Effects of Food Price Shocks on Dietary Composition of Farm Households’ in Nigeria
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Adekunle, Chioma Patricia
Department of Agricultural Economics and Farm management, Federal University of Agriculture, Abeokuta, Nigeria. Email: chiomaadekunle@gmail.com; phone: +2348060174180

Kutu, Augustine Adebayo
School of Accounting, Economics & Finance, University of KwaZulu-Natal, South Africa. Corresponding Author Email: kutuA@ukzn.ac.za; phone: +27712387450

Alori, David Alaba
Department of Agricultural Economics & Extension, Federal University of Agriculture, P.M.B 65, Minna, Nigeria. Email: david_alori@yahoo.com; phone: +2348060095870

Abstract

This study examined the effects of food price spikes on the quantity and quality of the dietary composition of farm households in Nigeria using the 2010/2011, 2012/2013 and 2015/2016 household survey panel data. The fixed effects models were estimated while controlling for participation in non-farm livelihood activities. Analysis indicated that seasonal comparisons of the average per capita daily calorie intake is lowest in the post-harvest season of 2011 (2511.44 kilocalories), which is higher than the average recommended intake. Results showed that rural households had lower per capita calorie intake and dietary diversity than urban households, which may be an indication of a shift in the calorie inadequacy from urban to rural farm households’ setting in Nigeria. In addition, income and education improvement are crucial for raising food calories and satisfaction of hunger needs among households with total food consumption expenditure of 79.0%. The study recommended that a combination of policy strategies, including income growth, agricultural development and targeted food distribution programs could reduce the problems of inadequate calorie consumption among farm households.

Keywords: Food price increase, household, nutrition, dietary diversity

Introduction

The 2007/2008 dramatic global food price upsurges and the recovery of 2010/2011 have kept the welfare effects of food price shocks at the center of policy discussions across the world. The effects of the food price shocks on household’s nutrition
motivated additional discussions in this regard. When food prices increase, poorest households, that spend a large share of their income on food, are forced to reduce the quality and the quantities of food consumed leading to increased food insecurity at the household level (Headey and Martin, 2016). Studies have identified food shortage as a major consequence of food price inflation and price volatility (Jolliffe, Seff and De La Fuente, 2018). However, if the situation is coupled with the decline in global economic conditions, the world agricultural prices will continue to maintain an upward and volatile trend thereby exerting considerable hardship on the poor. For instance, Food and Agriculture Organization (2017) estimates that of the 7.6 billion people in the world in 2016, about 815 million people or 10.7% experience hunger. About 705 million people were estimated to be living in extreme poverty as a result of rising food prices (Roser and Ortiz-Ospina, 2017).

Food inflation in Nigeria has witnessed dramatic changes in recent years, reaching a peak of 20.32% in September 2017 (CBN, 2018). In Nigeria, food price inflation is a critical economic problem posing threats to the poor who constitute the large proportion of the population. The average household in Nigerian spends about 73% of their income on food and beverage (Onyewuchi, 2016). Similarly, recent figures reported by Egwuma, Ojeleye and Adeola (2019) show that Nigerians spend 58.9% of their household income on food. This figure, which represents the highest in the world, implies that Nigerians spend over half of their income on food. This situation is made worse in the light of recent economic deterioration which has placed Nigeria as the country with the largest number of extremely poor people, with about 87 million people living in extreme poverty and six people becoming extremely poor every minute (Kharas, Hamel and Hofer 2018). Thus, the ability of the poor households to hedge against inflation especially regarding agricultural commodities is significantly hampered, leading to productivity losses and severe malnutrition (Rehman and Khan 2015).

Juarez-Torres (2015) posited that the burden of food price increases is borne more by the poor and vulnerable households who spend up to 80% or more of their earnings on foodstuffs. When households face massive negative price or income shocks, reduction in food budget is often their most immediate response (Baker, 2015). This manifests in compromised dietary intakes in terms of quantity and quality, which ultimately engender higher vulnerability to food insecurity, malnutrition, poverty and related issues. Available statistics on malnutrition in Nigeria show that incidences of stunting among under-five children in the country are approximately 32% (Akombi, 2017), with the situation of hunger in the country still being rated as “serious” from international standpoint (Von Grebmer, Bernstein, Prasai, Yin, and Yohannes, 2015). Previous studies have placed the incidence of household food insecurity in the country between 49% and 78% based
on food calorie intake (Abbasi, Ghoochani, Ghanian and Kitterlin, 2016). It has also been found that a large proportion of households in the country consume less varied and low-quality foods (Agada and Igbokwe, 2015 and Akerele, 2015). The fundamental challenges to human welfare and economic growth in Africa is food and nutrition security especially with the recent spikes of food prices, a situation that is making the world to be facing a worsening food crisis unseen in the last 30 years and having the potential of leading to catastrophe (Juarez-Torres, 2015 and Grabowski and Self, 2016).

The role of economic shocks, particularly as price shocks however, is only analyzed in a few studies that mostly compare differences in food and nutrition security indicators (Kalkuhl, 2016). There have been a few attempts to estimate the impacts of the 2007/08 price changes on food and nutrition security. Based on the FAO approach to calorie deficit consumption, Bühler, Grote, Hartje, Ker, Lam, Nguyen and Tong (2015) calculated the livelihood of an additional 63 million individuals that became undernourished as a result of global food price spikes. Also, Yu and Shimokawa (2016) analyzed the impact of price increases on the number of calorie deficit people in African countries. They found that in most of the considered countries (including Nigeria), calorie deficiency increased for the poorest households, both in rural and urban areas.

Alem (2015) investigated which types of households in urban Ethiopia changed their food consumption significantly in 2008 when food prices were very high by investigating changes in consumption over three survey rounds in 2000, 2004 and 2008. They found that the high food price inflation was the most adverse economic shock between 2004 and 2008, and that a significant proportion of households adjusted food consumption in response to the price shocks. Their results indicated that households with low asset levels and casual workers were particularly adversely affected by high food prices.

In Nigeria, a few studies have been done in this area. Shittu, Akerele, and Haile (2018) examined the welfare effects of food price spikes among households in the country. They found that higher spike in the price of cereals consistently has negative effect on food quantities (including calories) consumed, dietary diversity, and economic welfare of households, spikes of price of other staples such as animal proteins, fats and oils, fruits and vegetables exert heterogeneous influence. This study is related to the Shittu, Akerele, and Haile (2018) but differs in the number of waves used in the panel data and the food groups by disaggregating cereals into three groups which were rice, wheat and other cereals. This is because the country has a structural deficit for these two grains, rice and wheat. Besides, rice and wheat accounted for a larger mean value in cereal consumption expenditure. Adekunle (2020) examined the welfare effects of food price changes on farm households in Nigeria and found a larger share (79%) of farm households as net food buyers and cereal was identified as food for which the households were most vulnerable to price shocks. When adjustments are allowed,
households can adapt their consumption and production patterns resulting in lower deteriorations in welfare with significant differences across quintiles.

Conversely, little is known about the potential impacts of food price spikes on nutritional outcomes of households in Nigeria. Such information is crucial for the development of policies and programmes targeted at the improvement of well-being of households in the country. The highlighted concerns provide the motivation for this study. The study therefore seeks to address the following specific questions: How do food commodity price spikes affect the nutritional outcomes of households in Nigeria? Findings from this study can provide useful information for redesigning existing policy actions and programmes or for the introduction of new ones for improved living conditions of farm households in Nigeria.

This study becomes very important by the fact that the Sustainable Development Goals (SDGs) adopt nutritional status as one of the key indicators of poverty and hunger. It serves as the first step in recognizing that policies, programmes and processes to improve nutritional outcomes have a role to play in global development. Nutritional assessment in the community is essential for accurate planning and implementation of intervention programmes to reduce morbidity and mortality associated with undernutrition (Ghosh-Jerath, Singh, Lyngdoh, Magsumbol, Kamboj and Goldberg, 2018).

The rest of the article is as follows. The next section describes the data and presents the empirical model. Section 3 presents the results and discussion and section 4 concludes the study.

**Methodology**

This study was conducted in Nigeria using two types of quantitative data: household level panel consumption data from the general household survey (GHS) for 2010/2011, 2012/2013 and 2015/2016 post-planting and post-harvest agricultural seasons. The World Bank/National Bureau of Statistics (NBS), Nigeria, collected the data. The panel survey was targeted to cover a total of 5,000 households selected from rural and urban areas of the 36 states of the country. The data covered different aspects of household livelihoods. Parts of the data that are relevant to this study are data on socioeconomic characteristics of the household and household head namely household size, age, sex, marital status, education of household head, location (rural-urban), season (post-planting or post-harvest seasons), whether or not a household engages in agriculture as main source of income.
The second data set were price data, which are monthly time series (2010:1 – 2016:12). The retail commodity prices across the panel of 36 states and the Federal Capital Territory obtained from National Bureau of Statistics Headquarter office in Abuja. The food items are imported rice, local rice, maize, sorghum, millet, beef (meat), fish, egg, yam, garri, beans, and palm oil. These specific food items are very critical to household food security in the country, as they constitute important components of household diets.

In order to construct a measure of dietary diversity, food items were grouped into fourteen (14) which is in line with Hirvonen, Taffesse and Hassen (2016). The study follows a more conventional approach to eliminate bias and exclude all observations within each commodity group with values below and above the 1st and the 99th percentiles. However, only 23,039 households with the complete set of information required after data cleaning and appearing in all the three rounds of data collection were included in this study. Hence, the final panel was made up of 23,039 observations, consisting of data collected from 3921 households in 2010, 3886 households in 2011, 3869 households in 2012, 3959 households in 2013, 3734 households in 2015 and 3670 households in 2016 respectively.

**Variables Measurement**

**Index of food consumption**: Expenditure weighted food price index, and all items price index were required to divide the value of food consumed and the total household expenditure respectively. Consequently, Tornqvist-Theil price index (which is an expenditure-weighted price index) was computed for food as well as all items (food and non-food items). Doing this could better adjust for the possible varying effects that changes in prices might have on household food consumption over time and across locations. The Tornqvist-Theil index is a superlative index that utilizes expenditure and price data from the two domains, i.e. base location/time period and new location/time period. As mentioned earlier, food items were classified into fourteen groups and non-food items were classified into 2 groups. This makes a total of sixteen groups for all items (food and non-food items).

Two key dependent variables used in this study were the quantities of food calories consumed and dietary diversity.

**Calorie intake estimation**: Data on the quantities of food consumed were standardized per kilogramme using the conversion factors of local units to the standard unit provided in the Living Standard Measurement Study (LSMS) survey manual. The calorie content
of each food was obtained by multiplying food quantity by its calorie conversion factor (per kg). According to Rischke, (2015), the per capita calorie consumption of household ‘j’ in time ‘t’ is estimated given the specified formula:

\[ CI_{it} = \sum_{k=1}^{n} \frac{B_{kit}A_{k}}{Z_{it}} \]  

(1)

where \( CI_{it} \) is the per capita daily quantity (amount) of calorie consumed by household ‘i’ at time t, \( B_{kit} \) is weight in kilogramme of food item ‘k’ consumed by household i at time t, and \( A_{k} \) is the standardized nutrient content per kilogramme of food commodity k. \( Z_{it} \) is the number of people in household ‘i’ at time ‘t’. In line with Harttgen, Klasen and Rischke (2016), households with per capita daily calorie consumption below 500 and above 12000 kilocalories were removed as outliers.

**Food consumption dietary diversity:** The Dietary Diversity Score (DDS) estimates the rate at which the variety of food consumed by households differs in terms of nutrient intakes over a given period. The DDS suggests that households gain satisfaction on food consumed not only because “more is better”, but because ‘variety’ is the spice of life. According to Ogundari (2017), higher DDS implies that a household is consuming a diet that is diverse, and whose quality is sufficient to supply important micronutrients. According to Obwanga, Rurangwa, van Duijn, Soma and Kilelu (2018), DDS represents the food utilization dimension of the food security. Desiere, D’Haese and Niragira (2015) also referred to DDS as an effective indicator of food utilization for two reasons. First, it reflects consumption of both macro- and micronutrients. Second, demand theory suggests that individuals or households that allocate their resources towards higher-quality foods when they have satisfied their basic caloric needs attain a higher level of utility. However, a major limitation of DDS is that it fails to account for the quantity of nutrient consumed by the households.

Household demand for Dietary Diversity (DD) has always been modeled in the literature with either DDS, which represents count data, or DDI that represents indices. For example, Vadiveloo, Parkeh and Mattei (2015) and Kim, Shin, Guevarra, Lee, Kim, Seol and Isaacson (2017) employed DDS, while Freire and Rudkin (2019) utilized DDI in their respective studies. Similar to the work of Freire and Rudkin (2019), the study constructed the DDI from household food expenditure, which has been aggregated across food groups using Berry index as follows:
where DDI is as defined earlier and ranges between 0 and 1, with a value towards 1 implying higher dietary diversity and $S_{git}$ is the food consumption share of food group ‘g’ in the total value of food consumed by household ‘i’ in time ‘t’ and ‘F’ is the total number of food groups (Tian and Yu, 2015).

**Food price spikes:** The study constructed the expenditure weighted price spikes for the fourteen (14) aggregated food groups. First, a measure of price spike was constructed (for each food group) using the log return of the food group prices, across specific states, months and years that correspond to the household panel survey. The log return of the price of each food group was estimated for each of the months in each of the years covering January, 2010 to December, 2016. Following Tadasse, Algieri, Kalkuhl and Von Braun (2016), the price spike is specified as follow in a panel model:

$$Q_{imt} = \ln\left(\frac{P_{imt}}{P_{imt-1}}\right) = \ln(P_{imt}) - \ln(P_{imt-1})$$  \hspace{1cm} (3)

Where $P_{imt}$ is the average/proxy price for a food group in month ‘m’ of year ‘t’ in a given state of the country. $P_{imt-1}$ is the average/proxy price of the food group in the preceding month in the same state and year. Thereafter, each estimated spike ($Q_{imt}$) was weighted by the share of each food group in the total value of food consumed by each household as follows:

$$D_{ijt} = S_{ijt} * R_{mt}$$  \hspace{1cm} (4)

Where $D_{ijt}$ the expenditure is weighed spike which better reflect the variation in price spikes across households over time. Data on food price spikes are merged (mapped) with the household panel data by state, year, and month of data collection.

**Estimation model:** To measure the impacts of price spikes on the quantity of food calories consumed and dietary diversity the quantity of food calories consumed and dietary diversity of the household at the individual level was specified as a function of the price spikes, as well as household and individual characteristics. The model specification of the quantity of food calories consumed and dietary diversity of household $i$, at year $t$, denoted by $Y_{it}$, as follows:

$$DDI_{it} = 1 - \sum_{g=1}^{F} S_{git}^2$$  \hspace{1cm} (2)
\[ Y_{it} = \beta_0 + \beta_1 P_{it} + \beta_2 X_{it} + \mu_t + \eta_t + \varepsilon_{it} \]  

(5)

Where \( Y_{it} \) is a measure of the logarithmic value of dimensions of household nutritional outcomes (the quantity of food calories consumed and dietary diversity), \( P \) is a vector of price spikes variables, \( X_{it} \) is a vector of variables of households’ characteristics, \( \mu_t \) represents the time-invariant household’s fixed effects (such as for example eating habits or food preferences), \( \eta_t \) represents the year effects, and \( \varepsilon_{it} \) is the error term. This study assume that households’ fixed effects can be captured by a separate constant for each household: the use of time-invariant household fixed effects is necessary to remove unobserved time-invariant factors at the household level. The failure to control for these household-specific attributes will produce omitted variable bias if the omitted factors are correlated with observed covariates.

The key control variables, other relevant explanatory variables were operationalized as follows: natural log of non-food price (computed from the non-food price); surplus season dummy (1 if data were collected from household in the harvest/post-harvest season, 0 if data were collected during lean/post-planting season); Urban dummy (1 if household is located in urban area, zero, otherwise); sex of household head dummy (1 if household head is a male, 0 otherwise); marital status of household head (1 if household head is married, 0 otherwise); age of household head (years); household headed by master/PhD degree holder (1 if household head has Master/PhD degree, 0 otherwise); household headed by First Degree holder (1 if highest formal educational attainment of household head is Higher National Diploma or First degree, 0 otherwise); household headed (1 if highest formal educational attainment of household head is ND/NCE holder, 0 otherwise); household headed by secondary school certificate holder (1 if highest formal educational attainment of household head is secondary school, 0 otherwise); household headed by primary (1 if highest formal educational attainment of household head is primary school, 0 otherwise) and income status of the household.

There were three dummy variables for the income status, reflecting relatively low, middle and high income status. For the middle income status dummy, household is assigned 1 if the household classified as belonging to middle income, otherwise zero. Likewise, for the high-income status group, household is assigned if the household is classified as having relatively high income, otherwise zero. For the dummy variables on education, the “no formal education household head group” was dropped, while for the income status dummies, the relatively “low income household group” was dropped from analysis. Participation in non-farm livelihood activities and proportion of households with under five year children. There are six dummy variables for the geo-political zone as also included in the analysis.
Results and Discussion

Some Dimensions of Household Nutritional Outcomes

Measures of nutritional outcomes were a measure of dietary diversity (a count of food groups covered by a household’s reported consumption during the past week), and the number of calories consumed per Adult Equivalent Unit (AEU) over the past week. According to Akerele, Sanusi, Fadare and Ashaolu (2017), dietary diversity is a good predictor of child nutrition in particular as it captures the quality of the diet regarding micronutrients in contrast to using calorie consumption.

The amount of calories consumed are used as measures of the quantity dimension of food security (access to food). Dietary diversity is used as a proxy of the quality of diets. (Table 1). Food consumption, in terms of kcal/person/day, is the key variable used for measuring and evaluating the evolution of the world food situation. With respect to amount of calorie consumption, the observed pattern across seasons over the years covered by the study is that the average per capita daily calorie consumption is consistently higher during the post-harvest seasons than in the post-planting seasons for the three waves. The amount of dietary diversity seems to be higher in the first waves compared to the other two waves. Notwithstanding, the metric of dietary diversity for the post-harvest season (year 2011) is slightly higher (0.78) than that (0.76) of post-planting (2010) season.

The somewhat high dietary diversity recorded during the post-harvest period (of year 2011), might be associated with relatively higher real value of food consumed during the same season. This is because the quantity of food calories consumed during the period (season) is lower than the quantity consumed in the other periods. In the study of Akerele (2015), the direct correlation between consumption of more varied foods and likelihood of adequate intakes of food calories and nutrients in Nigeria is documented.

Seasonal comparisons of the average per capita daily calorie consumption indicated that it is lowest in the post-harvest season of 2011 (2511.44 kilocalories), this amount is higher than the average per capita daily calorie intake (2428 kilocalories) reported by Sibhatu and Qaim (2017) and lower than that the average per capita daily calorie intake (2936.99 kilocalories) of Shittu, Akerele, and Haile (2018). This is an evidence of dynamics in calorie supply/consumption in Nigeria in recent times compared to the past years. Likewise, the estimated calorie consumption (2511.44 kilocalories) per capita daily calorie consumed is still higher than the average recommended daily per capita calorie of 2500 kilocalories for developing countries (FAO, 2017). This means that on the average, household calorie consumption may still be adequate and enjoying higher level of dietary diversity during the period (2011) compared to other seasons (periods).
### Table 1: Dimensions of household nutritional outcome

| Year 2010 (Post-Planting Season) | Mean   | Standard Deviation | Minimum | Maximum |
|----------------------------------|--------|--------------------|---------|---------|
| Daily Per capita calorie consumed | 2814.42| 2208.17            | 484.04  | 15854.11|
| Dietary Diversity                | 0.76   | 0.14               | 0.00    | 0.84    |
| Weekly Real per capita Value of Food consumed | 897.43 | 816.79            | 42.18   | 16482.22|
| Weekly Real per capita total expenditure | 2112.02 | 5284.06          | 48.01   | 68112.28|
|                                  |        |                    |         |         |
| Total number of households in 2010 = 3921 |       |                    |         |         |

| Year 2011 (Post-Harvest Seasons) | Mean   | Standard Deviation | Minimum | Maximum |
|----------------------------------|--------|--------------------|---------|---------|
| Daily Per capita calorie consumed | 2511.44| 2444.36            | 482.08  | 14632.16|
| Dietary Diversity                | 0.78   | 0.14               | 0.00    | 0.84    |
| Weekly Real per capita Value of Food consumed | 1158.54 | 883.72            | 51.28   | 19912.34|
| Weekly Real per capita total expenditure | 2153.32 | 4846.91          | 55.04   | 56332.82|
|                                  |        |                    |         |         |
| Total number of households in 2012 = 3886 |       |                    |         |         |

| Year 2012 (Post-Planting Season) | Mean   | Standard Deviation | Minimum | Maximum |
|----------------------------------|--------|--------------------|---------|---------|
| Daily Per capita calorie consumed | 2919.16| 2172.28            | 612.44  | 12696.01|
| Dietary Diversity                | 0.70   | 0.15               | 0.00    | 0.82    |
| Weekly Real per capita Value of Food consumed | 1021.76 | 1005.04          | 44.26   | 16129.38|
| Weekly Real per capita total expenditure | 2562.82 | 5134.24          | 47.18   | 55199.06|
|                                  |        |                    |         |         |
| Total number of households in 2012 = 3869 |       |                    |         |         |

| Year 2013 (Post-Harvest Seasons) | Mean   | Standard Deviation | Minimum | Maximum |
|----------------------------------|--------|--------------------|---------|---------|
| Daily Per capita calorie consumed | 2886.18| 2242.24            | 498.19  | 13441.36|
| Dietary Diversity                | 0.73   | 0.15               | 0.00    | 0.82    |
| Weekly Real per capita Value of Food consumed | 930.13 | 911.26            | 44.43   | 16672.65|
| Weekly Real per capita total expenditure | 2041.93 | 5152.43          | 49.16   | 56331.63|
|                                  |        |                    |         |         |
| Total number of households in 2013 = 3959 |       |                    |         |         |

| Year 2015 (Post-Planting Season) | Mean   | Standard Deviation | Minimum | Maximum |
|----------------------------------|--------|--------------------|---------|---------|
| Daily Per capita calorie consumed | 2922.15| 2246.43            | 508.47  | 12655.68|
| Dietary Diversity                | 0.72   | 0.16               | 0.00    | 0.82    |
| Weekly Real per capita Value of Food consumed | 1004.83 | 963.66            | 51.13   | 15322.67|
| Weekly Real per capita total expenditure | 1593.53 | 2217.01          | 52.28   | 61563.45|
|                                  |        |                    |         |         |
| Total number of households in 2015 = 3959 |       |                    |         |         |

| Year 2016 (Post-Harvest Season) | Mean   | Standard Deviation | Minimum | Maximum |
|----------------------------------|--------|--------------------|---------|---------|
| Daily Per capita calorie consumed | 2651.96| 2142.52            | 511.17  | 12882.13|
| Dietary Diversity                | 0.64   | 0.16               | 0.00    | 0.82    |
| Weekly Real per capita Value of Food consumed | 954.24 | 913.44            | 50.56   | 14849.75|
| Weekly Real per capita total expenditure | 2435.64 | 4490.45          | 53.32   | 69112.23|
|                                  |        |                    |         |         |
| Total number of households in 2016 = 3670 |       |                    |         |         |

Note: The percent of households with zero (0) values for dietary diversity are negligible.
Effects of Food price spikes on Real Value of Food, Dietary Diversity and Calorie Consumed

In Table 2, results of the influence of food price spike and other control variables on the real value of household and per capita calorie and dietary diversity are presented. Higher spikes in the price of rice, wheat, fats and oils, vegetables, eggs, fish, sweeteners, other foods among others, have significant reduction effects on per capita calorie household food consumption, whereas greater spikes in the price of pulses, fruits, can substantially increase per capita calorie food consumption.

Although higher spikes in the price of some foods such as other cereals, pulses, fish may affect the real value of food consumed, this may not necessarily reflect in improved diets as extreme food price shocks may constrain poor people (households) to shift to less-varied diets, which could have a harmful effect on their nutritional status in the short and long run (Weber, 2015). Findings suggest that higher spike in the price of meat is unlikely to influence the real value of food and calorie consumed whereas, the spikes in the price of fish influence the real value of food consumed due to improved diet. This can be attributed to cost (meat is more expensive) of meat compared to fish.

The results suggest that increase in general price of non-food items may have positive effects on the real value of food consumed but a negative effect on both the calorie consumed and dietary diversity. For households who are not into sales of non-food items, higher non-food prices is expected to diminish the purchasing power of the household. However, consumption of food may increase, particularly if household can no longer afford consumption of some non-food commodities due to reduced income as noted by Kalkuhl (2016).

The results of the effects of food price spikes and related factors on food consumption variety revealed that higher spikes in the price of wheat, other cereals, pulses, fruits, vegetables, meat, milk/dairy products, sweeteners, other foods enhance food consumption diversity while higher price in the price of rice, roots and tubers, fat and oil, eggs, and fish will reduce food consumption variety. Greater spikes in the price of roots and tubers, rice, and egg had a negative and significant relationship with household dietary diversity. However, the price spike of meat and beans/pulses established a positive and significant relationship with household dietary diversity.

On the basis of the estimated coefficients, the results suggest that food consumption diversity is more sensitive to changes in the price of roots and tubers than cereals and eggs. Roots and tubers and rice are generally calorie-rich foods which are needed to meet hunger needs of the people. Hunger satisfaction (through food calories) is arguably a fundamental (food) need, and an average household would first seek to gratify this before fulfilling other food nutrient needs such as proteins and vitamins.
Besides, rice, root and tubers accounted for 15.62% and 12.59% respectively in total food consumption expenditure of 79.0%. This implies that households are still generally less willing to trade-off food diversity (quality) for quantity (calories) in the face of extreme spikes of rice and root and tubers price.

Ideally, the real per capita value of food consumed in urban households is unlikely to be substantially lower than rural households. However, results suggest that urban households had lower per capita calorie consumption than rural households. This may be indicative of a shift in the locus of caloric inadequacy from urban to rural farm households setting in the country. Rural households have a higher food consumption diversity than their urban counterpart. The rural households are expected to have direct access to what they cultivate and this could enhance dietary diversification (Weber, 2015), while urban households (non-agricultural households) can only purchase what is available in the market (Famine Early Warning Systems Network, 2017). Hence, households that are engaged primarily in agricultural production consumed more variety of foods than non-agricultural households as this is the case of rural farm households in Nigeria. The results further show that the demand for dietary diversity significantly increases among households headed by farmers in the study, in contrast to what was obtained by Freire and Rudkin (2019), where being a farmer means having decreased food diversity.

The coefficient associated with the male headed households has a statistically significant and negative effect on real value of food consumed and dietary diversity but a positive on food calorie consumed. This implies that the male headed households have a substantially higher real per capita food consumption compared to their female headed household counterparts. This is due to the fact that male is the head of the household and saddle with more responsibility that benefit all household members.

The coefficient of marital status of household head (married dummy) was statistically significant but has a positive impact on the real value of household per capita food consumption. This means that an average household headed by a married person has a higher value of per capita food consumption compared to other household groups. This is possible if such household has larger members, and, who do contribute substantially to raising household income. The coefficient of marital status of the household head is statistically significant with negative impact on household per capita daily calorie intake. This points out that households whose heads are married have lower per capita daily calorie consumption than the other household groups.

The results further show that age of household head has a significantly negative effect on calorie intake and positive effect on the real value of household per capita food consumption. This is probably because older people may not be able to afford adequate
nutrition intake. That is, the capacity to access sufficient calories declines with age. The results further show that the lower the age of the household head, the higher their demand for higher dietary diversity compared to household heads who are older. This is possible, because those that are younger have to consume as many nutritional foods as possible to be in the possession of the physical and mental capacities needed to engage in the education and labour activities (Ogundari, 2017).

The coefficients of educational dummies were positive, and statistically significant, pointing to the potential role that formal education gains could play in improving food consumption. The educational level of the household head, which can be taken as a proxy for consumer dietary knowledge and ability to process dietary information, had a significant negative relationship with the demand for dietary diversity. The positive and significant coefficients of educational dummies imply that having access to formal education is likely to promote food consumption diversity. This finding is in line with previous findings that found households whose heads are educated to have higher demand for dietary diversity as greater access to formal education is expected to reflect in better appreciation of the roles of more varied foods in enhancing diet quality (Pauzé, Batal, Philizaire, Blanchet and Sanou, 2016). Similar result of this study is found in Ogundari (2017). This outcome suggests that other factors beyond access to formal education might need to be considered to raise diversity of household diets.

The coefficient of surplus/post-harvest season is statistically significant and positively related to the real value of household per capita food consumption, calorie intake and dietary diversity. This implies that households also consumed more diverse foods during the post-harvest season (Farayibi and Owuru, 2016). This is because, supply and availability of food in the market are linked to seasonality and this can in turn influence food consumption patterns.

In addition, households classified in middle and relatively high income group had higher real per capital consumption than households in low income group. It is worthy of note that although price spikes may lead to reduction in food calories and real per capita value of food consumption (Genoni, Baez and Salazar, 2015), richer households are more likely to reduce calories than poorer ones (D'Souza and Jolliffe, 2016). Nevertheless, extremely poor households (whose lives are characterized by inadequate calorie intakes) may be unable to substantially cut-back food quantity (calories), and would rather adjust the compositions of their diets to sustain their calorie intake (energy) levels. Again, household dietary diversity increases significantly as household income increases. High-income households have access to more varied diets than low income households (Daniel, 2016). At very low level of income, households spend a substantial amount of their income on necessities including staple foods. However, they tend to allocate more of their income to more diverse foods, and other goods and services as
their incomes increase. D’Souza and Jolliffe (2016), noted that richer households do normally consume more varied diet (of relatively more expensive foods); thus providing the opportunity to bias consumption towards cheaper (alternative) foods as prices increase (D’Souza and Jolliffe, 2016). This is consistent with the finding of Shittu, Akerele, and Haile (2018) for households in Nigeria, Kenya, and Ethiopia respectively. Our finding, according to Freire and Rudkin (2019), is consistent with the hypothesis

Table 2: Food price shocks on food calorie consumption and dietary diversity

| Variable                                           | Coeff. | t-value | Coeff. | t-value | Coeff. | t-value |
|----------------------------------------------------|--------|---------|--------|---------|--------|---------|
| Spike in price of Rice                             | -0.18*** | 4.24    | -0.26*** | -2.55 | -0.22*** | -2.92 |
| Spike in price of Wheat                            | -0.09*  | 2.45    | -0.18*** | 4.27  | 0.12***  | 2.88  |
| Spike in price of other cereals                    | 0.02    | 1.63    | 0.16***  | 2.17   | 0.03**   | 2.16  |
| Spike in price of roots and tubers                 | -0.26*** | -3.13  | -0.21*** | -4.23 | -0.22*** | -2.83 |
| Spike in price of pulses                           | 0.32*** | 2.56    | 0.15***  | 2.88   | 0.24***  | 2.51  |
| Spike in price of fats and oils                    | -1.11*** | -2.14  | 0.28     | 1.49   | 0.21     | 1.01  |
| Spike in price of fruits                           | 1.21*** | 2.89    | -4.18*** | -5.12 | 1.15***  | 2.18  |
| Spike in price of vegetables                       | -1.82*** | -3.42  | -2.65*** | -2.94 | 2.03***  | 4.14  |
| Spike in price of egg                              | -0.06** | -2.34  | 1.21     | 1.23   | -1.25*** | -2.93 |
| Spike in price of meat                             | 0.04    | 1.33    | 0.02     | 1.43   | 0.03*    | 1.94  |
| Spike in price of fish                             | -0.02*  | -1.80  | 0.23***  | 3.44   | -0.02    | -0.16 |
| Spike in price of milk/dairy                       | 0.21    | 1.11    | -1.68**  | -2.48  | 0.32**   | 4.94  |
| Spike in price of sweeteners                       | -1.03*** | -4.12  | -1.26*   | -1.94  | 1.23**   | 2.33  |
| Spike in price of other foods                      | -1.15*** | -2.66  | 2.22     | 1.18   | 1.31**   | 2.28  |
| Natural log of non-food price                      | -0.08** | -2.18  | 0.11***  | 4.16   | -0.08*   | -1.86 |
| Location (Urban dummy)                             | 0.02**  | 2.44    | -0.04*** | -5.54  | -0.02*** | -2.74 |
| Sex (Male Headed Household dummy)                  | 0.24*** | 4.16    | -0.18*** | -5.62 | -0.21*** | -3.13 |
| Marital status of household head (married dummy)   | -0.02*** | -3.22  | 0.08***  | 4.04   | 0.33     | 0.43  |
| Age of household head                              | -0.18*** | -3.14  | 0.02***  | 3.83   | -0.03*   | -1.88 |
| Household headed by master/PHD holder               | 0.03    | 1.12    | -0.02    | -0.11  | -0.02*   | -1.76 |
| Household headed by First Degree holder             | 0.14    | 0.68    | 0.02**   | 2.16   | -0.02**  | -2.54 |
| Household headed by OND holder                      | 0.03    | 1.33    | 0.01***  | 3.12   | -0.04*** | -3.67 |
| Household headed by secondary school certificate holder | 0.02*** | 2.52    | 0.02***  | 3.88   | -0.02*** | -2.55 |
| Household headed by primary school holder           | 0.16*** | 2.02    | 0.21     | 0.44   | -0.03*** | -4.74 |
| Season (Post-Harvest/Surplus)                       | 0.03*** | 3.43    | 0.02***  | 3.33   | 0.01*    | 2.23  |
| Middle income household                            | 0.22**  | 2.35    | 0.16***  | 3.27   | 0.02**   | 4.12  |
| High income household                              | 0.13*** | 2.02    | 1.18***  | 2.88   | 0.04***  | 3.91  |
| Non-Farm (Participation dummy)                     | -0.22*** | -2.18  | 0.23**   | 2.22   | 0.25***  | 2.84  |
| Dependency dummy (households with under five year children) | 1.26*** | 2.73    | 0.44***  | 4.17   | 0.24***  | 2.24  |
| Zone (North central dummy)                         | -0.24** | 2.22    | -0.01    | 1.23   | 0.08*    | 1.94  |
| Zone (North west dummy)                            | 0.15    | 1.63    | 0.08     | 0.24   | 0.11     | 0.92  |
| Zone (South west dummy)                            | 0.47*   | 1.99    | -0.14**  | 2.02   | -0.02*   | -2.18 |
| Zone (South east dummy)                            | 0.13*** | 4.48    | 0.02**   | 2.42   | 0.09***  | 3.13  |
| Zone (South south dummy)                           | 0.18*** | 2.56    | 0.33***  | 5.13   | 0.26**   | 2.98  |
| Constant                                           | 1.63*** | 3.28    | 1.18***  | 41.22  | 1.14***  | 44.17 |
| F-Value                                            | 116.34*** | 132.82*** | 156.71*** |
| Prob>F                                             | 0.000  | 0.000   | 0.000    |
that consumption evolves along hierarchical order as income increases. The substantial growth in the real per capita income of the household from agriculture are from related sources or from non-farm sources. Farm households are engaged in some income generating activities with the intention to boost future income and food consumption (Aiyede, Sha and Olawale, 2017). This result might be arising from the additional sources of income that may be accruing to the household in a situation where most of the members of the household are involved in some other income-generating activities apart from farming. Therefore, involvement of household head in non-farm activities improves households’ dietary diversity. It thus becomes imperative to give more serious attention to the concerns relating to dietary diversity, particularly given the relationship between farm households’ nutritional outcomes and food price shocks in Nigeria.

Conclusion and Recommendations
On the average, household food consumption variety is fairly high. The observed patterns of food calorie consumption over the years indicate higher consumption during the post-harvest season than in the post-planting periods. Spikes in cereal prices (rice and wheat) are generally, associated with a lower number of meals consumed in a household and a lower likelihood of the number of calories being consumed. However, spikes in the price of wheat are associated with higher dietary diversity. As a food deficit and wheat and rice importing country, food prices in Nigeria are closely linked to global prices and price shocks. Hence, excessive international as well as domestic food price variability is a concern for domestic food security. Households with high income and educational status can substantially raise calorie consumption, the real value of food consumed and dietary diversity. Spikes in the price of cereals generally hold negative consequences for food quantity consumption (in terms of calorie and real value of food consumed), and affect dietary diversity of households. Increases in the price of non-food items could advance the real value of food intake of households. Beyond, increased consumption of food calories, greater access to higher education and income seems likely to enhance food consumption diversity and the real value of food consumed by the household. Therefore, income and education improvement is crucial for raising food calories and satisfaction of hunger needs among households. A combination of policy strategies, including income growth, agricultural development and targeted food distribution programs could reduce the problems of inadequate calorie consumption among farm households.

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