$

$$P(B_t \in dy, B_s B_s > 0 \text{ for } s < t | B_0 = x) = \underbrace{\{P_t(x, y) - P_t(-x, y)\} dy}_{\text{in probability}}$$



$\rightarrow$   $* B_t^2 - t$  is a martingale

$* M_t^2 - [M]_t$

### Theorem

Suppose  $f: [0, \infty) \times \mathbb{R}^d \rightarrow \mathbb{R}$  is  $C^{1,2}$  satisfies an exponential growth bound; for some  $k > 0$

$$\sup_{t,x} e^{-k|x|} \left[ |f(t,x)| + \left| \frac{\partial f}{\partial t}(t,x) \right| + \sum_i \left| \frac{\partial f}{\partial x_i}(t,x) \right| \right.$$

~~$\left| \frac{\partial^2 f}{\partial x_i \partial x_j}(t,x) \right|$~~   $\left. + \sum_{ij} \left| \frac{\partial^2 f}{\partial x_i \partial x_j}(t,x) \right| \right] < \infty$

Define

$$(Lf)(t,x) = \frac{\partial f}{\partial t} + \frac{1}{2} \Delta f$$

Then

$$\mathbb{E}^x f(t, B_t) = f(0, x) + \mathbb{E}^x \left[ \int_0^t Lf(s, B_s) ds \right]$$

(Notation)  $\mathbb{E}^x(\cdot) = \mathbb{E}(\cdot | B_0 = x)$

Remark if  $Lf = 0$ , then we have  $f(t, B_t)$  is a mart.

- you check that  $f(t, x) \equiv |x|^2 - dt$  satisfies

$$Lf = 0$$

$$- f(t, x) = \exp \left( a \cdot x - \frac{1}{2} |a|^2 t \right)$$

Proof we have to observe that (just do  $d=1$  for simplicity)

$$-\frac{\partial}{\partial t} P_t(x,y) + \frac{1}{2} \frac{\partial^2}{\partial x^2} P_t(x,y) = 0$$

we'll consider for some  $\varepsilon > 0$  (fabini)

$$\mathbb{E}^x \left[ \int_{\varepsilon}^t Lf(s, B_s) ds \right] = \int_{\varepsilon}^t \mathbb{E}^x [Lf(s, B_s)] ds$$

$$= \int_{\varepsilon}^t \left\{ \int_{\mathbb{R}_s} P_s(x, y) Lf(s, y) dy \right\} ds$$

$$= \int_{\varepsilon}^t \left\{ \int_{\mathbb{R}_s} P_s(x, y) \left( \frac{\partial f}{\partial s} + \frac{1}{2} \frac{\partial^2 f}{\partial y^2} \right)(s, y) dy \right\} ds$$

Integration  
by  
parts

$$= \int_{\varepsilon}^t \left\{ \int_{\mathbb{R}_s} \left( P_s(x, y) \frac{\partial f}{\partial s} + \frac{1}{2} \frac{\partial^2 P_s}{\partial y^2} f(s, y) \right) dy \right\} ds$$

truncation @  $\varepsilon$

$$= \int_{\varepsilon}^t \left( \int \left\{ P_s(x, y) \frac{\partial f}{\partial s} + f(s, y) \frac{\partial P_s}{\partial s}(x, y) \right\} dy \right) ds$$

$$= \int \int_{\varepsilon}^t \frac{\partial}{\partial s} (P_s(x, y) f(s, y)) dy ds$$

$$= \int \left\{ P_t(x, y) f(t, y) - P_\varepsilon(x, y) f(\varepsilon, y) \right\} dy$$

$$= \mathbb{E}^x [f(t, B_t)] - \mathbb{E}^x [f(\varepsilon, B_\varepsilon)]$$

Now we let  $\varepsilon \downarrow 0$ : use dominated convergence and we get

$$\mathbb{E}^x [f(t, B_t)] - f(0, x) = \mathbb{E}^x \left[ \int_0^t (Lf)(s, B_s) ds \right]$$

Corollary for any  $f$  satisfying the exponential growth condition and also  $C^{1,2}$  property,

$f(t, B_t) - \int_0^t (Lf)(s, B_s) ds$  is a martingale

$$= f(0, B_0) + \int_0^t f'(s, B_s) dB_s$$

$$\frac{\partial f}{\partial x} \quad ; \quad \dot{f} = \frac{\partial f}{\partial t} \quad ; \quad f' = \frac{\partial f}{\partial x}$$