Public buffer stocks as agricultural output price stabilization policy in Ghana

Emmanuel Abokyi1,2*, Henk Folmer2,3 and Kofi Fred Asiedu1

Abstract

Background: Food price volatility poses widespread risks, from farmers to consumers, notably in developing countries. Because of its devastating effects on sustainable farming and food security, particularly for the poor, it continues to be a crucial policy priority. Governments have applied various methods of stabilizing domestic food prices including publicly held buffer stocks, import and export tariffs and production supports. This study evaluates Ghana's agricultural output price stabilization policy implemented in the context of the National Buffer Stock Program.

Results: Based on data from the Ministry of Food and Agriculture, we apply the coefficient of variation and the corrected coefficient of variation to analyze the volatility of output prices of maize and rice. The results show that the price volatility of maize and rice has declined in the markets where the policy has been implemented but not in the non-policy markets.

Conclusion: There is, therefore, empirical evidence that the policy has been successful.

Keywords: Agricultural output, Price volatility, Price stabilization policy, Buffer stock, Coefficient of variation, Augmented Dicky–Fuller test, Ghana

Background

Surges in global food prices continue to be of serious concern to governments all over the world, especially in developing countries, because of the devastating effects it has on food security for the poor and on sustainable production of agricultural commodities [1]. When food price levels increase, household purchasing power is lowered, especially for the poor [2]. However, farmers are often delighted with increases in price levels for their farm products with the expectation that it provides them with higher income [3]. The opposite holds when prices fall. While consumers prefer low and producers high price levels, volatility—magnitude and direction—ultimately results in harmful effects on both producers and consumers [4]. High food price volatility, therefore, is very detrimental to every country because it can cause serious political, economic and social problems [5].

Price volatility in agricultural commodity markets is normal [6]. According to Prakash [7], it is natural and instrumental that, when there is a shortage in the supply of a particular agricultural commodity, its price will increase, consumption will decline, and investment in the production of the commodity is stimulated. Therefore, some level of volatility is needed for commodity markets to function well [8]. However, it becomes of much concern when the magnitude and the frequency of the volatility are such that producers and consumers find it difficult to cope with them. Sumpsi [9] demonstrates that agricultural commodity volatility has negative impacts on food security and health. For example, when food prices rise sharply, households are likely to respond by eating cheaper and less nutritious foods, which can have potential lifelong effects on the social, physical, and mental well-being of millions of people. When children are malnourished, it exposes them to the risk of stunting, underweight, morbidity and mortality. In developing countries, the risk posed by high food price volatility is precarious for poor households because they spend a substantial proportion of their income on food, and hence price volatility poses a very high risk to them. For instance, in
Sub-Saharan Africa, poor households spend about 60% of their income on food [8]. The nature and magnitudes of the effects that food price volatility poses in developing countries vary depending inter alia on the specific country poverty levels [10, 11].

Food price volatility also poses the risk of unsustainable farming because of under- or overinvestment. High levels of volatility in food prices deter smallholder farmers—who are often risk averse—from making the necessary investments to increase productivity and production [3]. Furthermore, banks and other financial institutions tend to shy away from giving loans to farmers due to high financial risks and default. Consequently, adoption of efficient technologies will be low which has consequences for agricultural modernization, irrigation, and good agricultural practices. The effect is low productivity and poor income levels for farmers [12] which ultimately may have consequences for economic growth.

Low prices for agricultural commodities and food price volatility can be periodical [1] and thus reduce farmer’s household income during periods of glut or harvest time. This is especially for crops for which farmers have low or no storage capacity or lack the possibility to process into other products. In such situations, rural farmers often need to sell their farm produce at harvest time which results in the underpricing of farm products and makes farmers poor [13].

Food price volatility can also lead to political instability [14]. Arezki and Brückner [15] in their study across 120 countries conclude that in low-income countries, when food prices increase, political institutions deteriorate which is likely to increase the probability of civil conflicts and other forms of civil strife, such as anti-government demonstrations and riots. Berazneva and Lee [16] found that some of the political and social unrest in Africa and Asia in 2007 and 2008 was the result of the food price crises at the time. Bellemare [17] and Schneider [18] report that rising food prices have caused political unrest across Africa, Latin America, Asia, and the Middle East.

Food price volatility may also have macroeconomic implications, particularly for imports and inflation. During periods of high food prices, countries often import food to supplement local supply which can lead to increased demand for foreign currency and consequently unfavorable exchange rates. According to Clapp [19], high volatility of food prices can result in the devaluation of the currency of a country. In Sub-Saharan Africa, this risk is aggravated by the fact that most of the countries are net importers of food.

Because of its far-reaching implications, reducing food price volatility has been a core issue in agricultural policy. According to Rashid [20], one of the most widely debated agricultural output price stabilization policy options is dual pricing implemented through buffer stock operations. The dual pricing mechanism involves the setting up of ceiling and floor prices in a commodity market, i.e., a price band, such that the government can intervene when prices are outside of the band/range [21]. Within the band, prices are allowed to fluctuate. Outside the band, the government intervenes by purchasing produce from farmers, usually during the harvesting period, at a fixed price within the band which is higher than the prevailing market price. In addition, during lean periods when prices are high, the government intervenes by selling produce to consumers at a lower price than the prevailing price such that prices (return into) are kept within the band. This policy was adopted by many Asian countries during the Green Revolution. In many African countries, stabilization of agricultural output prices through buffer stock operations tends to focus on grains because of their storability and because they constitute a large share of the food requirements of most rural and urban households.

In Indonesia, the implementation of the buffer stock operation is done by a special food and logistics agency called the Bandan Urusan Logistik (BULO). BULO manages a nationwide set of local agencies at the district level: the Depot Logistics (DOLOGs). The main function of the DOLOGs is to store rice for BULO. BULO procures paddy rice from farmers’ cooperatives as well as from private traders. Individual farmers are encouraged to establish village cooperatives from which rice is purchased to lift the price in rural markets to the floor price. BULO has also been given monopolistic power to import rice when domestic production is low and to export it when is high [22].

In Zambia, the Zambia Food Reserve Agency (FRA) was set up by the government in 1996 to stabilize grain prices. FRA, a parastatal strategic food reserve agency, buys maize at territorially determined prices that typically exceed wholesale market prices in the major maize-producing areas during periods of glut. FRA then stores the maize and at the appropriate time, exports it or sells it domestically. Prices are determined by tender, auction, or administratively. In years when there is a deficit in production, the FRA imports maize which it sells to selected large-scale millers at prices that are below the market prices.

In India, buffer stocks are operated by a food grains management system. The government is the dominant agency in most market operations of grains, such as procurement, storage, transport, and distribution. All these activities are carried out by the Food Corporation of India (FCI), a government agency. However, the procurement price is set by another government agency, the Commission for Agricultural Costs and Prices (CACP).
The CACP bases the price on production costs and includes a return to land and family labor.

In Ghana, the 2008 world food crisis led to food security concerns and brought to the fore persistent hunger [23, 24], agricultural production issues and welfare of the citizens as a result of the drastic increases in prices of some commonly traded foods such as rice, maize, and wheat which rose by 50–75% over a matter of weeks [25]. As a response, Ghana introduced its agricultural output price stabilization policy consisting of buffer stock operations and a dual pricing mechanism. The policy has been in operation for the past 8 years.

This paper seeks to provide an empirical evaluation of this policy in Ghana. The basic question it addresses is: has the volatility, highs and lows, of prices of maize and rice been reduced, following the implementation of buffer stocking and dual pricing? Accordingly, the paper shows that buffer stocking can be applied as a means to ensuring price stabilization in the agricultural food systems in Ghana. The paper presents econometric tests for the efficacy of public buffer stock operations as reflected by reduced price volatility measured by the (corrected) coefficient of variation. The intervention involves government purchasing maize and rice from farmers through agents during periods of glut at a fixed price and re-selling when prices are high.

The structure of the rest of the paper is as follow. “Ghana’s agricultural output price stabilization policy” section presents a description of the agricultural output price stabilization policy implemented in Ghana. “Methods” section presents the data, methods and the empirical results. Summary and conclusions follow in “Conclusion” section.

**Ghana’s agricultural output price stabilization policy**

During the 2007/2008 world food price crisis, the government of Ghana had to import food from the international market as domestic food prices were soaring. In response, in 2009 the current agricultural output price stabilization policy was implemented across the country. The policy seeks to stabilize the prices of two key agricultural outputs: rice and maize. The implementation of the policy saw the setting up of the National Buffer Stock Company (NAFCO). NAFCO was primarily designed to stabilize prices of cereals (maize and rice) by smallholder farmers, particularly to reduce potential annual gluts that characterize maize and rice production. NAFCO purchases excess outputs of these cereals at a fixed (floor) price set by government above the open market price during the glut period and sells them during the lean period. The aim is to insulate farmers against losses, provide them with assured income, help increase their yields and stimulate the expansion of agricultural land and inputs to improve production. It is also to provide consumers with a respite when prices of cereals are getting too high.

The determination of the floor price is done annually by the Ministry of Food and Agriculture (MoFA) through the Post-Harvest Committee. This committee is made up of representatives of farmer associations, NAFCO and MoFA. In determining the floor price, the committee takes into account the cost of production and ensures that a minimum of 15 percent or more profit margin is added. The purchase of maize and rice is made across the different maize growing areas spread across five regions in the country with potential spatial differentials in the cost of production. However, the committee does not take into account local price differentials when setting the floor price [26]. The determination of the ceiling price is done by NAFCO taking into consideration the prevailing open market price. The purchase of the cereals is done by Licensed Buying Agents (LBCs) which are private companies. The LBCs purchase the cereals on behalf of NAFCO at the floor price determined by the Post-Harvest Committee on a commission basis in the various rural communities. For every 100 kg bag of maize purchased from farmers by LBCs and delivered to the NAFCO warehouse, NAFCO pays a percentage of the fixed price to the LBCs as a commission. The percentage of the commission is determined by taking into consideration the distance of the location where the maize was purchased from.

Storage of the cereals is done by NAFCO with the use of the GrainPro Cocoon technology. GrainPro Cocoons are airtight unsupported rectangular structures made of lightweight UV-resistant PVC. The simple two-piece Cocoon consists of a top cover, and a floor piece joined with a PVC tongue and groove zipper similar to those used to close environmental safety suits. Insects trapped in the bagged grain expire in a matter of days as a result of an increase in carbon dioxide and reduction of oxygen. Cocoons are packed folded in a carry bag for transport and can be made ready for use in minutes. The technology can store bagged agricultural outputs such as grains, seeds, cocoa and coffee beans, and others.

Farm price volatility is the extent to which prices of farm produce rise or fall beyond the expectations of consumers and farmers [27]. In other words, it is the unpredictable change in price [28] defined over a specific period [29]. Von Braun and Tadesse [30] define price volatility as the deviation of a price series from its mean. This study focuses on two key agricultural outputs: maize and rice. The methods we adopt in measuring the output price volatility are the coefficient of variation and the corrected coefficient of variation.
Methods

The ADF test
The data analyzed are real monthly wholesale prices for maize and rice reported by the Ministry of Food and Agriculture of Ghana. In preparation for the analyses, we tested each series for the presence of a unit root using the augmented Dickey–Fuller (ADF) test [31]. The ADF test is specified as:

\[
\Delta y_t = \alpha + \beta y_{t-1} + \delta t + \zeta_1 \Delta y_{t-1} + \zeta_2 \Delta y_{t-2} + \ldots + \zeta_k \Delta y_{t-k} + \varepsilon_t
\]

where \( y \) is price, \( t \) is time, \( k \) is the number of lags, \( \beta \) and \( \delta \) are unknown parameters. Since the price series are monthly data, we used \( k = 12 \) lags [32] with a time trend.

The main hypothesis of interest is \( H_0: \beta = 0 \) versus \( H_1: \beta < 0 \). Failure to reject \( H_0 \) indicates that the data series have a unit root and ought to be differenced. The test is one sided. The asymptotic distribution of the usual t-statistic under \( H_0 \) is the augmented Dickey–Fuller distribution with critical values tabulated by Dicky and Fuller [31], Banerjee et al. [33], Cheung and Pascual [34] among others.

When the data are non-stationary, we follow Yang et al. [35] and use the difference of the logarithm of the prices to generate log price series. The log-differenced series is:

\[
P R_t = \ln(P_t) - \ln(P_{t-1})
\]

where \( PR_t \) is the log-differenced series of the price series. With the log series, the CV is estimated in percentages.

Coefficient of variation (CV) and the corrected coefficient of variation (CCV)

Common approaches to measuring price volatility of agricultural outputs or food are the standard deviation and the coefficient of variation (CV) [36, 37]. The CV is the ratio of the standard deviation to the mean defined as [38]:

\[
CV = \frac{\text{standard deviation}}{\text{mean}} = \frac{\sqrt{\sum_{i=1}^{n} (P_i - \bar{P})^2}}{\bar{P}}
\]

where \( \bar{P} \) is the mean of the (in the present case, price) series \( P_t \), and \( n \) is the number of observations or data points.

The CV is a unit-free measure of variability [39] and is expressed as a percentage. It is applied in many disciplines [40]. The CV is also used to compare the variability of two series. Particularly, let period 1 be the policy-off period and period 2 the policy-on period. The hypothesis to be tested is \( CV_1 = CV_2 \) against—in the present case study—the alternative \( CV_1 > CV_2 \), i.e., the CV for the policy-off period (market without buffer stock policy) is greater than the CV for the policy-on period (market with buffer stock policy).

The test statistic \( Z \) is [41]:

\[
Z = \frac{a \left( \frac{S_1^2}{\bar{P}_1} - \frac{S_2^2}{\bar{P}_2} \right)}{\sqrt{\frac{2}{n} \left( \frac{\gamma_1^2}{2} + \frac{\gamma_2^2}{2} \right)}}^{1/2}
\]

where \( a = \sqrt{(n/(n-1))} \) with \( n \) the number of observations, \( S_1 \) and \( S_2 \) the standard deviations of price series 1 and 2, respectively, \( \bar{P}_1 \) and \( \bar{P}_2 \) their means, and \( \gamma_1^2 \) and \( \gamma_2^2 \) their variances. When the number of observation between the series is different, \( n \) is the average number of observations of the two series. The \( Z \)-statistic follows the standard normal distribution. Although we expect a dampening policy effect on price volatility, we nevertheless apply a two-sided test.

Although the CV is widely used, its ability to fully capture price volatility is limited as it assumes that the variance of the price series is constant over time [42]. This particularly so, if there is non-stationarity due to a unit root or random-walk behavior, then the use of the CV can lead to over-estimation of volatility. There are two options to overcome the non-stationarity issue: removing it by differencing, as outlined above, or by applying the corrected coefficient of variation (CCV) to the non-stationary data. The CCV, also called the Cuddy-Della Valle instability index [43–45], is a transformation of the CV. Bediane and Odjo [46] described the CCV as the trend-corrected CV. According to Huchet-Bourdon [38], the CCV is defined as:

\[
CCV = CV \sqrt{1 - R^2}
\]

where \( R^2 \) is the coefficient of determination obtained from the simple linear regression of the price series on time \( t \) using the same period for which the CV is estimated. That is, \( R^2 \) is the coefficient of determination of the time trend regression model [45]:

\[
\log(P) = a + \beta t + \varepsilon \quad \text{where} \ t = 1, 2, \ldots, T
\]

where \( a \) is the constant, \( \beta \) the slope, \( \varepsilon \) the disturbance term and \( T \) denotes the endpoint of the estimation period. Ordinary least square (OLS) can be applied to estimate Eq. (6).

The difference between the CCVs of the price series for the policy-on and policy-off period can be tested by using the approaches developed by Deb and Pramanik [47] and Singh and Byerlee [48] which are based on Kendal and Stewart [49]. In particular, \( Z \) is defined as [50]:

\[
Z = \frac{a \left( \frac{S_1^2}{\bar{P}_1} - \frac{S_2^2}{\bar{P}_2} \right)}{\sqrt{\frac{2}{n} \left( \frac{\gamma_1^2}{2} + \frac{\gamma_2^2}{2} \right)}}^{1/2}
\]
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\[ Z = \frac{(CCV_1 - CCV_2)}{D} \]

where \( D \) is:

\[ D = CCV \left\{ \left[ \left( 1 + 2CCV^2 \right) / 2 \right] \left( 1/n_1 + 1/n_2 \right) \right\}^{0.5} \]

where \( CCV_1 \) and \( CCV_2 \) relate to periods 1 (policy-off) and 2 (policy-off), respectively; \( CCV \) is the \( CCV \) for the parent population, that is, the series involving both the policy-off and policy-on periods; \( n_1 \) and \( n_2 \) are the numbers of observations for period 1 and period 2, respectively. Deb et al. [50] showed that \( Z \) follows a standard normal distribution. As in the case of \( CV \), we apply a two-sided test.

In the application below, we apply both the \( CCV \) to the non-stationary data and the \( CV \) to the detrend data.

**Empirical results**

We analyze real monthly wholesale prices of maize and rice reported by the Ministry of Food and Agriculture of Ghana from three major cereal markets. These markets are Techiman and Tamale where the stabilization policy has been implemented, and Ho where this is not the case. The data cover the period January 2006–April 2015 giving a total of 112 data points. For Techiman and Tamale, the period from January 2006–2010 is the pre-buffer stock operation (policy-off) era and the period January 2011–April 2015 the buffer stock era (policy-on). For the entire period January 2006–April 2015, there are no missing observations. The data are presented in Figs. 1, 2, 3, 4, 5 and 6. For each series, we present two graphs: graphs \( a \) (at the left-hand side) show the (non-stationary) nominal level data and graphs \( b \) (at the right-hand side) the (stationary) data after log differencing.

The outcomes of the test results presented in Table 3 support the results in Table 2. The changes in the \( CVs \) of maize in Techiman and Tamale, and of rice in Tamale are significant at 1%, and that of rice in Techiman at 5%. The changes in the \( CVs \) of both commodities in the Ho market are insignificant. This means that in the Ho market, the volatility has not changed between the two periods. Comparison of the effects for the two crops in the Techiman market shows that the policy had a greater effect on the price volatility of rice (11.80) than on maize price.

The results of the augmented Dicky–Fuller (ADF) test are presented in Table 1.

The results of the ADF unit root test are presented in Table 1. The results indicate that the raw (level) data for all the price series for both maize and rice are non-stationary. After first differencing, all the series are stationary (1% level).

We first compare the policy-on and policy-off periods and markets by way of the \( CV \) for the differenced time (stationary) series (see section on methods). The between periods–within markets \( CVs \) are presented in Table 2 and the corresponding one-tailed test results in Table 3. Table 4 presents the test results for the policy-on periods between markets. Comparisons by way of the \( CCV \) for the raw data are presented in Tables 5 and 6.

The results presented in Tables 2 and 3 indicate that the between period price volatility in the Techiman and Tamale markets declined for both commodities. For instance, in the Techiman market, the \( CV \) of the price of maize declined from 22.10% in period 1, 15.01% in period 2, and of rice from 16.77 to 4.97. For the Ho market, the between period \( CVs \) of both commodities also declined, though marginally only: of maize from 20.93 to 19.05 and of rice from 12.58 to 12.0.

The results presented in Table 4 support the results in Table 2. The changes in the \( CVs \) of maize in Techiman and Tamale, and of rice in Tamale are significant at 1%, and that of rice in Techiman at 5%. The changes in the \( CVs \) of both commodities in the Ho market are insignificant. This means that in the Ho market, the volatility has not changed between the two periods. Comparison of the effects for the two crops in the Techiman market shows that the policy had a greater effect on the price volatility of rice (11.80) than on maize price.

**Fig. 1 a, b** Price of Maize from January 2006–April 2015 in Techiman market
For the Tamale market, the difference is smaller. Table 4 confirms the policy effects found in Tables 2 and 3. For both commodities, the differences in CVs between the Techiman and Tamale markets on the one hand and of the Ho market on the other are significant at 1%, except for maize between Techiman and Ho where it is significant at 5%.

The results of the CCV analyses presented in Tables 5 and 6 are globally in line with the results of CV analyses in Tables 2, 3 and 4. Table 6 shows that there is a significant (1% level) difference in the CCV between the policy-on and policy-off for both crops in the Techiman market. For the Tamale market, the difference is significant (5%) for maize and marginally significant for rice. For the Ho market, the differences are insignificant. (Note that the time regression model (6) which is only needed for the $R^2$ in the CVV, is presented in Appendix 1).

**Discussion**

Overall, the study provides evidence that the buffer stock policy and dual pricing system have had mitigating effects on the price volatility of maize and rice. For both produces, we estimated the correlation coefficients (CV) based on the detrended time series and the corrected correlation coefficients (CCV) estimated on the raw data. For the Techiman market, the differences between the policy-on and the policy-off periods for both rice and
Fig. 4 a, b Price of Rice for January 2006–April 2015 in Tamale market

Fig. 5 a, b Price of Maize for January 2006–April 2015 in Ho market

Fig. 6 a, b Price of Rice for January 2006–April 2015 in Ho market
maize were found to be significant at 1% (in both the CV and CCV test). For the Tamale market, the difference between both periods for maize was significant at 1%, for rice at 5% (CV test); for maize, the difference was significant at 5%, for rice at slightly less than 15% (CCV test). Apparently, the CCV test based on raw data is less conclusive than the CV test based on detrended data. Similar results were found by [44, 46]. For the Ho market without buffer stocking, we failed to reject the hypothesis of no difference between periods 1 and 2 for both rice and maize (both CV and CCV tests). We also compared the Techiman and Tamale markets on the one hand and the Ho market on the other for period 2 for both commodities. The hypotheses of no difference between the policy-on and the policy-off markets for both produces were rejected for both types of markets at 1% (both tests).

The results of the CV estimates provide evidence that the effect of the policy has generally been greater on rice than on maize as indicated by the higher CV changes for rice. Table 3 shows that the CV changes for maize and rice are 7.09 and 11.80% in the Techiman market, respectively, and 13.19 and 15.18% in the Tamale market. The average proportions of maize and rice that enter the buffer stocks yearly are estimated at 7.2 and 12.5%, respectively [26]. With these proportions, the buffer stock operations are able to affect the supply and price of rice more than those of maize.

Table 3 furthermore indicates that the policy has affected the price volatility of both crops in Tamale (CV change for rice: 15.18%, for maize: 13.19%) more than in Techiman where the corresponding percentages are 7.09 and 11.80. The differential policy effects between Tamale (Northern Region) and Techiman (Brong Ahafo Region) could have been affected by the fact that Techiman is centrally located in the country with relatively better roads to the farming communities compared to Tamale which is located in the northern part of the country with more remote farming communities. Due to its better infrastructure and location, imports of both produces can reach the Techiman market more easily than the Tamale market which reduces the effects of buffer stocking in the former. Furthermore, the agro-ecological conditions are more favorable in Techiman than in Tamale for the production of both commodities, especially maize [51]. This means that production volumes in Tamale are subject to more fluctuations than in Techiman with, ceteris paribus, larger policy impacts in the former.

| Table 1 Results of the ADF test |
|---------------------|-----------------|-------------------|
| Market              | Price series     | ADF test          |
|                     |                  | Levels            | First difference |
| Techiman            | Maize            | −2.357            | −5.005*           |
|                     | Rice             | −1.869            | −4.844*           |
| Tamale              | Maize            | −2.380            | −4.379*           |
|                     | Rice             | −2.189            | −5.988*           |
| Ho                  | Maize            | −2.344            | −4.394*           |
|                     | Rice             | −1.485            | −4.200*           |

Asymptotic critical values for unit root t test with linear time trend, 12 lags: 10%: −2.58; 1%: −3.51 (Banerjee et al. [33])

*Significant at 1%. A t statistic larger than the critical value implies rejection of \( H_0 \) and presence of a unit root

**Table 2 Coefficients of variation (CV): between periods–within markets**

<table>
<thead>
<tr>
<th>Period</th>
<th>Market</th>
<th>Crop</th>
<th>Policy-off period CV</th>
<th>Policy-on period CV</th>
<th>Change in CV</th>
<th>Z-statistics</th>
</tr>
</thead>
<tbody>
<tr>
<td>Policy-off period</td>
<td>Techiman</td>
<td>Maize</td>
<td>22.10</td>
<td>15.01</td>
<td>7.09</td>
<td>5.220**</td>
</tr>
<tr>
<td></td>
<td></td>
<td>Rice</td>
<td>16.77</td>
<td>4.97</td>
<td>11.80</td>
<td>9.172**</td>
</tr>
<tr>
<td>Policy-on period</td>
<td>Tamale</td>
<td>Maize</td>
<td>19.47</td>
<td>6.28</td>
<td>13.19</td>
<td>11.488**</td>
</tr>
<tr>
<td></td>
<td></td>
<td>Rice</td>
<td>22.61</td>
<td>7.43</td>
<td>15.18</td>
<td>4.810*</td>
</tr>
<tr>
<td></td>
<td>Ho</td>
<td>Maize</td>
<td>20.93</td>
<td>19.05</td>
<td>1.88</td>
<td>0.114</td>
</tr>
<tr>
<td></td>
<td></td>
<td>Rice</td>
<td>12.58</td>
<td>12.20</td>
<td>0.38</td>
<td>0.025</td>
</tr>
</tbody>
</table>

**Significant at 1%; *Significant at 5%**

*For the Ho market, the first and second periods are not distinguished by policy intervention*
Our result is in line with Bryan's [52] analysis of price volatility in 14 countries following the 2008 food crisis. For instance, the Ethiopian government released its grain stocks directly to consumers through consumers’ associations organized at local levels to help mitigate the effects of the rising food prices at the time. Although the quantity was not sufficient to reduce overall domestic price levels in the entire country, the release significantly lowered local prices in several parts of the country, thus reducing price volatility.

The results are also in line with the findings of David et al. [13] who used the CV approach for food price volatility analysis. The authors found a negative relationship between locally traded volumes of locally produced crops and the volatility of food prices in some a few Eastern African countries such as Kenya, Malawi, Zambia, and Zimbabwe using strategies such as public buffer stock holdings and management. The results imply that countries that have created an economic environment that allows for rapid supply responses to demand tend to be successful in reducing price volatility of staple food crops. The results corroborate the findings of Abbott [53] who reported that after the 2008 food crisis, buffer stock operations that were introduced by countries in Sub-Saharan Africa have helped to reduce the volatility of some staple food crops, especially cereals.

However, the study by Minot [8] raises questions about the effectiveness of food price stabilization programs like buffer stock operations being implemented by large state-owned enterprises in sub-Sahara African countries like Zambia and Zimbabwe, especially for maize. In spite of the operations, price volatility of maize was significantly higher in some of these countries than in other African countries.

### Table 4 One-sided CV tests between markets (policy-on periods only)

<table>
<thead>
<tr>
<th>Crop</th>
<th>Techiman</th>
<th>Tamale</th>
<th>Ho</th>
<th>Difference</th>
<th>Z-Statistics</th>
</tr>
</thead>
<tbody>
<tr>
<td>Maize</td>
<td>15.01</td>
<td>6.28</td>
<td>–</td>
<td>8.73</td>
<td>5.216**</td>
</tr>
<tr>
<td>–</td>
<td>15.01</td>
<td>19.05</td>
<td>4.04</td>
<td>1.690*</td>
<td></td>
</tr>
<tr>
<td>–</td>
<td>6.28</td>
<td>19.05</td>
<td>12.77</td>
<td>5.689**</td>
<td></td>
</tr>
<tr>
<td>Rice</td>
<td>4.79</td>
<td>7.43</td>
<td>–</td>
<td>2.64</td>
<td>0.535</td>
</tr>
<tr>
<td>–</td>
<td>4.79</td>
<td>12.20</td>
<td>7.41</td>
<td>5.195**</td>
<td></td>
</tr>
<tr>
<td>–</td>
<td>7.43</td>
<td>12.20</td>
<td>4.77</td>
<td>2.811**</td>
<td></td>
</tr>
</tbody>
</table>

### Table 5 Corrected coefficients of variation (CCV) of raw data: between periods–within markets

<table>
<thead>
<tr>
<th>Market</th>
<th>Crop</th>
<th>Techiman market</th>
<th>Tamale market</th>
<th>Ho market</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td>Maize</td>
<td>Rice</td>
<td>Maize</td>
</tr>
<tr>
<td></td>
<td>Coefficient</td>
<td>0.451</td>
<td>0.404</td>
<td>0.067</td>
</tr>
<tr>
<td></td>
<td>S.E</td>
<td>0.067</td>
<td>0.060</td>
<td>0.067</td>
</tr>
<tr>
<td></td>
<td>R²</td>
<td>0.435</td>
<td>0.438</td>
<td>0.497</td>
</tr>
<tr>
<td></td>
<td>CCV</td>
<td>0.319</td>
<td>0.223</td>
<td>0.287</td>
</tr>
<tr>
<td></td>
<td>Coefficient</td>
<td>– 0.066</td>
<td>0.740</td>
<td>0.921</td>
</tr>
<tr>
<td></td>
<td>S.E</td>
<td>0.077</td>
<td>0.068</td>
<td>0.131</td>
</tr>
<tr>
<td></td>
<td>R²</td>
<td>0.014</td>
<td>0.703</td>
<td>0.499</td>
</tr>
<tr>
<td></td>
<td>CCV</td>
<td>0.148</td>
<td>0.095</td>
<td>0.219</td>
</tr>
</tbody>
</table>

**Significant at 1%; *Significant at 5%**

### Table 6 One-sided tests of CCV₁ > CCV₂ of raw data: between periods–within markets

<table>
<thead>
<tr>
<th>Market</th>
<th>Crop</th>
<th>Policy-off CCV</th>
<th>Policy-on CCV</th>
<th>Change in CCV</th>
<th>Z-statistics</th>
<th>P value</th>
</tr>
</thead>
<tbody>
<tr>
<td>Techiman</td>
<td>Maize</td>
<td>0.319</td>
<td>0.148</td>
<td>0.171</td>
<td>5.191**</td>
<td>0.000</td>
</tr>
<tr>
<td>Techiman</td>
<td>Rice</td>
<td>0.223</td>
<td>0.095</td>
<td>0.128</td>
<td>5.863**</td>
<td>0.000</td>
</tr>
<tr>
<td>Tamale</td>
<td>Maize</td>
<td>0.287</td>
<td>0.219</td>
<td>0.068</td>
<td>1.889*</td>
<td>0.029</td>
</tr>
<tr>
<td>Tamale</td>
<td>Rice</td>
<td>0.269</td>
<td>0.311</td>
<td>0.042</td>
<td>1.520</td>
<td>0.064</td>
</tr>
<tr>
<td>Ho</td>
<td>Maize</td>
<td>0.290</td>
<td>0.302</td>
<td>0.012</td>
<td>0.279</td>
<td>0.780</td>
</tr>
<tr>
<td>Ho</td>
<td>Rice</td>
<td>0.145</td>
<td>0.168</td>
<td>0.023</td>
<td>1.071</td>
<td>0.285</td>
</tr>
</tbody>
</table>

**Significant at 1%; *Significant at 5%**
countries with little or no price stabilization programs or policies for maize. This could be because the former countries had inherently more unstable food markets [8].

Conclusion
The objective of this paper was to analyze the effects of buffer stock operations on the price volatility of maize and rice in some selected markets in Ghana. Based on times series data obtained from the Ministry of Food and Agriculture spanning the period from January 2006 to April 2015, we employed the (corrected) coefficient of variation as measures of volatility. The results show that price volatility of maize and rice was stabilized in the markets where the output price stabilization policy was implemented. For the same crops, price volatility was persistent in a market where the policy was absent. With these findings, we conclude that the buffer stock operations have been quite successful to control the price volatility of maize and rice. The findings of this study provide an appeal for scaling up of the policy to other parts of the country, especially remote areas, where maize and rice production are intensive, yet access to these areas by buyers are challenging.

A crucial condition for the policy to be effective is the proportion of the production volume that enters into buffer stock. Too small a proportion renders the policy ineffective. Benin et al. [26] argued that the proportion is critical for the policy to be effective and sustainable. As a rule of thumb, the authors recommend about 27% for Ghana.

An important side effect of the buffer stock operations is that they provide traders, individual farmers and their organizations with secure, and reliable storage which is crucial for the quality and price of the produce sold when prices are high. Moreover, reliable and safe storage provides farmers with credible title to their produce which enables them to obtain credit for their activities from the government but also from the private sector [54–56]. Reliable storage creates a warehouse receipt system (WRS) that enables smallholder farmers to access credit due to improved security for loan recovery.

The study also highlights the importance of buffer stocks as a tool for implementing crop insurance programs [57], especially with the involvement of the private sector. To hedge farmers against the risks posed by adverse weather conditions, destruction by pests such as armyworms and disease outbreaks which have the potential to cause farmers huge losses, weather index-based crop insurance programs may be facilitated with buffer stock providing storage systems. If farmers have access to market based-insurance tools, variations in the production and prices may not require immediate policy response by government unless characterized by catastrophic events. Moreover, with the involvement of the private sector, farmers will have different packages of insurance programs to choose from to minimize risk. It is expected that with stable output prices, the income of farmers will improve and stimulate smallholder farmers to invest into increasing their farm outputs [58].

Finally, the results of the study lend insight into directions for future research, particularly to find out the extent to which the gains made through the stabilizations of prices of maize and rice affect the welfare of the smallholder farmers and their families including their health. It is after this that more comprehensive conclusions can be drawn on the real impacts of the policy-on smallholder farming in Ghana.

Abbreviations

Authors’ contributions
This paper was conceived and developed by EA with significant guidelines and intellectual contributions from HF and KFA. All authors read and approved the final manuscript.

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Competing interests
The authors declare that they have no competing interests.

Availability of data and materials
The data that support the findings of this study are available from the Statistical, Research and Information Division (SRID) of the Ministry of Food and Agriculture, Ghana. Data are, however, available from the authors upon reasonable request.

Consent for publication
Not applicable.

Ethics approval and consent to participate
Not applicable.

Funding
Not applicable.

Appendix 1
See Table 7.
Table 7 The regressions underlying the CCV calculations

<table>
<thead>
<tr>
<th>Market</th>
<th>Crop</th>
<th>Constant</th>
<th>Coeff.</th>
<th>( R^2 )</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Policy-off period</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Techiman</td>
<td>Maize</td>
<td>14.322***</td>
<td>0.451***</td>
<td>0.435</td>
</tr>
<tr>
<td></td>
<td></td>
<td>(2.366)</td>
<td>(0.067)</td>
<td></td>
</tr>
<tr>
<td></td>
<td>Rice</td>
<td>23.430***</td>
<td>0.404***</td>
<td>0.438</td>
</tr>
<tr>
<td></td>
<td></td>
<td>(2.106)</td>
<td>(0.060)</td>
<td></td>
</tr>
<tr>
<td>Tamale</td>
<td>Maize</td>
<td>15.522***</td>
<td>0.506***</td>
<td>0.497</td>
</tr>
<tr>
<td></td>
<td></td>
<td>(2.342)</td>
<td>(0.067)</td>
<td></td>
</tr>
<tr>
<td></td>
<td>Rice</td>
<td>20.003***</td>
<td>0.319***</td>
<td>0.326</td>
</tr>
<tr>
<td></td>
<td></td>
<td>(2.111)</td>
<td>(0.060)</td>
<td></td>
</tr>
<tr>
<td>Ho</td>
<td>Maize</td>
<td>22.246***</td>
<td>0.599***</td>
<td>0.442</td>
</tr>
<tr>
<td></td>
<td></td>
<td>(3.079)</td>
<td>(0.088)</td>
<td></td>
</tr>
<tr>
<td></td>
<td>Rice</td>
<td>29.440***</td>
<td>0.576***</td>
<td>0.686</td>
</tr>
<tr>
<td></td>
<td></td>
<td>(1.794)</td>
<td>(0.051)</td>
<td></td>
</tr>
<tr>
<td><strong>Policy-on period</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Techiman</td>
<td>Maize</td>
<td>57.856***</td>
<td>-0.066***</td>
<td>0.014</td>
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<td></td>
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<td>(2.360)</td>
<td>(0.077)</td>
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</tr>
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<td>Rice</td>
<td>57.192***</td>
<td>0.740***</td>
<td>0.703</td>
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<td></td>
<td></td>
<td>(2.071)</td>
<td>(0.068)</td>
<td></td>
</tr>
<tr>
<td>Tamale</td>
<td>Maize</td>
<td>39.593***</td>
<td>0.921***</td>
<td>0.499</td>
</tr>
<tr>
<td></td>
<td></td>
<td>(3.976)</td>
<td>(0.131)</td>
<td></td>
</tr>
<tr>
<td></td>
<td>Rice</td>
<td>19.224**</td>
<td>2.555***</td>
<td>0.672</td>
</tr>
<tr>
<td></td>
<td></td>
<td>(7.698)</td>
<td>(0.252)</td>
<td></td>
</tr>
<tr>
<td>Ho</td>
<td>Maize</td>
<td>49.206***</td>
<td>1.734***</td>
<td>0.472</td>
</tr>
<tr>
<td></td>
<td></td>
<td>(8.083)</td>
<td>(0.260)</td>
<td></td>
</tr>
<tr>
<td></td>
<td>Rice</td>
<td>65.926***</td>
<td>1.908***</td>
<td>0.632</td>
</tr>
<tr>
<td></td>
<td></td>
<td>(6.267)</td>
<td>(0.206)</td>
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</tr>
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</table>

Price series as the dependent variable and time trend as the independent variable

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