

Lecture 1 (MAE 271B, Boelter 5272)

Sunday, January 4, 2026 7:13 PM

- what is a Brownian Motion?

consider a particle of interest in a fluid medium: (Brown pollen Experiment)

- what is the path travelled by this particle in unit time?

the sample path is a result of many small, independent collisions of the particle with the fluid molecules [Einstein's argument]; and these collisions must have zero-mean because the overall velocity is zero.

- How can we model this (stochastic) dynamics?

- We know that the mean-square displacement of the particle per unit time must remain constant, say $= \sigma$, and only depends on temperature.
- collisions are independent, identically distributed (i.i.d.) displacements ξ_n

so lets fix the time interval $t \in [0, 1]$, and assume there is N collisions, so that:

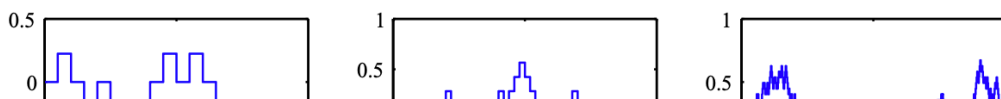
(particle position at time t)
$$x_t(N) = x_0 + \sum_{n=1}^{\lfloor Nt \rfloor} \xi_n, \text{ for } t \in [0, 1]$$

Then
$$\sigma = \mathbb{E} (x_1(N) - x_0)^2 = \text{Var} \left(\sum_{n=1}^N \xi_n \right) = N \text{Var}(\xi_n)$$

which implies an important information about.

$$\boxed{\text{Var}(\xi_n) = \sigma N^{-1}}$$

- How does this process looks like as $N \rightarrow \infty$: [Figure from Handel '07]



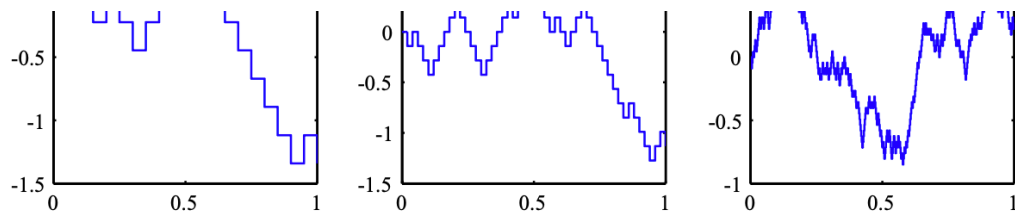


Figure 0.1. Randomly generated sample paths $x_t(N)$ for the Brownian motion model in the text, with (from left to right) $N = 20, 50, 500$ collisions per unit time. The displacements ξ_n are chosen to be random variables which take the values $\pm N^{-1/2}$ with equal probability.

lets rewrite this process in order to use the central Limit Theorem:

$$X_t(N) = X_0 + \sqrt{\delta t} \cdot \frac{\sum_{n=1}^{\lfloor Nt \rfloor} \Xi_n}{\sqrt{Nt}} \quad \text{where } \Xi_n := \xi_n \cdot \sqrt{N\delta t^{-1}}$$

$\uparrow$
 i.i.d, zero-mean, unit variance

For fixed t , by CLT: as $N \rightarrow \infty$, $X_t(N) \xrightarrow{d} \mathcal{N}(0, \delta t)$

Question: As $N \rightarrow \infty$, does the limit of the stochastic process $t \mapsto X_t(N)$ exist (in a suitable sense)?

[by Wiener] yes, it is the "Brownian Motion", $\{W_t : t \in [0, 1]\}$.
Wiener process

But it takes some serious effort to establish this!

Applications of the Wiener process W_t :

• Stochastic control:

$$\min_{\{u_t\}} \mathbb{E} \left[\int_0^1 (W_t + Z_t)^2 dt \right] + \mathbb{E} \left[\int_0^1 u_t^2 dt \right]$$

s.t. $\frac{dZ_t}{dt} = u_t$

• Stock market ("Geometric Brownian Motion"):

$$S_t = S_0 \exp \left\{ \left(\mu - \frac{\sigma^2}{2} \right) t + \sigma W_t \right\}$$

$\uparrow$
 stock price

(a unit stock price) at time t return rate μ
volatility σ

$$\Rightarrow E[S_t] = S_0 e^{\mu t} \quad (\text{why?})$$

- Estimation: measurement at continuous time t
 $y_t = x_t + \zeta_t$ x_t is positions of a particle
 ζ_t is a "white noise".

integrated measurement signal $Y_t = \int_0^t y_s ds = \int_0^t x_s ds + W_t$

Find the best MSE estimate $\hat{x}_t$ of x_t s.t.

$$\{\hat{x}_s\}_0^t = \arg \min_{\{\bar{x}_s\}_0^t} E \left[\int_0^t (y_s - \bar{x}_s)^2 ds \right]$$

We have a serious issue with W_t !

Suppose $X_t = X_0 + W_t$

what is the derivative of X_t , $\frac{dX_t}{dt}$?

It does not exist a.s. because, even though $\{W_t\}$ is a well-defined process, it is not differentiable a.e.!

But we still like to make sense of " dW_t " s.t.

$$dX_t = dW_t \quad \equiv \quad X_t = \int_0^t dX_s = \int_0^t dW_s = W_t$$

more generally, we write

$$dX_t = b(X_t, t) dt + \sigma dW_t$$

to mean that

$$X_t = \int_0^t b(x_s, s) ds + \sigma W_t$$

and we can show that ~~this~~ is a unique solution
by Picard iterations

$$\varphi_0(t) = X_0, \quad \varphi_{k+1}(t) = X_0 + \int_0^t b(\varphi_k(s), s) ds + \sigma W_t$$

and use Banach Fixed-Point theorem to argue that

$$\varphi_k(\cdot) \rightarrow X(\cdot) \text{ as } k \rightarrow \infty.$$

However, we still have some non-trivial challenges:

$$\text{what } \frac{d}{dt} X_t^2 = ?!$$

$$(\text{naively}) \quad \frac{d}{dt} X_t^2 = 2X_t \frac{dX_t}{dt} = 2X_t b(X_t, t) + 2\sigma X_t dW_t$$

but now what does $\int_0^t X_t dW_t$ really mean?!

This means we need some new calculus!

Most generally, we define the stochastic Differential Eq.

$$dX_t = b(X_t, t) dt + \sigma(X_t, t) dW_t$$

by defining a new type of integral, called Ito Integral

that can make sense of

$$\int_0^t \sigma(x_s, s) dw_s$$

Also, it results in a "new chain rule":

$$\text{for any } f \in C^2, \quad df(w_t) = \underbrace{f'(w_t) dw_t}_{\text{usual term}} + \underbrace{\frac{1}{2} f''(w_t) dt}_{\text{2nd-order correction}}$$

Using this formula, we can double check the variance of w_t as we designed to t :

$$\begin{aligned} \text{Var}(w_t) &= E(w_t^2) = E\left[\int_0^t \frac{d}{ds} w_s^2\right] \\ &= E\left[\int_0^t [2w_s dw_s + ds]\right] \\ &= E\left[\underbrace{2\int_0^t w_s dw_s}_{\text{zero-mean process (in fact a "martingale")}} + \int_0^t ds\right] = t \end{aligned}$$