

Brownian motion

Definition (Brownian motion / Wiener process)

A process $\{W_t : t \geq 0\}$ on a probability space $(\Omega, \mathcal{F}, \mathbb{P})$ is called a **Brownian motion** if:

- (i) $\mathbb{P}(W_0 = 0) = 1$.
- (ii) **Independent increments:** if $0 = t_0 \leq t_1 \leq \dots \leq t_k$, then for any Borel sets $H_1, \dots, H_k$,

$$\mathbb{P}\left(W_{t_i} - W_{t_{i-1}} \in H_i \text{ for } i = 1, \dots, k\right) = \prod_{i=1}^k \mathbb{P}(W_{t_i} - W_{t_{i-1}} \in H_i).$$

- (iii) **Gaussian increments:** for $0 \leq s < t$, $W_t - W_s \sim \mathcal{N}(0, t - s)$.
- (iv) **Continuous paths:** for almost every ω , the map $t \mapsto W(t, \omega)$ is continuous.

Brownian motion: notes

Gaussian increment distribution: for $0 \leq s < t$ and a Borel set $H \subset \mathbb{R}$,

$$\mathbb{P}(W_t - W_s \in H) = \frac{1}{\sqrt{2\pi(t-s)}} \int_H \exp\left(-\frac{x^2}{2(t-s)}\right) dx.$$

Note: if $0 < t_1 < \dots < t_k$, then the joint density of $(W_{t_1}, W_{t_2} - W_{t_1}, \dots, W_{t_k} - W_{t_{k-1}})$ is the product of the corresponding normal densities.

Filtration associated with Brownian motion

Note:

1. Brownian motion is continuous.
2. Brownian motion is nowhere differentiable.

Filtration:

1. For $0 \leq s \leq t$, $\mathcal{F}_s \subseteq \mathcal{F}_t$.
2. $W(t)$ must be adapted to the filtration.

Brownian motion is a martingale

Let $\mathcal{F}_t := \sigma(W_s : s \leq t)$. Then for any $s < t$,

$$\mathbb{E}[W_t \mid \mathcal{F}_s] = W_s.$$

Proof sketch:

$$\mathbb{E}[W_t \mid \mathcal{F}_s] = \mathbb{E}[W_t - W_s + W_s \mid \mathcal{F}_s] = \mathbb{E}[W_t - W_s \mid \mathcal{F}_s] + \mathbb{E}[W_s \mid \mathcal{F}_s] = \mathbb{E}[W_t - W_s] + W_s = W_s,$$

since $W_t - W_s$ is independent of $\mathcal{F}_s$ and has mean 0.

Statement

The increments of Brownian motion are stationary; for $0 \leq s < t$,

$$W_t - W_s \stackrel{d}{=} W_{t-s} - W_0.$$

Proof sketch: since $W_0 = 0$ and $W_t - W_s \sim \mathcal{N}(0, t - s)$, we have

$$W_{t-s} - W_0 = W_{t-s} \sim \mathcal{N}(0, t - s),$$

hence $W_t - W_s \stackrel{d}{=} W_{t-s} - W_0$.

Expectation and variance

Consider $0 \leq s \leq t$ and $W_t = (W_t - W_s) + W_s$.

$$W_s W_t = W_s (W_t - W_s) + W_s^2$$

$$\mathbb{E}[W_s W_t] = \mathbb{E}[W_s (W_t - W_s)] + \mathbb{E}[W_s^2]$$

$$\mathbb{E}[W_s W_t] = \mathbb{E}[W_s] \mathbb{E}[W_t - W_s] + \mathbb{E}[W_s^2]$$



We know that

$$\mathbb{E}[W_s] = 0$$

$$\mathbb{E}[W_t - W_s] = 0$$

$$\therefore \mathbb{E}[W_s] \mathbb{E}[W_t - W_s] = 0$$

If $t \leq s$ then

$$\mathbb{E}[W_s W_t] = \mathbb{E}[W_t^2] = \text{var}(W_t) = t$$

$$\therefore \mathbb{E}[W_s W_t] = \min(s, t) = \text{cov}(W_s, W_t)$$

Markov Property

Consider $\{W_t : t \geq 0\}$ on a probability space $(\Omega, \mathcal{F}, \mathbb{P})$ is Brownian motion.

for $t_0 \geq 0$

$$\mathbb{P}(W_{t+t_0} \mid \mathcal{F}_{t_0}) = \mathbb{P}(W_{t+t_0} \mid W_{t_0}).$$

To know W_{t+t_0} , all we need the current W_{t_0} . The entire history adds no information.

$$W'_t := W_{t_0+t} - W_{t_0}$$

$$W_{t_0+t} = W_{t_0} + W'_t$$

$$W'_t \sim \mathcal{N}(0, t)$$

Markov Property (cont.)

1. $(W'_{t_1}, \dots, W'_{t_k})$ is independent of $\mathcal{F}_{t_0}$:

$$\mathcal{F}_{t_0} = \sigma(W_s : s \leq t_0)$$

consider

$$0 \leq s_1 \leq \dots \leq s_j \leq t_0 \quad \text{and} \quad 0 \leq t_1 \leq \dots \leq t_k$$

and $u_i = t_0 + t_i$.

Now

$$(W'_{t_1}, W'_{t_2} - W'_{t_1}, \dots, W'_{t_k} - W'_{t_{k-1}}) = (W_{u_1} - W_{t_0}, \dots, W_{u_k} - W_{u_{k-1}}). \quad (1)$$

Finite Dimensional Distributions

Finite Dimensional Distributions

Consider $(X_t)_{t \in T}$ $T = [0, 1]$ (or $\mathbb{R}$)

i.e, $X_t(\omega)$ or $X(t, \omega)$, $\omega \in \Omega$ and $t \in T$

fix an outcome ω ,

$$X(\cdot, \omega) : T \longrightarrow \mathbb{R}$$

For an outcome, it is a function in space $\mathbb{R}^T$.

For all outcomes, functions of functions

Joint probability law of random variables

Joint probability law of random variables

Fix a probability space $(\Omega, \mathcal{F}, \mathbb{P})$.

Given $X_1, X_2, \dots, X_n$ defined with respect to $\mathcal{F}$, their joint probability law is the probability measure

$$P_{X_1, \dots, X_n} : \mathcal{B}(\mathbb{R}^n) \rightarrow [0, 1]$$

defined as

$$P_{X_1, \dots, X_n}(B) = \mathbb{P}\left(\{\omega \in \Omega : (X_1(\omega), \dots, X_n(\omega)) \in B\}\right), \quad B \in \mathcal{B}(\mathbb{R}^n).$$

Finite-dimensional distributions (continued)

To define stochastic process usually describe in terms of distributions.

Consider

1. k -tuple $(t_1, t_2, \dots, t_k)$ of distinct elements of T .
2. random vector $(X_{t_1}, \dots, X_{t_k})$ has over $\mathbb{R}^k$ some measure $\mu_{t_1, t_2, \dots, t_k}$.

$$\begin{aligned}\mu_{t_1, t_2, \dots, t_k}(B) &= \mathbb{P}(\{\omega \in \Omega : (X_{t_1}(\omega), \dots, X_{t_k}(\omega)) \in B\}), \quad B \in \mathcal{B}(\mathbb{R}^k). \\ &= P_{X_{t_1}, \dots, X_{t_k}}(B).\end{aligned}$$

The collection of probability measures $\mu_{t_1, t_2, \dots, t_k}$ (or $P_{X_{t_1}, \dots, X_{t_k}}$) are the Finite dimensional distributions of the stochastic process $[X_t : t \in T]$.

FDD implies 2 properties

The FDD implies 2 properties

1. Suppose $B = B_1 \times B_2 \times \cdots \times B_k$ ($B_j \in \mathcal{B}(\mathbb{R}^1)$)
consider a permutation π of $(1, 2, \dots, k)$

$$\mu_{t_1, \dots, t_k}(B_1 \times B_2 \times \cdots \times B_k) = \mu_{t_{\pi(1)}, \dots, t_{\pi(k)}}(B_{\pi(1)} \times B_{\pi(2)} \times \cdots \times B_{\pi(k)}).$$

2. The second consistency condition is

$$\mu_{t_1, \dots, t_{k-1}}(B_1 \times B_2 \times \cdots \times B_{k-1}) = \mu_{t_1, \dots, t_{k-1}, t_k}(B_1 \times B_2 \times \cdots \times B_{k-1} \times \mathbb{R}^1).$$

FDD consistency: explanation

Explanation:

$$\mu_{t_1, \dots, t_{k-1}, t_k} (B_1 \times B_2 \times \dots \times B_{k-1} \times \mathbb{R}^1)$$

$$= \mathbb{P}(X_{t_1} \in B_1, \dots, X_{t_{k-1}} \in B_{k-1}, X_{t_k} \in \mathbb{R}).$$

We know that

$$\mathbb{P}(Y \in B) = \mathbb{P}(\{\omega \in \Omega : Y(\omega) \in B\}).$$

$$\therefore \mu_{t_1, \dots, t_k} (B_1 \times B_2 \times \dots \times B_{k-1} \times \mathbb{R}^1) = \mathbb{P}\left(\bigcap_{i=1}^{k-1} \{\omega \in \Omega : X_{t_i}(\omega) \in B_i\}\right) \cap \{\omega \in \Omega : X_{t_k}(\omega) \in \mathbb{R}\}.$$

$$= \mathbb{P}(X_{t_1} \in B_1, X_{t_2} \in B_2, \dots, X_{t_{k-1}} \in B_{k-1}) = \mu_{t_1, \dots, t_{k-1}} (B_1 \times B_2 \times \dots \times B_{k-1}).$$

Kolmogorov's Existence Theorem

Kolmogorov's Existence Theorem

If $\mu_{t_1, \dots, t_k}$ are a system of distributions satisfying the consistency conditions, then there exists on some probability space $(\Omega, \mathcal{F}, \mathbb{P})$ a stochastic process $[X_t : t \in T]$ having the $\mu_{t_1, \dots, t_k}$ as its finite dimensional distributions.

Existence of Brownian motion

Existence of Brownian motion:

Let $\mu_{t_1, \dots, t_k}$ be the distribution on $(S_1, \dots, S_k)$

$$\text{where } S_i = \sum_{j=1}^i X_j \quad \text{and} \quad X_i \sim \mathcal{N}(0, t_i - t_{i-1}) \quad \forall t \geq 2, \quad X_1 \sim \mathcal{N}(0, t_1).$$



$$\mu_{t_1, \dots, t_k} = \mathbb{P}(S_1 \in B_1, S_2 \in B_2, \dots, S_k \in B_k), \quad B_i \in \mathcal{B}(\mathbb{R}).$$

$$= \mathbb{P}\left(\bigcap_{i=1}^k \{\omega \in \Omega : \mathcal{S}_i \in B_i\}\right).$$

Consider a permutation π

$$\mu_{t_{\pi(1)}, \dots, t_{\pi(k)}} = \mathbb{P}(S_{\pi(1)} \in B_{\pi(1)}, \dots, S_{\pi(k)} \in B_{\pi(k)}).$$

$$= \mathbb{P}\left(\bigcap_{i=1}^k \{\omega \in \Omega : \mathcal{S}_{\pi(i)} \in B_{\pi(i)}\}\right).$$

Permutation Consistency Cont.

Since π is a permutation, as j runs 1 to k , $i = \pi(j)$ also runs from 1 to k .

$$\therefore \mu_{t_1, \dots, t_k} = \mu_{t_{\pi(1)}, \dots, t_{\pi(k)}}.$$

Marginalization consistency

Marginalization consistency: Consider a function g

$$g(x_1, x_2, \dots, x_k) = (x_1, \dots, x_{i-1}, x_{i+1}, \dots, x_k).$$

$$\therefore g(S_1, S_2, \dots, S_k) = (S_1, \dots, S_{i-1}, S_{i+1}, \dots, S_k).$$

Without loss of generality

$\mu_{t_1, \dots, t_{i-1}, t_{i+1}, \dots, t_k}$ has distribution for $S_1, \dots, S_{i-1}, S_{i+1}, \dots, S_k$

So

$$\mu_{t_1, \dots, t_{i-1}, t_{i+1}, \dots, t_k} = \mathbb{P}(S_1 \in B_1, \dots, S_{i-1} \in B_{i-1}, S_{i+1} \in B_{i+1}, \dots, S_k \in B_k).$$

Marginalization consistency

$$\mu_{t_1, \dots, t_k} = \mathbb{P}(S_1 \in B_1, \dots, S_i \in \mathbb{R}, \dots, S_k \in B_k).$$

$$\therefore \mu_{t_1, \dots, t_k} = \mu_{t_1, \dots, t_{i-1}, t_{i+1}, \dots, t_k}.$$

$\therefore$ A process $[W_t : t > 0]$ considering $W_t = 0$ at $t = 0$ exists a some $(\Omega, \mathcal{F}, \mathbb{P})$.

Markov property

Consider $\{W_t : t \geq 0\}$ on a probability space $(\Omega, \mathcal{F}, \mathbb{P})$ is Brownian motion.
For $t_0 \geq 0$,

$$\mathbb{P}(W_{t+t_0} \mid \mathcal{F}_{t_0}) = \mathbb{P}(W_{t+t_0} \mid W_{t_0}).$$

To know W_{t+t_0} , all we need the current W_{t_0} . The entire history adds no information.

Define

$$W'_t := W_{t_0+t} - W_{t_0}.$$

$$W'_t \sim \mathcal{N}(0, t).$$

1. $(W'_{t_1}, \dots, W'_{t_k})$ is independent of $\mathcal{F}_{t_0}$:

$$\mathcal{F}_{t_0} = \sigma(W_s : s \leq t_0).$$

Consider $0 \leq s_1 \leq \dots \leq s_j \leq t_0$ and $0 \leq t_1 \leq \dots \leq t_k$, and $u_i = t_0 + t_i$.

$$\left(W'_{s_1}, W'_{s_2}, \dots, W'_{s_j}, W'_{t_1}, W'_{t_2}, \dots, W'_{t_k} \right) \sim \left(W_{u_1}, W_{u_2}, \dots, W_{u_j}, W_{u_{j+1}}, W_{u_{j+2}}, \dots, W_{u_{j+k}} \right) \quad (1)$$

Markov property (cont.)

and also

$$W_{s_1}, W_{s_2} - W_{s_1}, W_{s_3} - W_{s_2}, \dots, W_{s_j} - W_{s_{j-1}}. \quad (2)$$

From (??), the intervals

$$[s_1, s_2], \dots, [s_{j-1}, s_j] \quad \text{end before } t_0,$$

and

$$[t_0, u_1], \dots, [u_{k-1}, u_k] \quad \text{start at or after } t_0.$$

So, the above are all disjoint intervals; hence, from the above properties, the increments over disjoint intervals are independent, i.e.,

$$(W'_{t_1}, \dots, W'_{t_k}) \perp (W_{s_1}, \dots, W_{s_j}),$$

and therefore

$$(W'_{t_1}, \dots, W'_{t_k}) \perp \mathcal{F}_{t_0}.$$

Markov property (cont.)

Consider, for $B \in \mathcal{B}(\mathbb{R})$,

$$\mathbb{P}(W_{t_0+t} \in B \mid \mathcal{F}_{t_0}) = \mathbb{E} \left[\mathbf{1}_{\{W_{t_0+t} \in B\}} \mid \mathcal{F}_{t_0} \right] = \mathbb{E} \left[\mathbf{1}_{\{W_{t_0} + W'_t \in B\}} \mid \mathcal{F}_{t_0} \right].$$

Here W_{t_0} is $\mathcal{F}_{t_0}$ -measurable and W'_t is independent of $\mathcal{F}_{t_0}$. Hence,

$$\mathbb{E} \left[\mathbf{1}_{\{W_{t_0} + W'_t \in B\}} \mid \mathcal{F}_{t_0} \right] = \varphi(W_{t_0}),$$

for some (Borel) function φ .

Markov property (cont.)

Now consider

$$\mathbb{E}\left[\mathbf{1}_{\{W_{t_0+t} \in B\}} \mid \sigma(W_{t_0})\right] = \mathbb{E}\left[\mathbf{1}_{\{W_{t_0} + W'_t \in B\}} \mid \sigma(W_{t_0})\right].$$

Since $\sigma(W_{t_0}) \subset \mathcal{F}_{t_0}$ and W'_t is independent of $\sigma(W_{t_0})$, we again get $\varphi(W_{t_0})$.

Therefore,

$$\mathbb{E}\left[\mathbf{1}_{\{W_{t_0} + W'_t \in B\}} \mid \sigma(W_{t_0})\right] = \varphi(W_{t_0}).$$

$$\therefore \mathbb{P}(W_{t_0} + W'_t \in B \mid \mathcal{F}_{t_0}) = \mathbb{P}(W_{t_0} + W'_t \in B \mid \sigma(W_{t_0})),$$

and hence

$$\therefore \mathbb{P}(W_{t_0+t} \in B \mid \mathcal{F}_{t_0}) = \mathbb{P}(W_{t_0+t} \in B \mid \sigma(W_{t_0})).$$

Markov property (cont.)

The other way: consider

$$\mathbb{P}(W_{t_0+t} \in B \mid \mathcal{F}_{t_0}) = \mathbb{P}(W_{t_0} + (W_{t_0+t} - W_{t_0}) \in B \mid \mathcal{F}_{t_0}).$$

Since $\sigma(W_{t_0}) \subset \mathcal{F}_{t_0}$ and $(W_{t_0+t} - W_{t_0}) = W'_t$ is independent of $\mathcal{F}_{t_0}$ (hence independent of $\sigma(W_{t_0})$),

$$\therefore \mathbb{P}(W_{t_0+t} \in B \mid \mathcal{F}_{t_0}) = \mathbb{P}(W_{t_0+t} \in B \mid \sigma(W_{t_0})).$$

Strong Markov Property

Consider τ is a stopping time and also

$$W_t^*(\omega) = W_{\tau(\omega)+t}(\omega) - W_{\tau(\omega)}(\omega), \quad t \geq 0.$$

Consider τ has countable range V , i.e., $V \subset [0, \infty)$.

Let t_0 be some general point of V .

$$\mathcal{F}_\tau = \{M \in \mathcal{F} : M \cap \{\omega : \tau(\omega) \leq t\} \in \mathcal{F}_t, \forall t \geq 0\}.$$

Consider

$$\{\omega : W_t^*(\omega) \in B\} = \bigcup_{t_0 \in V} \{\omega : W_{t_0+t}(\omega) - W_{t_0}(\omega) \in B, \tau(\omega) = t_0\}.$$

Now $A \in \mathcal{F}_\tau$.

Strong Markov Property (cont.)

Now $A \in \mathcal{F}_\tau$.

$$\begin{aligned}
 \mathbb{P}(\{(W_{t_1}^*, \dots, W_{t_k}^*) \in B\} \cap A) &= \sum_{t_0 \in V} \mathbb{P}(\{(W_{t_1}^*, \dots, W_{t_k}^*) \in B\} \cap A \cap \{\tau = t_0\}) \\
 &= \sum_{t_0 \in V} \mathbb{P}(\{(W_{t_1}^*, \dots, W_{t_k}^*) \in B\}) \mathbb{P}(A \cap \{\tau = t_0\}) \\
 &= \mathbb{P}(\{(W_{t_1}^*, \dots, W_{t_k}^*) \in B\}) \sum_{t_0 \in V} \mathbb{P}(A \cap \{\tau = t_0\}) \\
 &= \mathbb{P}(\{(W_{t_1}^*, \dots, W_{t_k}^*) \in B\}) \mathbb{P}(A).
 \end{aligned}$$

Strong Markov Property (cont.)

The above holds for τ having countable range. Now for any general τ , consider

$$\tau_n = \begin{cases} k \cdot 2^{-n} & \text{if } (k-1)2^{-n} < \tau \leq k \cdot 2^{-n}, \quad k = 1, 2, \dots, \\ 0 & \tau = 0. \end{cases}$$

This is dyadic approximation which is used to approximate real numbers using dyadic rationals, where a number x is approximated by finding an integer k such that

$$\frac{k}{2^n} \leq x < \frac{k+1}{2^n}.$$

Strong Markov Property (cont.)

At $\tau_n(\omega) = k \cdot 2^{-n}$ we know that

$$\tau(\omega) \leq k \cdot 2^{-n}, \quad \therefore \tau(\omega) \leq \tau_n(\omega),$$

and

$$\tau(\omega) > (k-1)2^{-n} \implies \tau(\omega) + 2^{-n} \geq k \cdot 2^{-n}.$$

Therefore

$$\therefore \tau(\omega) \leq \tau_n(\omega) \leq \tau(\omega) + 2^{-n}.$$

Hence, as $n \rightarrow \infty$,

$$\tau_n(\omega) \longrightarrow \tau(\omega).$$

Strong Markov Property (cont.)

τ_n is a stopping time: if $k \cdot 2^{-n} \leq t < (k+1) \cdot 2^{-n}$ then

$$\{\tau_n \leq t\} = \{\tau \leq k \cdot 2^{-n}\} \in \mathcal{F}_{k \cdot 2^{-n}} \subset \mathcal{F}_t.$$

Suppose $A \in \mathcal{F}_\tau$ and $k \cdot 2^{-n} \leq t < (k+1) \cdot 2^{-n}$ then

$$A \cap \{\tau_n \leq t\} = A \cap \{\tau \leq k \cdot 2^{-n}\} \in \mathcal{F}_{k \cdot 2^{-n}} \subset \mathcal{F}_t.$$

Hence all such A 's lie in $\mathcal{F}_{\tau_n}$, i.e.,

$$A \in \mathcal{F}_\tau \implies A \in \mathcal{F}_{\tau_n}.$$

Strong Markov Property (cont.)

As $\tau_n(\omega) \rightarrow \tau(\omega)$ and Brownian paths are continuous,

$$W_t^{*(n)}(\omega) \longrightarrow W_t^*(\omega).$$

Therefore,

$$\begin{aligned} \mathbb{P}\left(\{(W_{t_1}^{*(n)}, \dots, W_{t_k}^{*(n)}) \in B\} \cap A\right) &= \mathbb{P}\left(\{(W_{t_1}^*, \dots, W_{t_k}^*) \in B\} \cap A\right) \\ &= \mathbb{P}\left(\{(W_{t_1}^*, \dots, W_{t_k}^*) \in B\}\right) \mathbb{P}(A). \end{aligned}$$

$$\therefore (W_{t_1}^*, \dots, W_{t_k}^*) \perp \mathcal{F}_\tau.$$

$$\therefore \mathbb{P}(W_{\tau+t} \in B \mid \mathcal{F}_\tau) = \mathbb{P}(W_{\tau+t} \in B \mid W_\tau).$$

Reflection principle

Reflection principle: $(W_t)_{t \geq 0}$ is a Brownian motion on $(\Omega, \mathcal{F}, \mathbb{P})$.

Let τ be the stopping time and define the reflected process W_t'' by

$$\begin{aligned} W_t'' &= \begin{cases} W_t, & t \leq \tau, \\ W_\tau - (W_t - W_\tau), & t \geq \tau, \end{cases} \\ &= \begin{cases} W_t, & t \leq \tau, \\ 2W_\tau - W_t, & t \geq \tau. \end{cases} \end{aligned}$$

W'' is Brownian motion.

The sample path for $\{W_t'' : t \geq 0\}$ is the same as the sample path for $\{W_t : t \geq 0\}$ up to τ . After that, it is reflected through the point W_τ .

Reflection principle

Consider τ has some countable range V .

Consider two times s, t for $\{\tau = t_0\}$.

Case 1: $s \leq t \leq t_0$

$$\begin{aligned} W_s'' &= W_s, & W_t'' &= W_t \\ W_s'' &\stackrel{d}{=} W_s, & W_t'' &\stackrel{d}{=} W_t. \end{aligned}$$

Case 2: $s \leq t_0 \leq t$

$$W_s'' = W_s, \quad W_t'' = 2W_{t_0} - W_t.$$

Consider

$$W_t = W_{t_0} + (W_t - W_{t_0}).$$

$$W_t - W_{t_0} \sim \mathcal{N}(0, t - t_0)$$

Reflection Principle

and Gaussian increments are symmetric, so

$$W_t \stackrel{d}{=} W_{t_0} - (W_t - W_{t_0}) = W_t''.$$

Hence

$$W_s \stackrel{d}{=} W_s'', \quad W_t \stackrel{d}{=} W_t''.$$

Case 3: $t_0 \leq s \leq t$

$$W_s'' = 2W_{t_0} - W_s, \quad W_t'' = 2W_{t_0} - W_t.$$

By the same argument in Case 2,

$$W_s \stackrel{d}{=} W_s'', \quad W_t \stackrel{d}{=} W_t''.$$

$$\therefore (W_{t_1}, \dots, W_{t_k}) \stackrel{d}{=} (W_{t_1}'', \dots, W_{t_k}'').$$

Reflection principle: distribution of the maximum

Now consider,

$$\tau = \inf \{s : W_s \geq x\},$$

the first time W_s hits x and consider

$$M_t = \sup_{s \leq t} W_s.$$

M_t is the highest point that Brownian paths reaches before time t .

$$M_t \geq x \iff \exists s \in [0, t] \text{ such that } W_s \geq x.$$

The path reaches x before time t and τ is the first time path reaches level x .

If the path ever exceed x then the path must hit x before time t .

$$\mathbb{P}(M_t \geq x) = \mathbb{P}(\tau \leq t).$$

Reflection principle: distribution of the maximum

$$\mathbb{P}(M_t \geq x) = \mathbb{P}(\tau \leq t, W_t \leq x) + \mathbb{P}(\tau \leq t, W_t \geq x)$$

$$= \mathbb{P}(\tau \leq t, W_t'' \geq x) + \mathbb{P}(\tau \leq t, W_t \geq x) \quad (\text{by reflection principle})$$

$$= \mathbb{P}(\tau \leq t, W_t \geq x) + \mathbb{P}(\tau \leq t, W_t \geq x)$$

since

$$\mathbb{P}(W_t \geq x) = \frac{1}{\sqrt{2\pi}} \int_{x/\sqrt{t}}^{\infty} e^{-u^2/2} du.$$

$$\mathbb{P}(M_t \geq x) = \mathbb{P}(W_t \geq x) = \frac{2}{\sqrt{2\pi}} \int_{x/\sqrt{t}}^{\infty} e^{-u^2/2} du.$$