

Probability Theory 大数定理，中心极限定理

$$\bar{Y} = \frac{1}{N}(x_1 + x_2 + \dots + x_N)$$

$$\bar{Y} = N(\mu, \frac{\sigma^2}{N})$$

$$\lim_{N \rightarrow \infty} P(|\bar{Y} - \mu| < \epsilon) = 1$$

变量： X_i : 随机变量

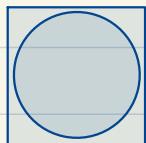
iid 独立同分布

Brownian Motion

$$\langle x^2 \rangle - \langle x \rangle^2 \approx 2Dt$$

大数定理 / 中心极限定理

布丰投针实验



$$\frac{\pi}{4} \approx \frac{N_c}{N}, \quad N_c: \text{在圆内的针数}$$

N: 总投针次数

统计力学 ~ 求平均值

Brownian Motion

$$1905 \quad Einstein \rightarrow \frac{\partial P}{\partial t} = \frac{\partial^2 P}{\partial x^2} D$$

$$P(x) \propto e^{-\frac{x^2}{4Dt}}$$

$$\langle x(t) \rangle - \langle x(0) \rangle^2 = 2Dt$$

1908 Langevin

$$m \ddot{\overrightarrow{x}} = -\gamma \dot{\overrightarrow{x}} - \nabla V(\vec{x}) + \overrightarrow{g}(t)$$

$\overrightarrow{g}(t)$, 随机力

方程的解是不可寻的。

Wiener 过程

Ito 依赖清晰

| Ito 引理

| 数值求解

Brown Motion $\rightarrow P(x, t)$

$$P(x_0, t+dt) = \int d\xi P(x+\xi, t) w(\xi) = \int d\xi \left(p(x, t) + \frac{\partial p}{\partial x} \xi + \frac{1}{2} \frac{\partial^2 p}{\partial x^2} \xi^2 \right) w(\xi)$$

$$\int d\xi w(\xi) = 1, \quad \int d\xi w(\xi) \xi = 0, \quad \int d\xi \xi^2 w(\xi) = \bar{\xi}^2$$

$$\Rightarrow P(x, t+dt) = P(x, t) + \frac{\bar{\xi}^2}{2} \frac{\partial^2}{\partial x^2} P(x, t)$$

$$\frac{\partial P(x, t)}{\partial t} dt = \frac{\bar{\xi}^2}{2} \frac{\partial^2}{\partial x^2} P(x, t)$$

↓

$$\frac{\partial P(x, t)}{\partial t} = D \frac{\partial^2}{\partial x^2} P(x, t) \quad \bar{\xi}^2 = 2Ddt$$

Ref: Langevin 1908 翻译本, Lemmon, Gythiel. 1997 "on the Brown Motion"

谐波方程

$$m\ddot{x} = -\eta \dot{x} + \xi$$

$$m \frac{d^2x}{dt^2} = -\eta \frac{dx}{dt} + \xi$$

$$m \times \frac{d^2x}{dt^2} = -\eta \times \frac{dx}{dt} + \xi x$$

$$\Rightarrow \frac{m}{2} \frac{d^2(x^2)}{dt^2} - m \left(\frac{dx}{dt} \right)^2 = -\frac{\eta}{2} \frac{d}{dt} x^2 + \xi x$$

此时的处理有一定的问题，

微分式法则不再成立（实际上）

$$\text{定义} Z = \langle \frac{dx}{dt} x^2 \rangle_e \quad \leftarrow e, \text{系统平均}$$

$$\Rightarrow \frac{m}{2} \frac{dZ}{dt} - \langle m v^2 \rangle = -\frac{1}{2} Z + \langle \xi x \rangle \\ = k_B T = 0$$

$$\Rightarrow \frac{m}{2} \frac{dZ}{dt} - k_B T = -\frac{1}{2} Z$$

$$\Rightarrow Z = (e^{-\frac{1}{m}t} + \frac{2k_B T}{\eta})$$

$$\text{长时间之后} \Rightarrow Z = \frac{2k_B T}{\eta} \Rightarrow \langle \frac{dx}{dt} x^2 \rangle = \frac{2k_B T}{\eta}$$

$$\Rightarrow \langle x^2 \rangle = \underbrace{\frac{2k_B T}{\eta} \cdot t}_{= 2D t}$$

$$D = \frac{k_B T}{\eta} \quad \leftarrow$$

$$= \underline{2D t}$$

Black-sholes Equation

$$ds = (\gamma + \xi) S dt \Rightarrow d(\ln S) = (\gamma + \xi) dt$$

$$\Rightarrow \ln S - \ln S_0 = \int_0^t (\gamma + \xi) dt$$

$$\Rightarrow S = S_0 e^{\gamma + \int_0^t \xi dt'}$$

似乎总是赚钱，与生活不符。

对于此类问题的求解需要改变。

$$\dot{x} = a + b \xi(t), \quad a, b \text{ 常数}, \text{ 而 } \langle \xi(t) \rangle = 0. \quad \langle \xi(t) \xi(t') \rangle = \delta(t-t')$$

$$x = x_0 + at + b \int_0^t \xi(t') dt'$$

$$\text{定义 } w(t) = \int_0^t \xi(t') dt'$$

$$\begin{aligned} \text{由 } \langle w(t) w(t') \rangle &= \int_0^t dt' \int_0^{t'} ds' \langle \xi(t') \xi(s') \rangle \\ &= \int_0^t dt' \int_0^{t'} ds' \delta(t'-s') \\ &= \int_0^{\min(t,s)} dt' = \min(t,s) \end{aligned}$$

$$\langle w^2(t) \rangle = t \Rightarrow \bar{x} = x_0 + at$$

$$\langle x^2 \rangle - \langle x \rangle^2 = b^2 t$$

离散化后如何处理。

$$\int_0^t \xi(t') dt' = \bar{\xi}_n \delta t \xi_n$$

$$\text{其中 } \bar{\xi}_n = 0, \text{ 而 } t = \bar{\xi}_n \delta t^2 \langle \xi_n^2 \rangle$$

$$\Rightarrow \langle \xi_n^2 \rangle = \frac{1}{\delta t}, \text{ 上式才能成立。}$$

$$k_1 \quad X = x_0 + at + \frac{N}{\pi} st \xi_n. \quad \xi_n \sim N(0, \frac{1}{st})$$