## Discrete Time Hedging Error Cihui PAN

## 1 Numerical experiment

We choose the parameters to do simulate the term  $M_T^n$  and compare the simulation with the bound we have obtained.

Under the conditions that r = 0,  $\sigma(X_t) = \sigma = constant > 0$ ,  $f(x) = (x - K)^+$  recall that

$$\begin{split} M_T^n &= \int_0^T \frac{\partial u}{\partial x}(t, X_t) d\widetilde{X}_t - \int_0^T \frac{\partial u}{\partial x}(\varphi(t), X_{\varphi(t)}) d\widetilde{X}_t \\ &= (X_T - K)^+ - \mathbb{E}[(X_T - K)^+] - \int_0^T \frac{\partial u}{\partial x}(\varphi(t), X_{\varphi(t)}) dX_t \\ &= (X_T - K)^+ - x_0 N(d_1(0)) + K N(d_2(0)) - \sum_{i=0}^{n-1} N(d_1(t_i))(X_{t_{i+1}} - X_{t_i}) \\ &= (X_T - K)^+ - \frac{x_0}{\sqrt{2\pi}} \int_{-\infty}^{\frac{\log(\frac{x_0}{K}) + \frac{1}{2}\sigma^2 T}{\sigma\sqrt{T}}} e^{-\frac{v^2}{2}} dv + \frac{K}{\sqrt{2\pi}} \int_{-\infty}^{\frac{\log(\frac{x_0}{K}) - \frac{1}{2}\sigma^2 T}{\sigma\sqrt{T}}} e^{-\frac{v^2}{2}} dv \\ &- \sum_{i=0}^{n-1} \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\frac{\log(\frac{X_{t_i}}{K}) + \frac{1}{2}(T - t_i)}{\sigma\sqrt{T} - t_i}} e^{-\frac{v^2}{2}} dv (X_{t_{i+1}} - X_{t_i}) \end{split}$$

What we need to simulate are the price  $X_{t_i}$ , with i = 1, 2, ..., n. The algorithm is as follows:

## Algorithm 1 Simulation of Geometric Brownian Motion

**Step1**: Generate  $2 \times n \times m$  independent random variables that are uniformly distributed on the interval [0,1], here m is the Monte Carlo samples.

**Step2**: Generate a normal distributed random variable from two independent uniformly distributed random variables using Box Muller transform. Form a matrix NRV[m][n] of size  $m \times n$ . formed by  $m \times n$  independent normal distributed random variables.

**Step3**: Construct the Brownian motion matrix BM[m][n] by setting  $BM[i][j] = BM[i][j-1] + \sqrt{\frac{T}{n}}NRV[i][j]$  with the initial condition  $BM[i][1] = \sqrt{\frac{T}{n}}NRV[i][1]$ ,  $1 \le i \le m, 1 \le j \le n$ .

**Step4**: Construct the matrix PX[m][n] of price  $X_t$  by  $PX[i][j] = x_0 exp(-\frac{1}{2}\sigma^2j\frac{T}{n} + \sigma BM[i][j]), 1 \le i \le m, 1 \le j \le n.$ 

**Step5**: For each i, calculate  $M_T^n[i]$  using the expression above. Count the numbers s that  $M_T^n[i]$  is greater or equals to  $\gamma$ , calculate the simulated probability as  $P_1 = prob = \frac{s}{m}$ . Step6: Compare  $P_1 = prob$  with  $P_2 = L(p*)$ .