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Columbia University
IEOR4703 – Monte Carlo Simulation (Hirsa)
Assignment 4 – Due 17:40 on Thursday March 22nd, 2018

Problem 1 (Pricing Heston Utilizing Various Discretization Schemes): Extend the matlab code *exampleSimulationOfSDE.m* for Heston stochastic volatility model

$$\begin{aligned}dS_t &= (r - q)S_t dt + \sqrt{v_t}S_t dW_t^1, \\dv_t &= \kappa(\theta - v_t)dt + \lambda\sqrt{v_t}dW_t^2\end{aligned}$$

with the following parameter set: spot price of $S_0 = \$100$, strike price of $K = \$110$, risk-free rate $r = 0.025$, continuous dividend rate of $q = 0.0125$, volatility of $\sigma = 0.30$, maturity of $T = 1$ year, $\Theta = \{\theta = 0.0625, \kappa = 2.75, \lambda = 0.0125, v_0 = 0.05, \rho = -0.65\}$.

Problem 2 (Calculating Greeks via Simulation): Extend the matlab code *exampleGreek_BMS.m* for (a) Digital Call, (b) Up-and-Out Call, (c) Average Option Call with the following parameter set: spot price of $S_0 = 100$, strike price of $K = \$115$, risk-free rate $r = 0.04$, continuous dividend rate of $q = 0.015$, volatility of $\sigma = 0.28$ for all options, maturity of $T = 1$ year for all options, barrier of $H = 130$ for up-and-out call, $m = 12$ monitoring times for average option.

P₁. $Z_1, Z_2 \sim N(0,1)$, $Z_2 = \rho Z_1 + \sqrt{1-\rho^2} \cdot Z_3$

Euler $\begin{cases} S_{t_{j+1}} = S_{t_j} + (r-q)S_{t_j} \Delta t + \sqrt{V_{t_j}} S_{t_j} \cdot \sqrt{\Delta t} \cdot Z_1 \\ V_{t_{j+1}} = V_{t_j} + \kappa(\theta - V_{t_j}) \Delta t + \lambda \sqrt{V_{t_j}} \cdot \sqrt{\Delta t} \cdot Z_2 \end{cases}$

Milsten $\begin{cases} S_{t_{j+1}} = S_{t_j} + (r-q)S_{t_j} \Delta t + \sqrt{V_{t_j}} S_{t_j} \cdot \sqrt{\Delta t} \cdot Z_1 \\ \quad + \frac{1}{2} \sqrt{V_{t_j}} S_{t_j} \cdot \sqrt{V_{t_j}} \cdot \Delta t (Z_1^2 - 1) \\ V_{t_{j+1}} = V_{t_j} + \kappa(\theta - V_{t_j}) \Delta t + \lambda \sqrt{V_{t_j}} \cdot \sqrt{\Delta t} \cdot Z_2 \\ \quad + \frac{1}{2} \lambda \sqrt{V_{t_j}} \cdot \frac{1}{\sqrt{V_{t_j}}} \cdot \Delta t (Z_2^2 - 1) \end{cases}$

Runge-Kutta $\begin{cases} \tilde{S}_{t_j} = S_{t_j} + (r-q)S_{t_j} \Delta t + \sqrt{V_{t_j}} S_{t_j} \cdot \sqrt{\Delta t} \\ S_{t_{j+1}} = S_{t_j} + (r-q)S_{t_j} \Delta t + \sqrt{V_{t_j}} S_{t_j} \cdot \sqrt{\Delta t} \cdot Z_1 \\ \quad + \frac{1}{2\sqrt{\Delta t}} (\sqrt{V_{t_j}} (\tilde{S}_{t_j} - S_{t_j})) \cdot \Delta t (Z_1^2 - 1) \\ \tilde{V}_{t_j} = V_{t_j} + \kappa(\theta - V_{t_j}) \Delta t + \lambda \sqrt{V_{t_j}} \cdot \sqrt{\Delta t} \\ V_{t_{j+1}} = V_{t_j} + \kappa(\theta - V_{t_j}) \Delta t + \lambda \sqrt{V_{t_j}} \cdot \sqrt{\Delta t} \cdot Z_2 \\ \quad + \frac{1}{2\sqrt{\Delta t}} (\lambda (\sqrt{\tilde{V}_{t_j}} - \sqrt{V_{t_j}})) \cdot \Delta t (Z_2^2 - 1) \end{cases}$

(Please find output in the end)

P₂. Likelihood Score function for Digital Option and Up & Out :

$$g(x) = \frac{1}{x \sigma \sqrt{T}} \cdot \phi(\xi(x)) \quad , \quad \xi(x) = \frac{\log(x/s_0) - (r - q - \sigma^2/2)T}{\sigma \sqrt{T}}$$

$$\begin{aligned} \frac{dg(x)/ds_0}{g(x)} &= \frac{\frac{1}{x \sigma \sqrt{T}} \phi(\xi(x)) \cdot (-\xi(x)) \cdot \frac{d\xi(x)}{ds_0}}{g(x)} \\ &= \frac{\log(x/s_0) - (r - q - \frac{\sigma^2}{2})T}{s_0 \sigma^2 T} = \tau(x) \end{aligned}$$

$$\begin{aligned} \frac{d^2g(x)/ds_0^2}{g(x)} &= \frac{\frac{1}{x \sigma \sqrt{T}} \phi(\xi(x)) \cdot (-\xi(x)) \cdot \frac{d\xi(x)}{ds_0} \cdot \tau(x)}{g(x)} \\ &\quad + \frac{g(x)}{g(x)} \cdot \left(-\frac{1 + \log(x/s_0) - (r - q - \frac{\sigma^2}{2})T}{s_0^2 \sigma^2 T} \right) \\ &= \tau(x)^2 - \frac{1}{s_0^2 \sigma^2 T} - \frac{\tau(x)}{s_0} \end{aligned}$$

$$x = s_T \quad , \quad \xi(x) = z$$

$$\Rightarrow \frac{dg(x)/ds_0}{g(x)} = \frac{z}{s_0 \sigma \sqrt{T}}$$

$$\frac{d^2g(x)/ds_0^2}{g(x)} = \frac{z^2 - 1 - \sigma \sqrt{T} \cdot z}{s_0^2 \sigma^2 T}$$

$$\begin{aligned} \frac{dg(x)/ds_0}{g(x)} &= \left(-\frac{1}{x \sigma^2 \sqrt{T}} \cdot \phi(\xi(x)) + \frac{1}{x \sigma \sqrt{T}} \cdot \phi(\xi(x)) \cdot (-\xi(x)) \cdot \frac{d\xi(x)}{ds_0} \right) / g(x) \\ &= \frac{1}{\sigma} (z^2 - \sigma \sqrt{T} z - 1) \end{aligned}$$

① Digital Option : $e^{-rT} \cdot 1_{\{S_T > K\}}$

② Up & Out : $e^{-rT} (S_T - K)^+ 1_{\{S_T < H\}}$

③ Average Option : $e^{-rT} (\bar{S} - K)^+ , \bar{S} = \frac{1}{m} \sum_{t=1}^m S_{t_i}$

Pathwise for Average Option :

$$\frac{\partial h}{\partial S_0} = \frac{\partial h}{\partial \bar{S}} \cdot \frac{\partial \bar{S}}{\partial S_0} = e^{-rT} \cdot 1_{\{S_T > K\}} \cdot \frac{\partial \bar{S}}{\partial S_0}$$

$$* \frac{\partial E[h]}{\partial S_0} = E \left[\frac{\partial h}{\partial S_0} \right] = E \left[e^{-rT} \cdot 1_{\{\bar{S} > K\}} \cdot \frac{\bar{S}}{S_0} \right]$$

$$\frac{\partial h}{\partial \sigma} = \frac{\partial h}{\partial \bar{S}} \cdot \frac{\partial \bar{S}}{\partial \sigma} = e^{-rT} \cdot 1_{\{\bar{S} > K\}} \cdot \frac{\partial \bar{S}}{\partial \sigma}$$

$$S_{t_i} = S_0 \cdot e^{(r-q-\frac{1}{2}\sigma^2)t_i} \cdot e^{\sigma \sqrt{\Delta t} \cdot \sum_{j=1}^i Z_j}$$

$$\frac{\partial S_{t_i}}{\partial \sigma} = S_{t_i} \cdot (-\sigma t_i + \sqrt{\Delta t} \cdot \sum_{j=1}^i Z_j)$$

$$* \frac{\partial E[h]}{\partial \sigma} = E \left[\frac{\partial h}{\partial \sigma} \right] = E \left[e^{-rT} \cdot 1_{\{\bar{S} > K\}} \cdot \frac{1}{m\sigma} \sum_{i=1}^m \left(S_{t_i} \left(\log \frac{S_{t_i}}{S_0} - (r-q+\frac{\sigma^2}{2})(t_i) \right) \right) \right]$$

Likelihood Score function for Average option :

$$g(S_{t_1}, \dots, S_{t_m}) = g(S_{t_1} | S_{t_0}) g(S_{t_2} | S_{t_1}) \dots g(S_{t_m} | S_{t_{m-1}})$$

$$g(S_{t_i} | S_{t_{i-1}}) = \frac{1}{S_{t_i} \sigma \sqrt{\Delta t}} \cdot \phi \left(\mathcal{L}(S_{t_i} | S_{t_{i-1}}) \right)$$

$$\mathcal{L}(S_{t_i} | S_{t_{i-1}}) = \frac{\log(S_{t_i}/S_{t_{i-1}}) - (r-q-\frac{1}{2}\sigma^2) \cdot \Delta t}{\sigma \sqrt{\Delta t}}$$

$$\frac{dg(S_{t_1}, \dots, S_{t_m})/dS_0}{g} = \frac{dg(S_{t_1}|S_0)/dS_0}{g(S_{t_1}|S_{t_0})} = \frac{Z_1}{S_0 \cdot \sigma \cdot \sqrt{\Delta t}}$$

$$\frac{d^2g/dS_0^2}{g} = \frac{Z_1^2 - 1 - \sigma \sqrt{\Delta t} \cdot Z_1}{S_0^2 \sigma^2 \Delta t}$$

$$\begin{aligned} \frac{dg/d\sigma}{g} &= \frac{\sum_1^m \frac{dg(S_{t_i}|S_{t_{i-1}})}{d\sigma} \cdot \prod_{j=i}^m g(S_{t_j}|S_{t_{j-1}})}{\prod g(S_{t_j}|S_{t_{j-1}})} \\ &= \sum_1^m \frac{dg(S_{t_i}|S_{t_{i-1}})}{d\sigma} / g(S_{t_i}|S_{t_{i-1}}) \\ &= \sum_1^m \frac{1}{\sigma} (Z_i^2 - \sigma \sqrt{\Delta t} Z_{i-1}) \end{aligned}$$

(Please find output in the end)

Output (Please find attached code for Q1 & Q2)

```
>> hw4_q1
```

```
ans =
```

```
" Euler: 6.179451
  Milstein: 6.182216
  Runge-Kutta: 6.182250
"
```

The results generated by different schemes are close to each other, and are close to the exact solution.

```
ans =
```

```
" Exact solution: 6.188100
"
```

```
>> hw4_q2
```

```
=====
```

```
Delta
```

```
-----
```

```
    Digital Call
```

```
W/ CRN:0.011578 W/O CRN:0.011606 Pathwise Estimator:NaN Likelihood Ratio:0.011744
```

```
    Up and Out Call
```

```
W/ CRN:0.020365 W/O CRN:-0.0013269 Pathwise Estimator:NaN Likelihood Ratio:0.025625
```

```
    Average Option Call
```

```
W/ CRN:0.25114 W/O CRN:0.32545 Pathwise Estimator:0.25117 Likelihood Ratio:0.25402
```

```
=====
```

```
Gamma
```

```
-----
```

```
    Digital Call
```

```
W/ CRN:-0.012106 W/O CRN:0.14988 Pathwise Estimator:NaN Likelihood Ratio:0.00011442
```

```
    Up and Out Call
```

```
W/ CRN:0.013658 W/O CRN:0.9037 Pathwise Estimator:NaN Likelihood Ratio:-0.000577
```

```
    Average Option Call
```

```
W/ CRN:0.018926 W/O CRN:-0.97191 Pathwise Estimator:NaN Likelihood Ratio:0.018797
```

```
=====
```

```
Vega
```

```
-----
```

```
    Digital Call
```

```
W/ CRN:0.33243 W/O CRN:-0.033628 Pathwise Estimator:NaN Likelihood Ratio:0.32038
```

```
    Up and Out Call
```

```
W/ CRN:-2.2634 W/O CRN:-2.5855 Pathwise Estimator:NaN Likelihood Ratio:-1.6156
```

```
    Average Option Call
```

```
W/ CRN:19.6766 W/O CRN:27.0959 Pathwise Estimator:19.6771 Likelihood Ratio:19.7701
```