



Student Research Group 'Stochastic Volatility Models'

# **Methods of Simulation of the Heston Model: A Review**

Artemy Sazonov, Danil Legenky, Kirill Korban

Lomonosov Moscow State University, Faculty of Mechanics and Mathematics

November 5, 2022



## Heston Model Definition

Assume that the spot asset at time  $t$  follows the diffusion

$$dS(t) = \mu S(t)dt + \sqrt{v(t)}S(t)dZ_1(t), \quad (1)$$

$$dv(t) = \left( \delta^2 - 2\beta v(t) \right) dt + 2\delta \sqrt{v(t)}dZ_2(t), \quad (2)$$

where  $Z_1, Z_2$  are the correlated Wiener processes with  $dZ_1 dZ_2 = \rho dt$ .



# Table of Contents

- Introduction to Monte-Carlo Methods
  - Statistical Estimation
  - Random and Pseudo-Random Numbers
  - A Flashback from the Probability course
- Euler Simulation Method
- Broadie-Kaya Simulation Method
- Andersen Simulation Method
- Computation Examples
- Greeks Computation
- Conclusion



## Statistical Estimation

Let  $X, X_1, X_2, \dots, X_n$  be a series of independent and identically distributed random variables with a distribution function  $F(x)$ . Unbiased consistent estimators of  $\mathbb{E}[X]$  and  $\text{var}[X]$  are:

- The sample mean is the average of the sample values:

$$\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i \quad (3)$$

- The sample variance is the average squared deviation from the mean:

$$S^2 = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2 \quad (4)$$

# Statistical Estimation



## Lemma 1

*Let  $X_1, X_2, \dots, X_n$  be a series of independent and identically distributed random variables, and  $h : \mathbb{R} \rightarrow \mathbb{R}$  be a borel function. Then  $h(X_1), h(X_2), \dots, h(X_n)$  is a series of independent and identically distributed random variables.*

Thus, we could write the unbiased consistent estimator of  $\mathbb{E} [h(X)]$  as follows:

$$\widehat{\mathbb{E} [h(X)]} = \frac{1}{n} \sum_{i=1}^n h(X_i). \quad (5)$$



# Random and Pseudo-Random Numbers

## Definition 2

A random sequence of numbers is a sequence (in time or in space) with no discernable pattern.

## Definition 3

Pseudorandom number generator is an algorithm that generates a sequence of numbers that has no internal pattern. However, the series requires a starting seed, and if the algorithm is started repeatedly with the same seed, it will go through precisely the same sequence of numbers.

Check out the lecture by A.N.Shiryaev (Probability Theory Department x Vega Institute Foundation Seminar, Oct 19) for more details.



## Law of Large Numbers

### Theorem 4 (Khinchin)

Let  $X_1, X_2, \dots, X_n$  be a sequence of independent and identically distributed random variables with a distribution function  $F(x)$  and  $\mathbb{E}X_i = \mu$ . Then

$$\mathbb{P}\text{-}\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=1}^n X_i = \mu. \quad (6)$$

### Theorem 5 (Kolmogorov)

Let  $X_1, X_2, \dots, X_n$  be a sequence of independent and identically distributed random variables with a distribution function  $F(x)$  and  $\mathbb{E}X_i = \mu$ . Then

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=1}^n X_i \stackrel{a.s.}{=} \mu. \quad (7)$$



# Central Limit Theorem

## Theorem 6 (Lindeberg-Lévy)

*Let  $X_1, \dots, X_n$  be a sequence of i.i.d. random variables with  $\mathbb{E}[X_i] = \mu$  and  $\text{var}[X_i] = \sigma^2$ . Then as  $n$  approaches infinity, the random variables  $\sqrt{n}(\bar{X}_n - \mu)$  converge in law to a normal distribution  $\mathcal{N}(0, \sigma^2)$ , i.e.*

$$\sqrt{n}(\bar{X}_n - \mu) \xrightarrow{d} \mathcal{N}(0, \sigma^2). \quad (8)$$





# Table of Contents

Introduction to Monte-Carlo Methods

Statistical Estimation

Random and Pseudo-Random Numbers

A Flashback from the Probability course

Euler Simulation Method

Broadie-Kaya Simulation Method

Andersen Simulation Method

Computation Examples

Greeks Computation

Conclusion





# Table of Contents

Introduction to Monte-Carlo Methods

Statistical Estimation

Random and Pseudo-Random Numbers

A Flashback from the Probability course

Euler Simulation Method

**Broadie-Kaya Simulation Method**

Andersen Simulation Method

Computation Examples

Greeks Computation

Conclusion





# Table of Contents

Introduction to Monte-Carlo Methods	
Statistical Estimation	
Random and Pseudo-Random Numbers	
A Flashback from the Probability course	
Euler Simulation Method	
Broadie-Kaya Simulation Method	
Andersen Simulation Method	
Computation Examples	
Greeks Computation	
Conclusion	





# Table of Contents

Introduction to Monte-Carlo Methods

Statistical Estimation

Random and Pseudo-Random Numbers

A Flashback from the Probability course

Euler Simulation Method

Broadie-Kaya Simulation Method

Andersen Simulation Method

Computation Examples

Greeks Computation

Conclusion







# Table of Contents

Introduction to Monte-Carlo Methods

Statistical Estimation

Random and Pseudo-Random Numbers

A Flashback from the Probability course

Euler Simulation Method

Broadie-Kaya Simulation Method

Andersen Simulation Method

Computation Examples

Greeks Computation

Conclusion





# Table of Contents

Introduction to Monte-Carlo Methods

Statistical Estimation

Random and Pseudo-Random Numbers

A Flashback from the Probability course

Euler Simulation Method

Broadie-Kaya Simulation Method

Andersen Simulation Method

Computation Examples

Greeks Computation

Conclusion



## Conclusion

We introduced the three most common simulation methods for dynamics of the Heston stochastic volatility model:

1. Euler scheme;
2. Broadie-Kaya scheme;
3. Andersen scheme.

