

## Data Article

**Title:** Volatility Modelling of the JSE Top40 Index: Assessing the GAS Framework Against GARCH and Hybrid GARCH–XGBoost

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## Abstract

This paper studies the volatility dynamics of the JSE Top40 Index by estimating a univariate GAS model with time-varying location, scale, and shape parameters (identity score scaling) and comparing its density and point-forecast performance against a standalone ARMA(3,2)-EGARCH(1,1) model and a hybrid ARMA(3,2)-EGARCH(1,1)-XGBoost framework. The GAS model is estimated on 3,515 daily observations, and several conditional densities are examined. The Student-t GAS model (GAS-STD) obtains the lowest information criteria within the GAS family (AIC = 10,188.142; BIC = 10,243.626) and exhibits statistically significant persistence in location and scale dynamics. Statistical diagnostics provide evidence of correct density calibration (Normalised Log Score = 1.1932; Uniform score = 0.4417), although residual skewness remains (IID-Test skewness  $p = 0.0134$ ). Out-of- sample analysis shows that GAS-STD performs strongly in density and risk forecasting, producing accurate 5% VaR and ES paths and passing coverage backtests (Kupiec LRuc 13  $p = 0.8414$ ; DQ  $p = 0.2281$ ). However, short-horizon point forecasts are best delivered 14 by the hybrid ARMA-EGARCH-XGBoost model (RMSE = 0.1386), with Diebold-Mariano tests confirming a transitive ranking: Hybrid > ARMA-EGARCH > GAS-STD. Simulation experiments highlight the sensitivity of tail behaviour to degrees-of-freedom (e.g., kurtosis  $v = 5 \approx 7.32$ ). Overall, GAS-STD is a strong density and risk model for the JSE Top40, while the hybrid framework excels in short-term volatility forecasting. The data are stored in an Excel file.

## Specifications Table

Subject area	<i>Volatility Modelling and Short-Term Forecasting</i>
More specific subject area	<i>Financial Market Modelling and Forecasting</i>
Type of data	<i>Excel file</i>
How data was acquired	<i>Provided and from the internet</i> <a href="https://za.investing.com/indices/89ftse-jse-top-40-historical-dat">https://za.investing.com/indices/89ftse-jse-top-40-historical-dat</a>
Data format	<i>Filtered and analysed.</i>
Experimental factors	<i>N/A</i>

Experimental features	<i>N/A</i>
Data source location	<i>FTSE/JSE Top 40 index Data portal webpage</i>
Data accessibility	<i>The data is hosted on GitHub <a href="https://github.com/csiqaake">https://github.com/csiqaake</a></i>
Related research article	Volatility Modelling of the JSE Top40 Index: Assessing the GAS Framework Against GARCH and Hybrid GARCH–XGBoost

### **Value of the Data**

The data can be used for financial time series forecasting in cross-sectional, temporal, or cross-temporal frameworks.

### **Data**

The data used in this study is from Wall Street Journal Markets, FTSE/JSE Top 40 index, and it can be accessed from [https://za.investing.com/indices/ 89 ftse-jse-top-40-historical-dat](https://za.investing.com/indices/89-ftse-jse-top-40-historical-dat). The data comprises Date and Close price.

### **Experimental Design, Materials, and Methods**

Data used in the study is from FTSE/JSE Top 40 index.

### **Acknowledgements**

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### **Funding sources**

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### **References**

The FTSE/JSE Top 40 index from [https://za.investing.com/indices/ 89 ftse-jse-top-40-historical-dat](https://za.investing.com/indices/89-ftse-jse-top-40-historical-dat) (accessed on 25 April 2025).