

BRIEF ARTICLE

1. SOLUTION TO HOMEWORK 1

use equations 15.11 and 15.12 from the textbook. for the investor in Japan, the value of the call is

$$c = S_0 e^{-r_D T} N(d_1) - K e^{-r_J T} N(d_2),$$

where

$$d_1 = \frac{\ln(S_0/K) + (r_J - r_D + \sigma^2/2)T}{\sigma\sqrt{T}},$$

$$d_2 = \frac{\ln(S_0/K) + (r_J - r_D - \sigma^2/2)T}{\sigma\sqrt{T}},$$

r_D denotes the risk-free interest rate in US currency, r_J denotes the risk-free interest rate in Japanese currency, S_0 is the value of one dollar at time $t = 0$. In our case $S_0 = 90$ Yen, and K is the strike price, which equals S_0 for an ATM call. Hence (in Yen)

$$c = 90 e^{-r_D T} N(d_1) - 90 e^{-r_J T} N(d_2),$$

where

$$d_1 = \frac{(r_J - r_D + \sigma^2/2)T}{\sigma\sqrt{T}}, \quad d_2 = \frac{(r_J - r_D - \sigma^2/2)T}{\sigma\sqrt{T}}$$

similarly for US investor, (in USD)

$$p = e^{-r_D T} N(-\tilde{d}_2) - e^{-r_J T} N(-\tilde{d}_1)$$

where

$$\tilde{d}_2 = -d_1, \quad -\tilde{d}_1 = d_2.$$

by comparing the equations we see that the value of an ATM call on one dollar is the same as the value of an ATM put.

2. SOLUTION TO HOMEWORK 2

Suppose that the returns of two stocks (A and B) have the same standard deviation (30%) and correlation matrix:

$$M = \begin{pmatrix} 1 & 0.7 \\ 0.7 & 1 \end{pmatrix}.$$

a) Computer the eigenvalues and eigenvectors of the matrix.

Solving $\det(M - \lambda I) = 0$, we get $\lambda_1 = 0.3$ and $\lambda_2 = 1.7$. Substituting the eigenvector in $(M - \lambda I)\vec{X} = 0$. we get the the eigenvector corresponding to 0.3 is $\frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ -1 \end{pmatrix}$ and the eigenvector corresponding to 0.7 is $\frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 1 \end{pmatrix}$.

b) Describe how to set up a Monte Carlo simulation for stocks A and B that takes into account their volatilities and correlation.

We assume that stocks A and B follow the following processes

$$(1) \quad dA = rA dt + 0.3A dX_1,$$

$$(2) \quad dB = rB dt + 0.3B dX_2,$$

with correlation matrix $\begin{pmatrix} 1 & 0.7 \\ 0.7 & 1 \end{pmatrix}$.

is a symmetric matrix whose eigenvalue decomposition is

$$M = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix} \begin{pmatrix} 1.7 & 0 \\ 0 & 0.3 \end{pmatrix} \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix}.$$

We can generate Brownian motions $dX_1 dX_2$ with correlation matrix M by letting

$$\begin{pmatrix} dX_1 \\ dX_2 \end{pmatrix} = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix} \begin{pmatrix} \sqrt{1.7} & 0 \\ 0 & \sqrt{0.3} \end{pmatrix} \begin{pmatrix} dW_1 \\ dW_2 \end{pmatrix},$$

where W_1 and W_2 are two standard normal variables. Multiplying out, we get

$$(3) \quad dX_1 = \sqrt{0.85} dW_1 + \sqrt{0.15} dW_2$$

$$(4) \quad dX_2 = \sqrt{0.85} dW_1 - \sqrt{0.15} dW_2$$

By solving equations (1), (2), we can set up a Monte Carlo Simulation for stocks A and B. Define $A(t)$ as the stock price of A at time t, $B(t)$ as the stock stock price of B at time t. W_1, W_2 are two independent normal variables. Then

$$A(t) = A(t - \Delta t) \exp\left(\left(r - \frac{.3^2}{2}\right)\Delta t + 0.3\sqrt{\Delta t}X_1\right),$$

$$A(t) = A(t - \Delta t) \exp\left(\left(r - \frac{.3^2}{2}\right)\Delta t + 0.3\sqrt{\Delta t}X_1\right),$$

where

$$dX_1 = \sqrt{0.85} dW_1 + \sqrt{0.15} dW_2,$$

$$dX_2 = \sqrt{0.85} dW_1 - \sqrt{0.15} dW_2.$$

3. SOLUTION TO HOMEWORK 3

3.1. Explain why the implicit method is called "Implicit". Explicit methods calculate the state of a system at a later time from the state of a system at the current time. Mathematically, if $Y(t)$ is the state of that system, and $Y(t + \Delta t)$ is the state of that system at a later time then, for an explicit method,

$$(5) \quad Y(t + \Delta t) = F(Y(t)),$$

where F is a function of the state Y . In other words, we can solve for $Y(t + \Delta t)$ explicitly. when we don't have equation of the form (5) but we can still find a method where both $Y(t)$ and $Y(t + \Delta t)$ are related via an equation of the form

$$(6) \quad G(Y(t), Y(t + \Delta t)) = 0$$

The method is called "implicit" because we can no longer solve for $Y(t + \Delta t)$ explicitly.

3.2. Suppose that you have a random number generator for a uniform distribution: a) How do you use it to generate a normal variable ?

We can use Box-Muller method. Let x_1 and x_2 be random variables uniformly distributed on the interval $(0, 1)$. Then

$$X_1 = \sqrt{-2 \ln x_1} \cos(2\pi x_2), \quad X_2 = \sqrt{-2 \ln x_1} \sin(2\pi x_2)$$

are two independent standard normal random variables.

b) How do you simulate a pair of normal variables X and Y with correlation ρ ?

First simulate two independent standard normal variables X_1 and X_2 as in a). We can prove that variables $X = X_1, Y = \rho X_1 + \sqrt{1 - \rho^2} X_2$ are standard normal with correlation ρ . First, Y is normal since sum of two normal variables is a normal variable. Next, $EY = \rho EX_1 + \sqrt{1 - \rho^2} EX_2 = 0$ and $var(Y) = \rho^2 EX_1^2 + (1 - \rho^2) EX_2^2$ since X_1 and X_2 are independent. It follows,

$$Corr(X, Y) = \frac{cov(X, Y)}{\sqrt{var(X)var(Y)}} = \rho.$$

4. SOLUTION TO HOMEWORK 4

1) A random variable X follows an Exponential distribution with parameter λ if its density is given by

$$f(x) = \lambda e^{-\lambda x} \mathbf{1}_{x>0}.$$

Suppose that you obtain a sample x_1, \dots, x_n from the random variable X . How would you estimate λ using maximum likelihood?

The likelihood

$$\begin{aligned}\mathcal{L}(\lambda|x_1, \dots, x_n) &= f(x_1, \dots, x_n|\lambda) = \prod_i f(x_i|\lambda) = \lambda^n e^{-\lambda \sum_i x_i} 1_{\{x_1>0, \dots, x_n>0\}} \\ &= e^{n \ln \lambda - \lambda \sum_i x_i} 1_{\{x_1>0, \dots, x_n>0\}}\end{aligned}$$

the average log-likelihood is

$$\begin{aligned}\hat{l} &= \frac{\ln \mathcal{L}(\lambda|x_1, \dots, x_n)}{n} = \ln \lambda - \lambda \frac{\sum_i x_i}{n} \\ \frac{d\hat{l}}{d\lambda} &= \frac{1}{\lambda} - \frac{\sum_i x_i}{n}\end{aligned}$$

As a function of λ , \hat{l} increases for $\lambda < \frac{\sum_i x_i}{n}$ and decreases when $\lambda > \frac{\sum_i x_i}{n}$. hence it take maximum at $\lambda = \frac{\sum_i x_i}{n}$. The maximum likelihood estimator of λ is $\frac{\sum_i x_i}{n}$.