

Ito formula in high dimensions

Ito in 1d

$$f(t, B_t) = f(0, B_0) + \int_0^t \frac{\partial f}{\partial s}(s, B_s) ds + \int_0^t \frac{\partial f}{\partial x}(s, B_s) dB_s + \frac{1}{2} \int_0^t \frac{\partial^2 f}{\partial x^2}(s, B_s) ds.$$

$$df(t, B_t) = \frac{\partial}{\partial t} f(t, B_t) dt + \frac{\partial}{\partial x} f(t, B_t) dB_t + \frac{1}{2} \frac{\partial^2 f}{\partial x^2}(t, B_t)$$

d-dimension

$$\vec{B}_t = (B_t^1, \dots, B_t^d)$$

d indep 1d BMs.

$$df(t, \vec{B}_t)$$

$$= \underbrace{\frac{\partial}{\partial t} f(t, \vec{B}_t)}_{\text{wavy}} dt + \underbrace{\nabla f(t, \vec{B}_t)}_{\text{double underline}} \cdot \underbrace{d\vec{B}_t}_{\text{wavy}} + \frac{1}{2} \underbrace{\Delta f(t, \vec{B}_t)}_{\text{wavy}} dt$$

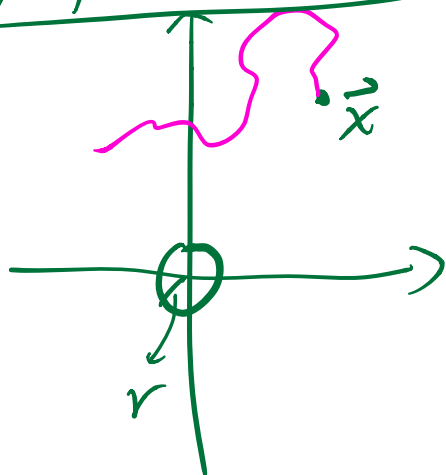
$$\int_0^t \nabla f(s, \vec{B}_s) \cdot d\vec{B}_s \text{ is mart.}$$

$$\left(\begin{array}{l} \nabla f(t, \vec{x}) = \left(\frac{\partial f}{\partial x_1}(t, \vec{x}), \dots, \frac{\partial f}{\partial x_d}(t, \vec{x}) \right) \\ \quad \quad \quad (x_1, \dots, x_d) \\ \Delta f(t, \vec{x}) = \frac{\partial^2 f}{\partial x_1^2} + \dots + \frac{\partial^2 f}{\partial x_d^2} \end{array} \right)$$

prop: $M_t = f(t, \vec{B}_t)$ is mart
 if $\frac{\partial f}{\partial t} = -\frac{1}{2} \Delta f$

(special case: $f(\vec{x})$ $\Delta f = 0$
 $\Rightarrow f$ is harmonic function)

Application



In \mathbb{R}^2 $P_{\vec{x}}(\tau_r < \infty) = 1$
Recurrent

In \mathbb{R}^d $d \geq 3$ $P_{\vec{x}}(\tau_r < \infty) < 1$
Transient

$$\tau_r = \inf \{t : |\vec{B}_t| = r\}$$

$$d=2, \quad f(\vec{x}) = \log |\vec{x}| = \log \sqrt{x_1^2 + x_2^2}$$

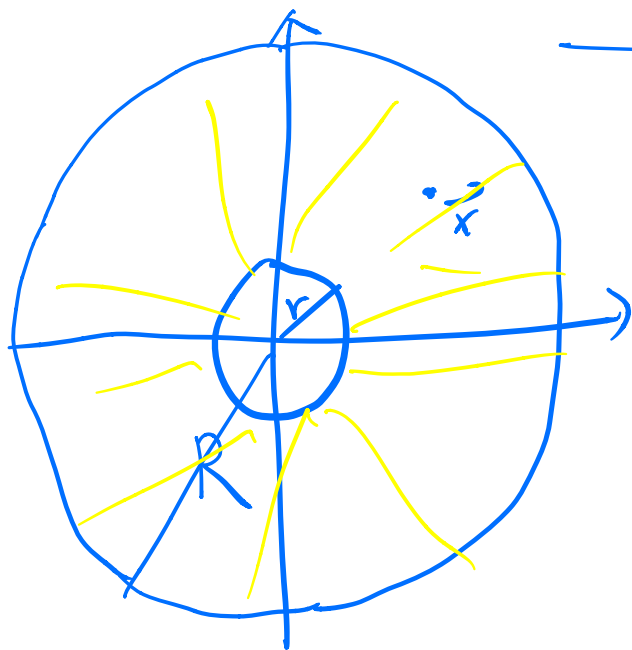
$$\underline{\Delta f} = 0, \quad \forall \vec{x} \neq 0.$$

$$d \geq 3$$

$$f(\vec{x}) = |\vec{x}|^{2-d}$$

$$\Delta f = 0, \quad \forall \vec{x} \neq 0$$

$$\Delta f = 0 \quad \forall x \neq 0.$$



$$\begin{cases} \Delta h = 0 \\ h(\vec{x}) = 1 & |\vec{x}| = r \\ h(\vec{x}) = 0 & |\vec{x}| = R \end{cases}$$

$h(\vec{B}_t)$ is mart.

$$\mathbb{E}[h(\vec{B}_t)] = \mathbb{E}[h(\vec{B}_0)] = \mathbb{E}[h(\vec{x})] = \underline{h(\vec{x})}$$

$$\tau = \min(\underline{\tau_r}, \tau_R)$$

$$\mathbb{E}[h(\vec{B}_t)] = P(|\vec{B}_t| = r) \cdot 1 + P(|\vec{B}_t| = R) \cdot 0$$

$$= P(|\vec{B}_t| = r) = \underline{P(\tau_r < \tau_R)}$$

$$\boxed{h(\vec{x})} = P(\tau_r < \tau_R)$$

Send $R \rightarrow \infty$

2D $f(\vec{x}) = \log |\vec{x}|$ $\Delta f = 0$.

$$h(\vec{x}) = \frac{\log R - \log |\vec{x}|}{\log R - \log r} \Rightarrow \Delta h = 0$$

$$\begin{aligned} \mathbb{P}(\tau_r < \infty) &= \lim_{R \rightarrow \infty} \frac{\log R - \log |\vec{x}|}{\log R - \log r} \\ &= 1 \end{aligned}$$

~~2D~~
 ~~$d \geq 3$~~ $f(\vec{x}) = |\vec{x}|^{2-d}$ $\Delta f = 0$.

$$h(\vec{x}) = \frac{R^{2-d} - |\vec{x}|^{2-d}}{R^{2-d} - r^{2-d}} \quad \Delta h = 0$$

$$\begin{aligned} \mathbb{P}(\tau_r < \infty) &= \lim_{R \rightarrow \infty} \frac{R^{2-d} - |\vec{x}|^{2-d}}{R^{2-d} - r^{2-d}} \\ &= \left(\frac{r}{|\vec{x}|} \right)^{d-2} < 1 \end{aligned}$$

Since $|\vec{x}| > r$

$d=2$ $P(\tau_r < \infty) = 1$ drunk man
Can always find home

$d \geq 3$ $P(\tau_r < \infty) < 1$.

drunk bird
Can possibly
get lost.