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Stock price + Portfolio Price.

Return to construction of stock price.

for any t we can write

$$S(t) \approx S(0) \left(1 + \frac{r}{N} + \sigma \frac{\epsilon_1}{\sqrt{N}}\right) \cdots \left(1 + \frac{r}{N} + \sigma \frac{\epsilon_n}{\sqrt{N}}\right)$$

for $\frac{n}{N} \leq t < \frac{n+1}{N}$.

$$P^*(\epsilon_i = 1) = P^*(\epsilon_i = -1) = 1/2.$$



$$\log \frac{S(t)}{S(0)} = \sum_{i=1}^n \frac{r}{N} - \frac{1}{2} \sigma^2 \sum_{i=1}^n \frac{\epsilon_i^2}{N} + \sigma \sum_{i=1}^n \frac{\epsilon_i}{\sqrt{N}}$$

$$\frac{r}{N} = rt - \frac{1}{2} \sigma^2 t + \sigma \sqrt{t} \sum_{i=1}^n \frac{\epsilon_i}{\sqrt{N}}$$

↑
 $Nt \sim n$

$$\sum_{i=1}^n \frac{\epsilon_i}{\sqrt{N}} \sim W$$

for $W \sim N(0, 1)$.

But.

$$\mathbb{E}^* \left\{ S(0) \left(1 + \frac{r}{N} + \sigma \frac{\varepsilon_1}{\sqrt{N}}\right) \dots \left(1 + \frac{r}{N} + \sigma \frac{\varepsilon_n}{\sqrt{N}}\right) \right\}$$

$$= S(0) \left(1 + \frac{r}{N}\right)^n \approx S(0) e^{rt}$$

By usual derivation in Binomial model.

\therefore Pass to the limit \rightarrow

$$\mathbb{E}^* \{ S(t) \} = S(0) e^{rt}$$

ie $e^{-rt} S(t)$ is a martingale.

BROWNIAN MOTION.

$$W_t \equiv \text{B.M.}$$

Properties:

$\frac{W_t}{\sqrt{t}}$ is $N(0,1)$ Random variable.

$$\mathbb{P} \left(a \leq \frac{W_t}{\sqrt{t}} \leq b \right) = \int_a^b \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx$$

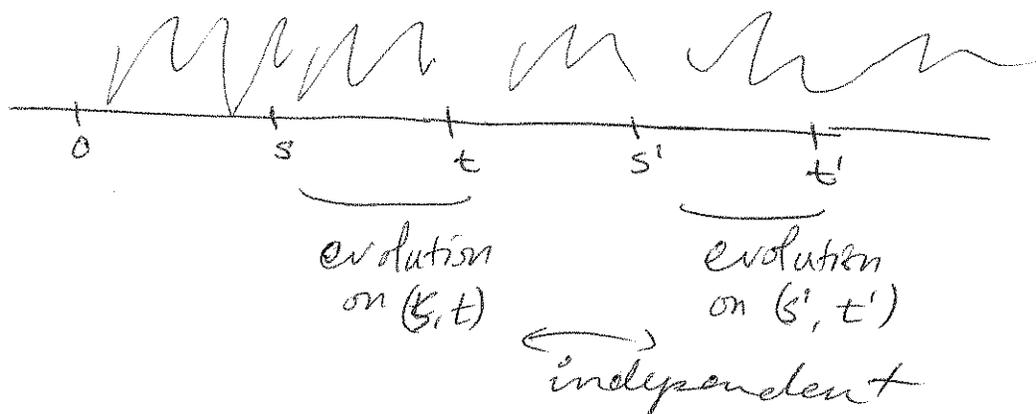
for $s < t < s' < t'$

$W_t - W_s$ is independent of $W_{t'} - W_{s'}$

AND,

$\frac{1}{\sqrt{t-s}}(W_t - W_s)$ is $N(0,1)$ Random Variable,

i.e., process 'starts over' at every time and forgets previous path - "Markov property"



Since

Since $\frac{1}{\sqrt{t-s}}(W_t - W_s)$ is $N(0, 1)$ +
 independent of path up to time s ,
 -ie independent of \mathcal{F}_s -info upto time s .

$$\begin{aligned} E(W_t | \mathcal{F}_s) &= E(W_t - W_s + W_s | \mathcal{F}_s) \\ &= \underbrace{E(W_t - W_s | \mathcal{F}_s)}_0 + \underbrace{E(W_s | \mathcal{F}_s)}_{W_s} \\ &= W_s. \end{aligned}$$

Fact: Gaussian processes (process such that time marginals are Gaussian).

are completely characterized by covariances,

$$\text{Cov}(W_t, W_s) = \min(s, t)$$

how do we see this?

Consider the limiting process ~~is~~ $s = \frac{n}{N}$, $t = \frac{m}{N}$, $s < t$
 $W_s \approx \sum_{i=1}^n \epsilon_i \frac{1}{\sqrt{N}}$ $W_t \approx \sum_{i=1}^m \epsilon_i \frac{1}{\sqrt{N}}$, $P(\epsilon_i = 1) = 1/2$,
 $P(\epsilon_i = -1) = 1/2$.

$$\begin{aligned} E(W_s W_t) &= E \sum_{i=1}^n \sum_{j=1}^m \epsilon_i \epsilon_j \frac{1}{N} = \frac{n}{N} = s \checkmark \\ E \epsilon_i \epsilon_j &= \begin{cases} 1 & i=j \\ 0 & i \neq j \end{cases} \end{aligned}$$

Path properties,

* W_t is continuous,

$$\lim_{s \rightarrow t} W_s = W_t$$

* \sqrt{t} fluctuations,

- for $\alpha \geq \frac{1}{2}$

← (blow up)

$$\sup_{0 \leq s < t \leq 1} \left| \frac{W_t - W_s}{(t-s)^\alpha} \right| = \infty$$

- for $\alpha < \frac{1}{2}$

$$\sup_{0 \leq s < t \leq 1} \left| \frac{W_t - W_s}{(t-s)^\alpha} \right| < \infty$$

"With probability 1",

* W_t is nowhere differentiable

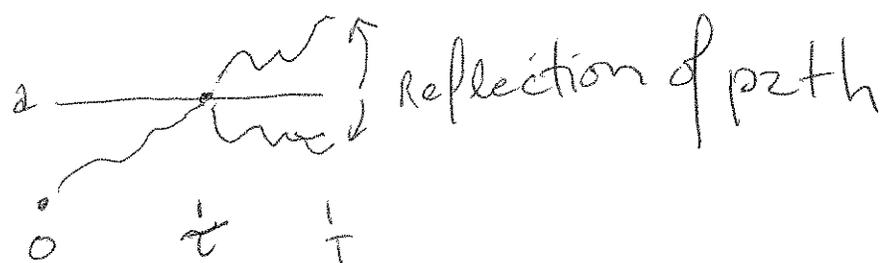
- holds since behavior at every point



Q: W_t starts @ 0 what is probability
 W_t hits $a > 0$ by time T ?

Reflection principle.

let $\tau = \min\{t: W_t \geq a\}$.



$$\text{let } U_t = \begin{cases} W_t, & t \leq \tau \\ a - W_t, & t > \tau \end{cases}$$

Then U_t is again a Brownian Motion -

Since $a - W_t$ still has Gaussian marginals & same covariance fn.

$$\therefore \cancel{P(U_T > a)} \cdot P(U_T > a) = P(W_T > a)$$

* But $U_T > a$ implies $W_T < a$

$$\therefore \{U_T > a\} \cup \{W_T > a\} = \{\tau < T\}.$$

$$P(\tau < T) = 2P(W_T > a)$$

$$\therefore \mathbb{P}(\tau < T) = 2 \mathbb{P}(W > a/\sqrt{T})$$

for $W \sim N(0,1)$

$$\mathbb{P}(W > a/\sqrt{T}) = \int_{a/\sqrt{T}}^{\infty} e^{-x^2/2} \frac{dx}{\sqrt{2\pi}}$$

$$f_{\tau}(T) = \frac{d}{dT} \mathbb{P}(\tau < T) = 2 \left(-\frac{e^{-x^2/2}}{\sqrt{2\pi}} \Big|_{a/\sqrt{T}} \right) \left(-\frac{a}{2T^{3/2}} \right)$$

$$= \frac{2}{\sqrt{2\pi}} \frac{a}{T^{3/2}} e^{-a^2/2T}$$