

Scientific Diary

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Chapter 1

Probability

Theorem 1.0.1 (Kolmogorov).

Пусть $\nu_{t_1 t_2 t_3 \dots t_k}$ для $\forall t_1, t_2, t_3 \dots t_k \in T$, $k \in \mathbb{N}$ являются вероятностными мерами на \mathbb{R}^{k^n} такими, что:

$$\nu_{t_{\sigma(1)}, \dots, t_{\sigma(k)}}(F_1 \times \dots \times F_k) = \nu_{t_1, \dots, t_k}(F_{\sigma^{-1}(1)} \times \dots \times F_{\sigma^{-1}(k)}) \quad (1.1)$$

для всех перестановок $\sigma \in S_k$

$$\begin{aligned} \nu_{t_1, \dots, t_k}(F_1 \times \dots \times F_k) \\ = \nu_{t_1, \dots, t_k, t_{k+1}, \dots, t_{k+m}}(F_1 \times \dots \times F_k \times \mathbb{R}^n \times \dots \times \mathbb{R}^n) \end{aligned} \quad (1.2)$$

Тогда $\exists(\Omega, \mathcal{F}, P)$ и случайный процесс $\{X_t\}$ на Ω , $X_t : \Omega \rightarrow \mathbb{R}^n$

$$\nu_{t_1, \dots, t_k}(F_1 \times \dots \times F_k) = P[X_{t_1} \in F_1, \dots, X_{t_k} \in F_k] \quad (1.3)$$

Exercise 1.0.1 (Irwin-Hall distribution).

Try to derive Irwin-Hall distribution formula for pdf convolution formula.

Chapter 2

Monte-Carlo Methods

Code Base

Heuristic 2.0.1.

There is one useful thing that i found while making Monte-Carlo simulations. Sometimes it is computationally costly to generate random variable from normal distribution and very good approximation for it appeared to be Irwin-Hall distribution.

$$X_n = \sum_{i=0}^n U_k \quad \text{where } U_k \text{ are independent random variables drawn from uniform distribution } U(0,1) \quad (2.1)$$

The density function is given by:

$$f_X(x; n) = \frac{1}{2(n-1)!} \sum_{k=0}^n (-1)^k \binom{n}{k} (x-k)^{n-1} \text{sgn}(x-k) \quad (2.2)$$

This pdf is basically piecewise polynomial function with $\mu = \frac{n}{2}$ and $\sigma = \frac{n}{12}$.

For $n = 12$ it gives good approximation for normal distribution pdf.

$$\phi(x) \approx \sqrt{\frac{12}{n}} (f_X(x; n) - \frac{n}{2}) - 6 \quad (2.3)$$

$$\phi(x) \approx f_X(x; n) - 6 = \sum_{i=0}^{12} U_k \quad (2.4)$$

2.1 Financial Instrument Pricing

2.1.1 Black-Scholes Model

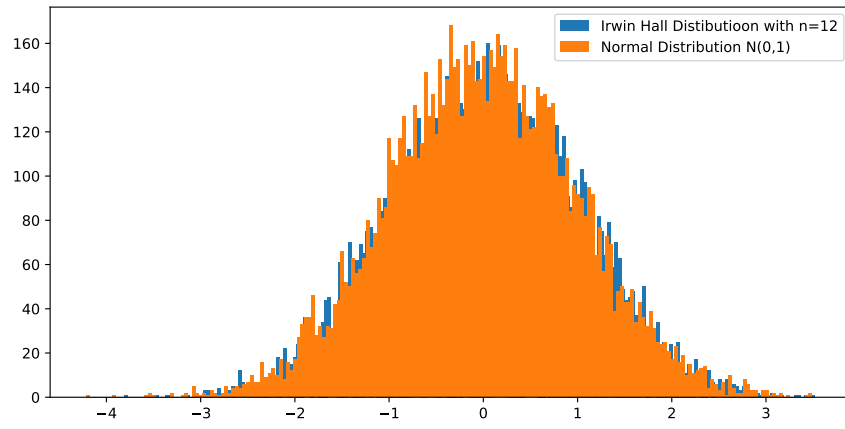


Figure 2.1:

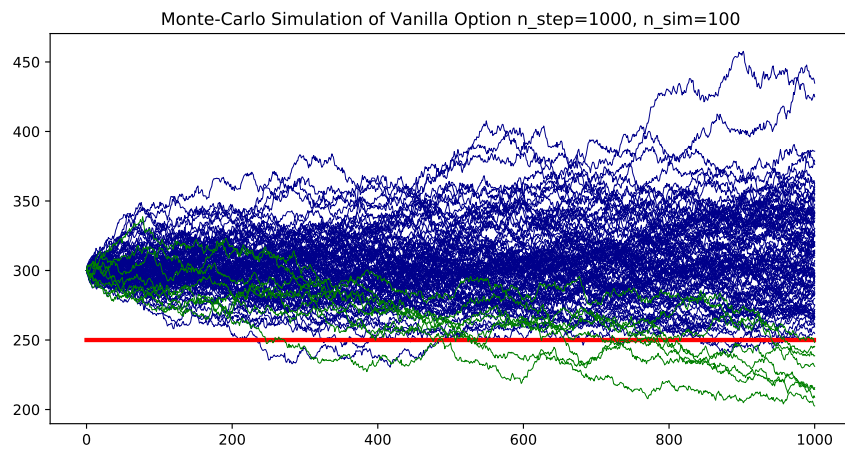


Figure 2.2:

Chapter 3

Matrix Models

Chapter 4

Stochastic Processes

	dt	dW _t
dt	0	0
dW _t	0	dt

Exercise 4.0.1 (Shreve 4.6).

Let $S_t = S(0) \exp((\alpha - \frac{\sigma^2}{2})t - \sigma W_t)$ be a geometric Brownian motion.
Let p be a positive constant. Compute $d(S_t^p)$

Solution 4.0.1.

First of all lets apply Ito-Doebelin formula to the function $f(x) = x^p$

$$df(S_t) = \partial_x f(S_t) dS + \frac{1}{2} \partial_{xx}^2 f(S_t) (dS_t)^2 \quad (4.1)$$

For $dS(t)$ we have

$$dS_t = \alpha S_t dt + \sigma S_t dW_t \quad (4.2)$$

To compute the $(dS_t)^2$ we apply mnemonic multiplication rule for stochastic differentials.

$$(dS_t)^2 = (\sigma S_t dW_t + \alpha S_t dt)^2 = \sigma^2 S_t^2 (dW_t)^2 + 2\sigma\alpha S_t^2 dt dW_t + \alpha^2 S_t^2 (dt)^2 = \sigma^2 S_t^2 dt \quad (4.3)$$

Substituting derivatives and differentials of all orders to the formula we get:

$$d(S_t^p) = pS_t^{p-1} \left(\sigma dW_t + \left(\alpha + \frac{p-1}{2} \sigma^2 \right) dt \right) \quad (4.4)$$

Exercise 4.0.2 (Shreve 4.7).
i Compute dW_t^4 and then write $W^4(T)$ as the sum of the ordinary (Lebesgue) integral with respect to time and an Ito integral.

ii Take expectations on both sides of the formula you obtained in i, use the fact that $\mathbb{E}[W_t^2] = t$, and derive formula $\mathbb{E}[W_T^4] = 3T^2$.

iii Use the method of i and ii to derive a formula for $\mathbb{E}[W_T^6]$.

Solution 4.0.2. i

$$\begin{aligned}
 df(W_t) &= \partial_x f(W_t) dW_t + \frac{1}{2} \partial_{xx}^2 f(W_t) (dW_t)^2, \text{ where } f(W_t) = W_t^4. \\
 dW_t^4 &= 4W_t^3 dW_t + \frac{4 \cdot 3}{2} W_t^2 (dW_t)^2 \\
 \int_0^T dW_t^4 &= \int_0^T 4W_t^3 dW_t + \int_0^T 6W_t^2 dt \\
 W_T^4 &= \int_0^T 4W_t^3 dW_t + \int_0^T 6W_t^2 dt
 \end{aligned} \tag{4.5}$$

ii

$$\mathbb{E}[W_T^4] = 4 \cdot \mathbb{E}\left[\int_0^T W_t^3 dW_t\right] + 6 \cdot \mathbb{E}\left[\int_0^T W_t^2 dt\right] \tag{4.6}$$

I will allow myself to be more detailed in this part because the following calculations are very useful and appear in many other places.

Firstly, let's handle the first summand and prove that it equals zero. One can recall theorem from stochastic calculus that states that integral of quadratically integrable process w.r of Wiener process is martingale and if we prove that W_t^3 is quadratically integrable the calculation of expectation becomes straightforward. Let's show it for more general case W_t^n .

$$\mathbb{E}[X^n] = \begin{cases} 0 & n \text{ is odd} \\ \sigma^n (n-1)!! & n \text{ is even} \end{cases} \quad \text{where } X \sim \mathcal{N}(0, \sigma^2) \tag{4.7}$$

$$\mathbb{E}\left[\int_0^t W_s^n ds\right] = \int_0^t \mathbb{E}[W_s^n] ds = \begin{cases} 0 & n \text{ is odd} \\ (n-1)!! \int_0^t s^{n/2} ds & n \text{ is even} \end{cases} \tag{4.8}$$

So the quadratica variation of $W_t^3 : \int_0^T W_s^6 ds < \infty$ From this follows that

$$\mathbb{E}\left[\int_0^T W_t^3 dW_t | \mathcal{F}_0\right] = 0 \tag{4.9}$$

$$\mathbb{E}[W_T^4] = 6 \cdot \int_0^T \mathbb{E}[W_t^2] dt = 6 \cdot (1)!! \cdot \frac{1}{2} T^2 = 3T^2 \tag{4.10}$$

iii

$$\begin{aligned}
 dW_t^6 &= 6W_t^5 dW_t + 15W_t^4 dt. \\
 W_t^6 &= 6 \int_0^T W_t^5 dW_t + 15 \int_0^T W_t^4 dt \\
 \mathbb{E}[W_T^6] &= 6 \cdot \mathbb{E}\left[\int_0^T W_t^5 dW_t\right] + 15 \cdot \mathbb{E}\left[\int_0^T W_t^4 dt\right] = 15 \cdot \mathbb{E}\left[\int_0^T W_t^4 dt\right] = 15T^3
 \end{aligned} \tag{4.11}$$