

# Simulation Methods for Barrier and Look-back Options

Course: Simulation Methods for Finance  
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# 1 Generate Random Number

There are two methods we used to generate random numbers. The first one is to generate a uniform distribution and then transform it into Standard Normal Distribution. The second method we tried is the system build-in function ***Randon***, which can directly generate a variable follows Standard Normal Distribution.

To generate a uniform distribution we tried two ways: one can use the system build-in methods ***rand***. The function will return a number between 1 and  $2^{15} - 1$  randomly. The range is around 30,000, way below 100,000, the sample size we plan to generate. In other words, when we use ***rand*** to simulate 100,000 entries, there will be numbers appear more than once, which will create bias. We prefer the second way to generate a uniform distribution: linear congruential generator.

$$n_i = (an_{i-1}) \bmod m$$

for  $i = 1, 2, \dots, 10,000$ , and we let  $a = 7^5$ , and  $m = 2^{31} - 1$ , which gives about 2 billion points. We run 100,000 times and will only use 0.005% of all points. Theoretically, no pattern should appear.

To transform the uniform distribution we got into Standard Normal Distribution, we have tried three methods. By Central Limit theory,

$$Z_n = \frac{\sum_{i=1}^n X_i - n\mu}{\sqrt{n}\sigma}$$

where  $X_i$  is from the uniform distribution we generated previously.  $Z_n$  converges in distribution to SND, however, it requires  $n$ , the number of uniform distribution, to be sufficiently large. Therefore this method requires significant large simulations and speed is slow consequently.

We also tried Box-Muller methods

$$Z_1 = \sqrt{-2\ln X_1} \sin(2\pi X_2), \quad Z_2 = \sqrt{-2\ln X_1} \cos(2\pi X_2)$$

Since its simulation involves computation of *sine* and *cosine*, the speed is slow. We prefer the last method, Marsaglia Polar method.

$$\text{let } V_1 = 2U_1 - 1, \quad V_2 = 2U_2 - 1.$$

where  $U_1$  and  $U_2$  are two independent uniform distribution. Let  $W = V_1^2 + V_2^2$ . If  $W > 1$ , return to the beginning. Otherwise,

$$N_1 = \sqrt{\frac{(-2\log W)}{W}} V_1, \quad N_2 = \sqrt{\frac{(-2\log W)}{W}} V_2$$

As the computation does not involve sine and cosine, it is generally faster than Box-Muller.

The following is the simulation result of Standard Normal distribution and time used by different methods.

Method	Mean	Variance	Time
<i>rand</i> generated U + CLT	cell5	cell6	..
<i>rand</i> generated U + Box-Muller	cell8	cell9	..
<i>rand</i> generated U + Marsaglia Polar	cell8	cell9	..
linear congruential generator + CLT	cell8	cell9	..
linear congruential generator + Box-Muller	cell8	cell9	..
linear congruential generator + Marsaglia	cell8	cell9	..
<i>Randon</i>	cell8	cell9	..

Error Analysis... For the tasks in this project, we use the combination of linear congruential generator and Marsaglia methods to simulate random variable.

## 2 Basic Task

The asset price in a risk neutral probability space  $(\Omega, \mathcal{F}, (\mathcal{F}_t)_{0 \leq t \leq T}, P)$  follows Geometric Brownian Motion,

$$dS_t = rS_t dt + \sigma S_t dW_t, \quad 0 \leq t \leq T$$

with initial price  $S_0 = S$ , where  $r$  is riskless interest rate,  $\sigma$  volatility, and  $W_t$  the standard Brownian motion. A European call option price at time  $t$  with maturity time  $T$  is given by

$$C_t = E[e^{-r(T-t)}(S_T - K)^+ | \mathcal{F}_t]$$

For the basic task, we let  $S_0 = 100$ ,  $K = 100$ , interest rate  $r = 0.05$ , volatility  $\sigma = 0.4$ , maturity time  $T = 1$ . We use Monte Carlo method to simulate the path of

$$S_T = S_0 e^{(r - \frac{1}{2}\sigma^2)T + \sigma\sqrt{T}Z}$$

and get sample distribution of  $(S_T - K)^+$ . By Black-Scholes formula,

$$C_{bs}(S_0, K) = N(d_1)S_0 - N(d_2)Ke^{-rT} \quad (1)$$

where

$$d_1(S_0, K) = \frac{1}{\sigma\sqrt{T}}[\ln(\frac{S_0}{K}) + (r + \frac{\sigma^2}{2})T], \quad d_2(S_0, K) = d_1 - \sigma\sqrt{T} \quad (2)$$

The closed-form Greeks are calculated the following way:

$$\delta_{bs} = \frac{\partial C}{\partial S_0} = \Phi(d_1) \quad (3)$$

$$\gamma_{bs} = \frac{\partial^2 C}{\partial S_0^2} = \frac{\Phi'(d_1)}{S_0\sigma\sqrt{T}} \quad (4)$$

$$\nu_{bs} = \frac{\partial C}{\partial \sigma} = \Phi'(d_1)\sqrt{T} \quad (5)$$

To calculate the Greeks from simulation, we compare Likelihood ratio method and Pathwise method. As the Call option price is given by

$$C = e^{-rT} E[(S_T - K)^+] = e^{-rT} \int (S_T - K)^+ h_{S_0}(S_T) dS_T$$

where  $h_{S_0}(S_T)$  is the probability density function of  $(S_T - K)^+$ . Then by Likelihood ratio methods, the partial derivative of  $C$  with respect to  $S_0$  is

$$\frac{\partial C}{\partial S_0} = \int (S_T - K)^+ \frac{d}{dS_0} h_{S_0}(S_T) dS_T = E[(S_T - K)^+ \frac{h'_{S_0}(S_T)}{h_{S_0}(S_T)}]$$

And the lognormal density function of  $S_T$  is given by

$$h(x) = \frac{1}{x\sigma\sqrt{T}} \phi(\xi(x)), \quad \xi(x) = \frac{\ln(x/S_0) - (r - \frac{1}{2}\sigma^2)T}{\sigma\sqrt{T}}$$

Therefore,

$$\Delta_{LR} = \frac{\partial C}{\partial S_0} = E[e^{-rT} (S_T - K)^+ \frac{Z}{S_0\sigma\sqrt{T}}]$$

where  $Z \sim N(0, 1)$ .

$$\Gamma_{LRLR} = \frac{\partial^2 C}{\partial S_0^2} = E[e^{-rT} (S_T - K)^+ (\frac{Z^2 - 1}{S_0^2\sigma^2 T} - \frac{Z}{S_0^2\sigma\sqrt{T}})]$$

$$\text{Vega}_{LR} = \frac{\partial C}{\partial \sigma} = E[e^{-rT} (S_T - K)^+ \left( \frac{Z^2 - 1}{\sigma} - Z\sqrt{T} \right)]$$

By pathwise methods, the greeks are given as following<sup>1</sup>, with  $Z \sim N(0, 1)$ :

$$\Delta_{PW} = \frac{\partial C}{\partial S_0} = E[e^{-rT} \mathbf{1}_{S_T > K} \frac{S_T}{S_0}]$$

$$\Gamma_{LRPW} = \frac{\partial^2 C}{\partial S_0^2} = E[e^{-rT} \mathbf{1}_{S_T > K} \frac{KZ}{S_0^2\sigma\sqrt{T}}]$$

$$\Gamma_{PWLRL} = \frac{\partial^2 C}{\partial S_0^2} = E[e^{-rT} \mathbf{1}_{S_T > K} \frac{S_T}{S_0^2} \left( \frac{Z}{\sigma\sqrt{T}} - 1 \right)]$$

$$\text{Vega}_{PW} = \frac{\partial C}{\partial \sigma} = E[e^{-rT} \mathbf{1}_{S_T > K} S_T (\sqrt{T}Z - \sigma T)]$$

The following are the results calculated from the closed-form formulae and the simulations:

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<sup>1</sup>For gamma, there are two different methods.

100 000 MC simulations	<b>Error Mean</b>	<b>Error Variance</b>	<b>Time (seconds)</b>
<b>Option Price</b>	na	na	5.0 e-6
<b>Delta LR</b>	4.1 e-3	9.7 e-6	1.4 e-3
<b>Delta PW</b>	1.8 e-3	1.8 e-6	9.9 e-4
<b>Gamma PWLR</b>	6.1 e-5	2.0 e-9	1.4 e-3
<b>Gamma LRPW</b>	3.5 e-5	7.4 e-10	1.4 e-3
<b>Gamma LRLR</b>	1.9 e-4	2.1 e-8	4.7 e-3
<b>Vega LR</b>	7.7 e-1	3.3 e-1	4.6 e-3
<b>Vega PW</b>	2.4 e-1	3.2 e-2	1.6 e-3

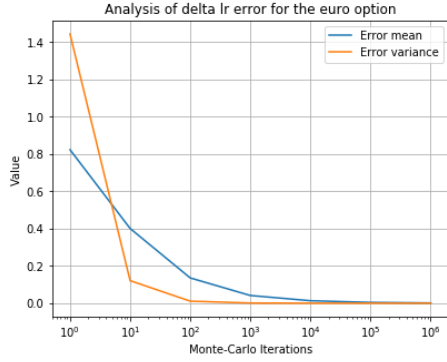
Table 1: Sample from our simulation dataset with 100 000 Monte-Carlo simulations. Note that the run time depends on the computer (here a 3.1 GHz Intel Core i5).

		<b>Black -Scholes</b>	<b>Monte Carlo Simulation</b>		
			<b>1,000</b>	<b>10,000</b>	<b>100,000</b>
<b>Option Price</b>		18.023	19.0162	17.6732	18.0014
<b>Delta</b>	<b>LR</b>	0.627409	0.662561	0.612362	0.627889
	<b>PW</b>		0.64485	0.623239	0.627453
<b>Gamma</b>	<b>PW LR</b>	0.0094605	0.0101213	0.00919597	0.00945029
	<b>LR PW</b>		0.00994418	0.00930474	0.00944593
	<b>LR LR</b>		0.00975465	0.00918059	0.0095502
<b>Vega</b>	<b>LR</b>	37.842	39.0186	36.7224	38.2008
	<b>PW</b>		40.4851	36.7839	37.8012

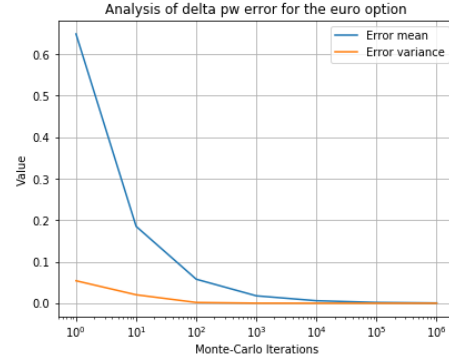
We can see from above table that simulation results improve and get closer to the theoretical value as simulation number increase. Hence to compare the results from different methods, we will only analyze the simulation result from the most simulation number - 100,000 in this case. This is even more clear on the graphs (Figure 2).

We now have a closer look at 100,000 simulations, which is the industry standard and which is above the 1,000 simulations threshold we have noticed on the graphs. For delta, although the fastest methods is LR, it has a significantly lower accuracy. Gamma LRLR is the most accurate but it is the slowest, so gamma PWLR seems to be a good compromise. Vega PW is both faster and more accurate than vega LR.

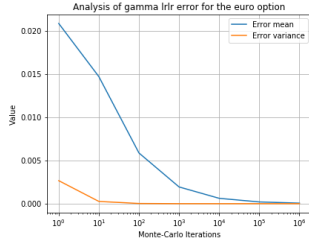
For the following section, PW methods will be infeasible to do for Barrier Option. Hence, we will use Likelihood ratio method when calculating the greeks.



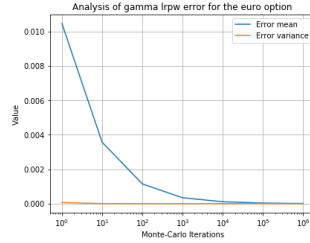
(a) Delta with LR method



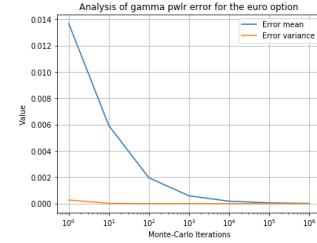
(b) Delta with PW method



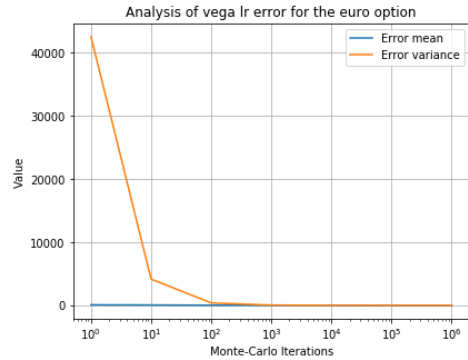
(c) Gamma with LRLR method



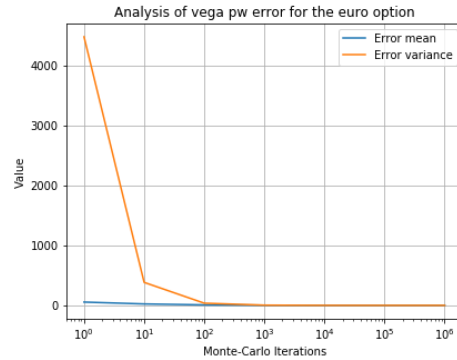
(d) Gamma with LRPW method



(e) Gamma with PWLR method



(f) Vega with LR method



(g) Vega with PW method

Figure 1: A graphical analysis of the error of the greeks depending on the number of simulations. We notice that there is a significant increase above 1000 simulations in all cases. This is very clear on our vega which tends to have a very high error variance below 1000 simulations. In order to generate these graphs, we have done one thousand simulations for each value, at every power of 10.

### 3 Main Task - Barrier Option

For an up-and-out barrier call option  $A_T = (S_T - K)^+ 1_{\max_{0 \leq t \leq T} S_t \leq B}$ , where  $B$  is a barrier level and  $1_S$  is an indicator function. The closed form formula for the option price is

$$UOC(S_0, K, B) = 1_{B > K} \{ C_{bs}(S_0, K) - C_{bs}(S_0, B) - (B - K)e^{-rT} \Phi[d_1(S_0, B)] \\ - \frac{B}{S_0}^{\frac{2v^2}{\sigma^2}} \left[ C_{bs}\left(\frac{B^2}{S_0}, K\right) - C_{bs}\left(\frac{B^2}{S_0}, B\right) - (B_0 - K)e^{-rT} \Phi[d_1(S_0, B)] \right] \} \quad (6)$$

Where  $C_{bs}$  and  $d_1$  are as stated in the Black-Scholes formula (1) and (2), and  $v = r - \frac{\sigma^2}{2}$ .

The closed-form formulas for the greeks are the following:

$$DOC\delta = \Phi\left(\frac{\log \frac{S_0}{K} + (r + \frac{\sigma^2}{2})T}{\sigma\sqrt{T}}\right) - \left(\frac{B}{S_0}\right)^{r/\sigma^2 - 1} \\ \times \left(-\frac{B^2}{S_0} \Phi\left(\frac{\log \frac{B^2}{S_0 K} + vT}{\sigma\sqrt{T}} + \sigma\sqrt{T}\right) - \frac{2vC_{BS}(B^2/S_0, K)}{(S_0\sigma^2)}\right)$$

$$UOC\delta = \delta_{BS}(S_0, K) - \delta_{BS}(S_0, B) - \frac{B - K}{\sigma S_0 \sqrt{T}} e^{-rT} \Phi(d_2(S_0, B)) \\ + \left(\frac{2v}{\sigma^2 S_0} \left(\frac{B}{S_0}\right)^{2v/\sigma^2}\right) \\ \times \left(C_{BS}\left(\frac{B^2}{S_0}, K\right) - C_{BS}\left(\frac{B^2}{S_0}, B\right) - (B - K)e^{-rT} \Phi(d_2(B, S_0))\right) \\ - \left(\frac{B}{S_0}\right)^{2v/\sigma^2} \left(\left(\frac{-B}{S_0}\right)^2 \delta_{BS}\left(\frac{B^2}{S_0}, K\right) + \left(\frac{B}{S_0}\right)^2 \delta_{BS}\left(\frac{B^2}{S_0}, B\right)\right) \\ + \left(\frac{B - K}{\sigma S_0 \sqrt{T}}\right) e^{-rT} \Phi(d_2(B, S_0));$$

$$DOC\gamma = \frac{\phi(d_2)}{S_0 \sigma \sqrt{T}} - \left(\frac{B}{S_0}\right)^{2v/\sigma^2} \left(\frac{4v^2 + 2v\sigma^2}{S_0^2 \sigma^4} C_{BS}(B^2/S_0, K) + \gamma_{BS} - \frac{4v\delta_{BS}}{S_0 \sigma^2}\right)$$

We simulate the path of  $S_T$  by taking the maturity time  $T$  into 10,000 steps, and we simulate 5,000 such paths. As standard normal distribution is symmetrically distributed, 5,000 paths can be treated as 10,000 paths by adding negative sign and creating the other 5,000.

To determine the greeks for Barrier Option, we first assessed the pathwise method. It is impossible to find the partial derivative of the indicator function  $1_{\max_{0 \leq t \leq T} S_t \leq B}$



100 000 MC simulations	<b>Error Mean</b>	<b>Error Variance</b>	<b>Time</b>
<b>Option Price</b>			
<b>Delta</b>			
<b>Gamma</b>			
<b>Vega</b>			

Table 2: Sample from our simulation dataset with 100 000 Monte-Carlo simulations.

with respect to  $S_0$ . Hence pathwise method is eliminated by us. For the likelihood ratio method, we find the joint cdf of  $S_T$  and the  $M_T = \max_{0 \leq t \leq T} S_t$  to be \*\*

\*\*closed form for greeks

To analyze further the accuracy of our simulation, we plot the simulated option prices and greeks versus their theoretical value as barrier increase.

\*\*graph

\*\*graph

\*\*graph

## 4 Look-back Option

Let  $T$  denote option expiration time and  $[0, T]$  lookback period. For  $T_0 \leq t \leq T$  denote by

$$m_0^t = \min_{0 \leq u \leq t} S_u, \quad M_0^t = \max_{0 \leq u \leq t} S_u$$

minimum (maximum) value of realized asset. Lookback call option with fixed strike price  $K$  has payoff  $(M_0^T - K)^+$ . The call option price at time  $t$  is

$$c(S_0, K, t) = e^{-r(T-t)} E[(\max(M_0^t, M_t^T) - K)^+ | \mathcal{F}_t]$$

The closed-form formula for fixed strike lookback call option at time 0 is

$$c(S_0, K, 0) = C_{bs}(S_0, K) + \frac{S_0 \sigma^2}{2r} \{ \Phi[d_2(S_0, K)] - e^{-rT} \frac{S_0}{K}^{-\frac{2r}{\sigma^2}} \Phi[-d_1(K, S_0)] \}$$

\*\*greeks closed form and simulation formulas

	<b>Closed-form Formula</b>	<b>Mean</b>	<b>Error</b>	<b>Variance</b>	<b>Time</b>
<b>Option Price</b>					
<b>Delta</b>					
<b>Gamma</b>					
<b>Vega</b>					

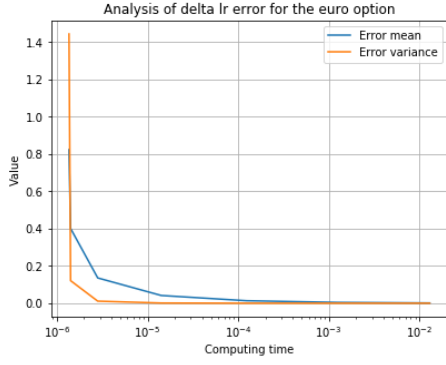
## References

[1] A

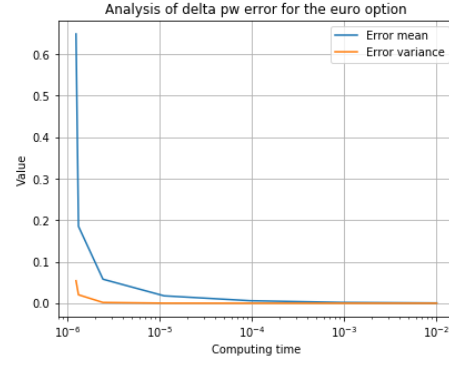
[2] B

**A Appendix: R code of part A**

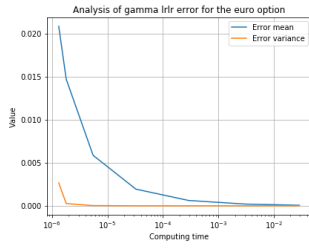
**B Analysis of the error with time**



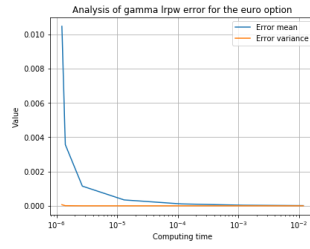
(a) Delta with LR method



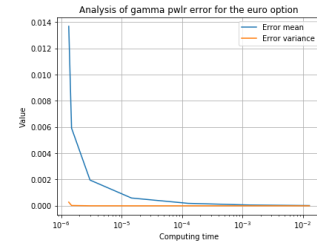
(b) Delta with PW method



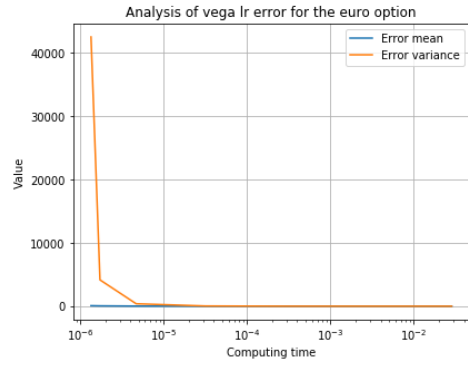
(c) Gamma with LRLR method



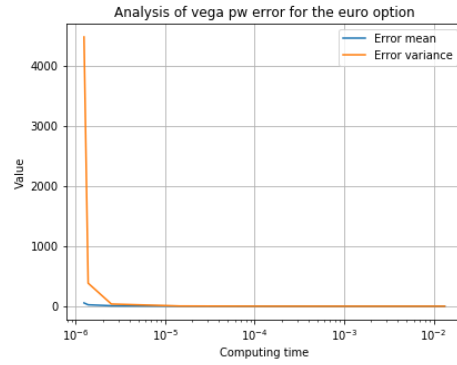
(d) Gamma with LRPW method



(e) Gamma with PWLR method



(f) Vega with LR method



(g) Vega with PW method

Figure 2: A graphical analysis of the error of the greeks depending on the time of computation. In order to generate these graphs, we have done one thousand simulations for each value, at every power of 10.