

# Monte Carlo Valuation of European Call Options

Technical Documentation Mathematical Model

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## 1 Introduction

### 1.1 Project Objective

The primary objective of this project is to estimate the fair value of a **European Call Option** on Apple Inc. (AAPL) utilizing Monte Carlo simulation methods.

### 1.2 Option Definitions

- **European Call Option:** Grants the holder the right, without obligation, to purchase the underlying asset at a fixed Strike Price ( $K$ ) only on the expiration date ( $T$ ).
- **American Call Option:** Differing from the European style, this option allows for exercise at any point up to maturity, offering distinct advantages for dividend capture strategies.

## 2 Methodology: Monte Carlo Simulation

The Monte Carlo simulation is a computational algorithm that utilizes random sampling to model stochastic processes. It quantifies uncertainty by generating a distribution of possible outcomes to derive a probabilistic fair value.

### 2.1 Key Advantages

Monte Carlo methods are preferred over analytical models (such as Black-Scholes) in three specific scenarios:

#### 1. Path Dependency

Analytical models often look only at the starting and end points. Monte Carlo simulates the entire price trajectory, making it essential for instruments like *Asian Options*, where the payoff depends on the average price over time.

#### 2. Complex Payoff Rules

It adeptly handles exotic options with conditional execution, such as *Barrier Options* (active only if price hits  $X$  or stays below  $Y$ ). Deriving closed-form mathematical formulas for these conditional constraints is often impossible.

#### 3. High Dimensionality

When valuing *Basket Options* (a derivative based on a portfolio of assets like AAPL, GOOGL, and MSFT), the mathematics involves solving differential equations for multiple variables. Monte Carlo simulations efficiently manage these multi-variable environments without exponential complexity.

## 3 The Payoff Function

The value of a call option at maturity ( $T$ ) depends on where the stock price ( $S_T$ ) finishes relative to the strike price ( $K$ ). The payoff formula is:

$$C_T = \max(S_T - K, 0) \tag{1}$$

- If  $S_T > K$ : The option is *In-The-Money* (Profit =  $S_T - K$ ).
- If  $S_T \leq K$ : The option is *Out-Of-The-Money* (Worthless, Payoff = 0).

## 4 Stochastic Process: Geometric Brownian Motion

We assume the stock price follows a **Geometric Brownian Motion (GBM)**. This is the standard model used in the Black-Scholes framework. The continuous-time Stochastic Differential Equation (SDE) is:

$$dS_t = \mu S_t dt + \sigma S_t dW_t \quad (2)$$

Where:

- $\mu$  is the drift (expected return, replaced by risk-free rate  $r$  for valuation).
- $\sigma$  is the volatility of the stock.
- $W_t$  is a Wiener process (Brownian motion).

## 5 Monte Carlo Discretization

To simulate this on a computer, we apply **Itô's Lemma** to solve the SDE and discretize it over small time steps ( $\Delta t$ ). The formula used in the Python script is:

$$S_{t+\Delta t} = S_t \cdot \exp \left( \left( r - \frac{1}{2}\sigma^2 \right) \Delta t + \sigma \sqrt{\Delta t} Z \right) \quad (3)$$

Where  $Z$  is a random draw from the standard normal distribution:  $Z \sim \mathcal{N}(0, 1)$ .

## 6 Valuation Methodology

The simulation performs the following steps 10,000 times:

1. Generate a random price path from  $t = 0$  to  $t = T$ .
2. Calculate the price  $S_T$  at maturity.
3. Calculate the payoff  $\max(S_T - K, 0)$ .

The fair option price is the average of these payoffs, discounted back to the present value:

$$V_0 = e^{-rT} \cdot \frac{1}{N} \sum_{i=1}^N \text{Payoff}_i \quad (4)$$

