

# A Comprehensive Study of Stock Market Volatility: Types, Determinants, and Measurement Methods

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## Abstract

The Volatility, a critical aspect of financial markets, quantifies the uncertainty associated with asset price movements. This paper provides a comprehensive review of volatility, encompassing its definition, types, determinants, and measurement methods. Various factors, including macroeconomic indicators, market sentiment, liquidity, global events, and corporate actions, influence volatility. A range of methods, such as standard deviation, historical volatility, implied volatility, GARCH models, EWMA, and range-based volatility, are employed to measure volatility. Understanding volatility is essential for investors, traders, and risk managers to make informed decisions and manage risk effectively. Future research can delve into advanced volatility models, volatility spillovers, investor behavior, and the relationship between volatility and asset pricing.

**Keywords:** Stock Market, The Volatility, GARCH

## INTRODUCTION

The volatility of stock returns is a critical aspect of financial markets that has garnered significant attention from researchers and practitioners alike (Orabi & Alqurran, 2015). The Volatility, defined as the variance of stock returns, serves as a measure of the uncertainty associated with the returns and is essential for portfolio management, risk assessment, and derivative pricing. Recent studies have highlighted the importance of understanding the dynamics of volatility in emerging stock markets, which often exhibit unique characteristics compared to their developed counterparts (D. D. Bhowmik, n.d.). The stock market plays a pivotal role in today's world economic activities, named a "barometer" and "alarm" for economic and financial activities in a country or region. (Nwosa, 2011).

The Volatility is the tendency for prices to change unexpectedly, however, all kinds of volatility is not bad. At the same time, financial market volatility has also a direct impact on macroeconomic and financial stability. Important economic risk factors are generally highly valued by governments around the world. Therefore, research on the volatility of financial markets has always been the focus of financial economists and financial practitioners. Nowadays, a large part of the literature has studied some characteristics of the stock market, such as the leverage effect of volatility, the short-term memory of volatility, and the GARCH effect, etc., but some researchers show that when adopting short-term memory by the GARCH model, there is usually a confusing phenomenon, as the sampling interval tends to zero.

The characterization of the tail of the yield generally assumes an ideal situation, that is, obeys the normal distribution, but this perfect situation is usually not established. (R. Bhowmik & Wang, 2020).

## LITERATURE REVIEW

**(Engle et al., 2013)** In the study of stock market volatility, the work of Engle et al. Significantly contributes to understanding the interplay between macroeconomic fundamentals and financial market fluctuations. Their research highlights the critical role that macroeconomic variables play in influencing stock market dynamics, particularly through the lens of models like GARCH (Generalized Autoregressive Conditional Heteroskedasticity). They identify key macroeconomic indicators—such as interest rates, inflation, and GDP growth—that serve as determinants of volatility. Engle et al.'s work thus serves as a foundational reference in the field, illustrating the importance of macroeconomic fundamentals in stock market volatility analysis. Their contributions underline the necessity of adopting multifaceted models that account for external economic influences when examining volatility, setting the stage for future research and methodology developments. market sentiment and investor behavior, ultimately impacting stock prices.

**(Orabi & Alqurran, 2015)** Orabi and Alqurran investigate how changes in volatility impact the performance of the Jordanian stock market, offering a comprehensive analysis of the relationship between volatility and market returns. Their findings indicate that increased volatility often leads to heightened uncertainty among investors, which can result in significant fluctuations in stock prices. This relationship underscores the broader implications of volatility in emerging markets, where investor sentiment is particularly sensitive to market changes. The authors employ advanced econometric techniques, including GARCH models, to analyze volatility changes effectively. This methodological approach is critical for understanding the intricacies of volatility in a market characterized by external shocks and limited liquidity. The study provides essential implications for both investors and policymakers in Jordan and similar emerging markets. By elucidating the effects of volatility on market behaviour, the research offers insights that can inform investment strategies and regulatory policies. In summary, the research by Orabi and Alqurran enriches the discourse on stock market volatility by examining its specific effects in the context of an emerging market like Jordan. Incorporating their findings will enhance your comprehensive analysis, emphasizing the multifaceted nature of volatility and its implications in financial markets. In the exploration of stock market volatility, various models and frameworks have been developed to enhance forecasting accuracy and understand the underlying determinants. The GARCH-MIDAS (Generalized Autoregressive Conditional Heteroskedasticity – Mixed Data Sampling) approach is one such advanced methodology that effectively captures both long-term and short-term volatility dynamics by integrating mixed-frequency data.

**(Rjumohan Asalatha, 2019)** emphasizes that stock markets are integral to capital allocation, with their impacts differing across countries based on unique structural and governance attributes. The historical evolution of stock markets, from informal trading practices to formal exchanges, has been documented, showcasing significant milestones such as the establishment of the Bombay Stock Exchange as Asia's first stock exchange. Advances in technology have also transformed market operations, enabling electronic trading and enhancing accessibility for individual investors, which has been critical for market efficiency and globalization.

**(Hewamana et al., 2022)** delve into the behavioral aspects that affect stock price volatility, categorizing determinants into macroeconomic, company-specific fundamentals, and investor psychology. Their

findings reveal that emotional biases and cognitive errors can lead to irrational trading behaviors, contributing to market fluctuations and inefficiencies.

**(Kariuki, 2022)** explores the influence of external factors such as oil prices, government expenditure, and exchange rates on stock market return volatility. This critical literature review emphasizes how global economic conditions can significantly affect local stock exchanges, thereby influencing investor sentiment and trading patterns.

**(Ganie et al., 2022)** examines the impact of the COVID-19 outbreak on stock markets across select economies, highlighting significant volatility and shifts in investor behavior as a result of the pandemic. The authors synthesize findings from various studies that explore similar themes, emphasizing that the abrupt nature of the crisis led to heightened uncertainty and rapid declines in stock prices globally.

**(Akash Chaurasia et al., 2023)** examine the macroeconomic determinants of stock market development, identifying factors such as interest rates, inflation, and GDP growth as critical influencers of market stability and investor confidence. Their systematic review underscores the interconnectedness of economic indicators and stock market performance, suggesting that macroeconomic stability is essential for fostering a healthy investment environment.

**(Safiq & Yulianti, 2023)** The research emphasizes that dividend payout ratios can significantly influence stock price volatility. This aligns with the broader literature, which suggests that higher dividend payouts often correlate with reduced volatility, as they signal stability and financial health to investors. Safiq and Yulianti's findings support this notion, indicating that a higher Dividend Payout Ratio negatively impacts stock price volatility, suggesting a stabilizing effect on stock prices during the Omnibus Law era. The study identifies several determinants of dividend policies, including ownership structure (managerial and institutional), profitability measures (ROA and ROI), and regulatory influences (specifically the Omnibus Law). The paper also touches on how investor perceptions are shaped by dividend policies, suggesting that favorable policies can lead to increased investment interest and potentially lower volatility.

**(El Rifai et al., 2023)** study, Volatility Spill overs Among Major Tourism Stock Indices During the COVID-19 Pandemic provides critical insights into the interconnectedness of tourism stock markets during a period of unprecedented global disruption. The research emphasizes the unique challenges faced by the tourism sector, highlighting how the pandemic led to significant volatility spill overs among major tourism stock indices. The authors employ advanced econometric models to analyze the volatility transmission across different markets, revealing that shocks in one market can have pronounced effects on others. This finding aligns with previous research on volatility spill over effects, such as that of Wang and Liu, who investigated cross-market spill overs in various sectors during economic downturns. They extend this understanding by focusing specifically on the tourism industry, which is particularly sensitive to external shocks. Furthermore, the study underscores the importance of understanding these spill over dynamics for investors and policymakers. By identifying the channels through which volatility spreads, stakeholders can better manage risks associated with investments in tourism stocks. Additionally, this paper discusses the implications of their findings for portfolio diversification, suggesting that investors should consider the interdependence of tourism stocks when making investment decisions. This insight is crucial given the pandemic's impact on global markets, as highlighted in related literature that explores the benefits of diversification in volatile environments. Overall, this study contributes significantly to the understanding of volatility in the tourism sector, particularly during crisis periods. By integrating empirical analysis with theoretical frameworks, the paper offers valuable lessons for managing investments in a highly interconnected and volatile market environment.

(D. D. Bhowmik, n.d.) study, “Stock Market Volatility: An Evaluation,” provides a thorough examination of the multifaceted nature of stock market volatility, emphasizing its significance in financial analysis and investment strategies. The paper critically assesses various determinants of volatility, categorizing them into macroeconomic factors, market microstructure, and behavioural influences. Bhowmik highlights the role of macroeconomic variables such as interest rates, inflation, and economic growth in shaping volatility patterns. This aligns with the findings of Engle et al., who demonstrate that these fundamental indicators are crucial for understanding and forecasting market fluctuations. Their work underscores the necessity of incorporating macroeconomic fundamentals into volatility models to enhance predictive accuracy. Furthermore, Bhowmik delves into the psychological aspects of market behavior, exploring how investor sentiment and behavioral biases contribute to volatility. This perspective is supported by studies such as those by Akin and Akin, which investigate the effects of behavioral finance on market anomalies and price movements. Bhowmik’s integration of these psychological factors enriches the analysis of volatility, suggesting that understanding investor behavior is essential for a comprehensive evaluation.

(Market Volatility - Robert J. Shiller - Google Books, n.d.) Stock market volatility has been a pivotal area of research in financial economics, with significant contributions from various scholars. Robert J. Shiller, in his seminal work, highlighted the inherent unpredictability of stock prices and the implications of behavioral finance on market volatility. His analysis laid the groundwork for understanding how psychological factors can drive investor behavior and, consequently, contribute to price fluctuations. Recent studies have expanded on Shiller’s insights by identifying macroeconomic determinants of stock market volatility. The development of measurement methods has also evolved, particularly with the introduction of advanced econometric models. The GARCH (Generalized Autoregressive Conditional Heteroskedasticity) family of models has been widely adopted to capture the time-varying nature of volatility. In summary, the literature illustrates that stock market volatility is influenced by a complex interplay of psychological factors, macroeconomic conditions, and external shocks, with various methodologies developed to measure and analyze these dynamics effectively.

(Ajmal et al., 2023) Crude oil price volatility is a critical area of research due to its profound impact on global economies. They conducted a comprehensive analysis of factors influencing crude oil front volatility across major economies, including the USA, China, Spain, Germany, and India. Their study identifies several key determinants, including macroeconomic indicators, geopolitical events, and market sentiment, which collectively shape the dynamics of oil prices. Moreover, the authors underscore the importance of geopolitical risks, noting that conflicts and policy changes in oil-producing regions can lead to abrupt price fluctuations. In summary, this study contributes to the growing body of literature by offering a multi-faceted view of crude oil volatility, emphasizing the need for a holistic approach that considers both macroeconomic fundamentals and external factors to better understand price dynamics in the global oil market.

Study by (Pandey et al., 2024) offers crucial insights into the determinants of stock price volatility in a specific market context. They analyze various factors that impact stock price volatility in Nepalese commercial banks, focusing on financial metrics such as Earnings Per Share (EPS), Dividends Per Share (DPS), and the Price-Earnings Ratio (PER). Their findings illustrate how these financial indicators significantly influence market fluctuations, emphasizing the importance of firm-specific characteristics in determining stock volatility. The authors utilize bivariate correlation and regression models to assess the relationships between these financial metrics and stock price volatility. This methodological

framework provides a robust analytical foundation for understanding how specific factors contribute to volatility, making it a valuable reference for your analysis of measurement methods in stock market volatility. Their approach underscores the necessity of employing appropriate statistical tools to capture the dynamics of market behavior effectively. The study also highlights the implications of these findings for investors and policymakers. By identifying key factors that drive volatility, the research provides actionable insights that can inform investment strategies and regulatory frameworks.

**(Alajlani et al., 2024)** examines the significant impacts of the COVID-19 pandemic on stock markets globally, highlighting increased volatility and shifts in investor behavior across various regions. The review synthesizes findings from multiple studies, including those that analyze the pandemic's effects on Central and Eastern European stock markets and the responses of US and European markets to exogenous shocks. Additionally, the review discusses the role of government interventions in stabilizing markets during the crisis, emphasizing the need for a nuanced understanding of how such global events reshape financial landscapes. The insights provided can inform both academic research and practical investment strategies in tumultuous times.

**(Nisha Sharma & Dr. Priti Dubey, 2024)** explore the herding behavior of investors, revealing that demographic profiles and the financial information received significantly impact investment decisions, often leading to irrational market movements.

**(Wu et al., 2024)** The volatility of Bitcoin has emerged as a critical area of study, particularly as it relates to various external economic indicators. Wu and Hossain (2024) employed Ordinary Least Squares (OLS) regression to analyze Bitcoin price fluctuations from January 2014 to March 2023. Their research identified key factors influencing Bitcoin volatility, including economic policy uncertainty (EPU), oil prices, the NASDAQ index, and gold prices. The study concluded that EPU and the NASDAQ index positively impact Bitcoin volatility, whereas gold prices tend to stabilize it. Notably, the authors found no significant relationship between energy prices and Bitcoin volatility. As highlighted by Wu and Hossain, the implications of these findings extend to both investors and regulatory bodies, suggesting that careful consideration of the volatility factors is essential for informed decision-making in the cryptocurrency market.

## OBJECTIVES

The objectives of this research paper are to define stock market volatility, categorize its types, review measurement methods such as standard deviation, historical volatility, implied volatility, GARCH models, EWMA, and investigate the implications of volatility for portfolio management and risk assessment, particularly in emerging markets.

## RESEARCH METHODOLOGY

This research paper is solely based on secondary data, which was gathered from research papers and other secondary sources such as books, websites etc.

## LIMITATIONS

This comprehensive analysis of stock market volatility provides valuable insights, but it's important to acknowledge certain limitations. The accuracy and reliability of the findings depend heavily on the availability and quality of data. Limited or incomplete data can hinder the analysis of specific time periods or factors. Additionally, the models used to measure and analyze volatility, such as GARCH and EWMA,

have their own limitations and underlying assumptions. These models might not capture the full complexity of volatility dynamics, especially during extreme events or structural shifts. Furthermore, the study primarily focuses on stock market volatility, while volatility is also observed in other financial markets like commodities, bonds, and foreign exchange.

## **STOCK MARKET**

The stock market is a pivotal component of the global financial system, serving as a platform for the buying and selling of shares in publicly traded companies. It plays a crucial role in facilitating capital formation, providing businesses with necessary funds for expansion and innovation while offering investors opportunities for wealth accumulation. Investors participate in the stock market with the expectation of achieving financial returns, influenced by the performance of the companies in which they invest. In recent years, the dynamics of the stock market have evolved significantly, influenced by technological advancements, changes in regulatory frameworks, and shifts in investor behavior. The rise of online trading platforms has democratized access to the stock market, allowing a broader demographic to participate. However, this increased participation has also introduced challenges, as many investors lack sufficient understanding of market operations and the inherent risks involved. The stock market is characterized by several key features that influence its functioning and dynamics. One prominent characteristic is market volatility, which refers to the degree of variation in stock prices over time. This volatility can be driven by various factors, including macroeconomic indicators, investor sentiment, and external shocks, such as pandemics or geopolitical events. For instance, research by (Sahil Narang et al., 2023) demonstrates how COVID-19-induced fear and firm-specific characteristics affected stock market responses during the pandemic, highlighting the interplay between external events and market behavior. Additionally, the stock market is influenced by the dynamic characteristics of trading behaviors and investor decisions, which can be analyzed through big data techniques. (Yang & Hou, 2022) explore the evolving nature of stock market efficiency, emphasizing the need for continuous monitoring of market dynamics to understand its complexities. Together, these characteristics underscore the multifaceted nature of stock markets and the necessity for investors to remain informed about the factors that drive market movements.

## **THE VOLATILITY**

Volatility can be defined in several ways, often depending on the context in which it is used. According to (Hull, J. C., 2017) Volatility refers to the degree of variation of a trading price series over time, usually measured by the standard deviation of returns. It quantifies how much the price of a financial asset fluctuates, providing insight into its risk profile. Typically measured by the standard deviation of returns, volatility indicates the extent to which an asset's price deviates from its average over a specific period. High volatility signifies significant price swings, indicating greater uncertainty and risk, while low volatility suggests more stable prices and lower risk. This variability can arise from various factors, including market sentiment, economic news, and geopolitical events. Understanding general volatility is essential for investors, as it helps them gauge potential price movements and align their investment strategies with their risk tolerance. In practical terms, volatility can impact trading strategies, portfolio management, and risk assessment, influencing decisions such as the timing of trades and the choice of asset allocation. By analyzing historical volatility, investors can make more informed predictions about future price behavior, ultimately aiding in better investment outcomes.

In finance, volatility is a statistical measure of the dispersion of returns for a given security or market index, indicating the degree of risk associated with the asset's price movements (Black, F., & Scholes, M., 1973). It quantifies how much the returns on an asset deviate from its expected mean over a specific period. This measure is crucial in assessing the risk associated with an investment, as higher volatility indicates a greater range of potential outcomes and, therefore, a higher level of risk. Volatility is often expressed in terms of standard deviation, which provides a numerical value representing the extent of variation in returns (R. Bhowmik & Wang, 2020).

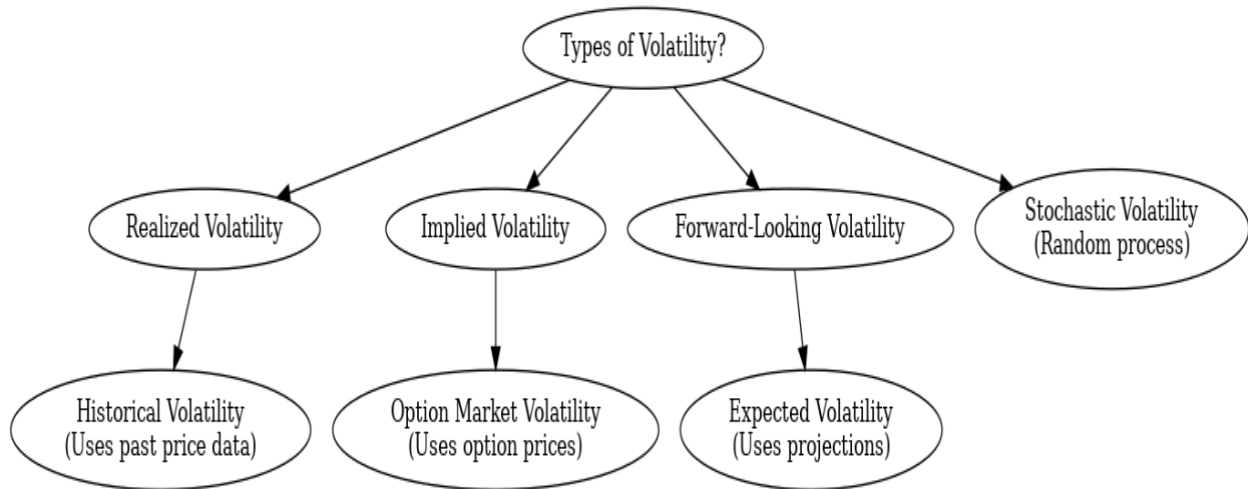
In finance, understanding volatility is essential for various applications, including portfolio management, option pricing, and risk assessment. Tools like the VIX index, often referred to as the "fear gauge," measure market expectations of future volatility, allowing investors to gauge market sentiment. Recent research has explored different aspects of volatility, such as its asymmetry in financial time series and the application of advanced models like **GARCH (Generalized Autoregressive Conditional Heteroskedasticity)** to better capture the dynamic of volatility (Marisetty, 2024). Understanding these complexities enhances investors' ability to manage risk and make informed decisions in volatile environments.

In statistics, volatility can refer to the variability of a dataset, particularly in terms of the standard deviation or variance, showcasing how much the data points deviate from the mean (Marisetty, 2024). The statistical definition of volatility refers to the variability or dispersion of a dataset, particularly in the context of financial time series. It is commonly measured using standard deviation or variance, which quantify how much individual data points deviate from the mean. In finance, this statistical approach helps investors and analysts understand the degree of uncertainty associated with an asset's returns. High volatility indicates that the returns of an asset can vary widely, reflecting increased risk and potential for significant price swings. Conversely, low volatility suggests more stable returns, indicating lower risk. This measure is particularly important in risk management, where understanding the distribution of returns helps in constructing portfolios and making investment decisions.

In recent studies, advanced statistical models like GARCH (Generalized Autoregressive Conditional Heteroskedasticity) have been employed to capture the dynamics of volatility more effectively. These models recognize that volatility is not static and can change over time, often clustering in periods of market stress or high trading activity. For instance, research has shown that stochastic volatility models can effectively analyze financial time series, accounting for nonlinearity and long memory effects. Additionally, circular statistics have emerged as a useful tool for analyzing high-volatility financial markets, particularly in assessing the tail behavior of probability distributions, which is critical for understanding extreme price movements.

Market volatility reflects the frequency and magnitude of price movements in a financial market, often influenced by market sentiment, economic indicators, and geopolitical events. (Malkiel, B. G., 2003).

**TYPES OF VOLATILITY**



**Figure: 1.1**

Volatility can be categorized into several types, each serving different analytical purposes in finance and economics.

- **Historical Volatility:** This measures the past price fluctuations of an asset over a specific period, typically calculated using standard deviation. It provides insight into how much the price of an asset has varied in the past.
- **Implied Volatility:** Derived from market prices of options, it reflects the market’s expectations of future volatility. Higher implied volatility suggests greater uncertainty or risk regarding the future price movements of the underlying asset. During crises, such as the COVID-19 pandemic, implied volatility spiked dramatically across equity markets. The VIX, often referred to as the “fear index,” reached historical highs as investors priced in uncertainty about the economic outlook. Under implied volatility, Option market volatility refers to the degree of fluctuation in the price of underlying assets, such as stocks, commodities, or currencies. It’s a crucial factor in determining the price of options contracts. Option contracts are financial instruments that give the buyer the right, but not the obligation, to buy or sell an underlying asset at a predetermined price (strike price) within a specified period. Two Main Types of Option Contracts:
  1. Call Options: These give the buyer the right to buy the underlying asset at the strike price.
  2. Put Options: These give the buyer the right to sell the underlying asset at the strike price.
- **Future Volatility:** This type is projected based on models and analysis of various factors, including market sentiment, economic indicators, and historical trends. It is often used in risk management and derivatives pricing. Generalized Autoregressive Conditional Heteroskedasticity (GARCH) models are commonly used to forecast future volatility. These models allow for the conditional variance of asset returns to change over time, making them suitable for capturing volatility clustering observed in financial markets.
- **Forward-Looking Volatility:** A general term that encompasses any type of volatility measure that looks ahead. This can include implied volatility, expected volatility, or forecasts based on models. It is used in investment decision-making, risk management, and asset allocation.
- **Realized Volatility:** This is the actual observed volatility over a specific interval, calculated using high-frequency trading data. It is often used to assess the accuracy of volatility forecasts.



- **Conditional Volatility:** This refers to the volatility that is conditional on past information. Models like GARCH (Generalized Autoregressive Conditional Heteroskedasticity) are used to estimate conditional volatility based on past returns.(Nugroho et al., 2024)
- **Stochastic volatility:** Stochastic volatility or Random volatility refers to a model where the volatility of a financial asset is treated as a stochastic process rather than a constant or deterministic function. This approach captures the unpredictable nature of market fluctuations and allows for more accurate modeling of asset prices.

## FACTORS AFFECTING STOCK MARKET VOLATILITY

The factors affecting volatility can be broadly categorized into several areas, often explored in financial literature:

- **Macroeconomic Indicators:** Economic variables such as GDP growth, inflation rates, and unemployment rates significantly impact market volatility. For instance, changes in macroeconomic conditions can lead to fluctuations in investor confidence and market sentiment.(Engle et al., 2013)
- **Market Sentiment:** Investor psychology and market sentiment play a crucial role in volatility. Events that affect investor perceptions, such as geopolitical tensions or major corporate announcements, can lead to increased market fluctuations.
- **Liquidity:** The level of liquidity in a market can influence volatility. In less liquid markets, even small trades can lead to larger price movements, resulting in higher volatility.(Safiq & Yulianti, 2023).
- **Global Events:** Events such as financial crises, natural disasters, or pandemics (e.g., COVID19) can trigger significant changes in market volatility due to their widespread impact on economic conditions. (Wu et al., 2024)
- **Corporate Actions:** Actions such as earnings announcements, dividends, and stock splits can also lead to changes in volatility, as they may significantly affect investors' perceptions of a company's value.
- **Interest Rates:** Changes in interest rates can significantly impact volatility. Higher interest rates often lead to increased borrowing costs, which can affect corporate profits and investor behavior, leading to heightened market fluctuations. (Ajmal et al., 2023)
- **Exchange Rates:** Fluctuations in currency exchange rates can create volatility in the markets, especially for companies with international operations. Changes in exchange rates can affect profits and competitiveness, leading to price swings.(Safiq & Yulianti, 2023)
- **Market Structure:** The characteristics of the market, such as the number of participants, the nature of trading (institutional vs. Retail), and the presence of derivatives, can influence volatility levels. More complex market structures may lead to greater volatility due to varied trading strategies.
- **Regulatory Changes:** New regulations or changes in existing laws can impact market volatility. For example, financial regulations affecting trading practices can lead to shifts in market dynamics, affecting investor behavior.
- **Technological Developments:** Innovations in trading technology and algorithmic trading can lead to increased volatility, as automated trading systems may contribute to rapid price changes in response to market signals.
- **Psychological Factors:** Behavioural finance theories suggest that cognitive biases, such as overconfidence and herd behavior, can lead to irrational trading behavior, increasing volatility during certain market conditions.(Safiq & Yulianti, 2023)

## Methods of Calculating Volatility

### Standard Deviation:

This is the most common method, calculating the dispersion of returns around the mean. It's appropriate for normally distributed returns and provides a basic measure of risk.

Formula:

$$\sigma = \sqrt{(\sum(R_i - \bar{R})^2 / N)}$$

where:

$\sigma$  = standard deviation

$R_i$  = individual return

$\bar{R}$  = average return

$N$  = number of returns

For example :

If daily returns are 2%, -1%, 3%, 1%, and -2%:

Calculate average return:  $\bar{R} = (2 - 1 + 3 + 1 - 2) / 5 = 0.6\%$ .

Calculate standard deviation:  $\sigma = \sqrt{((2 - 0.6)^2 + (-1 - 0.6)^2 + (3 - 0.6)^2 + (1 - 0.6)^2 + (-2 - 0.6)^2) / 5} \approx 1.87\%$ .

This example suggest that average return (0.6%) a slight overall positive trend in the returns. Standard deviation (1.87%) shows that individual returns deviate from the average by a relatively significant amount. This suggests a moderate level of volatility. A higher standard deviation would indicate greater volatility, meaning returns are more widely dispersed and unpredictable. A lower standard deviation would suggest less volatility, with returns clustering more closely around the average.

Limitations of Standard deviation

- Assumes normal distribution of returns, which is often unrealistic.
- Underestimates risk during periods of extreme volatility

### Historical Volatility:

This method uses past price data to calculate volatility over a specified period. It is useful for understanding how an asset has behaved historically, especially suited for long-term analysis.

Historical Volatility Formula:

$HV = \sigma * \sqrt{T}$  where:

HV = historical volatility

$\sigma$  = daily standard deviation

T = trading days in a year (usually 252)

If daily standard deviation is 1.5%,  $HV = 1.5\% * \sqrt{252} \approx 23.8\%$ .

This indicates an annualized historical volatility of approximately 23.8%.

Historical volatility is a measure of how much a security's price fluctuates over a given period. In this case, the daily standard deviation of 1.5% is annualized to 23.8%, which means that the price of the security is expected to fluctuate by an average of 23.8% per year. A higher historical volatility suggests greater risk. Investors who are comfortable with higher risk might be willing to accept a higher historical volatility in exchange for potentially higher returns. Conversely, investors who are risk-averse may prefer securities with lower historical volatility.

Limitations of historical volatility

- Relies solely on past data, which may not accurately predict future volatility.
- Can be biased by recent market events.

**Implied Volatility:**

Implied volatility is derived from the market price of options and reflects the market's expectations of future volatility. It is particularly useful in options pricing and for traders looking to gauge market sentiment.

Limitations of Implied volatility

- Influenced by market sentiment rather than actual volatility.
- May not reflect true risk in illiquid options markets.

**GARCH (Generalized Autoregressive Conditional Heteroskedasticity):**

GARCH models forecast future volatility based on past periods' returns and volatility. It is suitable for time series data with changing volatility patterns, commonly used in financial econometrics.

Formula:

$$\sigma_t^2 = \alpha_0 + \alpha_1 R_{t-1}^2 + \beta_1 \sigma_{t-1}^2 \text{ where:}$$

$\sigma_t^2$  = conditional variance  $R_{t-1}^2$  = previous return squared

$\alpha_0, \alpha_1, \beta_1$  = parameters

Limitation of GARCH method

- Complex estimation process and sensitivity to parameter choice.
- Computationally intensive and may struggle with sudden market shifts.

**EWMA (Exponentially Weighted Moving Average):**

This method assigns more weight to recent observations, making it responsive to recent price changes. It is appropriate for situations where more recent volatility is more relevant than historical averages.

Formula:

$$\sigma_t^2 = \lambda \sigma_{t-1}^2 + (1 - \lambda) R_t^2$$

where:

$\sigma_t^2$  = current variance

$\lambda$  = decay factor ( $0 < \lambda < 1$ )

$R_t^2$  = current return squared

**Example:**

If previous variance is 0.0025, current return is 0.03, and  $\lambda = 0.94$ :  $\sigma_t^2 = 0.94 * 0.0025 + (1 - 0.94) * 0.03^2 \approx 0.0026$ .

The calculated variance ( $\sigma_t^2$ ) of 0.0026 indicates a moderate level of volatility.

In the context of the EWMA (Exponentially Weighted Moving Average) model, this variance reflects the current estimated volatility based on the previous variance and the current return.

A higher variance would suggest greater volatility, meaning the price of the asset is experiencing more significant fluctuations.

A lower variance would indicate less volatility, with a more stable price.

In this case, the variance of 0.0026 falls somewhere in the middle, suggesting a moderate level of volatility. It's important to note that this is a relative measure, and the specific interpretation would depend on the context of the asset and the industry.

Limitations of EWMA method

- Slow to react to abrupt changes in volatility.
- Choice of decay factor can significantly impact results.

### Range-Based Volatility:

This method uses the high and low prices over a certain period to estimate volatility, which can be more robust to outliers than standard deviation. It's suitable for markets with infrequent trading.

Formula:

$R = (H - L) / L$  where:

R = range-based volatility

H = highest price

L = lowest price

### Example:

If highest price is \$120 and lowest is \$100:  $R = (120 - 100) / 100 = 0.2$  or 20%.

The calculation  $R = (120 - 100) / 100 = 0.2$  or 20% indicates a range-based volatility of 20%.

This means that the price of the asset fluctuated by 20% from its lowest to its highest point during the given period. This means that the price of the asset fluctuated by 20% from its lowest to its highest point during the given period.

A higher range-based volatility suggests greater price fluctuations within that period.

A lower range-based volatility indicates a more stable price. In this case, a 20% range-based volatility suggests a moderately volatile asset. However, it's important to note that this is a relative measure, and the specific interpretation would depend on the context of the asset and the industry.

Limitations of Range-based volatility

- Requires high and low-price data, which may not be available for all assets.
- Less effective in capturing volatility during rapid price changes.

## CONCLUSION AND RECOMMENDATION

The Volatility, a fundamental concept in financial markets, quantifies the uncertainty associated with asset price movements. This paper has provided a comprehensive analysis of volatility, covering its definition, types, determinants, and measurement methods. Key findings include the influence of various factors on volatility, the use of different measurement methods, and the importance of understanding volatility for informed decision-making and risk management.

Future research can explore advanced volatility models, volatility spill overs, investor behavior, and the relationship between volatility and asset pricing. By continuing to study and understand volatility, researchers, and practitioners can contribute to the development of more effective risk management tools and investment strategies.

## LIMITATIONS

This comprehensive analysis of stock market volatility provides valuable insights, but it's important to acknowledge certain limitations. The accuracy and reliability of the findings depend heavily on the availability and quality of data. Limited or incomplete data can hinder the analysis of specific time periods or factors. Additionally, the models used to measure and analyze volatility, such as GARCH and EWMA, have their own limitations and underlying assumptions. These models might not capture the full complexity of volatility dynamics, especially during extreme events or structural shifts.

Furthermore, the study primarily focuses on stock market volatility, while volatility is also observed in other financial markets like commodities, bonds, and foreign exchange.

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