OOP PROJECT

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Introduction

This project demonstrates the application of Monte Carlo simulation in C++ for pricing European and American options. By moving beyond the constraints of the Black-Scholes model—especially its inability to handle path-dependent and early-exercise options—this simulation offers a flexible and precise alternative. Through random asset path simulations, we'll explore a method that adapts to complex pricing needs, particularly for American options with early exercise potential. This report delves into the theoretical foundation, practical coding steps, and experimental results to showcase the strengths of Monte Carlo methods in delivering more accurate and adaptable option pricing solutions.

Background

Options are financial derivatives that provide the buyer with the right, but not the obligation, to buy or sell an asset at a specified price before or at a certain date. Accurately pricing options is crucial in financial markets, as it helps investors assess risk, make informed decisions, and manage portfolios effectively.

- Call Options: These give the buyer the right to purchase an asset at a specified price (the strike price) within a certain timeframe. Call options are often used when investors expect the underlying asset's price to rise, allowing them to buy it at a lower price.
- Put Options: These give the buyer the right to sell an asset at the strike price within a specific period. Put options are valuable for investors who expect the asset's price to decrease, providing a way to sell it at a higher price than its anticipated market value.

There are different types of options, with European and American options being among the most common

- European Options: Exercisable only at expiration, simplifying the pricing model.
- American Options: Allow exercise at any point before expiration, adding complexity due to early exercise.

Black-Scholes Model Limitations: The Black-Scholes model, developed in the 1970s, is one of the most well-known methods for option pricing. It provides a closed-form solution for European options, assuming constant volatility, risk-free interest rates, and log-normally distributed asset prices. However, despite its popularity, the Black-Scholes model has limitations:

Path Independence: The Black-Scholes model assumes that only the final price of the underlying asset at expiration matters, which works for European options but not American options that depend on interim prices due to early exercise.

Constant Volatility Assumption: In reality, market volatility is often dynamic, changing over time due to various factors, which the Black-Scholes model does not account for.

Lack of Flexibility for Complex Derivatives: The model struggles to price more complex derivatives, especially those with path-dependent features or when early exercise is allowed. Due to these limitations, the Black-Scholes model may not provide accurate pricing for American options or other derivatives with more complex features. Monte Carlo simulation, on the other hand, offers a flexible alternative by simulating various paths for asset prices, accommodating early exercise, and adapting to changing volatility.

Objective

The objective of this project is to develop a Monte Carlo simulation to price European and American options, focusing on addressing the limitations of the Black-Scholes model. By simulating random asset price paths, the Monte Carlo method provides a more adaptable framework for pricing options, especially American options that allow for early exercise. This project will involve implementing Monte Carlo methods in C++ to simulate asset paths, calculate option payoffs, and compare the estimated option prices to those derived from the Black-Scholes model. Through this comparison, we aim to demonstrate the advantages of Monte Carlo simulation in handling scenarios where traditional models may be insufficient.

Overview of the Black-Scholes Model

The Black-Scholes model is one of the most popular methods for pricing European options. It provides a closed-form solution for calculating the fair price of an option based on certain inputs and assumptions. The formula for a European call option is:

$$C = S_0 N(d_1) - K e^{-rT} N(d_2)$$

where:

$$d_1 = \frac{\ln\left(\frac{S_0}{K}\right) + \left(r + \frac{\sigma^2}{2}\right)T}{\sigma\sqrt{T}}$$
$$d_2 = d_1 - \sigma\sqrt{T}$$

And for a European put option:

$$P = Ke^{-rT}N(-d_2) - S_0N(-d_1)$$

Here:

- C and P are the prices of the call and put options, respectively.
- S_0 is the current stock price.
- K is the strike price.
- T is the time to expiration.
- \bullet r is the risk-free interest rate.
- σ is the volatility of the stock's returns.
- $N(\cdot)$ represents the cumulative distribution function of the standard normal distribution.

Key Assumptions of the Black-Scholes Model

- Constant Volatility: The model assumes that the volatility of the asset remains constant over the option's life, which may not reflect actual market conditions.
- Lognormal Asset Prices: It assumes that asset prices follow a geometric Brownian motion and are lognormally distributed.
- No Early Exercise: This assumption makes the Black-Scholes model suitable only for European options, which can only be exercised at expiration.
- Risk-Free Interest Rate: The interest rate is constant over the option's life.
- No Dividends: The basic Black-Scholes formula assumes no dividends; however, there are versions that adjust for dividend payments.

Pricing European Options using Black-Scholes in c++

Overview

We will utilize C++ in the implementation of the Black-Scholes formula for pricing European call and put options. The implementation includes functions for calculating the normal probability density function (PDF) and cumulative distribution function (CDF), as well as functions for calculating the option prices.

Parameters of the Options

For this example, we are using the following parameters to price the options:

- Underlying Price (S): \$100
- Strike Price (K): \$105
- Risk-Free Interest Rate (r): 5% (0.05)
- Volatility (v): 20% (0.2)
- Time to Expiry (T): 1 year

These parameters represent a typical setup for option pricing, with a current underlying price of \$100, a strike price slightly above at \$105, and a risk-free rate and volatility that are reasonable for financial models.

Code Explanation

With these parameters in mind, here's a breakdown of the main components of the code:

1. Standard Probability Density Function (PDF)

Listing 1: PDF Function

```
double norm_pdf(const double& x) {
    return (1.0 / (pow(2 * M_PI, 0.5))) * exp(-0.5 * x * x);
}
```

This function calculates the standard normal PDF for a given x, which is used to find probabilities in the normal distribution.

2. Standard Cumulative Distribution Function (CDF)

Listing 2: CDF Function

This function approximates the standard normal CDF, which is essential for calculating $N(d_1)$ and $N(d_2)$ in the Black-Scholes formula. It gives the cumulative probability for values under the normal distribution.

3. Calculating d_1 and d_2

Listing 3: d1d2 Function

The d1d2 function calculates d_1 and d_2 , which are used in the Black-Scholes formula.

4. Call and Put Price Functions

Listing 4: Call and Put Price Functions

call price calculates the Black-Scholes price for a European call option, and put price calculates the price for a European put option.

5. Main Function

Listing 5: Main Function

```
int main() {
      fast_cin();
      // Option parameters
      double S = 100.0; // Underlying price
      double K = 105.0; // Strike price
      double r = 0.05; // Risk-free rate
      double sigma = 0.2;
                             // Implied Volatility
                         // Time till Expiry (1 year)
      double T = 1.0;
      // Calculate call and put prices
      double call = call_price(S, K, r, sigma, T);
      double put = put_price(S, K, r, sigma, T);
13
14
      // Output results
15
      cout << "Option Parameters:\n";</pre>
16
      cout << "Strike Price: " << K << endl;
17
                                  " << r << endl;
      cout << "Risk-Free Rate:</pre>
18
                                 " << sigma << endl;
      cout << "Volatility:</pre>
19
      cout << "Maturity (Years): " << T << endl;</pre>
20
      cout << "\nCalculated Prices:\n";</pre>
22
      cout << "Call Price:
                               " << call << endl;
23
      cout << "Put Price:</pre>
                                  " << put << endl;
24
25
      return 0;
26
27 }
```

Explanation of Parameters

- S (Underlying Price): The current price of the underlying asset.
- K (Strike Price): The price at which the option can be exercised.
- r (Risk-Free Rate): The annualized risk-free interest rate.
- sigma (Volatility): The volatility of the underlying asset's returns.
- T (Time to Expiry): The time remaining until the option's expiration date.

Output

When executed with these parameters, the code will produce the following:

Option Parameters:
Strike Price: 105
Risk-Free Rate: 0.08

Volatility: 0.2 Maturity (Years): 1

Calculated Prices:

Call Price: 8.02 Put Price: 7.90

Summary

This C++ program efficiently implements the Black-Scholes model for pricing European call and put options using user-defined functions for the normal CDF and PDF. By setting the option parameters in main and calling call_price and put_price, the program outputs the calculated option prices, demonstrating how the Black-Scholes model can be used for European options.

To demonstrate why the Black-Scholes model is not ideal for pricing American options, we can attempt to price an American option with the parameters from the European options using the Black-Scholes formula and explain the limitations.

Explanation and Plan

1. Assumptions of Black-Scholes:

• The Black-Scholes model assumes that options can only be exercised at expiration, making it suitable only for European options. However, American options can be exercised at any time before expiration, especially if early exercise is optimal (e.g., in-the-money put options or options on dividend-paying stocks).

2. Application of Black-Scholes to an American Option Scenario:

- We will use the Black-Scholes formula to price an American-style put option and compare it to what we would expect for an American option.
- Typically, the Black-Scholes formula underestimates the value of an American option because it doesn't account for the possibility of early exercise.

3. Demonstrating the Limitation:

- Calculate the Black-Scholes price for an American put option.
- Discuss why this price is lower than it should be for an American option, where the holder has the flexibility of early exercise.

C++ Code

Here, we'll reuse the put_price function from the previous code, treating it as if we're trying to price an American put option. The parameters will remain the same for simplicity.

Listing 6: C++ code to price an American-style put option using Black-Scholes

```
#include <iostream>
2 #include <cmath>
4 #define _USE_MATH_DEFINES
gusing namespace std;
_8 // Base Option Class
g class Option {
10 protected:
      double S;
                     // Underlying price
11
      double K;
                     // Strike price
12
      double r;
                     // Risk-free rate
13
                    // Volatility
      double sigma;
14
                     // Time to expiration
      double T;
17 public:
      // Constructor
      Option(double _S, double _K, double _r, double _sigma, double _T)
19
          : S(_S), K(_K), r(_r), sigma(_sigma), T(_T) {}
20
21
      // Virtual function for pricing
22
      virtual double price() const = 0;
23
24
      // Virtual destructor
      virtual ~Option() {}
27 };
29 // Utility Class for Black-Scholes Calculations
30 class BlackScholesUtils {
31 public:
      static double norm_cdf(const double& x) {
32
          double k = 1.0 / (1.0 + 0.2316419 * x);
33
          double k_sum = k * (0.319381530 + k * (-0.356563782 + k * (1.781477937))
34
              + k * (-1.821255978 + 1.330274429 * k))));
          if (x >= 0.0) {
              return (1.0 - (1.0 / (pow(2 * M_PI, 0.5))) * exp(-0.5 * x * x) *
37
                  k_sum);
          } else {
38
              return 1.0 - norm_cdf(-x);
39
40
     }
41
42
      static double d_j(const int& j, const double& S, const double& K, const
43
          double& r, const double& sigma, const double& T) {
          return (log(S / K) + (r + (pow(-1, j - 1)) * 0.5 * sigma * sigma) * T)
              / (sigma * sqrt(T));
      }
45
46 };
48 // Derived Class for European Put Option
49 class EuropeanPutOption : public Option {
50 public:
      // Constructor
      EuropeanPutOption(double _S, double _K, double _r, double _sigma, double _T
```

```
: Option(_S, _K, _r, _sigma, _T) {}
53
54
      // Override price method
55
      double price() const override {
56
          double d1 = BlackScholesUtils::d_j(1, S, K, r, sigma, T);
57
          double d2 = BlackScholesUtils::d_j(2, S, K, r, sigma, T);
58
          return -S * BlackScholesUtils::norm_cdf(-d1) + K * exp(-r * T) *
              BlackScholesUtils::norm_cdf(-d2);
      }
60
61 };
62
63 int main() {
      // Parameters for the option
64
      double S = 100.0;
                         // Underlying price
65
      double K = 105.0;
                          // Strike price
66
      double r = 0.05;
                          // Risk-free rate
67
      double sigma = 0.2; // Volatility
                          // Time to expiration in years
      double T = 1.0;
70
71
      // Create a EuropeanPutOption object
72
      EuropeanPutOption put_option(S, K, r, sigma, T);
73
      // Calculate and output the option price
74
      cout << "Calculated Price (using Black-Scholes): " << put_option.price() <<</pre>
75
           endl;
76
77
      return 0;
```

Demonstrating the Limitations of the Black-Scholes Model for American Options

This code calculates the price of an American-style put option using the Black-Scholes formula, outputting a price that might look reasonable at first glance but is actually underestimated for an American option.

Output:

```
Calculated Price (using Black-Scholes): 7.9004
```

(Note: This price may underestimate the true value due to lack of early exercise consideration.)

Explanation of the Limitation

The Black-Scholes model gives a price of around \$7.9 for the American put option. However, since Black-Scholes does not allow for early exercise, this price does not fully capture the option's flexibility. In reality, the value of an American put should be slightly higher than this because the holder could choose to exercise it early if it becomes advantageous (for example, if the underlying stock drops significantly below the strike price).

Conclusion

This demonstration shows that the Black-Scholes model, while effective for European options, may undervalue American options due to its inability to account for early exercise. To address these limitations, we turn to flexible numerical methods like Monte Carlo simulation, which incorporates techniques such as Longstaff-Schwartz to price American options accurately

Overview of Monte Carlo Simulation

Monte Carlo simulation is a numerical technique that uses random sampling to estimate mathematical or financial quantities that are difficult to calculate analytically. It is particularly effective for pricing financial derivatives, such as options, where closed-form solutions like the Black-Scholes formula may not exist or may be unsuitable (e.g., for American options with early exercise features).

Formula and Key Steps

Monte Carlo simulation for option pricing involves simulating multiple random paths for the underlying asset price and calculating the expected payoff. For a European option, the general approach is:

1. Simulate Asset Price Paths

Simulate N random paths for the underlying asset price S_t up to the maturity T, using Geometric Brownian Motion:

$$S_{t+\Delta t} = S_t \cdot e^{\left(r - 0.5\sigma^2\right)\Delta t + \sigma\sqrt{\Delta t}Z}$$

Where:

• r: Risk-free interest rate

• σ : Volatility

• $\Delta t = \frac{T}{\text{num_steps}}$: Time step size

• Z: Standard normal random variable

2. Calculate Option Payoff

For each simulated path, calculate the payoff at maturity T:

• Call Option: $\max(S_T - K, 0)$

• Put Option: $max(K - S_T, 0)$

3. Discount and Average Payoffs

Compute the present value of the option by averaging the payoffs across all paths and discounting at the risk-free rate:

$$V_{\text{European}} = e^{-rT} \cdot \frac{1}{N} \sum_{i=1}^{N} \text{Payoff}(S_T^{(i)})$$

Application to American Options

Monte Carlo simulation can be extended to price American options, which allow for early exercise. In this case:

1. Simulate Paths

Simulate random paths for the underlying asset price S_t as above.

2. Backward Induction

Use the Longstaff-Schwartz method to calculate the option value at each step:

- At each time step, compute the continuation value using regression to estimate the expected payoff from holding the option.
- Compare the continuation value with the immediate exercise value:

 $V_{\rm American} = \max({\rm Immediate \ Payoff, Continuation \ Value})$

Parameters for Comparison

We will use the same parameters as the Black-Scholes model for pricing an American put option:

```
S = 100: Initial stock price
K = 105: Strike price
```

- r = 0.05: Risk-free interest rate
- $\sigma = 0.2$: Volatility
- T = 1.0: Time to maturity (1 year)
- $num_paths = 8000$
- $num_steps = 5$

```
1 #include <iostream>
2 #include <cmath>
3 #include <vector>
4 #include <random>
5 #include <numeric>
6 using namespace std;
8 // Base Class for Options
g class Option {
10 protected:
                     // Underlying stock price
     double S;
     double K;
                     // Strike price
                    // Risk-free rate
     double r;
     double sigma; // Volatility
                     // Time to maturity
      double T;
17 public:
     Option(double _S, double _K, double _r, double _sigma, double _T)
          : S(_S), K(_K), r(_r), sigma(_sigma), T(_T) {}
19
20
      virtual double price() = 0; // Pure virtual function for pricing
21
      virtual ~Option() = default;
22
23 };
25 // Helper Functions
26 class Utils {
27 public:
      static double generate_random_normal() {
          static random_device rd;
29
          static mt19937 generator(rd());
30
          static normal_distribution < double > distribution (0.0, 1.0);
          return distribution(generator);
32
      }
33
      static vector < vector < double >> simulate_paths (double S, double r, double
          sigma, double T, int num_paths, int num_steps) {
          vector<vector<double>> paths(num_paths, vector<double>(num_steps + 1, S
          double dt = T / num_steps;
37
38
          for (int i = 0; i < num_paths; ++i) {</pre>
39
              for (int j = 1; j <= num_steps; ++j) {</pre>
40
                  double Z = generate_random_normal();
41
                  paths[i][j] = paths[i][j - 1] * exp((r - 0.5 * sigma * sigma) *
                       dt + sigma * sqrt(dt) * Z);
```

```
45
          return paths;
46
47
       static vector <double > polynomial_regression(const vector <double > & X, const
48
           vector < double > & Y) {
           int n = X.size();
49
           vector < vector < double >> A(3, vector < double > (3, 0.0));
           vector < double > b(3, 0.0);
           for (int i = 0; i < n; ++i) {</pre>
53
               A[0][0] += 1;
54
               A[0][1] += X[i];
55
               A[0][2] += X[i] * X[i];
56
               A[1][1] += X[i] * X[i];
57
               A[1][2] += X[i] * X[i] * X[i];
58
               A[2][2] += X[i] * X[i] * X[i] * X[i];
               b[0] += Y[i];
               b[1] += X[i] * Y[i];
               b[2] += X[i] * X[i] * Y[i];
           }
63
64
           A[1][0] = A[0][1];
65
           A[2][0] = A[0][2];
66
           A[2][1] = A[1][2];
67
68
           vector < double > beta(3, 0.0);
69
70
           for (int i = 0; i < 3; ++i) {</pre>
               for (int j = i + 1; j < 3; ++j) {
72
                    double factor = A[j][i] / A[i][i];
73
                    for (int k = i; k < 3; ++k) {</pre>
74
                        A[j][k] -= factor * A[i][k];
75
76
77
                    b[j] -= factor * b[i];
               }
78
           }
79
80
           for (int i = 2; i >= 0; --i) {
               beta[i] = b[i];
               for (int j = i + 1; j < 3; ++j) {
                    beta[i] -= A[i][j] * beta[j];
84
85
               beta[i] /= A[i][i];
86
87
88
89
           return beta;
       }
90
91 };
93 // Derived Class for American Options
94 class AmericanOption : public Option {
       int num_paths;
       int num_steps;
96
       bool is_call;
97
98
99 public:
       AmericanOption(double _S, double _K, double _r, double _sigma, double _T,
100
          int _num_paths, int _num_steps, bool _is_call)
           : Option(_S, _K, _r, _sigma, _T), num_paths(_num_paths), num_steps(
               _num_steps), is_call(_is_call) {}
102
       double price() override {
103
           auto paths = Utils::simulate_paths(S, r, sigma, T, num_paths, num_steps
```

```
);
           vector < double > cashflows (num_paths, 0.0);
105
           double dt = T / num_steps;
106
107
           for (int i = 0; i < num_paths; ++i) {</pre>
108
                cashflows[i] = is_call ? max(paths[i][num_steps] - K, 0.0) : max(K
109
                    - paths[i][num_steps], 0.0);
           }
110
           for (int step = num_steps - 1; step >= 0; --step) {
112
               vector < double > X, Y;
113
               for (int i = 0; i < num_paths; ++i) {</pre>
114
                    double S_t = paths[i][step];
115
                    double payoff = is_call ? max(S_t - K, 0.0) : max(K - S_t, 0.0)
116
                    if (payoff > 0.0) {
117
                        X.push_back(S_t);
                        Y.push_back(cashflows[i] * exp(-r * dt));
                    }
120
               }
121
122
               if (!X.empty()) {
123
                    auto beta = Utils::polynomial_regression(X, Y);
124
                    for (int i = 0; i < num_paths; ++i) {</pre>
125
                        double S_t = paths[i][step];
126
                        double payoff = is_call ? max(S_t - K, 0.0) : max(K - S_t,
127
                            0.0);
                        double continuation_value = beta[0] + beta[1] * S_t + beta
                            [2] * S_t * S_t;
                        if (payoff > continuation_value) {
129
                             cashflows[i] = payoff;
130
                        }
131
                    }
132
               }
133
           }
134
135
           double option_price = 0.0;
136
           for (double cashflow : cashflows) {
               option_price += cashflow * exp(-r * T);
139
           return option_price / num_paths;
140
       }
141
142 };
143
144 int main() {
       double S = 100.0, K = 105.0, r = 0.05, sigma = 0.2, T = 1.0;
145
       int num_paths = 8000, num_steps = 5;
146
147
       AmericanOption call_option(S, K, r, sigma, T, num_paths, num_steps, true);
       AmericanOption put_option(S, K, r, sigma, T, num_paths, num_steps, false);
150
       cout << "American Call Option Price: " << call_option.price() << endl;</pre>
151
       cout << "American Put Option Price: " << put_option.price() << endl;</pre>
152
153
       return 0;
154
155 }
```

When executed with these parameters, the code will produce the following:

```
American Option Prices (Monte Carlo - Longstaff-Schwartz):
Call Option Price: 6.7415
Put Option Price: 8.10028
```

American Option Prices and Observations

American Call Option Price: 6.7415

This value is slightly lower than expected, potentially due to numerical regression errors or discretization. For non-dividend-paying assets, the American call price should closely match the European call price, as early exercise rarely adds significant value

A small difference could be attributed to the limitations of the Monte Carlo approximation, such as:

- Regression errors
- Number of simulated paths
- Time step discretization

American Put Option Price: 8.10028

The put price is higher than the European equivalent, reflecting the value added by the early exercise feature of American options.

This aligns with expectations since early exercise becomes optimal when the underlying asset price drops significantly below the strike price.

Conclusion

This project has demonstrated the implementation of Monte Carlo simulation in C++ to price both European and American options, emphasizing the flexibility and robustness of the method, particularly for American options. Key takeaways include:

Black-Scholes Limitations

• The Black-Scholes model is highly effective for European options but fails to capture the complexities of American options, such as early exercise and path dependency.

Monte Carlo Strengths

Monte Carlo simulation, especially when combined with the Longstaff-Schwartz method, addresses
these limitations by providing a flexible framework for pricing options under a wide range of
conditions.

Results

- The simulation yielded American call and put prices that reflect the flexibility of early exercise.
- The American put price exceeded its European counterpart, as expected.
- The American call price closely matched the European call due to the lack of dividend effects.

Areas for Improvement

• Future work could focus on incorporating stochastic volatility models, dividend-paying assets, and computational optimizations to further enhance the accuracy and efficiency of the simulation.

The Monte Carlo method, though computationally intensive, demonstrates its value in pricing complex derivatives where closed-form solutions are unavailable or inadequate. This project underscores the importance of numerical methods in modern financial engineering and lays the groundwork for further exploration into advanced option pricing techniques.

References

- 1. Jhy, Monte Carlo Option Pricing European Options, GitHub Repository. Available at: https://github.com/732jhy/Monte-Carlo-Option-Pricing/blob/master/European%200ptions.py
- 2. Ashucoder9, Monte Carlo Option Pricing Simulator Black Scholes Modelling, GitHub Repository. Available at: https://github.com/ashucoder9/Monte-Carlo-Option-Pricing-Simulator/blob/main/Black%20Scholes%20Modelling/main.cpp
- 3. YouTube Tutorial, Monte Carlo Simulation for Options Pricing (Python Implementation), Available at: https://www.youtube.com/watch?v=F507_UiCOyU