IEOR E4602: Quantitative Risk Management (Spring 2016)

Columbia University

Instructor: Martin Haugh Assignment 7: Due Monday 2nd May 2016

Answer Questions 3, 4 and two of the other three questions.

Question 1

This question explains how importance sampling can be extremely useful for rare-event simulation. Consider the problem of estimating

$$\theta(x) = \mathrm{E}[e^{\sqrt{Z}}I_{\{Z \ge x\}}]$$

for $x \geq 1$ and where $Z \sim \mathcal{N}(0,1)$. Let $X := e^{\sqrt{Z}} I_{\{Z \geq x\}}$.

- (a) Show that $\theta(x) \geq (1 \Phi(x))e^{\sqrt{x}}$ where $\Phi(.)$ is the CDF of a standard normal random variable.
- (b) Show that $E[X^2] \ge (1 \Phi(x))e^{2\sqrt{x}}$.
- (c) Consider doing importance sampling with a new random variable $W \sim N(\mu_x, 1)$. (That is, we shift the mean by μ_x , where μ_x depends on x.) Show that the application of the 'maximum principle' yields $\mu_x = x$.
- (d) Now set $\mu_x = x$. Use the importance sampling principle to show that $\theta = E[Y]$ where

$$Y = e^{x^2/2} e^{\sqrt{W}} e^{-Wx} I_{\{W > x\}}.$$

Show also that $\mathrm{E}[Y] \leq e^{-x^2/2} \mathrm{E}[e^{\sqrt{W}} I_{\{W>x\}}]$.

(e) Using the fact that W may be expressed as Z + x, show that

$$\theta(x) < e^{1-(x-1)^2/2}$$
.

Hence show that $\theta(x) \to 0$ as $x \to \infty$.

(f) We will now develop an upper bound for the second moment using importance sampling. Show that

$$E[Y^2] \le e^{-(x-1)^2} e^3.$$

(g) Using the above results, show that

$$\frac{\mathrm{Var}(X)}{\mathrm{Var}(Y)} \to \infty$$

1

as $x \to \infty$. (You should use L'Hopitals rule for computing the limit.) This means that the improvement factor of the importance sampling algorithm over the standard simulation algorithm tends to infinity as the event becomes rarer.

Question 2

Consider the cumulant generating function, $\psi(t)$, as defined in the paragraph immediately following Example 8 of the *Monte-Carlo Methods for Risk Management* lecture notes. We will adopt the same setting as described in that paragraph so that $S_n = \sum_{i=1}^n X_i$ where the X_i 's are IID with $E[X_1] < 0$. We wish to estimate $\theta := P(\tau_a < \infty)$ where $\tau_a := \min\{n \ge 0 : S_n \ge a\}$. Then the expression in the lecture notes states that

$$\theta = \mathrm{E}_t \left[I_{\{\tau_a < \infty\}} e^{-tS_{\tau_a} + \tau_a \psi(t)} \right].$$

We also assume that $\psi(t)$ is finite in a neighborhood of the origin.

- (a) Show that $\psi'(t) = E_t[X]$, where X is a generic random variable with the same distribution as the X_i 's.
- (b) Show that $\psi''(t) = \operatorname{Var}_t(X)$ so that $\psi(t)$ is convex.
- (c) Suppose there exists some t such that $0 < \psi'(t) < \infty$. Explain why this implies t > 0 and why we can therefore write $\theta = \operatorname{E}_t \left[e^{-tS_{\tau_a} + \tau_a \psi(t)} \right]$. (Drawing a graph of $\psi(t)$ helps.)
- (d) Suppose there exists a t > 0 such that $\psi(t) > 0$. Explain why there must be a unique $t^* > 0$ such that $\psi(t^*) = 0$. (Again a graph of $\psi(t)$ helps.)
- (e) Show that $\theta = \mathbb{E}_{t^*}\left[e^{-t^*S_{\tau_a}}\right]$ which in turn implies $\theta \leq e^{-t^*a}$. (In fact it can also be shown that $\theta = O(e^{-t^*a})$.)
- (f) It is clear that the second moment of the estimator, $e^{-t^*S_{\tau_a}}$, sampled using the tilted density f_{t^*} , satisfies $E_{t^*}\left[e^{-2t^*S_{\tau_a}}\right] \leq e^{-2t^*a}$. Argue that importance sampling using t^* is asymptotically optimal as $a \to \infty$. (This is just a one-line argument based on the fact that $E[Z^2] \geq E[Z]^2$ for all random variables, Z.)

Question 3

Consider an insurance company that receives total premia of \$c per unit time and incurs losses that arrive according to a Poisson process with arrival rate, λ . Assume that loss sizes are IID with CDF, F, and that they are independent of the Poisson arrival process. Assume also that the insurance company has a capital reserve of \$A. Let θ be the probability that the insurance company ever goes bankrupt. Describe a detailed importance sampling algorithm for estimating θ when the loss sizes have an exponential distribution with parameter, δ . (The results of Question 2 should guide your choice of importance sampling density.)

Question 4

Implement your algorithm from Question 3 and estimate θ when $\lambda = 5$, $\delta = 1/100$ (so the average claim size is \$100), c = \$600 and \$A = \$10,000. Use N = 100,000 paths and also compute an approximate 95% confidence interval for θ .

Question 5

Consider the problem of pricing an Asian call option on a stock with time t price given by S_t . You may assume that $S_t \sim GBM(r, \sigma)$ where r = .05 and $\sigma = .25$. Other relevant parameters are T = 1 year, $S_0 = 100$ and m = 6 averaging periods. We assume K = 90 so the option is in-the-money. We will use a stratification method which is useful for in-the-money options. It is convenient to write the payoff function as

$$h(Z_1,\ldots,Z_m) = \max\left(0,g(Z_1,\ldots,Z_m) - K\right)$$

where

$$g(Z_1, \dots, Z_m) = \frac{\sum_{i=1}^m S_{\frac{iT}{m}}}{m}$$

$$= \frac{\sum_{i=1}^m S_0 e^{(r-\sigma^2/2)iT/m + \sigma\sqrt{T/m}(\sum_{j=1}^i Z_j)}}{m}$$

where the Z_j 's are IID N(0, 1).

(a) Let $\mathbf{grad}(\mathbf{0})$ be a vector such that the i^{th} component of this vector contains the partial derivative of $g(z_1, \ldots, z_m)$ with respect to z_i , evaluated at $\mathbf{0} = (0, \ldots, 0)$. Let $a = e^{(r-\sigma^2/2)T/m}$ and let $b = \sigma \sqrt{T/m}$. Show that the i^{th} component of $\mathbf{grad}(\mathbf{0})$ is given by

$$[\mathbf{grad}(\mathbf{0})]_i = \frac{bS_0(\sum_{j=i}^m a^j)}{m}.$$

(b) Let $\mu = (\mu_1, \dots, \mu_m)$ be given by

$$\mu = rac{\operatorname{grad}(\mathbf{0})}{||\operatorname{grad}(\mathbf{0})||}$$

where ||.|| is the usual distance norm. Also, let $W = \sum_{i=1}^{m} \mu_i Z_i$. Show that W may be used as a stratification variable.

- (c) Write a program to implement a stratification method using the stratification variable W. Use 10 equiprobable strata and 1000 samples for each stratum to estimate the option price and build 99% confidence intervals using the stratification method.
- (d) How many samples will be needed by the usual simulation method to achieve the same accuracy as the stratification method of part (c)? To answer this part, you will need to write and run a regular Monte-Carlo program for pricing the option.