

GPU Monte Carlo Simulation for VaR Calculation

Xuefei He
Manchester Business School

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- The most general, powerful, and important numerical method for derivative pricing (simple option example)
 - Underlying stock: $d \log S = (\mu - \frac{1}{2}\sigma^2)dt + \sigma dX$
 - Option payoff at T is equal to $\max(S_T - K, 0)$
 - One path: $\log S_{\Delta t, i} = \log S_0 + (\mu - \frac{1}{2}\sigma^2)\Delta t + \sigma\sqrt{\Delta t}\epsilon_i(0, 1)$
 - Option price at now is $P = e^{-rT} \frac{1}{N} \sum_{i=0}^N \max(S_{T, i} - K, 0)$
- Time consuming — 10,000 to 100,000 paths due to slow convergence rate ($\frac{1}{\sqrt{N}}$)
- Suitable for GPU — independent paths and no frequent data transfer between host and device

Value at risk (VaR)

- VaR — maximum loss value in next M days with $X\%$ confidence level
- VaR of the example option
 - 1 Volatility σ is the only risk factor.
 - 2 How much will σ change in next M days? — distribution of σ ($\sigma_1, \sigma_2, \dots, \sigma_n$)
 - 3 Distribution of price P (P_1, P_2, \dots, P_n)
 - 4 The X percentile is VaR.

GPU implementation steps

- Start on device
- For every Δt , generate random number
- Update stock price
- At T , calculate payoff
- Calculate the average of discounted prices
- Retrieve the result P from the device

Any easy-use library?

- Random number generator and summation
- Easy to install without sudo password