Quantlib project

Project 1 : Use constant parameters in Monte Carlo engines

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1. Problem

In Monte Carlo engines, repeated calls to the process methods may cause a performance hit; especially when the process is an instance of the GeneralizedBlackScholesProcess class, whose methods in turn make expensive method calls to the contained term structures. The performance of the engine can be increased at the expense of some accuracy. Create a new class that models a Black-Scholes process with constant parameters (underlying value, risk-free rate, dividend yield, and volatility); then modify the MCEuropeanEngine class so that it still takes a generic Black-Scholes process and an additional boolean parameter. If the boolean is false, the engine runs as usual; if it is true, the engine extracts the constant parameters from the original process (based on the exercise date of the option; for instance, the constant risk-free rate should be the zero-rate of the full risk-free curve at the exercise date) and runs the Monte Carlo simulation with an instance of the constant process. Compare the results (value, accuracy, elapsed time) obtained with and without constant parameters and discuss.

2. Implementation

The Monte-Carlo process can be decomposed in few steps:

- Random number of generations (2 types: MersemneTwister (totally random) and SobolRsg (random according to the dimensionality and the size of the vector chosen))
- 2. Stochastic processes (dX = mu(t,X)*dt + sigma(t,X)dW)
 - \circ mu(t,X) = drift component
 - sigma(t,X) = diffusion component
 - dW = random component
- 3. Path generation (result of 1 and 2) (define the grid of time)
- 4. Path pricing
- 5. Simulations (result of 3 and 4)

Knowing the steps of the Monte-Carlo process, the idea is to reuse some components of the Black-Scholes process in order to get the Monte Carlo simulations of pricing better performance.

So, we created the ConstantBlanckScholesProcess class to reuse some components of the original process. This class takes constant parameters such as the underlying value, the risk-free rate, the volatility and the dividend. Moreover, this class defines 4 other methods necessary for the development of the stochastic process in our class. The draft of the class is shown below:

ConstantBlackScholesProcess.hpp

The next step was to add a boolean (named "isConstantBS" in our implementation) in the MCEuropeanEngine class :

- if the boolean is false, the engine runs as usual
- if the boolean is true, the engine runs with the implementation of our ConstantBlackScholesProcess class which takes constant parameters of the original process

The draft of the constructor with the boolean added is shown below:

```
// constructor
MCEuropeanEngine_2(
    const boost::shared_ptr<GeneralizedBlackScholesProcess>& process,
    Size timeSteps,
    Size timeStepsPerYear,
    bool brownianBridge,
    bool antitheticVariate,
    Size requiredSamples,
    Real requiredTolerance,
    Size maxSamples,
    BigNatural seed,
    bool isConstantBS);
```

Constructor of MCEuropeanEngine class with the added boolean parameter

Finally, we overrode the PathGenerator method inherited from the parent class MCVanillaEngine so that our implementation takes into account the boolean parameter added as we can see below:

```
ext::shared_ptrcpath_generator_typep pathGenerator() const override {

    Size dimensions = McVanillaEnginecSingleVariate, RNG, S>::process_->factors();
    TimeGrid grid = this->timeGrid();
    typename RNG::rsg_type generator =

        RNG::make_sequence_generator(dimensions * (grid.size() - 1), McVanillaEnginecSingleVariate, RNG, S>::seed_);

//bool isConstantBS_true;

if (isConstantBS_true;

if (isConstan
```

Override of the PathGenerator method

Thus, we run our implementation modifying the main and compiling our new version of ConstantBlackScholesProcess class in order to get the results obtained in the next chapter.

3. Results

For our tests we compute the NPV value of the option, and we also check the elapsed time and estimated error. Those values are computed for the general and constant Black-Scholes process. In order to analyze our results, by passing an argument to our main, we store results in a csy file.

```
(base) sprietthibault:project1/ (master*) $ ./main results/simu.csv
```

calling main with path to store results

```
bool write_results = argc > 1;
if(write_results){
    std::ofstream file;
    file.open(argv[1],std::ios::app);
    file << "step,sample,c_err,c_npv,c_time,err,npv,time,\n";
    file.close();
}</pre>
```

create results file with columns

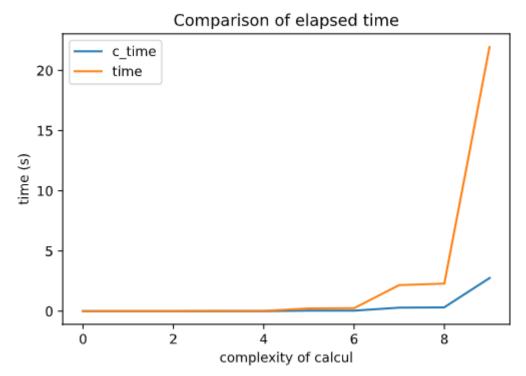
```
if(write_results){
    std::ofstream file;
    file.open(argv[1],std::ios::app);
    file << timeSteps << "," << n_samples << "," << errorEstimate1 << "," << NPV1 << ","
        << us1 << "," << errorEstimate2 << "," << NPV2 << "," << us2 << "," << std::endl;
    file.close();
}</pre>
```

write results in file

```
step, sample, c_err, c_npv, c_time, err, npv, time,
10,100,0.322852,4.13251,256,0.322852,4.13251,322,
10,1000,0.0994568,4.22306,349,0.0994568,4.22306,2541,
10,10000,0.0304028,4.17516,3527,0.0304028,4.17516,23431,
10,100000,0.00963383,4.16356,37449,0.00963383,4.16356,237482,
10,1000000,0.00304669,4.17073,315840,0.00304669,4.17073,2.28276e+06,
100,100,0.300936,4.32015,332,0.300936,4.32015,2298,
100,1000,0.0958765,4.18166,2843,0.0958765,4.18166,23083,
100,10000,0.030484,4.15086,35736,0.030484,4.15086,224465,
100,100000,0.00962119,4.17036,278741,0.00962119,4.17036,2.16902e+06,
100,1000000,0.00304455,4.17184,2.74929e+06,0.00304455,4.17184,2.19504e+07,
```

results

Firstly we compared the elapsed time. To do so, we sort time values in ascending order and we plot the two series on the same plot. We have done calculations for different timesteps and number of samples values.



In orange: time for general method; in blue: time for constant method.

We notice that the more timesteps and samples the more gap between methods is important.

We also compute the sum of elapsed time and the results are :

- General method is 786 times longer in calculation time
- The relative difference is 686%

Then we compare value of NPV and estimated error and we notice that they were the same

	c_err	err		c_npv	npv
0	0.003045	0.003045	0	4.13251	4.13251
1	0.003047	0.003047	1	4.15086	4.15086
2	0.009621	0.009621	2	4.16356	4.16356
3	0.009634	0.009634	3	4.17036	4.17036
4	0.030403	0.030403	4	4.17073	4.17073
5	0.030484	0.030484	5	4.17184	4.17184
6	0.095877	0.095877	6	4.17516	4.17516
7	0.099457	0.099457	7	4.18166	4.18166
8	0.300936	0.300936	8	4.22306	4.22306
9	0.322852	0.322852	9	4.32015	4.32015

With these results for the NPV, we conclude that there should be an error in our implementation for the calculation of the NPV and so the estimation of the error that we did not find.