Efficient Market Hypothesis in the Presence of Market Imperfections: Evidence from Selected Stock Markets in Africa

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This paper investigated the weak axiom of the efficient market hypothesis (EMH) as it applies to fifteen (15) leading stock markets in Africa. There are currently over twenty-nine stock exchanges in Africa with a significant degree of disparities ranging from market size, trading volume, number of listed companies, access to funds, access to information to market standardization etc. The article deviated from the conventional linear approach of testing efficient market hypothesis and the method of using the runs test for serial dependency to test the weak-form efficient market hypothesis. The paper adopted the wavelet unit root analysis-tool, which decomposed the stochastic processes into its wavelet components, with varying frequency band. The study found that institutional constraints have implications for the efficient market hypothesis and investment in the African stock market. The conclusions drawn from the study is the relevance of using past historical stock prices to predict the current earnings at stock markets in Africa, a negation of the efficient market hypothesis.

Keywords: Efficient market, information asymmetry, stock prices, wavelet unit root, Africa

JEL: C01, C23, D84, G14

The African continent has evolved from a description of being “The Hopeless Continent” in the periods between 2000 and 2002 (see Brook, 2000; Williams, 2004) to an optimistic story as the “Aspiring Africa” in 2013 (see Andersen and Jensen, 2014) where investors all over the world see the African continent as the next frontier of economic growth. Africa is an area where substantial growth has been predicted to take place in the stock market. A market where behavioral economists and financial analysts have argued that arbitragers with significant market information could predict future stock prices and profit earnings, also, wealth holders with insight into previous changes in the price level can predict future prices and current profit takings in the market (see Adam, Marcet and Beutel, 2017; Butt, et al., 2010; Dimpfl and Jank, 2016; Kafayat, 2014; Mallick, 2015; Yang, Jhang and Chang 2016; Zindel, Zindel and Quirino, 2014).

The reality, however, is that in a competitive market setting, current prices are known to adjust rapidly in real-time eliminating the ability of investors and arbitragers to use past information available to the reality, however, is that in a competitive market setting, current prices are known to adjust rapi-
dly in real-time eliminating the ability of investors and arbitragers to use past information available to predict current and future price outcome (Bhargava, 2014; Degutis and Novickytė, 2014; Tiwari and Kyophilavong, 2014).

Eugene Fama introduced efficient market hypothesis (EMH) in his 1960 dissertation. Accordingly, he asserted that at any given time, stock prices are a reflection of all available information in the capital market and are traded at their fair value at all times making it impossible for market participants to consistently choose stocks that will beat the returns of the overall market.

The efficient market hypothesis (EMH), although well-received by financial and behavioral economists from 1970s to 1990s; the theory came under criticism in the late 1990’s up until the period of global economic crisis of 2007–2008 following the series of events that happened globally that undermined the assumptions on which the EMH rested. The first of these events was the dot–com bubble and the technology bubble that occurred from 1995 to 2000 – a period of excessive speculation, rapid share price growth and high stock price valuation that allowed investors to make abnormal returns (see McAleer, Suen and Wong, 2016; Schubert et al., 2018).

The second is the United States sub–prime mortgage crisis, and the stock market crash of 2007–2010, which triggered the 2007–2008 global economic crises. Economic analysts questioned the relevance of the EMH on the premise that the dot–com bubble and the sub–prime mortgage crisis would not have occurred if the efficient market assumptions were fundamentally correct (Constâncio, 2014; Gilson and Kraakman, 2014; Ouarda, El Bouri and Bernard, 2013).

Despite the criticisms on the assumptions of the EMH, African stock market is on the increase, expanding rapidly and attracting private investment and integration into the global financial market. There are currently over twenty–nine (29) stock exchanges in Africa with a varying degree of disparities in terms of market size, the number of listed firms, trading volume, access to funds, access to information, and market standardization (Boamah, Watts and Loudon, 2017). These institutional constraints alongside the existence of information asymmetry, principal agency problems, regulatory constraints and the presence of weak financial institution have implications for the relevance of EMH and investment in the African stock market.

The gaps in the existing literature on EMH on the African continent prompted this study. Empirical studies conducted for Africa remain inconclusive with mixed reports. This contrary reports are evident in the works (Adigwe, Ugbomhe and Alajekwu, 2017; Bulla, 2015; Bundoo, 2000; Simons and Laryea, 2005; Van et al., 2013) which supported weak–form efficient market hypothesis, but many studies documented inefficient market hypothesis (Awiagah and Choi, 2018; Ayentimi, Mensah and Naa–Idar, 2013; Chikoko and Muparuri, 2013; Katabi and Raphael, 2018; Lawal, Somoye and Babajide, 2017;
Nwidobie, 2014; Smith, 2008; Zaman, 2019); while few studies reported mixed results (Abakah et al., 2018; Phiri, 2015; Vitali and Mollah, 2015). Additionally, Kelikume (2016) recorded strong-form efficient market hypothesis. The study prompted by the need to put to rest existing empirical divergences on the African stock market. Also, as the African stock market develops in the presence of imperfect information, investors, regulators and other participants require clarity on how efficient or inefficient the stock market is to avoid a crash shortly.

The study’s broad objective is to examine the relevance of the efficient market hypothesis in Africa in the presence of information asymmetry and market imperfections. The focus of the research will be to test the weak axiom of the efficient market hypothesis as it applies to selected stock markets in Africa. The study’s contribution to knowledge is four-fold. First, it improved on existing studies on EMH in Africa by considering a wider spectrum of 15 stock markets in Africa, which represents 50 percent expansion in sample size against the maximum of 10 stock markets investigated in previous studies (Lawal et al., 2018; Mlambo and Biekpe, 2007). Second, the study adds to empirical literature having adopted an improved method of data analysis (wavelet) against the conventional linear approach and the runs tests employed by Mlambo and Biekpe (2007). The wavelet analysis improves on other unit root tests by decomposing the stochastic processes into its wavelet components with a specific frequency band (Fan and Gencay, 2010). Third, the study contributed significantly to the debate on the relevance of EMH in Africa with important policy suggestions based on the ERS tests, Bai Perron, and variance ratio analysis. Fourth, the findings and policy implications of this study are relevant to economists, financial analysts and investors for stock market investment decisions in the presence of information asymmetry and other market imperfections.

Following the introductory section, the rest of the paper presents: the performance of the stock market in Africa in next sub-section, theoretical underpinnings and review of the empirical literature on EMH in section 2; the methodology is presented in section 3, while section 4 deals specifically with the results and discussion. Section 5, 6, and 7 present the conclusion, implications and future directions.

Stylized Facts
The equity market in Africa had developed and expanded in recent times. The number of operating stock exchange markets rose from just eight in 1989 to twenty-three in 2007 and twenty-nine in 2019, reaching a total market capitalization of over US$2.1 trillion. Many African stock market returns remain juicy to investors despite their small size, low liquidity and amidst current weak performance (see Figure 1).

African stock market has always been among the top 10 best–performing markets in the world since 1995. In 2004, six African countries (Ghana, Uganda, Kenya, Egypt, Mauritius, and Nigeria)
Figure 1. All Share Index (ASI) for Selected African Countries for June-2019

were among the world’s ten (10) best–performing stock markets, while in 2005 Egypt, Uganda and Zambia were in the top five (5) ranking. In 2006, Malawi outperformed every other market in the world. However, the top–performing stock market in order of ranking includes South Africa, Malawi, Nigeria, and Morocco since 2010, although these markets are currently experiencing a downturn (see Figure 2a), and of the nineteen countries, only Egypt and Zimbabwe remain positive (see Figure 2b).

Figure 2a. Year on Year Growth Rate of ASI for Top Four African Countries
LITERATURE REVIEW

Theoretical Underpinnings

The efficient market hypothesis (EMH) relates assets prices with information flow. The EMH is one financial market theory with rich historical development. Gyamfi, Kyei and Gill (2016) tracked the evolution of this theory to the earliest works of Cardano (1564) on the principle of equal conditions in gambling, the study of Brown (1828) on rapid oscillatory motion, and Regnault (1863) on stock price deviation and time relation. The efficient market was recognized when shares in the open market are publicly known: its value, as acquired, implies the judgment of the best intelligence concerning them (Gibson, 1889). Other studies that culminated in the evolution of EMH include (Einstein, 1905; Fama, 1965b; Friedman, 1953; Granger and Morgenstern, 1963; Harry, 1959; Keynes, 1923; Mandelbrot, 1963; Sharpe, 1964; Tussig, 1921; von Smoluchowski, 1906). Although Fama (1965b) provided the initial conceptual position of an efficient market, it was Samuelson (1965) who provided a formal economic argument for efficient markets from the standpoint of martingale, rather than a random walk.

Further, Fama and Blume (1966) submitted that measuring the direction and degree of dependence in price variations, serial correlation rule is sufficient. Ultimately, the empirical findings by Fama (1970) lend credence to the conclusion that the stock market is efficient. Fama (1970) posited that in an active market (with many well-informed and intelligent investors), appropriately priced securities reflect all available information, that is a market in which securities price reflect all possible information.
quickly and accurately so that it is impossible for market participants to earn abnormal profits. The theory was built on the assumption that returns distributions do not change over time.

Also, the degree of market efficiency depends on existing information conditions in the market environment; therefore, Fama (1970) categorized the information set into three forms (levels) namely the weak, semi-strong, and strong-form. The strong-form efficient market must have: many knowledgeable investors actively analyzing and trading stocks; information is widely available to all investors; events, such as labor strikes or accidents, tend to happen randomly; the fast and accurate reaction of investors to new information. When any of these conditions are absent in the market, we may have weak-form or semi-strong form. The weak-form EMH defines a market as being efficient if current prices fully reflect all information contained in previous stock prices. This form implies that historical prices cannot suffice as a forecasting tool for the stock price. Thus, it is impossible to make abnormal returns by using only past historical prices while the semi-strong form of the EMH states that current market prices reflect all publicly available information.

The strong-form EMH implies that private information (inside information) for making abnormal returns is hard to obtain as a result of the stiff competition amongst participants. However, in reality, some investors or market participants can make abnormal returns; thus, the strong-form EMH is not very likely to hold.

Empirical Literature

A large number of publications have examined the existence of the EMH in various developed and undeveloped markets with varying results (Ananzeh, 2014). Evidence from literature confirms the presence of the weak-form EMH in developed market (Anagnostidis, Varsakelis and Emmanouilides, 2016; Cootner, 1962; Fama and Blume, 1966; Mensi, Tiwari and Al–Yahyaeel, 2019; Williamson, 1972; Yang et al., 2019) while empirical evidence from studies conducted in emerging economies yielded mixed results, between accepting or rejecting the null hypotheses of weak-form EMH.

For instance, Dahel and Laabas (1999) documented that stock market in Kuwait displays weak-form EMH, and rejected the weak-form EMH for Bahrain, Kuwait, Saudi Arabia and Oman. Wheeler et al. (2002) did not support the weak form of EMH for the stock market of Warsaw (Poland). Also, Abeysekera (2001), and Abraham and Alsakran (2002) in their empirical finding, rejected the null hypothesis of weak-form efficiency for stock markets in Sri Lanka, Bahrain, Kuwait, and Saudi Arabia. On the other hand, Karemera, Ojah and Cole (1999) strongly support the weak form of EMH for the stock market in Turkey. Studies conducted by Iqbal and Mallikarjunappa (2008, 2010, and 2011) on Indian Stock Market, found that the stock market of India is not efficient in weak and semi-strong form.
Hou and Sun (2014) tested the weak-form market efficiency for Canada and China, and found a mixed result that differs from the sample period to sample period for both markets. Although, almost all testing techniques generated unfavorable results against the weak-form EMH for both countries, however, result from more recent data sample suggests that both markets are efficient. Furthermore, Awan and Subayyal (2016) studied six stock exchanges in the Gulf region, that included Bahrain, Kuwait, Oman, Saudi Arabia, UAE and Qatar for the five years, spanning 2011 to 2015. Their findings support evidence that the stock prices at the Gulf markets do not follow the random walk model.

In 2017, Hawaldar, Rohit and Pinto (2017) tested for the weak-form efficient market hypothesis in Bahrain Bourse using the Kolmogorov–Smirnov (K–S) goodness of fit test, runs test and autocorrelation test. Whereas the K–S test result concluded that the general stock price movement does not follow the random walk, results of the runs test revealed that share prices of seven companies do not follow random walk while the autocorrelation tests revealed that share prices exhibit low to moderate correlation varying from negative to positive values. With the show of mixed result from the different analysis, it was difficult for Hawaldar, Rohit and Pinto (2017) to ascertain the weak form of the efficiency of Bahrain Bourse.

Similarly, in Africa, empirical studies conducted remain inconclusive yielding weak–form efficient (Adigwe et al., 2017; Bulla, 2015; Bundoo, 2000; Simons and Laryea, 2005; Van et al., 2013), inefficient (Awiagah and Choi, 2018; Ayentimi et al., 2013; Chikoko and Muparuri, 2013; Katabi and Raphael, 2018; Lawal et al., 2017; Nwidobie, 2014; Smith, 2008; Zaman, 2019), mixed results (Abakah, et al., 2018; Phiri, 2015; Vitali and Mollah; 2015) and strong–form efficient (Kelikume, 2016) (see Table 1, Appendix–I).

From the empirical review of existing literature, it is obvious that stock market analysis in Africa is still far from harmony. To contribute to the ongoing debate in this research space requires the use of recent data on the stock market, the deployment of modern methods and expanded scope in terms of the numbers of stock markets examined. By these, this study contributes to the literature on EMH.

METHODOLOGY

–Sample and Data
The population of this study is the 29 stock markets in Africa. A sample of 15 stock markets was drawn for analysis. A convenient sampling technique was employed based on data availability. Monthly data for the sampled 15 stock markets in Africa spans the period of January 2010 to June 2018. Data collected from the online site of the African Markets (retrieved from https://www.african-markets.com/en/stock-markets). The study uses monthly all share index data for two reasons: (1) the
potential for thin trading in Africa stock markets and (2) using daily/weekly prices in a return series from infrequently traded stocks may lead to significant biases in the results (Mlambo and Biekpe, 2007). In line with the literature, the monthly index returns are computed as follows:

\[ r_t = \ln \left( \frac{P_t}{P_{t-1}} \right) \times 100\% \]  

Where \( r_t \) is a monthly market return for period \( t \), \( P_t \) and \( P_{t-1} \) denote market prices for period \( t \) and period \( t-1 \) respectively, while \( \ln \) connotes natural logarithm. The log transformation converts the data into continuously compounded rates. (see Table 2, Appendix–II for Descriptive Statistics)

Wavelet Unit Root Test

In line with Kelikume (2016), and Tiwari and Kyophilavong (2014) this study employed wavelet unit root technique to investigate the efficient market hypothesis for stock indices. The literature on unit root tests continues to develop one framework after another, with different assumptions and incorporating different levels of nonlinearity, volatility and structural breaks. While the traditional unit root tests in the literature are based on a time-domain analysis, this study uses a different test based on wavelet analysis. The wavelet analysis decomposes the stochastic processes into its wavelet components, with a specific frequency band. To develop the wavelet–based unit root techniques, Fan and Gençay (2010) decomposed the variance of the underlying processes into the variance in its low and high-frequency components through the discrete wavelet transformation (DWT).

Fan and Gençay (2010) defined \( \{X_t\}_{t=1}^{T} \) as a univariate time series which can be defined as

\[ x_t = \beta x_{t-1} + \varphi_t \]  

\( \varphi_t \) is a weakly stationary zero mean error with a strictly positive long-run variance which can be defined as

\[ \theta^2 = \delta_0 + 2 \sum_{j=1}^{\infty} \delta_j \]  

\[ \delta_j = E(\varphi_t \varphi_{t-j}) \]  

The test is only applicable to the linear trend and non-zero mean cases. Assuming that the process \( \{x_t\} \) can be defined as:

\[ x_t = \xi + \mu + x_t^\varepsilon \]  

\( x_t^\varepsilon \) is produced by model (1).

If the null hypothesis \( H_0: \beta = 1 \), then \( \{x_t^\varepsilon\} \) is a unit root process. On the other hand, if \( H_0: \beta < 1 \), then
\( \{x_t^i\} \) is a zero mean stationary process.

If \( \gamma = 0 \), then the demeaned series \( (x_t - \bar{x}) \) where

\[
\bar{x} = \frac{1}{T} \sum_{t=1}^{T} x_t
\]

defines the sample mean of \( \{x_t\} \).

If \( \gamma \neq 0 \), then the detrended series \( (x_t - \bar{x}) \) where

\[
\bar{x}_t = \sum_{j=1}^{T} (\Delta x_j - \Delta \bar{x})
\]

defines the sample mean of \( \{x_t\} \).

\( \bar{x} \) is the sample mean of \( \bar{x}_t \) where

\[
\Delta x_t = x_t - x_{t-1}
\]

And \( \Delta \bar{x}_t \) is the mean of \( \Delta x_t \).

Based on the unit scale DWT wavelet, Fan and Gençay (2010) introduced two test statistics, for the demeaned and the detrended series.

The test statistics for the demeaned series is defined as:

\[
\frac{\sum_{t=1}^{T/2} (B_{t,1}^\ell)^2}{\sum_{t=1}^{T} (x_t - \bar{x})^2}
\]

Where \( B_{t,1}^\ell \) stands for the scaling coefficient of the demeaned series.

The test statistics for the detrended series is defined as:

\[
\frac{\sum_{t=1}^{T/2} (B_{t,1}^d)^2}{\sum_{t=1}^{T} (x_t - \bar{x})^2}
\]

Where \( B_{t,1}^d \) denotes the scaling coefficients of the detrended series.

The two are used to test the null hypothesis, \( H_0: \beta = 1 \) against the alternative hypothesis, \( H_1: \beta < 1 \) in the model (1).
Figures 3a and 3b (see Appendix-V/VI) show the development of the African stock indices. African stock markets are on the increase, expanding rapidly and attracting private investment and integration into the global financial market. Currently, there are 29 stock markets in Africa with different degrees of differences in terms of market size, trading volume, and the number of listed companies.

RESULTS AND DISCUSSION

The wavelet unit root test is used to investigate the efficient market hypothesis for stock returns, and this follows previous studies on efficient market hypotheses for African stock (e.g., Kelikume, 2016; Lawal et al., 2018; Tiwari and Kyophilavong, 2014). The results of the wavelet-based unit root tests are summarized in Table 3 (see Appendix-III). The use of three different lags (10, 20 and 30) ensures the robustness of the results. Despite the different lag lengths, the test statistics accept the null hypothesis of stationarity at 1 percent, 5 percent and 10 percent significance level with the wavelet statistic greater than the corresponding critical values. In other words, stock return series are stationary.

To enhance the robustness of the empirical results, this study compares the wavelet results with a battery of traditional tests of efficient market hypothesis well established in the literature, such as Elliott–Rothenberg–Stock (ERS) unit root test, Bai–Perron breakpoint test, and variance ratio tests. From Table 4 (see Appendix-IV), all the tests show the fact that the return series are stationary, irrespective of the levels of the lag length.

CONCLUSION

The key conclusion drawn from the findings of this study is that arbitragers and wealth holders with significant market information and insight into previous changes in the price level can predict future prices and current profit takings in the market. In a competitive market setting, current prices are supposed to adjust rapidly in real-time to eliminate the ability of investors and arbitragers to use past information available to predict current and future price outcome (Bhargava, 2014; Degutis and Novickytė, 2014; Tiwari and Kyophilavong, 2014). However, this study has shown that, for African stock markets, the converse is the case. This study, therefore, debunks and negates Fama’s efficient market hypothesis. In sharp contrast to Fama’s efficient market hypothesis, the findings of this study has shown that it is possible for market participants in African stock markets to consistently choose stocks that will beat the returns of the overall market, thus, affirming earlier reports by Adigwe et al. (2017), Appiah-Kusi and Menyah (2003), Bundoo (2000), Lawal et al. (2017), Simons and Laryea (2005), and Smith (2008).
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The dot-com bubble and the technology bubble that occurred from 1995 to 2000 (a period of excessive speculation, rapid share price growth, and high stock price valuation that allowed investors to make abnormal returns) further illustrate the failure of the efficient market hypothesis. The United States subprime mortgage crisis and stock market crash of 2007–2010 are other pointers to the failure of the efficient market hypothesis. The 2007–2008 global economic crises, therefore, would not have occurred if the efficient market assumptions were fundamentally correct.

IMPLICATIONS

The findings of this study bear significant theoretical, social change, and practical implications for stock market participants. First, the African stock market performance does not support the efficient market hypothesis. The lack of support for the position of EMH could be justified by a lack of precision on what constitutes efficient price response information. Ball (2009) posits that comparing the returns earned from trading on the information with the returns otherwise expected from passive investing is a measure of efficiency; however, implementing the counterfactual in this way suffers from the bad model problem. The reason earlier studies relied on the capital asset pricing model.

Another implication of our findings is the silence of efficient market hypothesis on an exact sequence that makes an efficient price reaction in Africa. It also does not explain variations and magnitude in fall or recovery in the price of African stock. This is because the African stock market information is dynamic and changes with the interest rate, risk, security risk and risk premium. The hypothesis does not account for how changes in monetary policy, fiscal policy, demographics of investors, technological innovations, and labor productivity affect expected returns in an efficiently priced market. In Africa where technological changes, labor productivity, investors’ characteristics and discretionary policies are volatile, important and exogenous, the EMH remains significantly silent.

Implicit in the Fama’s EMH is the assumption that the observed level of risk (ex-post) risk level is the accurate level (Ball, 2009). However, from the result of this study, the EMH does not encompass the correct level of risk that makes security efficiently priced. Modeling and estimating risk parameters is cumbersome due to its dynamic nature (Ji et al., 2018). Also, African stock markets are susceptible to uncertainties occasioned by industrial and legal actions, competitive strategies, and major announcements of earnings (Auwal and Sanusi, 2016; Balcilar et al., 2019). These factors determine whether the market over-assesses or under-assesses risk (Tweneboah, Junior and Oseifuah, 2019), thus account for the weak-form market hypothesis, which departs sharply from the efficient market efficiency.

These research findings are also useful to institutional and individual investors in the African stock market from various perspectives. An understanding of stock market efficiency is relevant for corporate
executives whose decisions and actions affect the perceived value of companies. Also, stock market development models employed in this study are crucial for supervisory and operational benefits. Despite a paradigm shift towards behavioral financial theory in recent year, the EMH remains useful in analyzing stock returns globally.

LIMITATIONS AND FUTURE DIRECTIONS

The key limitation of this study is that it focused on only fifteen (15) stock markets in Africa to test the EMH. Studies interested in testing the validity of EMH in African stock market in the future should extend their analysis to all existing stock markets in the continent, where data availability is not a constraint; this is capable of yielding more robust empirical findings for the formulation of efficient stock market policies. Future study should also depart from the battery tests employed in this study. Other robust battery of unit root tests suggested included those that do not consider structural breaks in the data such as ADF, PP, ADF–GLS, NP4; and the others that allow for endogenously determined structural breaks in the data such as Lumsdaine and Papell (1997), and Clemente, Montañés and Reyes (1998). Clemente et al. (1998) tests are based on the framework of the innovative outlier and additive outliers.

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| Author and Year | Methodology Applied | Data / Country Examined | Key Findings |
|-----------------|---------------------|-------------------------|--------------|
| Bundoo (2000)   | Serial correlation test | Mauritius, 1992-1998   | Weak-form efficient |
| Appiah-Kusi and Menyah (2003) | EGARCH-M | Botswana, Egypt, Ghana, Ivory Coast, Kenya, Mauritius, Morocco, Nigeria, South Africa, Swaziland and Zimbabwe | Egypt, Kenya, Mauritius, Morocco and Zimbabwe are weak-form efficient |
| Simons and Laryea (2005) | K-S goodness of fit test, runs test, autocorrelation test, multiple VAR test, autoregression test | Egypt, Ghana, Mauritius, South Africa, 1990-2003 | Weak-form efficient |
| Smith (2008)    | Wright's joint VAR tests, Kim's wild bootstrap approach on Chow-Denning multiple VAR test | Botswana, Egypt, Ghana, Ivory Coast, Kenya, Mauritius, Morocco, Nigeria, South Africa, Tunisia and Zimbabwe, 2000-2006 | Not weak-form efficient |
| Ayentimi, Mensah, and Naa-Idar (2013) | Kolmogorov-Smirnov and runs test, Runs test and Autocorrelation test | Monthly for Ghana covering Jan. 2007 - Jun. 2012 | Not weak-form efficient |
| Chikoko and Maparuri (2013) | Threshold Autoregressive (TAR) model | Daily and weekly data for Zimbabwe covering Feb. 19, 2009 to Dec. 31, 2012 | Not weak-form efficient |
| Van, Rodrigues, Hockly, Lambert, Tjaart and Phiri (2013) | Augmented Dickey-Fuller | Weekly data for South Africa over 2000 and 2013 | Weak-form efficient |
| Nwidobie (2014) | Auto-correlation test, Ljung-Box Q test, LM serial correlation, unit roots test and Runs test | Monthly for Nigeria covering Jan. 2000 - Dec. 2012 | Not weak-form efficient |
| Vitali and Mollah (2015) | Unit root, auto-correlation, runs test and variance ratio | Monthly for Ghana covering Jan. 2007 - Jun. 2012 | Not weak-form efficient |
| Bulla (2015)    | Serial correlation, and Runs test | Monthly for Nigeria covering Jan. 2006 - Dec. 2011 | Not weak-form efficient |
| Phiri (2015)    | Classical augmented Dickey-Fuller tests, Two-regime threshold Autoregressive unit root tests and Three-regime unit root tests | Monthly for South Africa over Jan. 31, 2000 to Dec. 16, 2014 | Weak-form efficient |
| Kelikume (2016) | wavelet-based unit root tests | Monthly data for Nigeria Stock Market over the sample period 1985 to 2015 | Strong-form efficient |
| Adigwe, Ugbomhe, and Alajeckwa (2017) | Jarque-Bera statistics test and Augmented Dickey Fuller test wavelet-based unit root tests | Monthly data for 13 African Stock Exchange covering Jun. 2013 to Dec. 2015 | African stock markets are weak-form efficient |
| Lawal, Somoye and Babajide (2017) | Serial correlation test-The Ljung-Box test, Unit root tests, non-parametric runs test and the variance ratio test | Monthly data on seven African stock markets, Daily closing stock prices of the market index (All share Index-DSEI) for Dar es Salaam covering Jan. 2009 - Mar. 2015. | Weak form inefficient |
| Katabi and Raphael (2018) | Ljung-Box autocorrelation test, unit root tests, the runs test, and variance ratio tests (such as Wright’s rank and sign and Lo-Mac Kinlay) | Daily, weekly, monthly, and quarterly returns Ghana covering 1990 – 2017 | Inefficient at weak-form |
| Awiagah and Choi (2018) | Non-Linear Fourier unit root test | Weekly returns of S&P/IFC return indices for five African countries over the period 2000-2013 | South Africa, Nigeria and Egypt are weak-form efficient whilst Ghana and Mauritius are weak-form inefficient. Weak form inefficient |
| Abakah, Alagidede, Mensah, and Ohene-Asare (2018) | Descriptive statistics, Autocorrelation test, Run test | Daily data for two Bangladesh stock market covering Jan. 2013 – Aug. 2017 | Weak-form efficient |

Source: Authors’ Compilation

Table 1. Studies Conducted on EMH in Africa
**Table 2. Descriptive Statistics of Stock Returns**

| Stock Market | Mean  | Median | Maximum | Minimum | Std. Dev. | Skewness | Kurtosis | Jarque-Bera | Probability |
|--------------|-------|--------|---------|---------|-----------|----------|----------|-------------|-------------|
| BRVM         | 0.0015| 0.0029 | 0.0727  | -0.0740 | 0.0332    | 0.0116   | 2.4615   | 1.3797      | 0.5016      |
| BSE          | 0.0005| -0.0001| 0.0543  | -0.0964 | 0.0187    | -0.7353  | 8.9009   | 175.6696    | 0.0001      |
| BVC          | 0.0004| -0.0012| 0.0998  | -0.0663 | 0.0278    | 0.3108   | 3.5461   | 3.2518      | 0.1967      |
| BVMT         | 0.0041| 0.0033 | 0.0830  | -0.0889 | 0.0307    | -0.2211  | 4.6168   | 13.3463     | 0.0013      |
| DSE          | 0.0041| 0.0033 | 0.0830  | -0.0889 | 0.0307    | -0.2211  | 4.6168   | 13.3463     | 0.0013      |
| EGX          | -0.0064| -0.0074| 0.1772  | -0.2107 | 0.0637    | -0.2373  | 3.4762   | 2.1472      | 0.3418      |
| GSE          | -0.0006| 0.0165 | 0.5988  | -2.1939 | 0.2716    | -4.6634  | 39.0607  | 6589.9830   | 0.0001      |
| JSE          | -0.0002| 0.0080 | 0.0863  | -0.7802 | 0.0811    | -7.8550  | 76.5174  | 26845.1600  | 0.0001      |
| LUSE         | 0.0050| 0.0008 | 0.0742  | -0.1535 | 0.0318    | -0.7865  | 7.3880   | 103.2103    | 0.0001      |
| MSE          | 0.0159| 0.0113 | 0.2417  | -0.1259 | 0.0412    | 1.7823   | 12.0156  | 446.4432    | 0.0001      |
| NGSE         | 0.0025| -0.0052| 0.1734  | -0.1258 | 0.0506    | 0.5709   | 3.6113   | 7.9684      | 0.0186      |
| NSE          | 0.0058| 0.0104 | 0.0901  | -0.0946 | 0.0385    | -0.4112  | 2.9445   | 3.2275      | 0.1991      |
| SEM          | 0.0006| 0.0002 | 0.0855  | -0.0526 | 0.0205    | 0.8019   | 4.7103   | 26.1134     | 0.0001      |
| USE          | 0.0065| 0.0082 | 0.1221  | -0.1346 | 0.0464    | -0.2272  | 2.8879   | 1.0401      | 0.5945      |
| ZSE          | 0.0124| -0.0066| 0.4615  | -0.2914 | 0.0956    | 1.7204   | 9.4451   | 253.5476    | 0.0001      |

Source: Authors’ Computation
### Table 3. Wavelet Unit Root Tests

| Lags | 10     | 20     | 30     | 10     | 20     | 30     |
|------|--------|--------|--------|--------|--------|--------|
| BRVM | -42.31†| -43.33†| -28.39*| -155.65†| -101.58†| -44.43*|
| BSE  | -34.48*| -27.73*| -15.83  | -137.63*| -37.45*| -28.02  |
| BVC  | -67.33†| -54.13†| -49.18*| -244.83†| -239.80†| -205.18†|
| BVMT | -44.00†| -41.45†| -40.60†| -173.92†| -147.74†| -152.59†|
| DSE  | -46.45†| -43.81†| -40.32†| -154.13†| -147.08†| -121.21†|
| EGX  | -76.85†| -58.97†| -44.96†| -132.31†| -112.68†| -92.35†|
| GSE  | -45.85†| -40.03†| -30.81*| -112.30†| -47.29*| -24.10  |
| JSE  | -40.03†| -38.31*| -33.68*| -104.14†| -46.79*| -30.82*|
| LUSE | -44.22†| -41.66†| -34.02*| -138.86†| -109.62†| -97.53†|
| MSE  | -40.68†| -41.88†| -14.32  | -145.97†| -138.62†| -106.89†|
| NGSE | -65.33†| -50.52†| -40.43†| -131.92†| -121.99†| -99.71†|
| NSE  | -48.89†| -40.86†| -43.73†| -148.21†| -140.51†| -131.39†|

**Kenya**

| SEM  | -53.73†| -50.38†| -31.73*| -137.29†| -135.05†| -116.96†|
| USE  | -46.75†| -42.69†| -43.55†| -154.33†| -131.46†| -89.62†|

**Uganda**

| ZSE  | -48.62†| -32.59*| -36.69*| -179.70†| -153.63†| -141.32†|

**Critical Values**

|       |       |       |
|-------|-------|-------|
| 1%    | -40.38| -50.77|
| 5%    | -27.38| -36.54|
| 10%   | -21.75| -30.23|

Source: Authors' Computation

Note: *, and † denote level of significance of 5% and 10%, respectively.
Table 4. Traditional EMH Tests

|                | ERS Test | Bai-Perron Test | Variance ratio test |
|----------------|----------|-----------------|---------------------|
| BRVM           | 1.51†    | -8.62†          | 3.96†               |
| BSE            | 3.08*    | -8.56†          | 2.13                |
| BVC            | 1.59†    | -8.89†          | 3.85†               |
| BVMT           | 1.48†    | -9.28†          | 3.13†               |
| DSE            | 1.49†    | -9.29†          | 3.12†               |
| EGX            | 1.53†    | -8.63†          | 3.67†               |
| GSE            | 1.84†    | -8.88†          | 1.02                |
| JSE            | 158.87   | -26.13†         | 6.99†               |
| LUSE           | 2.82†    | -8.42†          | 3.45†               |
| MSE            | 4.04†    | -5.07†          | 1.70                |
| NGSE           | 1.98†    | -7.89†          | 3.04†               |
| NSE Kenya      | 2.92†    | -8.89†          | 4.37†               |
| SEM            | 1.93†    | -8.13†          | 2.63*               |
| USE            | 1.83†    | -9.33†          | 4.26†               |
| Uganda         |          |                 |                     |
| ZSE            | 0.93†    | -9.48†          | 2.25**              |

Source: Authors' Computation
Note: **, *, and † denote level of significance of 1%, 5% and 10%, respectively.
Appendix-V

a. 

![Graph A]

Source: Authors’ Computation

b. 

![Graph B]

Source: Authors’ Computation

Figure 3a & 3b. Development of African Stock Indices
Figure 4. The Evolution of Stock Returns across the 15 Stock Exchanges