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# mc\_antithetic

#### Input parameters:

- Number of iterations N
- Generator\_Type
- Increment inc
- Confidence Value

#### Output parameters:

- $\bullet$  Price P
- Error Price  $\sigma_P$
- Delta  $\delta$
- Error delta  $\sigma_{\delta}$
- Price Confidence Interval:  $IC_P = [Inf Price, Sup Price]$
- Delta Confidence Interval:  $IC_{\delta} = [Inf Delta, Sup Delta]$

### Description:

Computation for a Call - Put - CallSpread or Digit European Option of its Price and its Delta with the Antithetic Variables method with Monte Carlo or Quasi-Monte Carlo simulation. In the case of Monte Carlo simulation, the method also provides an estimation for the integration error and a confidence interval.

The underlying asset price evolves according to the Black and Scholes model, that is:

$$dS_u = S_u((r-d)du + \sigma dB_u), \quad S_{T-t} = s$$

then

$$S_T = s \exp\left((r - d - \frac{\sigma^2}{2})t\right) \exp(\sigma B_t)$$

where  $S_T$  denotes the spot at maturity T, s is the initial spot, t is the time to maturity.

The Price of an option at T-t is:

$$P = E\left[\exp(-rt)f(K, S_T, R)\right]$$

where f denotes the payoff of the option, K the strike and R the rebate (for Digit option only).

The Delta is given by:

$$\delta = \frac{\partial}{\partial s} E[\exp(-rt)f(K, S_T, R)]$$

Estimators with antithetic variables are expressed as:

$$\widetilde{P} = \frac{1}{2N} \exp(-rt) \sum_{i=1}^{N} (P(i) + P'(i))$$

$$\widetilde{\delta} = \frac{1}{N} \exp(-rt) \sum_{i=1}^{N} \frac{\partial}{\partial s} P(i) = \frac{1}{2N} \exp(-rt) \sum_{i=1}^{N} (\delta(i) + \delta'(i))$$

P(i) and P'(i) are antithetic payoffs obtained from  $S_T(i)$  and  $S_T'(i)$  given by:

$$\begin{cases} S_T(i) = s \exp\left[\left(r - d - \frac{\sigma^2}{2}\right)t\right] \exp(\sigma B_t(i)) \\ S_T'(i) = s \exp\left[\left(r - d - \frac{\sigma^2}{2}\right)t\right] \exp(\sigma B_t'(i)) \end{cases}$$

where  $B_t(i) = \sqrt{t}g_i$  and  $B_t'(i) = -\sqrt{t}g_i$  are antithetic brownian motions.  $g_i$  is a standard gaussian variable.

Values for P(i) and  $\delta(i)$  are detailed for each option in the following points. Values for P'(i) and  $\delta'(i)$  are obtained in the same way.

• **Put**: The payoff is  $(K - S_T)^+$ . We have:

$$P(i) = (K - S_T(i))^+$$

$$\delta(i) = \begin{cases} -\frac{\partial S_T(i)}{\partial s} = -\frac{S_T(i)}{s} & \text{if } P(i) > 0\\ 0 & \text{otherwise} \end{cases}$$

• Call: The payoff is  $(S_T - K)^+$ . The Call-Put Parity relations for price and delta are expressed by:

$$C = P + s \exp(-dt) - K \exp(-rt)$$
$$\delta_C = \delta_P + \exp(-dt)$$

where C and P respectively denotes the Call and the Put prices. They will be used for a Call simulation (which corresponds to a method of control variate and leads generally to a reduced variance for the estimator).

• CallSpread: The payoff is  $(S_T - K_1)^+ - (S_T - K_2)^+$ . We have:

$$P(i) = \left[ (S_T(i) - K_1)^+ - (S_T(i) - K_2)^+ \right]$$

$$\delta(i) = \begin{cases} \frac{\partial S_T(i)}{\partial s} = \frac{S_T(i)}{s} & \text{if } S_T(i) > K_1 \text{ and } S_T(i) < K_2 \\ -\frac{\partial S_T(i)}{\partial s} = -\frac{S_T(i)}{s} & \text{if } S_T(i) > K_2 \text{ and } S_T(i) < K_1 \\ 0 & \text{otherwise} \end{cases}$$

• **Digit**: The payoff is  $R1_{\{S_T-K\geq 0\}}$ . We have:

$$P(i) = R1_{\{S_T(i) - K \ge 0\}}$$

To have an estimation of the Delta in the case of a Digit option, we need to use the increment value inc at each iteration i as:

$$\delta_i = \begin{cases} \frac{R}{2s \cdot inc} & \text{if } S_T(i)(s(1+inc)) > K & \text{and } S_T(i)(s(1-inc)) < K \\ 0 & \text{otherwise} \end{cases}$$

 $S_T(i)(s(1+inc))$  is the spot value at T with initial value s(1+inc).  $S_T(i)(s(1+inc))$  and  $S_T(i)(s(1-inc))$  are computed with the same brownian motion for each iteration. Thus we always have  $S_T(i)(s(1+inc)) > S_T(i)(s(1-inc))$  and there is only one case for which  $delta_i > 0$ .

# Algorithm:

/\* Value to construct the confidence interval \*/

For example if the confidence value is equal to 95% then the value  $z_{\alpha}$  used to construct the confidence interval is 1.96. This parameter is taken into account only for MC simulation and not for QMC simulation.

Strike  $K_1$  and  $K_2$  used for a Call-Spread option.

/\*Median forward stock and delta values\*/

Computation of intermediate values we use several times in the program.

/\* Change a Call into a Put to apply the Call-Put parity \*/

In case of Call, we modify parameters of the option; they will be reinitialized at the end of the simulation program. Simulation will be done as for a put.

### • /\*MC sampling\*/

Initialization of the simulation: generator type, dimension, size N of the sample

/\* Test after initialization for the generator \*/

Test if the dimension of the simulation is compatible with the selected generator. (See remarks on QMC simulation, especially on dimension of low-discrepancy sequences). For standard Monte Carlo in the one-dimensional Black and Scholes model, we never have any problem with the dimension, fixed to 1 at the beginning of the programm.

Definition of a parameter which exprimes if we realize a MC or QMC simulation. Some differences then appear in the algorithm: for simulation of a gaussian variable and in results in the simulation.

### /\* Begin N iterations \*/

- /\* Simulation of a gaussian variable according to the generator type, that is Monte Carlo or Quasi Monte Carlo. \*/

Call to the appropriate function to generate a standard gaussian variable. See the part about simulation of random variables for explanations on this point. We just recall that for a MC simulation, we use the Gauss-Abramovitz algorithm, and for a QMC simulation we use an inverse method.

At the iteration i, we obtain  $S_T(i)$  and  $S_T'(i)$  from a simulation according to the selected generator of antithetics brownian motions  $B_t(i)$  as  $\sqrt{t}g_i$  and  $B_t'(i)$  as  $-\sqrt{t}g_i$  where  $g_i$  is a standard gaussian variable.

$$P(i) = \text{Payoff}(S_T(i), K)$$

$$P'(i) = \text{Payoff}(S'_T(i), K)$$

### - /\*Delta\*/

Calculation of Delta  $\delta_i$  and  $\delta'_i$  with formula previously detailed for each option.

Computation of the sums  $\sum \frac{1}{2}(P(i)+P'(i))$  and  $\sum \frac{1}{2}(\delta(i)+\delta'(i))$  for the mean price and the mean delta.

/\*Sum of squares\*/

Computation of the sums  $\sum (P(i) + P'(i))^2$  and  $\sum (\delta(i) + \delta'(i))^2$  necessary for the variance price and the variance delta estimations. (finally only used for MC estimation)

• /\*Price\*/

The price estimator is:

$$P = \frac{1}{N} \exp(-rt) \sum_{i=1}^{N} \frac{1}{2} (P(i) + P'(i))$$

The error estimator is  $\sigma_P$  with :

$$\sigma_P^2 = \frac{1}{N-1} \left( \frac{1}{N} \exp(-2rt) \sum_{i=1}^N \frac{1}{4} \left( P(i) + P'(i) \right)^2 - P^2 \right)$$

• /\*Delta\*/ 
$$\delta = \frac{1}{N} \exp(-rt) \sum_{i=1}^{N} \frac{1}{2} (\delta(i) + \delta'(i))$$

The error estimator is  $\sigma_{\delta}$  with:

$$\sigma_{\delta}^{2} = \frac{1}{N-1} \left( \frac{1}{N} \exp(-2rt) \sum_{i=1}^{N} \frac{1}{4} (\delta(i) + \delta'(i))^{2} - \delta^{2} \right)$$

• /\* Call Price and Delta with the Call Put Parity \*/

We now compute the price and the delta in case of a call, because call was considered as a put until now.

Parameters of the option are reinitialized. This step is necessary: if you want to begin an other call simulation just after a first one with the antithetic

method, the parameters must have been modified to specify that we really consider a Call and not a Put.

• /\* Price Confidence Interval \*/ The confidence interval is given as:

$$IC_P = [P - z_\alpha \sigma_P; P + z_\alpha \sigma_P]$$

with  $z_{\alpha}$  computed from the confidence value.

 $\bullet$  /\* Delta Confidence Interval \*/ The confidence interval is given as:

$$IC_{\delta} = [\delta - z_{\alpha}\sigma_{\delta}; \delta + z_{\alpha}\sigma_{\delta}]$$

with  $z_{\alpha}$  computed from the confidence value.

Confidence intervals are always computed, but for a QMC simulation they don't work, thus they don't appear in the results.

## References