

Thank you.

SAFE TRAVELS

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③ Wiener process

Properties

① w probability 1, the continuous mapping $t \rightarrow w(t)$
w/ $w(0) = w_0$

② if $0 = t_0 < t_1 < t_2 \dots < t_N = T$, then the increments
 $\Delta w_i = w(t_i) - w(t_{i-1})$ are independent
w/ gaussian variation

$$\{w(t), w(t)\} = t$$

③ $\forall t > s$, then $w(t) - w(s) \sim N(0, t-s)$

$$E(w(t) - w(s)) = 0$$

$$E((w(t) - w(s))^2) = t - s$$

that is

$$P(w_t - w_s \in \Gamma) = \int_{\Gamma} \frac{e^{-\frac{x^2}{2(t-s)}}}{\sqrt{2\pi(t-s)}} dx \quad (\Gamma \in \mathbb{R})$$

20/20

⑨ State and derive Feynman-Kac formula

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for smooth a, b, g, h and V , bounded functions.

$$u(t, x) = E \left[g(x(T)) e^{\int_t^T V(s, x(s)) ds} \mid x_t = x \right] =$$

$$+ E \left[- \int_t^T h(s, x(s)) e^{\int_t^s V(\tau, x(\tau)) d\tau} ds \right] \quad \checkmark$$

then u solve the PDE

$$L^* u = u_t + au_x + \frac{b^2}{2} u_{xx} + Vu = h \quad t < T \quad (1)$$

$$g(\cdot, T) = g(\cdot)$$

Correct
evaluate

Proof

① Define $\hat{u}$ to solve (1) then

$$L^* \hat{u} = h$$

and

let $G(s) = e^{\int_t^s V(\tau, x(\tau)) d\tau}$

↑ better to use another variable to avoid confusion with the upper limit s !

② $\hat{u}$ solves the Kolmogorov's BWD PDE

$$\begin{cases} L\hat{u} = u_t + au_x + \frac{1}{2}b^2 u_{xx} = 0 & t < T \\ \hat{u}(\cdot, T) = g(\cdot) \end{cases}$$

Apply Ito to $\hat{u}$ and use 1

③ Integrate from t to T and take expected values

Note $L^* \hat{u} + Vu = L^* \hat{u} = h$

$$L^* \hat{u} = h - V\hat{u}$$

pg 2 of 5

① cannot be solution to both these different PDEs (unless V and h are zero)

NO

In step 1 you defined $\hat{u}$ to solve another PDE

$$d\left(\hat{u}(t, x) e^{\int_t^T v(s, x(s)) ds}\right) = d(\hat{u} G)$$

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$$= G d\hat{u} + \hat{u} dG$$

$$= G [L\hat{u}dt + b\hat{u}_x dw] + \hat{u} V G dt$$

these should have been $(s, x(s))$

otherwise you have a constant term!

then from t to T

$$E[g(x(T)) | x_t = x] - u(t, x) =$$

$$= \int_t^T E[G\hat{u}ds] + \int_t^T E[bG\hat{u}_x dW_s] + \int_t^T E[\hat{u}VG]ds$$

$$= \int_t^T E[G\hat{u}ds] - \int_t^T E[\hat{u}VG]ds + \int_t^T E[\hat{u}VG]ds$$

$$u(t, x) = E(g(x(T)) | x_t = x) - E\left[\int_t^T G\hat{u}ds\right]$$

Missing (t, x)

Not correct

Does not correspond to what you enunciated before on the previous page.

20/40

STATE how to apply MC to compute an integral

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Define $I = \int_a^b g(x) dx$

$g(x) : [0,1]^d \rightarrow \mathbb{R}$

Represent the integral as an expected value.

this is different from your domain of integration!!

NOT CONSISTENT

$I = \int_a^b \frac{g(x)}{f_x(x)} f_x(x) dx$

$f_x > 0$

f_x is density function

$g(x) \neq 0$

$I = E \left[\frac{g(x)}{f_x(x)} \right]$

Assume $[a,b]^d \rightarrow \mathbb{R}$

$f_x = \begin{cases} \frac{1}{(b-a)^d} & a < x < b \\ 0 & \text{else} \end{cases}$

NOT TRUE

are you working with $d=1$ or with a general dimension?

$E(g(x)) = \frac{1}{b-a} \int_a^b g(x) dx = (b-a) \frac{1}{N} \sum_{i=1}^N g(x_i)$

unbiased estimator of μ is sample mean

there is no time dependence in the original problem, why should there be any such here?

$E(g(x(t))) = \frac{\sum_{j=1}^N g(\bar{x}(t;w_j))}{N}$

where $\bar{x}$ is approx of x

i.e. Euler method ???

Ex let $I = \int_{[0,1]^d} f(x) dx$

$I = \int_{[0,1]^d} f(x) p(x) dx$

$p(x)$ is uniform density function

$I = E(f(x))$

$I \approx \frac{\sum_{i=0}^N f(x_i)}{N}$

Error in MC method

FRADF

2 parts to error

① Time discretisation error (TD error)

② Statistical error

← come back

no #13

- There will be NO TD error when $\frac{?}{\text{in the case of this problem!!}}$ (converges quickly)

- Statistical error will go to zero b/c of CLT.

$$E(g(x(T))) - \frac{\sum_{i=1}^N g(\bar{x}(T; w_i))}{N} = E(g(x(T)) - g(\bar{x}(T))) + \frac{\sum_{i=1}^N E(g(\bar{x}(T)) - g(\bar{x}(T; w_i)))}{N}$$

no time dependence in the problem I asked!

Ex. computation of error from previous example NO TD ERROR

$$E_N = \int_{[0,1]} f^2(x) dx - \left(\int_{[0,1]} f(x) dx \right)^2$$

this is not the MC error!

$$= \int_{[0,1]} \left(f(x) - \int_{[0,1]} f(x) dx \right)^2 dx$$

this does not depend on N at all!

as $\sqrt{N} E_N \rightarrow 0$ by CLT $\sim N(0,1)$

$\sigma^2 =$

Not with your def. of E_N

