

FE545 Design, Patterns and Derivatives Pricing

# Strategies, Decoration, and Statistics

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# Overview

Design a Statistics Collection Class

Templates and Wrappers

Breakout Exercise: Compute More Statistics

A Convergence Table

Key Points

# Design a Statistics Collection Class

- ▶ One important aspect of the Monte Carlo Simulation is to check the convergence and additional analysis of the simulation quality.
- ▶ What we need to measure the standard error, a convergence table, or return all the path data and analyze them somewhere else.
- ▶ One of the important goals of the statistics gather is to be **reusable**.
  - ▶ For example, we might have many other Monte Carlo routines such as an exotics pricer or a BGM pricer for interest rate derivatives, etc.
  - ▶ If we are developing a risk management system, we might be more interested in the ninety-fifth percentile, or in the conditional expected shortfall, than in the mean and variance.

# Design a Statistics Collection Class

- ▶ How should the routine do? It must have two principal methods. The first should take in data for each path. The second must output the desired statistics.
  - ▶ We start with an abstract base class using the virtual methods, just as we did for the `PayOff`
  - ▶ We have to decide the precise interface for our two principal methods. They will be pure virtual functions declared in the base class and defined in the concrete inherited class.
  - ▶ Our second method will indeed return the results, and require a little more thought. We have to decide what sort of object to return the results in.
  - ▶ We may want to our results to be flexible to accommodate many forms so we may choose to use a vector of vectors.

# A Statistics Gatherer

- ▶ Virtual Function
  - ▶ StatisticsMC.h
  - ▶ StatisticsMC.cpp

## StatisticsMC.h

```
#ifndef __ch5_statistics_gatherer__MCStatistitcs__
#define __ch5_statistics_gatherer__MCStatistitcs__

#include <vector>

class StatisticsMC
{
public:
    StatisticsMC(){}

    virtual void DumpOneResult(double result)=0; // a pure virtual function
    virtual std::vector<std::vector<double>> GetResultsSoFar()const=0;
    virtual StatisticsMC* clone()const=0; // possibility of virtual copy constructor
    virtual ~StatisticsMC(){} // virtual destructor

private:
};
```

# Using the Statistics Gatherer

- ▶ Virtual Function
  - ▶ SimpleMC7.h
  - ▶ SimpleMC7.cpp
  - ▶ StatisticsMain1.cpp

## SimpleMC7.h

```
void SimpleMonteCarlo5(const VanillaOption& TheOption,
                      ...
                      unsigned long NumberOfPaths,
                      StatisticsMC& gatherer)
{
    ...
    double discounting = exp(-r.Integral(0, Expiry));

    for (unsigned long i = 0; i<NumberOfPaths; i++) {
        ...
        double thisPayOff = TheOption.OptionPayOff(thisSpot);
        // all work on accounting the mean is wrapped
        gatherer.DumpOneResult(thisPayOff*discounting);
    }
    return;
}
```

# Templates and Wrappers

- ▶ We have so far created a class hierarchy for gathering statistics. This hierarchy includes a *virtualconstructor*, *clone*, so we can copy these objects without knowing their types.
- ▶ The wrapper class will provide various functionalities which are inherited to make it act like a pointer to a single object but with added responsibilities.
  - ▶ If we copy the *Wrapper* object, the pointed-to object is also copied, so that each *Wrapper* object has its own copy of the pointed-to object.
  - ▶ When the *Wrapper* object ceases to exist because of going out of scope, or being deleted, the pointed-to object is automatically deleted as well.
  - ▶ If we set one *Wrapper* object to another, then the object previously pointed-to must be deleted, and then a copy of the new object must be created so each *Wrapper* still owns precisely on object.
  - ▶ It must possible to dereference the *Wrapper* to obtain the underlying object. For example, if you put *\*mywrapper* then you should obtain the object pointed to by *mywrapper*.

# Templates and Wrappers

- ▶ Template and Wrapper Class
  - ▶ Wrapper.h
  - ▶ Wrapper.cpp

## Wrapper.h

```
// specify class T later
// eg. Wrapper<MCStatistics> TheStatsGatherer;
template <class T>

class Wrapper
{
public:
    Wrapper(){ DataPtr = 0;} // the default constructor which point to not
    Wrapper(const T& inner) // constructor
    {
        DataPtr = inner.clone();
    }
    ~Wrapper()
    {
        if (DataPtr != 0) // if DataPtr Point to some memory, release it
        {
            delete DataPtr;
        }
    }
}
```



# Templates and Wrappers

- ▶ We start before each declaration with the command  
`template<class T>`
- ▶ The compiler will produce one copy of the code for each different sort of  $T$  that is used. Thus if we declare  
`Wrapper<MCStatistics> TheStatsGatherer;`
- ▶ This has side effects:
  - ▶ The first is that all the code for the *Wrapper* template is in the header file and there is no *Wrapper.cpp* file.
  - ▶ The second is that if we use the *Wrapper* class many times, we have to compile a lot more code than we might expect.
- ▶ We need to provide two different versions of the dereferencing  
`operator*`  
`operator->`  
to have both *const* and *non-const* versions.

# Breakout Exercise: Compute More Statistics

- ▶ Write a statistics gathering class that computes the first four moments of a sample.
  - ▶ Pair with another person as a team; one is writing the code and another is watching and then take turns to complete the design.
  - ▶ Use the `c++` project named `ch5 statistics gatherer` as the starting point for your exercise.
  - ▶ Run the simulation 100 times and gather the required results.

# A Convergence Table

- In order to specify how *accurate* a particular estimate  $\hat{\ell}$  is, that is, how close it is to the actual unknown parameter  $\ell$ , one needs to provide not only a point estimate  $\hat{\ell}$  but a confidence interval as well.
- By the central limit theorem  $\hat{\ell}$  has approximately a  $N(\ell, \sigma^2/N)$  distribution, where  $\sigma^2$  is the variance of  $H(\mathbf{X})$ . Usually  $\sigma^2$  is unknown, but it can be estimated with the *sample variance*

$$S^2 = \frac{1}{(N-1)} \sum_{i=1}^N (H(\mathbf{x}_i) - \hat{\ell})^2, \quad (1)$$

which (by the law of large numbers) tends to  $\sigma^2$  as  $N \rightarrow \infty$ .

- Consequently, for large  $N$  we see that  $\hat{\ell}$  is approximately  $N(\ell, S^2/N)$  distributed.

# Confidence Interval

- Thus, if  $z_\gamma$  denote the  $\gamma$ -quantile of the  $N(0, 1)$  distribution (this is the number such that  $\Phi(z_\gamma) = \gamma$ , where  $\Phi$  denotes the standard normal cdf; for example  $z_{0.95} = 1.645$ , since  $\Phi(1.645) = 0.95$ ), then

$$\mathbb{P} \left( \hat{\ell} - z_{1-\alpha/2} \frac{S}{\sqrt{N}} \leq \ell \leq \hat{\ell} + z_{1-\alpha/2} \frac{S}{\sqrt{N}} \right) \approx 1 - \alpha \quad (2)$$

- In other words, an approximate  $(1 - \alpha)100\%$  *confidence interval* for  $\ell$  is

$$\left( \hat{\ell} \pm z_{1-\alpha/2} \frac{S}{\sqrt{N}} \right), \quad (3)$$

where the notation  $(a \pm b)$  is shorthand for the interval  $(a - b, a + b)$ .

# Confidence Interval

- It is a common practice in simulation to use and report the *absolute and relative* widths of this confidence interval, defined as

$$w_a = 2z_{1-\alpha/2} \frac{S}{\sqrt{N}} \text{ and } w_r = \frac{w_a}{\hat{\ell}}, \quad (4)$$

respectively, provided that  $\hat{\ell} > 0$ .

- The absolute and relative widths may be used as stopping rules (criteria) to control the length of a simulation run. The relative width is particularly useful when  $\ell$  is very small.

For example, think of  $\ell$  as the unreliability (1 minus the reliability) of a system in which all the components are very reliable. In such a case  $\ell$  could be as small as  $\ell \approx 10^{-10}$ , so that reporting a result such as  $w_a = 0.05$  is almost meaningless, while in contract,  $w_r = 0.05$  is quite meaningful.

# Convergence Table

- ▶ One alternative method is to use a convergence table: rather than returning statistics for the entire simulation, we instead return them for every power of two to get an idea of how the numbers are varying.
- ▶ We could just write a class directly to return such a table for the mean, but since we might want to do this for any statistic, we do it in a reusable fashion. Our class must contain a statistics gather in order to decide for which statistics to create a convergence table.
- ▶ We therefore define a class *ConvergenceTable* which is inherited from *StatisticsMC* and has a wrapper of an *StatisticsMC* object as a data member.

# A Convergence Table

- ▶ A Convergence Table
  - ▶ ConvergenceTable.h
  - ▶ ConvergenceTable.cpp

## ConvergenceTable.h

```
class ConvergenceTable : public StatisticsMC
{
public:
    // constructor, take in a wrapper class which point to a StatisticMC o
    ConvergenceTable(const Wrapper<StatisticsMC>& Inner_);

    virtual StatisticsMC* clone() const;
    virtual void DumpOneResult(double result);
    virtual std::vector<std::vector<double>> GetResultsSoFar() const;

private:
    Wrapper<StatisticsMC> Inner;
    std::vector<std::vector<double>> ResultSoFar;
    unsigned long StoppingPoint;
    unsigned long PathsDone;
};
```

# Decoration

- ▶ The technique of this section is an example of a standard design pattern called the *decorator pattern*.
- ▶ We have added functionality to a class without changing the interface. This process is called *decoration*.
- ▶ We can decorate as many times as we wish. It would be syntactically legal (but not useful), for example, to have a convergence table of convergence tables.
  - ▶ If we have a stream of numbers defining a time series, we often want the statistics of the successive increments instead of the numbers themselves. A decorator class could therefore do this differencing and pass the difference into the inner class.
  - ▶ We might want more than one statistic for a given set of numbers; rather than writing one class to gather many statistics, we could write a decorator class which contains a vector of statistics gatherers and passes then gathered values to each one individually.



# Key Points

- ▶ Routines can be made more flexible by using the strategy pattern.
- ▶ Making part of an algorithm be implemented by an inputted class is called the strategy pattern.
- ▶ For code that is very similar across many different classes, we can use templates to save time in rewriting.
- ▶ If we want containers of polymorphic objects, we must use wrappers or pointers.
- ▶ Decoration is the technique of adding functionality by placing a class around a class which has the same interface; i.e. the outer class is inherited from the same base class.