Assignment

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1 Probability

2 Assignment-3

(Answer-1) Let $\langle S^N \rangle = \frac{b-a}{N} \sum_{i=0}^N f(x_i)$, where x_i are randomly chosen points from the interval [a,b]. Let X be a uniform distributed random variable on [a,b]. The expected value of $\langle S^N \rangle$ is as follows:

$$E(\langle S^N \rangle) = E(\frac{b-a}{N} \sum_{i=0}^{N} f(x_i))$$
$$= \frac{b-a}{N} \sum_{i=0}^{N} E(f(x_i))$$
$$= \frac{1}{N} \sum_{i=0}^{N} \int_{a}^{b} f(x) dx$$

By law of large number we have

$$P\left(\lim_{N\to\infty}\langle S^N\rangle=\int\limits_a^bf(x)dx\right)=1.$$

(Answer-2)

Let $f(x) = \frac{4}{1+x^2}$ and $\langle S^N \rangle = E(\frac{1}{N} \sum_{i=0}^N f(x_i))$. Then as seen in the above exercise,

$$E(\langle S^N \rangle) = \int_0^1 f(x)dx.$$

The error in the estimation is given by the standard deviation σ , computed as follows:

$$\sigma^{2} = \sigma^{2} \left(\frac{1}{N} \sum_{i=0}^{N} f(x_{i}) \right)$$

$$= \frac{1}{N^{2}} \sum_{i=0}^{N} \sigma^{2}(f(x_{i}))$$

$$= \frac{1}{N^{2}} \sum_{i=0}^{N} (E[f(x)^{2}] - E[f(x)]^{2})$$

$$= \frac{E[f(x)^{2}] - E[f(x)]^{2}}{N}$$

$$\Rightarrow \sigma = \frac{\sqrt{E[f(x)^{2}] - E[f(x)]^{2}}}{\sqrt{N}}.$$

(Answer-5)

- Let $dX = \mu dt + \sigma dB(t)$ where B(t) is the Canonical Brownian Motion.
 - (a) Let $f(X_t, t) = X_t^2$. In this case, we have $\frac{\partial f}{\partial t} = 0$, $\frac{\partial f}{\partial X_t} = 2X_t$, and $\frac{\partial^2 f}{\partial X_t^2} = 2$. Thus, by Ito's lemma, we obtain

$$df = \frac{\partial f}{\partial t}dt + \frac{\partial f}{\partial X_t}dX_t + \frac{1}{2}\frac{\partial^2 f}{\partial X_t^2}\sigma^2 dt$$

$$= \left(\frac{\partial f}{\partial t} + \frac{\partial f}{\partial X_t}\mu\right)dt + \frac{\partial f}{\partial X_t}\sigma dB(t) + \frac{1}{2}\frac{\partial^2 f}{\partial X_t^2}\sigma^2 dt$$

$$= (2X_t\mu + \sigma^2)dt + 2X_t\sigma dB(t).$$

(b) Let $f(X_t, t) = X_t^3$. In this case, we have $\frac{\partial f}{\partial t} = 0$, $\frac{\partial f}{\partial X_t} = 3X_t^2$, and $\frac{\partial^2 f}{\partial X_t^2} = 6X_t$. Thus, by Ito's lemma, we obtain

$$df = \left(\frac{\partial f}{\partial t} + \frac{\partial f}{\partial X_t}\mu\right)dt + \frac{\partial f}{\partial X_t}\sigma dB(t) + \frac{1}{2}\frac{\partial^2 f}{\partial X_t^2}\sigma^2 dt$$
$$= 3X_t(X_t\mu + \sigma^2)dt + 3X_t^2\sigma dB(t).$$

(c) Let $f(X_t, t) = log(X_t)$. Then $\frac{\partial f}{\partial t} = 0$, $\frac{\partial f}{\partial X_t} = \frac{1}{X_t}$, and $\frac{\partial^2 f}{\partial X_t^2} = \frac{-1}{X_t^2}$. Thus, by Ito's lemma, we obtain

$$df = \left(\frac{\partial f}{\partial t} + \frac{\partial f}{\partial X_t}\mu\right)dt + \frac{\partial f}{\partial X_t}\sigma dB(t) + \frac{1}{2}\frac{\partial^2 f}{\partial X_t^2}\sigma^2 dt$$
$$= \frac{1}{X_t}\left(\mu - \frac{1}{2X_t}\sigma^2\right)dt + \frac{1}{X_t}\sigma dB(t).$$

• Let $\frac{dS}{S} = \mu dt + \sigma dB(t)$ where B(t) is the Canonical Brownian Motion. For $f(X_t, t) \in \mathcal{C}^2(\mathbb{R}^2, \mathbb{R})$,

Ito's lemma states

$$df = \frac{\partial f}{\partial t}dt + \frac{\partial f}{\partial X_t}dX_t + \frac{1}{2}\frac{\partial^2 f}{\partial X_t^2}dX_t^2$$

$$= \frac{\partial f}{\partial t}dt + \frac{\partial f}{\partial X_t}(\mu X_t dt + \sigma X_t dB(t)) + \frac{1}{2}\frac{\partial^2 f}{\partial X_t^2}(\mu X_t dt + \sigma X_t dB(t))^2$$

$$= \frac{\partial f}{\partial t}dt + \frac{\partial f}{\partial X_t}(\mu X_t dt + \sigma X_t dB(t)) + \frac{1}{2}\frac{\partial^2 f}{\partial X_t^2}\sigma^2 X_t^2 dt$$

$$= \left(\frac{\partial f}{\partial t} + \frac{\partial f}{\partial X_t}\mu X_t + \frac{1}{2}\frac{\partial^2 f}{\partial X_t^2}\sigma^2 X_t^2\right) dt + \frac{\partial f}{\partial X_t}\sigma X_t dB(t)$$

(a) Let $f(S_t, t) = S_t^2$. In this case, we have $\frac{\partial f}{\partial t} = 0$, $\frac{\partial f}{\partial S_t} = 2S_t$, and $\frac{\partial^2 f}{\partial S_t^2} = 2$. Substituting these values in the above equation we get

$$df = (2\mu S_t^2 + \sigma^2 S_t^2)dt + 2\sigma S_t^2 dB(t)$$

= $S_t^2((2\mu + \sigma^2)dt + 2\sigma dB(t)).$

3 Assignment-4

(Answer-1)

Let M(t) denote the amount of money at time $t \in [0, T]$ after continuously compounding at the rate of r(t). Then it will follow the following differential equation:

$$\frac{dM(t)}{dt} = r(t)M(t).$$

Solving this we get

$$\frac{dM(t)}{M(t)} = r(t)dt$$

$$\int_{M}^{M(T)} \frac{dM(t)}{M(t)} = \int_{0}^{T} r(t)dt$$

$$\ln\left(\frac{M(T)}{M}\right) = \int_{0}^{T} r(t)dt$$

$$M(T) = Me_{0}^{\int_{0}^{T} r(t)dt}.$$