

1D Brownian motion.

Def 1D Brownian motion:

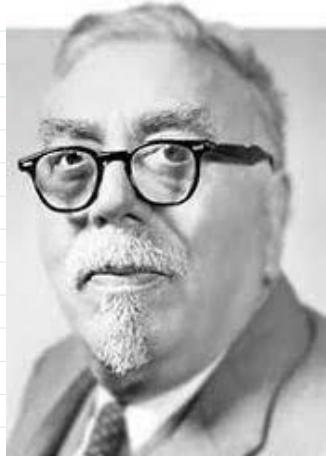
$\{B(t); 0 \leq t < \infty\}$ on $(\mathcal{R}, \mathcal{F}, P)$ (\mathcal{F} - σ -algebra, P -probability measure)

$B(t, \omega) : 0 \leq t < \infty, \omega \in \mathcal{R}$ s.t. s.f.s:

- a) $B(0, \omega) = 0 \quad \forall \omega$
- b) $B(t, \omega)$ - continuous in $t \quad \forall \omega$.
- c) $0 < t_1 < t_2 < \dots < t_n : B(t_1), B(t_2) - B(t_1), \dots, B(t_n) - B(t_{n-1})$ - independent, normal, mean zero variance: $t_1, t_2 - t_1, \dots, t_n - t_{n-1}$ - correspondingly.

$\left(c \right)$ is equivalent to requiring $E(B(t)) = 0 \quad \forall t,$
 $E(B(t)B(s)) = \min(t, s)$

Another point of view: Probability measure on $C([0, +\infty))$ such that $B(0) = 0$ a.s. and
c) is satisfied.



Norbert Wiener (1894-1964)

Theorem (Wiener) Brownian motion exists.

Properties:

1) (Uniqueness) Any two equivalent up to measure-preserving transformation between \mathcal{L} and \mathcal{L}_1 .

2) (Markov Property) $\Rightarrow (B_{t+s} - B_s)_{t \geq 0}$ - Brownian motion.

3) $-B_t$ - Brownian motion

4) (Brownian Scaling): $\forall c > 0: cB_{t/c^2}$ - Brownian motion.

5) (Time-reversal): Let $X_0 = 0, X_t = t B_{1/t}$. Then X_t - BM.

Proof Check covariance for $t > 0$. $E(t B_{1/t} \cdot s B_{1/s}) = st E(B_{1/t} B_{1/s}) = \min(s, t)$, continuity $t > 0$ - obvious.

For rational t , X_t distributed like B_t . So $\lim_{\substack{t \rightarrow 0 \\ t \in \mathbb{Q}}} X_t = \lim_{\substack{t \rightarrow 0 \\ t \in \mathbb{Q}}} B_t = 0$.

But since \mathbb{Q} is dense in \mathbb{R} , $\lim_{t \rightarrow 0} X_t = 0 = \lim_{t \rightarrow 0} B_t$.

5') $P(\lim_{t \rightarrow \infty} \frac{B(t)}{t} = 0) = 1$.

Proof: $\frac{B(t)}{t} = X\left(\frac{1}{t}\right)$.

Def Quadratic variation

Let (X_t) -real-valued process

X has a finite quadratic variation at time t if
for $\Delta_n = \{0 = t_0 < t_1 < \dots < t_n = t\}$ - partition of $[0, t]$.

$$|\Delta_n| = \max(t_i - t_{i-1}) \rightarrow 0. \quad \exists P\text{-lim}_{|\Delta_n| \rightarrow 0} \sum (X_{t_i} - X_{t_{i-1}})^2 = \langle X, X \rangle_t$$

notation.

$\langle X, X \rangle_t$ is an increasing (random) function.

Theorem $\langle B, B \rangle_t = t$.

We will use:



Gian Carlo Wick (1909-1992)

Wick's formula. Let X_1, \dots, X_n be centered normal variables.

$$\text{Then } E(X_1 \dots X_n) = \sum_{\substack{\text{perfect pairings} \\ \text{of } \{1, \dots, n\}}} E(X_{i_1} X_{j_1}).$$

$$\text{In particular: } E(X^4) = 3(E(X^2))^2$$

Proof of Wick's

Use characteristic functions.

If X is centered Normal, then $E(e^{tX}) = e^{-\frac{t^2\sigma^2}{2}}$,
where $\sigma^2 = E(X^2)$.

$$E(\prod e^{t_i X_i}) = E\left(e^{\sum t_i X_i}\right) = \exp\left(\frac{1}{2} E((\sum t_i X_i)^2)\right) = \exp\left(\frac{1}{2} \sum t_i t_j E(X_i X_j)\right)$$

$$\text{On one hand: } E(17 e^{t_i X_i}) = E(17((t_i X_i + t_i^2 X_i^2) + \dots)) =$$

$$1 + t_i t_{i-1} E(X_i) + \dots$$

On the other hand:

$$\exp\left(\frac{1}{2} \sum t_i t_j E(X_i X_j)\right) = \exp\left(\sum_{i < j} t_i t_j E(X_i X_j) + \frac{1}{2} \sum t_i^2 E(X_i^2)\right) =$$

$$\prod_{i < j} (1 + t_i t_j E(X_i X_j) + \dots) \prod (1 + \frac{1}{2} t_i^2 E(X_i^2) + \dots)$$

let us look for coefficient in front of $t_i t_{i-1} \dots t_n$.

Proof of Quadratic Variation:

$$E\left(\sum_i (B_{t_i} - B_{t_{i-1}})^2 - t\right)^2 = E\left(\left(\sum_i ((B_{t_i} - B_{t_{i-1}})^2 - (t_i - t_{i-1}))\right)^2\right) =$$

L²-norm!

$$E\left(\sum_i (B_{t_i} - B_{t_{i-1}})^4 - 2 \sum_i (B_{t_i} - B_{t_{i-1}})^2 (t_i - t_{i-1}) + \sum_i (t_i - t_{i-1})^2\right) = 3 \sum_i (t_i - t_{i-1})^2 - 2 \sum_i (t_i - t_{i-1})^2 + \sum_i (t_i - t_{i-1})^2 =$$

$$2 \sum_i (t_i - t_{i-1})^2 \leq 2 L_u |t| \rightarrow 0.$$

$\Leftrightarrow \sum_i (B_{t_i} - B_{t_{i-1}})^2 \xrightarrow{L^2} t \Rightarrow$ converges in Probability

Corollary: A.s., B.M has infinite variation on any interval $[a, b]$

Proof. Know: $\sum_i (B_{t_i} - B_{t_{i-1}})^2 \xrightarrow{P} \infty$ Select a subsequence so that converges a.s.

Observe: $\sum_i (B_{t_i} - B_{t_{i-1}})^2 \leq \sup_i |B_{t_i} - B_{t_{i-1}}| \cdot \text{Var}_a^{\infty} B$

Now pass to the limit: $\sup_i |B_{t_i} - B_{t_{i-1}}| \rightarrow 0$ (B-continuous).

So $(b-a) \leq 0 \cdot \text{Var}_a^{\infty} B$ if $\text{Var}_a^{\infty} B < \infty$ contradiction!

More is true (no proofs):

1) Thm (Levy): Let $h(t) = (2t \log \frac{1}{\epsilon})^{\frac{1}{2}}$.

Then $P\left(\lim_{t \rightarrow 0} \sup_{\substack{0 \leq t_1 \leq t_2 \leq 1 \\ |t_2 - t_1| < \epsilon}} \frac{|B_{t_2} - B_{t_1}|}{h(\epsilon)} = 1\right) = 1.$

2) Law of iterated logarithms:

$$\lim_{t \rightarrow 0} \frac{B_t}{\sqrt{2t \log \log \frac{1}{\epsilon}}} = 1 \text{ a.s.}$$

$$2') \lim_{t \rightarrow \infty} \frac{B_t}{\sqrt{2t \log \log t}} = 1 \text{ a.s. (apply 2) to } tB_{\frac{1}{t}}.$$

$$2'') \lim_{s \rightarrow 0} \frac{B_{t+s} - B_t}{\sqrt{2s \log \log s}} = 1 \text{ a.s. for given } t \text{ (by Markov).}$$