

MATH 565 Monte Carlo Methods in Finance

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Test

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Instructions:

- i. This test consists of FOUR questions. Answer all of them.*
- ii. The time allowed for this test is 75 minutes*
- iii. This test is closed book, but you may use 4 double-sided letter-size sheets of notes.*
- iv. Calculators, even of the programmable variety, are allowed. Computers, but only using MATLAB or JMP, are also allowed. No internet access.*
- v. Show all your work to justify your answers. Answers without adequate justification will not receive credit.*

1. (25 marks)

Let $X_1, \dots, X_n, X_{n+1}, \dots$ be independent and identically distributed random variables with mean μ and variance σ^2 . Define the sample mean of the first n random variables by

$$\bar{X} = \frac{1}{n}(X_1 + \dots + X_n).$$

a) Derive the mean and variance of \bar{X} . Is \bar{X} a biased or unbiased estimator of μ ?

Answer:

$$E[\bar{X}] = \frac{1}{n}E\{X_1 + \dots + X_n\} = \frac{n\mu}{n} = \mu,$$
$$\text{var}(\bar{X}) = \frac{1}{n^2} \text{var}(X_1 + \dots + X_n) = \frac{1}{n^2} [\text{var}(X_1) + \dots + \text{var}(X_n)] = \frac{n\sigma^2}{n^2} = \frac{\sigma^2}{n}.$$

Thus, \bar{X} is an unbiased estimator of μ .

b) Define the sample mean of the next m random variables X_i by

$$Y = \frac{1}{m}(X_{n+1} + \dots + X_{n+m}).$$

We may think of \bar{X} and Y as two Monte Carlo estimators of μ . Which of these has the smaller variance?

Answer: Since $\text{var}(Y) = \sigma^2/m$, then \bar{X} has the smaller variance iff $n > m$, and Y has the smaller variance iff $m > n$.

c) Let $Z = a\bar{X} + (1 - a)Y$ for some choice of constant a . This is a third Monte Carlo estimator of μ derived from first two estimators. Show that Z is unbiased, regardless of the choice of a .

Answer:

$$E[Z] = E[a\bar{X} + (1 - a)Y] = aE[\bar{X}] + (1 - a)E[Y] = a\mu + (1 - a)\mu = \mu.$$

- d) For what choice of a does the unbiased Monte Carlo estimator Z have minimum variance, and what is this minimum variance?

Answer:

$$\begin{aligned}
 \text{var}(Z) &= \text{var}(a\bar{X} + (1-a)Y) = a^2 \text{var}(\bar{X}) + (1-a)^2 \text{var}(Y) \\
 &= a^2 \frac{\sigma^2}{n} + (1-a)^2 \frac{\sigma^2}{m} = \sigma^2 \left[\frac{a^2}{n} + \frac{1-2a+a^2}{m} \right] \\
 &= \sigma^2 \left[a^2 \left(\frac{1}{n} + \frac{1}{m} \right) - \frac{2a}{m} + \frac{1}{m} \right] = \frac{\sigma^2}{m} \left[a^2 \left(\frac{n+m}{n} \right) - 2a + 1 \right] \\
 &= \sigma^2 \left(\frac{n+m}{nm} \right) \left[\left(a - \frac{n}{n+m} \right)^2 - \left(\frac{n}{n+m} \right)^2 + \frac{n}{n+m} \right] \\
 &= \sigma^2 \left(\frac{n+m}{nm} \right) \left[\left(a - \frac{n}{n+m} \right)^2 + \frac{mn}{(n+m)^2} \right] \\
 &= \sigma^2 \left[\left(\frac{n+m}{nm} \right) \left(a - \frac{n}{n+m} \right)^2 + \frac{1}{n+m} \right]
 \end{aligned}$$

The minimum variance occurs for $a = n/(n+m)$ and $1-a = m/(n+m)$, so

$$Z = a\bar{X} + (1-a)Y = \frac{1}{n+m} (X_1 + \cdots + X_n + X_{n+1} + \cdots + X_{n+m}),$$

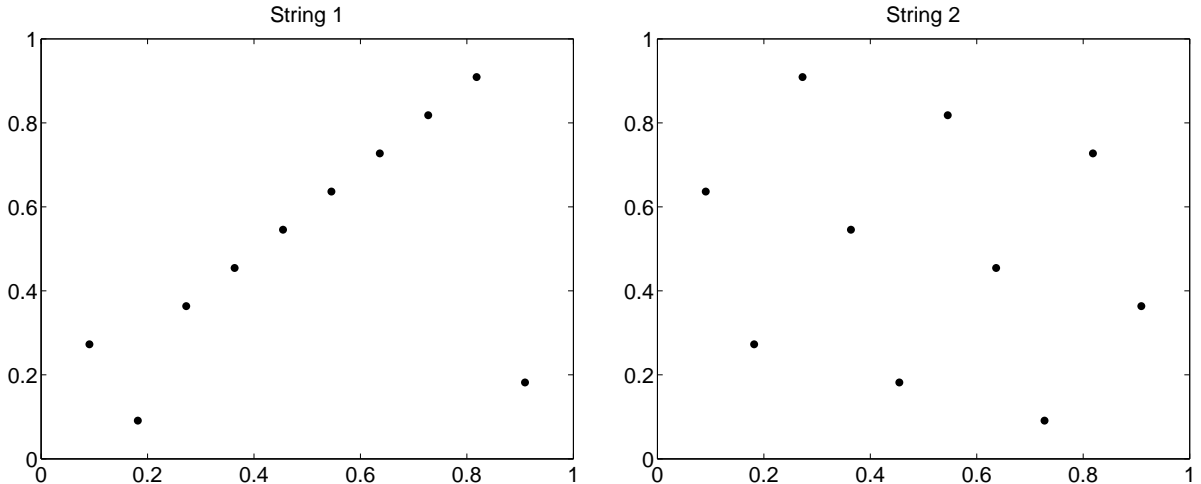
and $\text{var}(Z) = \sigma^2/(n+m)$.

2. (20 marks)

Consider the two periodic strings of uniform pseudorandom numbers. Although neither is perfect, which one is better and why?

| | | | | | | | | | | | | |
|----------|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|-----|
| String 1 | 0.364 | 0.455 | 0.545 | 0.636 | 0.727 | 0.818 | 0.909 | 0.182 | 0.091 | 0.273 | 0.364 | ... |
| String 2 | 0.364 | 0.545 | 0.818 | 0.727 | 0.091 | 0.636 | 0.455 | 0.182 | 0.273 | 0.909 | 0.364 | ... |

Answer: If we plot the ordered pairs (x_i, x_{i+1}) , $i = 1, 2, \dots$, the second string fills the square better than the first, in which most of these points lie on a single line. Thus, the second string is better.



3. (25 marks)

Consider the situation where the probability of having at least one taxi waiting at the taxi stand is 50%, in which case the waiting time is 0. If there is no taxi waiting at the taxi stand, then the waiting time for the next taxi to arrive is exponential with a mean of 2 minutes. This means that the cumulative distribution function for the waiting time for a taxi, T , is

$$F(t) = \text{Prob}(T \leq t) = \begin{cases} 0, & -\infty < t < 0. \\ 1 - \frac{1}{2}e^{-t/2}, & 0 \leq t < \infty. \end{cases}$$

Use the String 1 of pseudorandom numbers in the previous problem to produce *five pseudo-random numbers with this distribution*.

Answer: We may use the inverse cumulative distribution function method. Let $x = F(t)$, and $t = F^{-1}(x)$. Then

$$F^{-1}(x) = \begin{cases} 0, & 0 \leq x \leq 0.5, \\ -2 \log(2(1-x)), & 0.5 < x < 1. \end{cases}$$

Then, we compute $T_i = F^{-1}(X_i)$ to get

| | | | | | | | |
|----------|-------|-------|-------|-------|-------|-------|-----|
| String 1 | X_i | 0.364 | 0.455 | 0.545 | 0.636 | 0.727 | ... |
| | T_i | 0 | 0 | 0.189 | 0.635 | 1.210 | ... |

4. (30 marks)

There are two stocks whose prices right now are both \$100, and whose prices one year later will be $S_j(1) = 100e^{(0.03 - \sigma_j^2/2) + \sigma_j X_j}$ for $j = 1, 2$, where the X_1 and X_2 are i.i.d. $N(0, 1)$ random variables, and $\sigma_1 = 0.5$ and $\sigma_2 = 0.9$. A basket call option pays

$$\max(\max(S_1(1), S_2(1)) - 100, 0)e^{-0.03}$$

in today's dollars. Perform a Monte Carlo simulation to compute the fair price of this option with a relative error of no more than 1%, i.e., the 95% confidence interval should have a half width satisfying this tolerance. How many samples do you need? Write down your program or the pseudo-code for your algorithm as well as the results of your simulation.

Answer: The code for this program plus the output is given below. The value of n may be found by choosing a trial value, n_0 , obtaining a relative error estimate ε_0 , and then estimating $n = 1.2n_0(\varepsilon_0/0.01)^2$. The value $n = 100000$ is adequate.

```
% Basket option
r=0.03; %interest rate
sig1=0.5; sig2=0.9; %volatilities
n=1e3 %number of samples
tol=0.01;
%compute stock prices & payoff
stock1=100*exp((r-sig1*sig1/2)+sig1*randn(n,1));
stock2=100*exp((r-sig2*sig2/2)+sig2*randn(n,1));
payoff=max(max(stock1,stock2)-100,0)*exp(-r);
```

```

price=mean(payoff) %MC estimate of fair price
ciwidth=1.96*std(payoff)/sqrt(n) %confidence interval half-width
relerr=ciwidth/price %estimate relative error
nest=1.2*n*(relerr/tol)^2 %estimated n to meet tolerance

```

```

n =
    1000
price =
    53.3233
ciwidth =
    6.1874
relerr =
    0.1160
nest =
    1.6157e+05
n =
    200000
price =
    52.5749
ciwidth =
    0.4354
relerr =
    0.0083
nest =
    1.6458e+05

```