

Limit of Random Walks (functional central limit theorem)

$\{\xi_i\}_{i=1}^{\infty}$ i.i.d. $P(\xi_i = +1) = P(\xi_i = -1) = \frac{1}{2}$

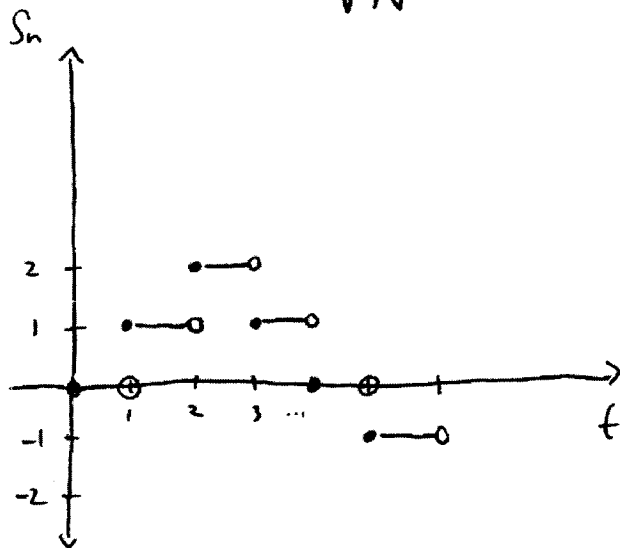
$S_n = \sum_{i=1}^n \xi_i, n=0, 1, 2, \dots$

by c.l.t.) $\frac{S_n}{\sqrt{n}} \Rightarrow X_1 \sim N(0, 1)$ as $n \rightarrow \infty$

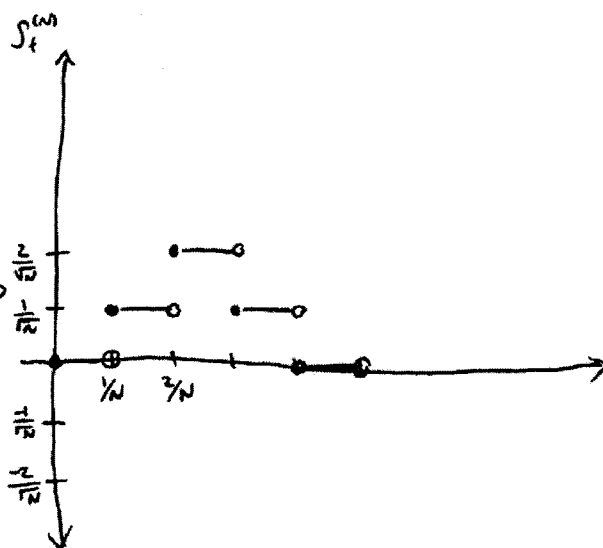
N pos integer,

$S_t^{(N)} = \frac{S_{\lfloor Nt \rfloor}}{\sqrt{N}}$

, $t \geq 0$ $\lfloor \cdot \rfloor$ integer part



→ Rescale →



$S_t^{(N)} = \{S_t^{(N)}, t \geq 0\} \Rightarrow \{B_t, t \geq 0\}$
as $N \rightarrow \infty$

1 dim standard Brownian Motion

[convergence in distribution]

In particular, $(S_{t_1}^{(N)}, S_{t_2}^{(N)}, \dots, S_{t_n}^{(N)}) \Rightarrow (B_{t_1}, \dots, B_{t_n})$

for any $0 \leq t_1 < t_2 < \dots < t_n < \infty$

recall: $B_t = B_t - B_0 \sim N(0, t)$

Binomial model in finance → Black-Scholes model

Defn A stochastic process $\{X(t), t \geq 0\}$ is a Gaussian process if for any $0 \leq t_1 < t_2 < \dots < t_n < \infty$, $(X_{t_1}, X_{t_2}, \dots, X_{t_n})$ has a multivariable normal distribution. Characterized by giving $\mu(t) = E(X_t)$, $t \geq 0$

Covariance Function:
 $R(s, t) = E[(X(s) - \mu(s))(X(t) - \mu(t))]$, $0 \leq s \leq t < \infty$

Brownian motion is a Gaussian Process

Let B be a standard 1-dim Brownian motion.

Fix $0 \leq t_1 < t_2 < \dots < t_n < \infty$

$$(B_{t_1}, B_{t_2}, \dots, B_{t_n}) = (B_{t_1} - B_{t_0}, (B_{t_2} - B_{t_1}) - (B_{t_1} - B_{t_0}), \dots, (B_{t_n} - B_{t_{n-1}}) + \dots + (B_{t_1} - B_{t_0}))$$

i.e. can write $(B_{t_1}, \dots, B_{t_n})$ in terms of linear transformation of

$$(B_{t_1} - B_0, B_{t_2} - B_{t_1}, \dots, B_{t_n} - B_{t_{n-1}}) \quad \text{where:}$$

$B_{t_1} - B_0, \dots, B_{t_n} - B_{t_{n-1}}$ are independent normal RV's with means of 0 and variances $t_1, t_2 - t_1, \dots, t_n - t_{n-1}$

Hint #3 on HW 2: $P(X_2 > 0 | X_1 > 0)$ where X is standard 1-dim BM
 $(X_2, X_1) = ((X_2 - X_1) + X_1, X_1)$ where $(X_2 - X_1, X_1)$ are indep N(0,1) RV.

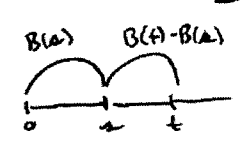
For BM:

$$\mu(t) = E[B(t)] = 0 \quad B(t) \sim N(0, t)$$

$$R(\Delta, t) = E[B(\Delta)B(t)] \quad 0 \leq \Delta \leq t < \infty$$

$$= E[B(\Delta)(B(t) - B(\Delta) + B(\Delta))] = E[B(\Delta)(B(t) - B(\Delta))] + E[B(\Delta)^2]$$

$$= E[B(\Delta)] E[B(t) - B(\Delta)] + \Delta = 0 + \Delta \Rightarrow \boxed{R(\Delta, t) = \Delta}$$



$$B(\Delta) \sim N(0, \Delta), \quad R(\Delta, t) = \Delta, \quad \Delta, t \geq 0$$

MARTINGALE PROPERTY

$\mathcal{F}_t = \sigma\{B_s, 0 \leq s \leq t\}$ = smallest σ -algebra containing $\sigma(B_s)$ for all $s \leq t$

$\{B_t, \mathcal{F}_t, t \geq 0\}$ is a continuous time Martingale

(i) B_t is \mathcal{F}_t -measurable for all $t \geq 0$

(ii) $E[|B_t|] < \infty$ since $B_t \sim N(0, t)$

(iii) $E[B_t | \mathcal{F}_s] = B_s \quad 0 \leq s < t < \infty$

To Prove (iii) Fix $0 \leq s < t < \infty$

$$E[B_t | \mathcal{F}_s] = E[(B_t - B_s) + B_s | \mathcal{F}_s]$$

linearity \rightarrow
$$= E[B_t - B_s | \mathcal{F}_s] + \underbrace{E[B_s | \mathcal{F}_s]}_{\hookrightarrow B_s}$$

$B_t - B_s$ indep of \mathcal{F}_s
& $B_s \in \mathcal{F}_s \rightarrow$

$$= E[B_t - B_s] + B_s$$

$$= 0 + B_s = B_s \quad B_t - B_s \sim N(0, t-s)$$

$$\therefore E[B_t | \mathcal{F}_s] = B_s$$

Exercise

$\{B_t^2 - t, \mathcal{F}_t, t \geq 0\}$ is a Martingale

(show $E[B_t^2 - t | \mathcal{F}_s] = B_s^2 - s$ for $0 \leq s < t < \infty$)

Thm: IF $X = \{X_t, t \geq 0\}$ is a 1-dim cts stochastic process s.t. $X_0 = 0$,

$\{X_t, \mathcal{F}_t, t \geq 0\}$ is a martingale & $\{X_t^2 - t, \mathcal{F}_t, t \geq 0\}$ is a martingale

where $\mathcal{F}_t = \sigma\{X_s: 0 \leq s \leq t\}$, $t \geq 0$

then X is a standard 1-dim Brownian Motion

MARKOV PROPERTY

$$P(B_t \in A \mid B_u : 0 \leq u \leq t) \text{ for } 0 \leq t < \infty$$

$$= P(B_t \in A \mid B_s) \text{ for any measurable set } A \text{ in } \mathbb{R}$$

In fact, enough to show

$$P(B_t \in A \mid B_{u_1} \in A_1, \dots, B_{u_n} \in A_n, B_s \in A') \text{ for } 0 \leq u_1 < u_2 < \dots < u_n < t$$

$$= P(B_t \in A \mid B_s \in A') \text{ for any meas. sets } A_1, \dots, A_n, A', A$$

$$\text{LHS} = \boxed{P(B_t \in A \mid B_{u_1} \in A_1, \dots, B_{u_n} \in A_n, B_s \in A')}$$

$$= P(B_t - B_s \in A - B_s \mid B_{u_1} \in A_1, \dots, B_{u_n} \in A_n, B_s \in A')$$

$$= P(B_t - B_s \in A - B_s \mid B_s \in A')$$

$$= \boxed{P(B_t \in A \mid B_s \in A')}$$