

MPI

Pricing Call Option

Pseudo and Sobol

Monte Carlo

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Monte Carlo Method

- Pricing a Call Option

- $I_N[f] = \frac{1}{N} \sum_{k=1}^N f(\text{sequence of random points}, K) = e^{-rT} \left[\frac{1}{N} \sum_{k=1}^N \max(S_T^k - K, 0) \right]$
- $I[f] = \text{Black Scholes formula}$

$$\varepsilon = |I[f] - I_N[f]|$$

- Slow convergence: $\varepsilon_N = \frac{\sigma(f)}{N^{1/2}}$
 - $f(\sigma)$ can be decreased by : antithetic variables or control variates.
 - But the rate the converge remains: $\varepsilon_N \sim \frac{1}{N^{1/2}}$

Monte Carlo Method

- MC with Sobol

- Improve convergence in Monte Carlo $\epsilon_N = \frac{O(\ln N)^n}{N}$
- Discrepancy is a measure of deviation of uniformity.
- Best uniformity of distribution for as N goes to infinity.
- Parallelization can be achieved by changing parameters in the “Preprocessing” part of the code.
- No inter-processor communication is needed.

Monte Carlo Method

- MC with Sobol

- Improve convergence in Monte Carlo $\varepsilon_N = \frac{O(\ln N)^n}{N}$
- If $N = 2^k$, k is integer
- Discrepancy is a measure of deviation of uniformity.
- Best uniformity of distribution for as N goes to infinity.
- Parallelization can be achieved by changing parameters in the “Preprocessing” part of the code.
- No inter-processor communication is needed.

Monte Carlo Method

Intel(R) Core (TM) i7-3612QM [CPU@2.10GHz](#) RAM 8.00GB 64 bit OS

Random/ Processors Number	Seconds	Exact Price	MC Price	NPaths	N	Absolute Error
Sobol/8	6.340026	19.620613	19.561750	2	1000000	5.886e-02
Pseudo/8	6.575098	19.620613	21.795881	2	1000000	2.175+00
Sobol /4	3.427575	19.620613	19.417058	1	1000000	2.036e-01
Pseudo/4	3.466962	19.620613	21.053336	1	1000000	1.433e+00

Monte Carlo Method

Intel Core 2 Duo [CPU@3.06GHz](#) RAM 4.00GB 32 bit OS

Random/ Processors Number	Seconds	Exact Price	MC Price	nPaths	N	Absolute Error
Sobol/4	12.831499	19.620613	19.731652	2	1000000	1.110e-01
Pseudo/4	12.721947	19.620613	25.817706	2	1000000	6.197e+00
Sobol /4	18.478064	19.620613	19.446532	3	1000000	1.741e-01
Pseudo/4	18.931408	19.620613	29.742907	3	1000000	1.012e+01

Monte Carlo Method

Precision Floating	Real Number $y = (1 + x)2^r$ $0 < x < 1$	The most significant	Binary representation exponent part	Binary representation of the mantissa part
Single	y is $d_0 d_1 \dots d_{31}$	$d_0 = 0$	$d_1 \dots d_8$ $r+127$ r is the floating point exponent $-126 < r < 127$	$d_9 \dots d_{31}$ $x2^{23}$
Double	y is $d_0 d_1 \dots d_{63}$	$d_0 = 0$	$d_1 \dots d_{11}$ r is the floating point exponent $-1022 < r < 1023$	$d_{12} \dots d_{63}$ $x2^{52}$

Monte Carlo Method

Atanassov's Algorithm Sobol Generator

Avoids the multiplication and conversion from integer to floating point

Single

1. $X \sim [0,1)$
2. Y is the mantissa
3. Y is stored as a 32 bit integer
4. If one xor's 001111111 to the nine most-significant bits of y
5. Remains in memory is the floating-point representation of $(1+x)$.

Double

4. If one xor's 001111111111 to the twelve most-significant bits of y

1. Input initial data:
 - if the precision is single, set the number of bits b to 32, and the maximal power of two p to 23, otherwise set b to 64 and p to 52;
 - dimension s ;
 - direction vectors $\{a_{ij}\}$, $i = 0, p, j = 1, \dots, s$ representing the matrices A_1, \dots, A_d (always $a_{ij} < 2^{i+1}$);
 - scrambling terms d_1, \dots, d_s - arbitrary integers less than 2^p , if all of them are equal to zero, then no scrambling is used;
 - index of the first term to be generated - n ;
 - scaling factor m , so the program should generate elements with indices $2^m j + n$, $j = 0, 1, \dots$.
2. Allocate memory for $s * l$ b -bit integers (or floating point numbers in the respective precision) y_1, \dots, y_s .
3. Preprocessing: calculate the twisted direction numbers v_{ij} , $i = 0, \dots, p - 1$, $j = 0, \dots, s$:
 - for all j from 1 to s do
 - for $i = 0$ to $p - 1$ do
 - if $i=0$, then $v_{ij} = a_{ij}2^{p-m}$, else

$$v_{ij} = v_{i-1j} \mathbf{XOR}(a_{i+m,j} * (2^{p-i-m}));$$

4. Calculate the coordinates of the n -th term of the Sobol' sequence (with the scrambling applied) using any known algorithm (this operation is performed only once). Add +1 to all of them and store the results as floating point numbers in the respective precision in the array y .
5. Set the counter N to $\lfloor \frac{n}{2^m} \rfloor$.
6. Generate the next point of the sequence:
 - When a new point is required, the user supplies a buffer x with enough space to hold the result.
 - The array y is considered as holding floating point numbers in the respective precision, and the result of subtracting 1. from all of them is placed in the array x .
 - Add 1 to the counter N ;
 - Determine the first nonzero binary digit k of N so that $N = (2M + 1)2^k$ (on the average this is achieved in 2 iterations);
 - consider the array y as an array of b -bit integers and updated it by using the k -th row of twisted direction numbers:
 - for $i = 1$ to d do
 - $y_i = y_i \mathbf{XOR} v_{ki}$.
 - return the control to the user. When a new point is needed, go to [6](#).

Monte Carlo Method

- Conclusion

- Monte Carlo Convergence with Sobol's sequences can be generated with speeds comparable or superior to those of pseudo random generators.
- MPI is straightforward to be implemented to do Monte Carlo with Sobol's sequences.
- Pricing Exotic Options can be easily implemented.

Monte Carlo Method

- Acknowledgments

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- References

- Atanassov E. I.: A new Efficient Algorithm for Generating the Scrambled Sobol's Sequence. (2003)
- URL: <http://parmac1.bas.bg/emanouil/sequence.html>.