A standard Brownian motion on  $\mathbb{R}^+$  for a filtration  $\{\mathcal{F}_t : t \in \mathbb{R}^+\}$  is an adapted process for which

- (i) all sample paths are continuous
- (ii)  $X(0, \omega) = 0$  for all  $\omega$
- (iii) for each pair s, t with  $0 \le s \le t$ ,

$$X_t - X_s$$
 is  $N(0, t - s)$  distributed independent of  $\mathcal{F}_s$ 

Properties (ii) and (iii) together imply: for each  $0 \le t_1 \le t_2 \le ... \le t_k$ , the random vector  $(X_{t_1}, ..., X_{t_k})$  has a multivariate normal distribution with zero means and covariances given by

$$cov(X_s, X_t) = min(s, t)$$

Abbreviate  $\mathbb{P}(\ldots \mid \mathcal{F}_t)$  to  $\mathbb{P}_t(\ldots)$ .

## Useful facts: some rigorous proofs to follow

- (i) For a fixed  $\tau \ge 0$  define  $Z_t = B_{\tau+t} B_{\tau}$  for  $t \ge 0$ . Then Z is a standard Brownian motion independent of  $\mathcal{F}_{\tau}$ .
- (ii) (Strong Markov property) Same assertion as in (i) except that  $\tau$  is a stopping time.
- (iii) (Time reversal) Define  $Z_t = tB_{1/t}$  for t > 0, with  $Z_0 = 0$ . Then  $\{Z_t : t \in \mathbb{R}^+\}$  is a also a standard Brownian motion.
- (iv) Both  $\{(B_t, \mathcal{F}_t) : t \in \mathbb{R}^+\}$  and  $\{(B_t^2 t, \mathcal{F}_t) : t \in \mathbb{R}^+\}$  are martingales.
- (v) For each real  $\theta$ , the process  $Y_t = \exp(\theta X_t \frac{1}{2}\theta^2 t)$  is a martingale. (For complex  $\theta$ , would it be a complex martingale?)
- <1> Lévy's martingale characterization of Brownian motion. Suppose  $\{X_t : 0 \le t \le 1\}$  is a martingale with continuous sample paths and  $X_0 = 0$ . Suppose also that  $X_t^2 t$  is a martingale. Then X is a Brownian motion.

Heuristics of the proof that  $X_1 \sim N(0, 1)$ . The two martingale assuptions give two properties of the increment  $\Delta X = X_t - X_s$ , for s < t:

$$\mathbb{P}_s \Delta X = 0$$
 and  $\mathbb{P}_s (\Delta X)^2 = t - s$ .

Let f(x,t) be a smooth function of two arguments,  $x \in \mathbb{R}$  and  $t \in [0,1]$ . Define

$$f_x = \frac{\partial f}{\partial x}$$
 and  $f_{xx} = \frac{\partial^2 f}{\partial x^2}$  and  $f_t = \frac{\partial f}{\partial t}$ .

Let h = 1/n for some large positive integer n. Define  $t_i = ih$  for i = 0, 1, ..., n. Write  $\Delta_i X$  for  $X(t_i + h) - X(t_i)$ . Then

$$\mathbb{P}f(X_{1}, 1) - \mathbb{P}f(X_{0}, 0) \\
= \sum_{i < n} \left( \mathbb{P}f(X_{t_{i} + h}, t_{i} + h) - \mathbb{E}f(X_{t_{i}}, t_{i}) \right) \\
\approx \sum_{i < n} \mathbb{P}\left( (\Delta_{i}X) f_{x}(X_{t_{i}}, t_{i}) + \frac{1}{2} (\Delta_{i}X)^{2} f_{xx}(X_{t_{i}}, t_{i}) + h f_{t}(X_{t_{i}}, t_{i}) \right) \\
= \sum_{i < n} \left( 0 + \frac{1}{2} h \mathbb{P}f_{xx}(X_{t_{i}}, t_{i}) + h \mathbb{P}f_{t}(X_{t_{i}}, t_{i}) \right) \\
\approx \int_{0}^{1} \left( \frac{1}{2} \mathbb{P}f_{xx}(X_{s}, s) + \mathbb{P}f_{t}(X_{s}, s) \right) ds \quad \text{if } h \text{ is small} \quad \text{by } <2>.$$

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<2>

We need to formalize the passage to the limit to get

$$\mathbb{P}f(X_1, 1) - \mathbb{P}f(X_0, 0) = \int_0^1 \left( \frac{1}{2} \mathbb{P}f_{xx}(X_s, s) + \mathbb{P}f_t(X_s, s) \right) ds.$$

Specialize to the case  $f(x, s) = \exp(\theta x - \frac{1}{2}\theta^2 s)$ , with  $\theta$  a fixed constant. By direct calculation,

$$f_x = \theta f(x, s)$$
 and  $f_{xx} = \theta^2 f(x, s)$  and  $f_t = -\frac{1}{2}\theta^2 f(x, s)$ 

Thus

$$\mathbb{P}e^{\theta X_1}e^{-\theta^2/2}-1=\int_0^1 0\,ds=0.$$

That is,  $X_1$  has the moment generating function  $\exp(\theta^2/2)$ , which identifies it as having a N(0, 1) distribution.

As you will see, we are effectively proving a martingale central limit theorem. Look at the handout *martingaleCLT.pdf* before we start on a rigorous proof.

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