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Help
#if defined(PremiaCurrentVersion) && PremiaCurrentVersion <</pre>
     (2007+2) //The "#else" part of the code will be freely av
    ailable after the (year of creation of this file + 2)
#else
#include <cmath>
#include "generator.h"
#include <vector>
#ifndef montecarlo2 h
#define montecarlo2 h
//Euler Scheme
//X(p)=X(p-1)+b(X(p), ph)h+sigma(X(p), ph)*sqrt(h)*random
    variable
//The functions b(), sigma() are differentes for differen
    tes models
//so we give the pointer to a model where these functions
    are described
//Also the vector of random variable can be different (wit
    h component correlated or non)
//so we give the pointer to a random vector.
//_nstep - number of partitions
std::vector<double> scheme_euler(rv_vector* _ptr_rv, model*
     _ptr_model, int _nstep,int generator)
{
  //h- step of discretisation
  double h=_ptr_model->T/_nstep;
  std::vector<double> x= ptr model->x0;
  for(int i=0; i<_nstep; i++)</pre>
      x=x+ ptr model->f b(x, ((double)i)*h)*h + ptr model-
    >f_sigma(x, ((double)i)*h)*sqrt(h)*_ptr_rv->get_rv();
    }
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return x;
//Ninomiya-Victoir Scheme (Kusuoka Scheme)
//The functions f 1, f 2 and the vector of random variable
    are differentes for differentes model
//_nstep - number of partitions
std::vector<double> scheme_kusuoka(rv_vector* _ptr_rv,
    model* _ptr_model, int _nstep,int generator)
{
  //h- step of discretisation
  double h=_ptr_model->T/_nstep;
  std::vector<double> x=_ptr_model->x0;
  rv bernoulli rv b(0.5,-1,1,generator);
  double rv_b_real;
  for (int i=0; i< nstep;i++)</pre>
      x = ptr_model \rightarrow exp_V0(0.5*h, x);
      rv_b_real=rv_b.get_rv();
      std::vector<double> y=_ptr_rv->get_rv();
      if (rv b real==1)
  x = ptr_model \rightarrow f_1(x, h, y);
      else
  x=ptr_model->f_2(x, h, y);
      x=_ptr_model->exp_V0(0.5*h, x);
  return x;
//the Monte Carlo Method
//we estimate the solution of stochastic differential equa
    tion by discretisation scheme (nres)
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//we calculate a function of this solution (x)
//then (the last step) we apply the Monte Carlo Method to
    x:
//we construct the sum of x ( niter times) (=nsum x)
//_nstep
             step of discretisation (for discretisation
    scheme)
         number of trajectories in the Monte Carlo Method
// niter
//_ptr_rv
            pointer to the vector of random variable (
    for discretisation scheme)
// function function of processus discretized
//_function_delta function of option delta
// schema
           name of the scheme
//_ptr_model
               pointer to a model
// nerror
           Monte Carlo Method error (_nerror*_nerror is
     a a variance of Monte Carlo method result)
            delta of option
// ndelta
//_nerror_delta
                  delta error
template < class A, class B, class C> double monte carlo
  (int _nstep, int _niter, rv_vector* _ptr_rv, A _function,
     B _function_delta, C _scheme, model* _ptr_model,int
                                                          generator,double& _
{
  std::vector<double> nres= ptr model->x0;
  double x=0.;
  double nsum x=0.;
  double nsum_x_x=0.;
  double ndelta=0.;
  double nsum delta=0.;
  double nsum_delta_delta=0.;
  double one n=(1./((double) niter));
  for (int i=0; i<_niter; i++)</pre>
      nres=_scheme(_ptr_rv, _ptr_model, _nstep,generator);
      x=_function(nres, _ptr_model);
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nsum x+=x;
     nsum x x+=x*x;
      ndelta=_function_delta(nres, _ptr_model);
      nsum delta+=ndelta;
      nsum delta delta+=ndelta*ndelta;
    }
  _nerror=sqrt(one_n*std::abs(nsum_x_x-one_n*nsum_x*nsum_x)
    /((double)(_niter-1)));
  ndelta=nsum delta*one n;
  _nerror_delta=sqrt(one_n*std::abs(nsum_delta delta-one n*
    nsum_delta*nsum_delta)/((double)(_niter-1)));
 return nsum_x*one_n;
}
//the Monte Carlo Method / Varince reduction
//we estimate the solution of stochastic differential equa
    tion by discretisation scheme (nres)
//we calculate a function of this solution (x)
//we construct a control variable (y)
//then we apply the varinace reduction technique
//In place of sum x/ niter we estimate (nsum x -alpha*nsum
    y)/_niter+alpha*nesp_y;
//this expression gives a mean empiric with a smaller
    variance.
//where alpha is calculed by (covariance(x,y)/variance(y))
//here x-variable estimed and y - control variable ;
//the parameters are the same that one for monte carlo
//_ncorr the correlation coefficient between variable es
    timated and control variable
template<class A, class B, class C> double monte_carlo2
```

```
(int _nstep, int _niter, rv_vector* _ptr_rv, A _function,
   B _function_delta, C _scheme, model* _ptr_model,int
                                                            generator, double&
  double &_ncorr)
double epsilon=DBL EPSILON;
std::vector<double> nres=_ptr_model->x0;
double x=0.;
double y=0.;
double nsum x=0.;
double nsum_y=0.;
double nsum_x_y=0.;
double nsum_y_y=0.;
double nsum_x_x=0.;
double ndelta=0.;
double nsum_delta=0.;
double nsum delta delta=0.;
//mean of a control variable
double nvar=0.;
double nesp_y=_ptr_model->f_esp(nvar);
double one_n=(1./((double)_niter));
for (int i=0; i< niter; i++)
  {
   nres=_scheme(_ptr_rv, _ptr_model, _nstep,generator);
   x= function(nres, ptr model);
   y=_ptr_model->f_control(nres);
   nsum x+=x;
   nsum_y+=y;
   nsum_x_y+=x*y;
   nsum_y_y+=y*y;
   nsum x x+=x*x;
    ndelta=_function_delta(nres, _ptr_model);
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nsum delta+=ndelta;
      nsum delta delta+=ndelta*ndelta;
    }
  double ncov x y=one n*nsum x y-one n*one n*nsum y*nsum x;
  double nvar_x=one_n*nsum_x_x-one_n*one_n*nsum_x*nsum_x;
  double nvar y=one n*nsum y y-one n*one n*nsum y*nsum y;
  double alpha=0.;
  alpha=(std::abs(nvar_y)<=epsilon)? 0.:ncov_x_y/nvar_y;</pre>
  _nerror=(std::abs(nvar_y)<=epsilon)? sqrt(one_n*std::abs(</pre>
    nvar_x)): sqrt(one_n*std::abs(nvar_x-ncov_x_y*ncov_x_y/nvar
    _y));
  _ncorr=((std::abs(nvar_y)<=epsilon) || (std::abs(nvar_x)</pre>
    =epsilon))? 0. : ncov_x_y/sqrt(std::abs(nvar_x*nvar_y));
  _ndelta=nsum_delta*one_n;
  _nerror_delta=sqrt(one_n*std::abs(nsum_delta_delta-one_n*
    nsum delta*nsum delta)/((double)( niter-1)));
  return one_n*nsum_x -alpha*one_n*nsum_y+alpha*nesp_y;
}
#endif
#endif //PremiaCurrentVersion
```

References