Problem Set 2 Monte Carlo Methods for Option Pricing

- This homework consists of three questions, each carrying equal marks. Your code will be graded by a Python script which compares your result against the baseline result. It is therefore important that your Python stack is identical to the baseline so that you avoid losing marks even though your code is correct.
- Submit three modules, BSMonteCarlo.py, MCStockPrices.py, MCOptionPrices.py and a short report in PDF format.

Problem 1. Write a function to price vanilla european-exercise options

where

- S0, K, T, sigma are the underlying price, the option strike, the maturity and the volatility respectively.
- 'checkpoints' is an ordered list of integer sample counts at which to return the running mean, standard deviation, and estimated error.
- 'rateCurve' is an InterestRateCurve stored as a numpy array.
- 'samples' is a numpy array of uniform random samples to use. If this input is None then your code should generate a fresh 1-dimensional sample array with M elements where M is the final entry in checkpoints.

Note that samples (when specified) must be a list with at least as many elements as the last entry in checkpoints. Raise an exception if this is not the case.

This function must return a dictionary with the following entries:

```
{ 'TV': , # The final value ( i.e. mean at checkpoints[-1] )
   'Means': , # The running mean at each checkpoint
   'StdDevs': , # The running standard deviation at each checkpoint
   'StdErrs': , # The running standard error at each checkpoint
}
```

Problem 2. Given uniform samples we can generate M paths of stock prices at given times t under the assumption of geometric brownian motion. Write a function

```
def MCStockPrices(SO, sigma, rateCurve, t,
```

samples, integrator):

where

- 'SO' is the stock prices at time t_0 .
- 'sigma' is the constant volatility.
- 'rateCurve' is an InterestRateCurve stored as a numpy array.
- 't' is an array of fixing times t_i , i = 1 ... N to simulate to.
- 'samples' is an array of uniform random samples to use. The length of samples should be $N \times M$ where N is the number of fixing times and M is the number of paths.
- integrator controls how the samples are generated according to the following value list
 - 'standard', where the paths are generated by using the solution of the Black-Scholes SDE step-by-step
 - 'euler', to use Euler-method integration of the Black-Scholes SDE
 - 'milstein', to use Milstein-method integration of the Black-Scholes SDE

The function should return a numpy array of simulated stock prices having the same dimensions as samples.

Problem 3. Combine your BSMonteCarlo and MCStockPrices functions from Questions 1 and 2 to write a function

to approximate the prices of European Options where

- 't' is an array of fixing times t_i , i = 1 ... N to simulate to.
- 'K' is the strike price.
- 'T' is the expiration date of the European option.
- 'rateCurve' is an InterestRateCurve stored as a numpy array.
- 'checkpoints' is an ordered list of integer sample counts in the range [1, M] at which to return the running mean, standard deviation, and estimated error.
- 'samples' is an array of uniform random samples to use. The length of samples should be $N \times M$ where N is the number of fixing times and M is the number of paths.
- integrator controls how the samples are generated according to the following value list
 - 'standard', where the paths are generated by using the solution of the Black-Scholes SDE step-by-step
 - 'euler', to use Euler-method integration of the Black-Scholes SDE
 - 'milstein', to use Milstein-method integration of the Black-Scholes SDE

This function must return a dictionary with the following entries:

Submit a brief report comparing the error properties of each integrator with the exact option price. Your studies should compare the convergence properties of each scheme as M and N vary.