

## Brownian motion.

Consider a particle moving ~~at~~ on the real line  $\mathbb{R}$ .

Position at time  $t \geq 0$   $X(t)$

It moves in discrete steps

$$t \in \mathbb{N}\tau = \{0, \tau, 2\tau, \dots\}$$

$$X_{\tau}(k+1) = X_{\tau}(k) + \delta \cdot U_k$$

where  $(U_k)_{k \geq 1}$  is a sequence of  
iid rv  $\mathbb{E}U_k = 0, \mathbb{E}U_k^2 = 1$ .

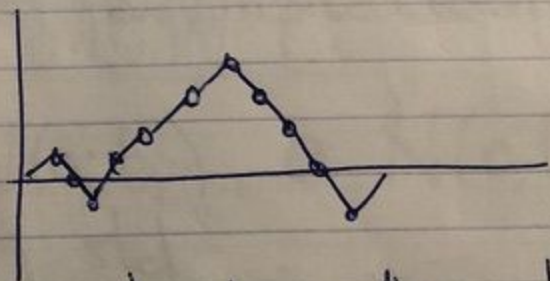
We interpolate linearly to get a  
random function

$$t \mapsto X_{\tau}(t)$$

Linear interpolation

For  $t \in [k\tau, (k+1)\tau]$

$$X_{\delta}(t) := \left( \frac{(k+1)\tau - t}{\tau} \right) X(k\tau) + \left( \frac{t - k\tau}{\tau} \right) X((k+1)\tau)$$



What does this look like when  $\delta, \tau \rightarrow 0$

Fix  $t > 0$

$$X_{\delta}(t) = \delta \sum_{k=1}^{\lfloor t/\tau \rfloor} U_k + \delta \cdot \left( \frac{t - \lfloor t/\tau \rfloor \cdot \tau}{\tau} \right) \cdot U_{\lfloor t/\tau \rfloor + 1}$$

$O(\delta)$

$$\frac{1}{\delta \lfloor t/\tau \rfloor^{1/2}} X_{\delta}(t) \xrightarrow{d} N(0, 1)$$

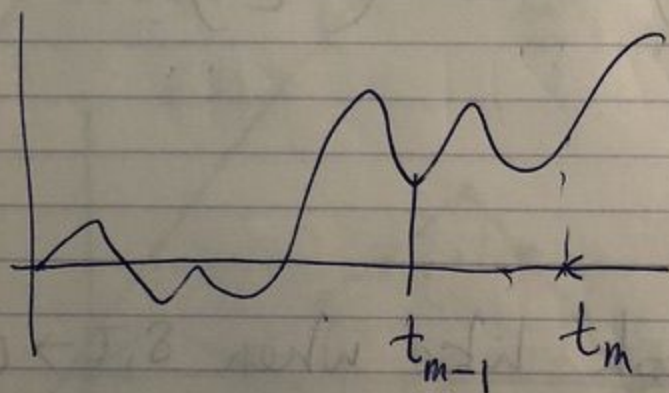
as  $\tau \rightarrow 0$   
any  $\delta$

$$\frac{\tau^{1/2}}{\delta} X_{\delta}(t) \xrightarrow{d} N(0, t)$$

Need to take  $\boxed{\delta = \sqrt{\tau}}$

What about  $m$  distinct times?

$$0 < t_1 < t_2 < \dots < t_m$$



Condition on  $(X_s(t))_{t \leq \lceil \frac{t_{m-1}}{\tau} \rceil \tau}$

$$\lim_{\delta \rightarrow 0} \mathbb{P}(X_s(t_{m-\delta}) - X_s(t_{m-1}) \leq s \mid (X_s(t))_{t \leq \lceil \frac{t_m}{\tau} \rceil})$$

$$\stackrel{?}{=} \mathbb{P}(N(0, t_m - t_{m-1}) \leq s)$$

What is the joint distr?

$$(X_s(t_1), X_s(t_2) - X_s(t_1), \dots, X_s(t_m) - X_s(t_{m-1}))$$

are asymptotically indep. Gaussian

$$N(0, t_1) \otimes N(0, t_2 - t_1) \otimes \dots \otimes N(0, t_m - t_{m-1})$$

A <sup>std</sup> Brownian Motion  $(X(t))_{t \geq 0}$

(Wiener process) is a random fct. satisfying 2 conditions

(1)  $(X(t))_{t \geq 0}$  is ~~as~~ continuous  
w prob = 1

(2)  $\forall 0 \leq t_1 < t_2 < \dots < t_m$

$$(X(t_1), X(t_2) - X(t_1), \dots, X(t_m) - X(t_{m-1}))$$

is a vector with indep Gaussian

$$\mathbb{E}[(X(t_i) - X(t_{i-1}))^2] = t_i - t_{i-1}$$

Equivalently  $\forall t \geq t_0$

$$P(X(t) \leq s \mid (X(u))_{u \leq t_0}) = P(N(0, t - t_0) \leq s)$$

It is a nontrivial theorem that this ~~of~~ prob distr over function can be constructed and is actually completely determined by the above 2 properties.

If  $(W_t)_{t \geq 0}$  is a std BM.

\* BM started at  $a$ :  $X_t = a + W_t$

\* BM started at  $a$ , with drift  $v$

$$X_t = a + vt + W_t$$

Definition Given a continuous time process  $(X_t)_{t \geq 0}$ .  $T$  is a stopping time of cont sample paths with respect to  $(X_t)$  if  $\forall t \in \mathbb{R}_{\geq 0}$

$$1_{T \leq t} = \mathcal{F}((X_s)_{s \leq t}).$$

Def  $Y_t$  is a MG wrt  $(X_t)_{t \geq 0}$  if  $\forall t \quad \mathbb{E}|Y_t| < \infty$  and

$$\mathbb{E}(Y_{t+h} | (X_s)_{s \leq t}) = Y_t.$$

Example  $(W_t)_{t \geq 0}$  BM started at  $a > 0$

$$T = \inf \{ t \geq 0 : W_t \in \{0, A\} \}$$

Hitting time at  $0, A$ .

$Y_t = e^{\theta W_t - \frac{\theta^2 t}{2}}$  is a MG wrt  $(W_t)$

$$\begin{aligned} \mathbb{E}\{Y_{t+h} \mid (W_s)_{s \leq t}\} &= \mathbb{E}\{e^{\theta(W_{t+h} - W_t) - \frac{\theta^2 h}{2}} Y_t \mid (W_s)_{s \leq t}\} \\ &= Y_t \mathbb{E}[e^{\theta(W_{t+h} - W_t)} \mid W_{\leq t}] e^{-\frac{\theta^2 h}{2}} = Y_t \end{aligned}$$

recall that

$W_{t+h} - W_t$  is indep. of  $W_{\leq t}$

$\sim N(0, h)$

[true also if you start at  $a$ , since  $W_t = W_t^0 + a$  ...]  
 $\tau$  started at 0

it can be proven.

$\Rightarrow T < \infty$  a.s.  $Y_{t \wedge T}$  is a MG as well

$$\mathbb{E}[Y_T] = \mathbb{E}(Y_0) = e^{\theta a}$$

$$\mathbb{E}\left[e^{\theta A - \frac{\theta^2 T}{2}} 1_{T_A < b} + e^{-\frac{\theta^2 T}{2}} 1_{b < T_A}\right]$$

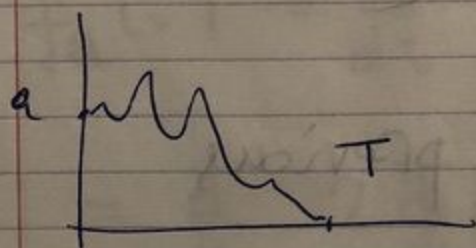
$$e^{\theta a} = \mathbb{E} \left[ \left( e^{\theta A} 1_{T_A < T_0} + 1_{T_0 < T_A} \right) e^{-\frac{\theta^2}{2} T} \right]$$

$$e^{-\theta a} = \mathbb{E} \left[ \left( e^{-\theta A} 1_{T_A < T_0} + 1_{T_0 < T_A} \right) e^{-\frac{\theta^2}{2} T} \right]$$

$$e^{\theta a} + e^{-\theta(A-a)} = (1 + e^{\theta A}) \mathbb{E} \left[ e^{-\frac{\theta^2}{2} T} \right]$$

$$\mathbb{E} \left[ e^{-\frac{\theta^2}{2} T} \right] = \frac{e^{\theta a} + e^{-\theta(A-a)}}{1 + e^{\theta A}}$$

or take limit  $A \rightarrow \infty$



$$T = T_0$$

$$\mathbb{E} \left[ e^{-\frac{\theta^2}{2} T} \right] = e^{-\theta a}$$

$$e^{\lambda a} \mathbb{E} \left[ e^{-\lambda T} \right] = e^{-a\sqrt{2\lambda}}$$

$f_a(t)$  density of  $T$

$$\int_0^{\infty} e^{-\lambda t} f_a(t) dt = e^{-a\sqrt{2\lambda}}$$