What Role Did Rising Demand Play in Driving Food Prices Up?

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Abstract

Average food inflation in India during the period 2006–2013 was one of the highest among emerging market economies and nearly double the inflation witnessed in India during the previous decade. An often-cited hypothesis argues that the surge in food inflation during this period was driven by rising demand for high-value food products due to higher per capita income and diversification of Indian diets. In this article, we test the validity of this hypothesis by estimating the expenditure elasticity and then calculating the aggregate demand using data from household survey conducted by the National Sample Survey Organisation (NSSO). Our results show that in recent years, estimated demand has exceeded supply of all major food products, barring fruits. Moreover, empirical estimates indicate that the demand–supply gap is an important driver of rise in food prices, along with other factors such as minimum support prices, global prices, fiscal deficit and agricultural wages.

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JEL Classification: E31, E37, Q11

Introduction

India experienced one of the highest rates of food inflation among emerging economies during the period 2006–2013. The average rate of inflation in food products in India was more than 9 per cent during this period. Moreover, the rate of increase in food prices during this period was nearly double of what was witnessed in the previous decade. While India has witnessed sporadic spurts in food inflation, episodes of such persistently high food inflation have been rare.

The welfare impact of such high rate of food inflation is bound to be significant, given that food constitutes a significant share of the consumption basket of the Indian households. According to the report on Household Consumption Survey by the National Sample Survey Organisation (NSSO), in 2011–2012, on average, food accounted for 48.6 per cent of overall expenditure in rural areas and 38.5 per cent in urban areas. The proportion is significantly higher for 269 million people, or 21.9 per cent of the total population, who lived in abject poverty. On average, the bottom three expenditure deciles in rural areas devoted 60.4 per cent of their total expenditure on food products while in urban areas the share was 56.5 per cent. Given that this section already spent a large proportion of their income on food, they were generally unable to divert additional expenditure on food to neutralize the impact of food inflation, thereby aggravating food and nutrition deficiency.

A variety of reasons have been put forward to explain the rise in food prices in this period. Cost escalation factors including rise in price of inputs such as agricultural wages and fuel have been identified as major drivers of surge in food prices. A sharp rise in minimum support prices (MSPs) during these years also contributed to food inflation, given that these prices acted as floor to market price. Negative production shocks such as the drought in 2009 and low production growth in 2008–2009 also propped up agricultural prices. Lax monetary and fiscal policies as well as policies related to stocking and futures trading and temporary trade policies have also contributed to rising food inflation.

Apart from these factors, an often-cited structural factor resulting in rising food prices in India is the increase in demand for various food products due to rising per capita income and population growth that was
not matched by a commensurate increase in supply due to low productivity. In fact, rising per capita income in emerging economies such as India and China, and the associated rise in demand for food, has been pointed out as a driver of global spike in food prices. Krugman (2008) points out that the rise in per capita income in emerging markets has shifted the dietary habits towards meat, which in turn has raised the demand for grains as animal feed. Similarly, Wolf (2008) argues that the shifts in land use in response to rising demand for meat and related animal feed have reduced the supply of cereals available for human consumption.

Studies focusing on food inflation in India have stressed the importance of rising demand of various food products playing a role in driving food prices up. Mishra and Roy (2012) argue that the rising demand–supply mismatch contributed to food inflation for all commodities, barring cereals. Kumar, Vashisht and Kalita (2010) also argue that the rising inequality in per capita income resulted in rise in demand for food products, which in turn led to a price rise.

Similarly, Gokarn (2011), Bandara (2013) and Gulati and Saini (2013) also point out that India has witnessed shifts in its food basket from cereals to protein- and vitamin-rich diets, such as pulses, milk, vegetables, egg, meat and fish, causing upward pressure on prices of these commodities. In contrast, Eapen and Nair (2012) argue that there is little evidence supporting the view that a consumption shift towards high-value agriculture products has been driving prices up, except in the case of milk.

A few papers have conducted empirical analyses to identify the primary drivers of food inflation in India. For example, Agrawal and Kumarasamy (2012) argue that demand–supply gap has resulted in food inflation in India. They forecast the demand for various food products by using an autoregressive distributive lag (ARDL) model and estimate income elasticity, cross-price elasticity and own price elasticity as well as point out supply-side measures that should be put in place to control food inflation. Similarly, Sasmal (2015) develops a two-sector general equilibrium model to conclude that food price inflation will exist in an economy where demand for food products increases at a higher rate in a growing economy but growth in agricultural production lags behind. Empirical evidence suggests that increase in per capita income has a significant positive impact on food prices, while there is a negative long-run relationship between food grains production and food prices. Using a panel of three South Asian economies, Carrasco and Mukhopadhyay (2012) find that rising per capita income has a positive impact on food prices, while a rise in agricultural production negatively impacts food prices, with the extent of impact being different in the various countries.
Our article adds to the existing literature in a couple of ways. First, we use the Quadratic Almost Ideal Demand System (QUAIDS) model to estimate the demand for key important commodities.\(^1\)\(^2\) The application of the QUAIDS model allows us to make use of household consumption data to estimate income elasticities for various food products, and thereby assess the evolution of demand of various commodities over time. Thus, the application of QUAIDS model significantly improves the modelling of consumption behaviour compared to studies that have used aggregate consumption as a proxy for underlying demand in the economy. Second, the use of the QUAIDS model allows us to calculate the demand–supply gap explicitly and estimate the impact of this gap on food prices, after controlling for other major determinants of price rise. This is again in contrast to most of the existing studies that have looked separately at the impact of demand (proxied by per capita income) and the impact of supply (proxied by volume of agriculture production or production of food grains). Finally, we use the 2009–2010 household survey data to compute the elasticities, with the survey period falling right in the middle of the high inflation years in India, and thereby providing us with important insights into how the change in consumption behaviour during this period has influenced food prices. Most of the existing studies including Mittal (2010) and Kumar, Kumar, Parappurathu and Raju (2011) have used data from surveys up to 2004–2005 to estimate demand for food products in India.

The rest of the article is structured as follows. The second section describes the change in consumption pattern over a 25-year period and across consumption expenditure deciles. The third section describes the model used to calculate the expenditure elasticities for various food products, evolution of the demand for these products with rising per capita income and the resulting demand–supply gap. In the fourth section, we estimate the extent to which this gap was responsible for a rise in food prices. Finally, the fifth section concludes by summarizing the main findings of the article.

\(^1\) The QUAIDS model has been primarily used to estimate future demand for various commodities (Mittal, 2010; Xie, Mittelhammer & Heckelei, 2004).

\(^2\) Xie et al. (2004) point out that the QUAIDS model has several advantages in modelling consumption behaviour for grouped commodities including approximating any demand system arbitrarily to first order, aggregating perfectly over consumers, satisfying the axioms of choice and being capable of testing the restrictions of homogeneity and Slutsky symmetry. The inclusion of the quadratic term in log income helps capturing the effects of non-linear Engel curves, as observed in various empirical demand studies.
Evolution of Food Products Consumption

The NSSO conducts a survey of consumer expenditure at regular intervals. We use data from these surveys to document the trend in share of household consumption on various food items since the late 1980s in Figure 1. We concentrate on six major food products, namely cereals; pulses; milk; eggs, meat and fish; fruits and vegetables. It is evident that there has been a steady decline in the share of income spent on consuming food both in rural and urban areas. Between 1993–1994 and 2011–2012, the share of expenditure on food fell by 14.6 percentage points in rural areas and 16.1 percentage points in urban areas. During this period, the real monthly per capita expenditure (MPCE) grew by an average annual rate of 1.8 per cent in rural areas and 2.3 per cent in urban areas, implying that higher MPCE was associated with a drop in share of food in the consumption basket. Thus, the evidence corroborates Engel’s law, which argues that expenditure share dedicated to food consumption declines as income (here proxied by expenditure) rises.

Despite a secular decline in the proportion of expenditure on food, within food expenditure, spending on selected food products witnessed an increase. While share of expenditure on cereals witnessed the biggest decline in rural and urban areas, in the case of pulses and vegetables, there was a decline in urban areas only. In contrast, a higher share of income was spent on milk, fruits and eggs, meat and fish, providing support to the hypothesis that consumption of protein- and vitamin-rich agriculture products has increased.

Cross-sectional data from the household consumer expenditure survey of 2011–2012 also validate a faster increase in consumption of protein- and vitamin-rich agricultural products such as milk, fruits and eggs, meat and fish relative to staples such as cereals and pulses. Figure 2 plots the household MPCE of six major food items across consumption deciles. For the rural households, the ratio of average consumption of the top two expenditure deciles to that of the bottom two deciles at 9.8 is highest for fruits. This is followed by milk at 7.4 and eggs, meat and fish at 4.2. In contrast, the ratio is less than 2.0 for cereals, pulses and vegetables.

A similar pattern is observed in the case of urban consumers where the ratio is again highest for fruits followed by milk and eggs, meat and fish. The ratio remains below 2.0 for cereals and pulses, while it is 2.2 for vegetables. Thus, increases in household income are associated with a significantly larger incremental expenditure on fruits, milk and eggs, meat and fish relative to cereals, pulses and vegetables.
Figure 1. Share of Food Expenditure in Monthly Per Capita Expenditure

Source: National Sample Survey Organisation, various reports.
Estimating Demand–Supply Gap for Food Commodities

Having corroborated a dietary shift towards products, which have contributed significantly to the food inflation in recent years, we focus on the change in aggregate demand resulting from this shift in diet.
We estimate expenditure elasticity of the above-selected food items using household consumer data. We cover the period from 2004–2005 to 2013–2014. During this period, three large household surveys were conducted in 2004–2005, 2009–2010 and 2011–2012. Given the possibility of expenditure elasticities changing over a period, we use the data from 2009–2010 household survey to compute these elasticities to mitigate the errors arising from this. The six selected food items comprise 76 per cent of average MPCE.

We compute aggregate demand as the sum of aggregate household demand and indirect demand requirements from industries using these food items as inputs (seed, feed and wastage [SFW]). We estimate per capita household demand and associated expenditure elasticities for the selected food items using the QUAIDS following Banks, Blundell and Lewbel (1997). For India, Mittal (2010) has also used the QUAIDS model to estimate the expenditure elasticity in India. However, the elasticities computed in Mittal (2010) are based on household expenditure data from surveys conducted until 1999–2000. We use more recent survey data to compute these elasticities to accurately capture the role played by rising demand in influencing food prices.

The QUAIDS is specified with expenditure shares as the dependent variable. A household’s expenditure share for good \( i \) is defined as

\[
w_i = \frac{p_i q_i}{m},
\]

where \( p_i \) is the unit price of good \( i \), \( q_i \) is the quantity of good \( i \) purchased or consumed and \( m \) is the total expenditure on all goods in the demand system. With this definition of \( m \), \( \sum_{i=1}^{K} w_i = 1 \), where \( K \) is the number of goods in the system. The functional form of the expenditure share under QUAIDS is as follows:

\[
w_i = \alpha_i + \sum_{j=1}^{K} \gamma_{ij} \ln p_j + \beta_i \ln \left\{ \frac{m}{P(p)} \right\} + \frac{\lambda_i}{b(p)} \left[ \frac{m}{P(p)} \right]^2,
\]

where \( p \) is the vector of all prices and \( b(p) \) is defined as \( b(p) = \prod_{i=1}^{K} p_i^{\beta_i} \).

The aggregate price index is defined as

\[
\ln P(p) = \alpha_0 + \sum_{i=1}^{K} \alpha_i \ln p_i + \frac{1}{2} \sum_{i=1}^{K} \sum_{j=1}^{K} \gamma_{ij} \ln p_i \ln p_j.
\]

The parameters are subject to the following restrictions:

\[
\sum_{i=1}^{K} \alpha_i = 1; \sum_{i=1}^{K} \beta_i = 0; \sum_{i=1}^{K} \gamma_{ij} = 0 \forall j;
\]
and Slutsky symmetry implies that $\gamma_{ij} = \gamma_{ji}$. The estimated expenditure elasticities for the selected food items are given in Table 1.

Table 1 shows that expenditure elasticities for selected major food items are found to be positive, suggesting a rise in total household expenditure would lead to stronger demand for these items. Expenditure elasticities for milk and milk products, vegetables and fruits are over 1, suggesting that a 1 per cent increase in household expenditure on food would lead to a more than 1 per cent increase in the demand for these items. The elasticity for meat and fish, although below unity, is high enough to cause significant rise in demand for these items as total expenditure on food increases. Compared to Mittal (2010), our elasticities are higher for cereals, vegetables and fruits, and lower for pulses, meat and fish.

We compute aggregate household demand using the following equation:

$$D_{i,t}^H = d_{i,0} N_t \prod_{n=1}^{t} (1 + g_n \times e_i),$$  \hspace{1cm} (4)$$

where $D_{i,t}^H$ is the aggregate household demand for commodity $i$ in year $t$, $d_{i,0}$ is the per capita demand for commodity $i$ in the base year, $N_t$ is the population in year $t$, $g_n$ is the per capita income growth rate in year $n$ where $n$ goes from 1 to $t$ and $e_i$ is the expenditure elasticity for the commodity $i$. We generate a historical household demand series for the period from 2004–2005 to 2013–2014.

### Table 1. Expenditure Elasticity

<table>
<thead>
<tr>
<th>Commodity</th>
<th>Elasticities</th>
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<tbody>
<tr>
<td>Cereals</td>
<td>0.226</td>
</tr>
<tr>
<td>Pulses</td>
<td>0.515</td>
</tr>
<tr>
<td>Vegetables</td>
<td>1.535</td>
</tr>
<tr>
<td>Fruits</td>
<td>2.21</td>
</tr>
<tr>
<td>Milk and Milk Products</td>
<td>2.185</td>
</tr>
<tr>
<td>Meat and Fish</td>
<td>0.796</td>
</tr>
</tbody>
</table>

Source: Authors’ estimates.

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3 Since our base year per capita household demand $d_{i,0}$ is estimated for 2009–2010, we generate the historical household demand series from 2004–2005 until 2013–2014 by iterating equation (4) backward and forward with respect to $d_{i,0}$. The data for GDP at factor cost at constant 2004–2005 prices and population are sourced from the RBI.
We compute aggregate demand using the following equation:

\[ D_{i,t} = \frac{D_{i,t}^H}{1 - x_i}, \]  

(5)

where \( D_{i,t} \) is the aggregate demand for commodity \( i \) in year \( t \) and \( x_i \) is the share of indirect demand in total demand for commodity \( i \). The share of indirect demand in total demand for the selected food products are sourced from the Planning Commission (2012) and are presented in Table 2. The report provides estimates of indirect demand for 2004–2005 and 2011–2012. We take an average of the estimates for these two periods to compute total demand for 2004–2005 to 2013–2014. In addition, an average of the indirect demand estimates for rice and wheat given in the report is used as a proxy for indirect demand for cereals.

In Figure 3, we compare the estimated aggregate demand for the selected commodities with the domestic supply during the period 2004–2005 to 2013–2014 to evaluate the demand–supply gap. We adjust the overall domestic production with post-harvest losses experienced by the various commodities to obtain the domestic supply available for consumption. Estimates of the losses have been taken from Nanda et al. (2012). These losses reduce the amount available for domestic consumption and take place during post-harvest on-farm operations, transportation from farm to the next destination and storage at various points in marketing channels for all crops.

For cereals, post-harvest loss-adjusted supply has been higher than estimated demand due to healthy production growth between 2004–2005 and 2008–2009 and again between 2010–2011 and 2013–2014. It was only in 2009–2010, when there was a severe drought, that supply fell marginally below estimated demand. In contrast, in pulses, the estimated demand has consistently outstripped supply, and the gap has widened in recent years. This gap has been met to some extent through import of

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<tbody>
<tr>
<td>Rice</td>
<td>12.97</td>
<td>13.43</td>
<td>13.20</td>
</tr>
<tr>
<td>Wheat</td>
<td>17.08</td>
<td>17.69</td>
<td>17.39</td>
</tr>
<tr>
<td>Cereals</td>
<td>15.03</td>
<td>15.56</td>
<td>15.29</td>
</tr>
<tr>
<td>Pulses</td>
<td>37.00</td>
<td>41.71</td>
<td>39.36</td>
</tr>
<tr>
<td>Milk</td>
<td>40.58</td>
<td>41.58</td>
<td>41.08</td>
</tr>
<tr>
<td>Fish and Meat</td>
<td>39.45</td>
<td>40.83</td>
<td>40.14</td>
</tr>
<tr>
<td>Vegetables</td>
<td>37.76</td>
<td>38.43</td>
<td>38.10</td>
</tr>
<tr>
<td>Fruits</td>
<td>81.47</td>
<td>82.90</td>
<td>82.19</td>
</tr>
</tbody>
</table>

these commodities, which in turn makes the domestic price vulnerable to changes in international prices.

In the case of protein-rich commodities, milk as well as meat and fish, the high expenditure elasticities have meant that estimated demand has been increasing at a significantly higher rate than supply. Consequently, while supply of milk and milk products was higher than estimated
demand in 2004–2005, estimated demand outstripped supply in 2008–2009, with the gap between estimated demand and supply increasing since then.

In contrast, in the case of meat and fish, estimated demand exceeded supply by 6.5 million tonnes even in 2004–2005. Since then the gap between the two has been steadily rising with the difference amounting to nearly 8 million tonnes in 2013–2014.

As can be seen from Table 2, both fruits and vegetables exhibit expenditure elasticities which are greater than one, implying a more than a proportionate increase in demand for these products for a given increase in expenditure. In case of vegetables, while aggregate supply exceeded estimated demand by 24.3 million tonnes in 2004–2005, the gap between the two had closed by 2010–2011, and in recent years, estimated demand has outstripped supply. In case of fruits, despite the high expenditure elasticity, we find that domestic supply has been higher than estimated demand throughout the entire period. Thus, factors other than demand–supply gap explain the rapid surges in fruit prices in recent years.

It is evident that shortfall in production has resulted in demand pressure for a number of commodities such as pulses, milk, fish and meat, and vegetables. The shortfall in production has been driven by limited gains in agricultural productivity in recent decades, with the latter in turn being driven by a myriad of factors including fragmented landholdings, outdated farming techniques, inadequate use of modern inputs, declining share of public investment in agriculture and rising share of subsidies, and lack of organized agricultural marketing.

**Did Rising Demand–Supply Gap Contribute to Food Inflation?**

We now turn our focus to whether the gap between demand and supply for the various commodities contribute to a rise in prices in India as has been claimed in a number of studies. We empirically estimate the impact of a demand–supply gap on food prices after controlling for some of variables found in the literature to have affected food prices. We focus on six food products described in the third section, that is, cereals, pulses, milk, meat and fish, vegetables and fruits. Data on various food product prices are sourced from the WPI series available with the Office of the Economic Advisor, Ministry of Commerce and Industry, while data on demand–supply gap, the main variable of interest, are calculated in the third section.
Other factors, which have been found to impact food prices, include global prices of various food commodities, MSPs, agricultural wages and fiscal deficit. Transmission of international prices to domestic retail prices could be a major factor driving inflation, especially given the surge in international prices of food commodities in 2008, and again in 2010. The high inflation during this period was driven by surge in prices of cereals, edible oils, sugar and some dairy products. Some of this rise in global prices is likely to have transmitted to domestic shores, given that the extent of agricultural sector’s integration with global market, calculated as the share of agriculture trade to agriculture GDP, increased nearly fourfold from 5.2 per cent in 1990–1991 to around 19 per cent in 2013–2014. India has been a net exporter of food grains during the last three decades, while imports are limited to edible oils, sugar and pulses. Table 3 highlights the extent of co-movement between domestic and global inflation rates during 2005–2014 and indicates that only in the case of sugar and edible oils, there was significant positive co-movement between the two.

Rise in global price of a commodity can affect its domestic price through trade channel, subject to trade policies. First, when global price of a commodity rises, exporters find it more attractive to supply in the global market, causing lower supply in the domestic market, followed by a rise in domestic prices. The extent of the impact on domestic supply and the resultant effect on domestic price of the commodity will depend on the extent of tradability of the product, as well as short-run trade policy responses to global price hike, such as export ban, etc. Second, surge in global price of a commodity, if it is included in the import basket of the consumers, raises the price of its domestic substitute via expenditure switching channel. Again, the extent of transmission of global price rise into its domestic counterpart of a commodity depends on whether the domestic price is administered or not.

For instance, India is the largest milk producer in the world, contributing to 17 per cent of the global milk production (Rajeswaran, Naik, & Dhas, 2014). However, the large spikes observed in global milk prices do not seem to get transmitted to domestic milk prices (Bhattacharya & Sen Gupta, 2015). The Indian milk and dairy industry is highly segmented with Gujarat Cooperative Milk Marketing Federation (GCMMF) being the largest organized player in the dairy sector and the largest exporter in

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4 For a detailed review of the literature discussing factors driving food prices in India, please refer to Mishra and Roy (2012).
Table 3. Extent of Co-movement between Global and Domestic Price Changes

<table>
<thead>
<tr>
<th>Product</th>
<th>Co-movement</th>
</tr>
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<tbody>
<tr>
<td>Food</td>
<td>-0.0883</td>
</tr>
<tr>
<td></td>
<td>(0.3373)</td>
</tr>
<tr>
<td>Meat</td>
<td>-0.1047</td>
</tr>
<tr>
<td></td>
<td>(0.2553)</td>
</tr>
<tr>
<td>Dairy</td>
<td>0.0904</td>
</tr>
<tr>
<td></td>
<td>(0.3261)</td>
</tr>
<tr>
<td>Cereals</td>
<td>-0.0386</td>
</tr>
<tr>
<td></td>
<td>(0.6753)</td>
</tr>
<tr>
<td>Edible Oils</td>
<td>0.6380***</td>
</tr>
<tr>
<td></td>
<td>(0.0000)</td>
</tr>
<tr>
<td>Sugar</td>
<td>0.5634***</td>
</tr>
<tr>
<td></td>
<td>(0.0000)</td>
</tr>
</tbody>
</table>

Source: FAOSTAT and Office of Economic Advisor, Ministry of Commerce.

Notes: ***, ** and * imply significance at 1%, 5% and 10%, respectively.

the country (Rajeswaran et al., 2014). The study finds that the price of milk paid by GCMMF, which can be taken as a proxy for farm gate price of milk, is found to have seen a sharp rise in post-2006 period in the backdrop of a sharp rise in exports of skim milk powder in response to a rise in global milk prices and the rapid depletion of its domestic buffer stock. In the pre-2006 period, maintenance of buffer stock along with ban on exports of skim milk powder in India has been used as a tool to moderate domestic milk prices. In this scenario, occasional export ban on skim milk powder and building of domestic buffer stock have been adopted again in post-2006 period, which eventually moderated the transmission of spikes in global milk prices to its domestic counterpart. This indicates that a part of the milk prices in India is administered and its dynamics can vary in response to global price changes, exogenously to trade channel.

A rise in MSP also fuels food inflation, given that it is meant to be the floor price for various crops, that is, the minimum price at which the government stands to procure crops from farmers. The wholesale prices are typically higher than these floor prices, and if the floor price keeps rising, as has been the case in India, then it leads to a rise in wholesale prices as well. Figure 4 shows that the weighted average of annual inflation in MSP (WPI weights) during the period 2007–2008 to 2012–2013 was 13 per cent compared to only 3.9 per cent during the preceding four
years. This was associated with WPI inflation in these products increasing from 3.7 per cent to 10.1 per cent. Subsequently, when MSP inflation fell to an average 3.8 per cent in 2013–2014 and 2014–2015, it corresponded with WPI inflation declining to 3.3 per cent. The data on cropwise MSP are compiled from Ministry of Agriculture. In the case of vegetables, fruits, milk, and meat and fish, for which there is no MSP, we set the MSP at zero.

A sharp increase in rural wages, by raising the cost of production, can push up food prices. Alternately, by providing a higher purchasing power, higher wages can bolster the demand for food items. Agricultural wages grew by an average annual rate of 17.3 per cent between 2008–2009 and 2012–2013, nearly four times higher than the average annual growth between 2003–2004 and 2007–2008 (Figure 5). Moreover, during the latter period, the rise in agricultural wages was considerably higher compared to non-agricultural wages, which grew at an average rate of 15.6 per cent. The introduction and the subsequent expansion of the National Rural Employment Guarantee Scheme, which guarantees 100 days of wage employment, have been postulated as driving rural wages up. However, the empirical evidence indicates that the impact is limited. While Imbert and Papp (2015) find that National Rural

![Figure 4. Relationship between Minimum Support Price and Wholesale Price Inflation](image)

Source: Ministry of Agriculture and Office of Economic Advisor, Ministry of Commerce.
Employment Guarantee Act (NREGA) raised wage income by 4.5 per cent, Berg, Bhattcharyya, Durgam and Ramachandra (2012) argue that the boost to daily wages was 5.3 per cent. Data on agricultural wages are taken from the database on the Indian Economy, RBI.

Finally, the link between fiscal deficit and inflation has been extensively explored in the literature from several perspectives. The first argument is based on the rationale that in developing economies with low tax base and tax evasion, fiscal deficit is financed by borrowing from the banking system (Dornbusch, Sturzenegger, & Wolf, 1990; Sargent & Wallace, 1981). This in turn results in higher money growth leading to high inflation. A second argument points out that high level of deficit may lead to higher demand, which will raise prices of products whose output cannot be immediately increased. For India, Gulati and Saini (2013) find fiscal deficit to be the biggest driver of food inflation. In India there was a sharp increase in the borrowing program of the government to finance the deficit (Figure 6).

In Table 4, we examine the relationship between commodity prices and the demand–supply gap in these commodities in a panel regression framework for the period 2004–2005 to 2013–2014. The regression specification is outlined as

\[ Y_{i,t} = \alpha_i + \beta X_{i,t} + \gamma Z_{i,t} + \mu_{i,t}, \]  

(5)

where \( i \) refers to the different food products, namely cereals, pulses, milk, fruits, vegetables, and meat and fish, and \( t \) represents the time
We use the fixed effects estimator accounting for heteroscedasticity. Here, $Y$ is the dependent variable, measured as change in food product prices. Among the explanatory variables, $X$ is the main variable of interest, that is, demand–supply gap for various food products, while $Z$ is a vector of other variables found in the literature to have influenced food prices, including MSPs, global prices, fiscal deficit as a percentage of GDP and agricultural wage growth. While column I focuses on the relationship between demand–supply gap and food prices, in columns II–V, we control for other factors. Initially, we introduce these factors one at a time, to evaluate their role in determining food prices. In column VI, we focus on the various major drivers of food prices under one specification. The summary statistics of the different variables are given in Table A1 in the Appendix.

We find that the demand–supply gap has a significant and positive impact on food prices across all the specifications, although there is considerable variation in the size of the coefficient. Thus, the results indicate that even after controlling for factors that have been found in the literature to drive food inflation up, demand–supply gap has a positive impact on food prices. An additional gap of 1 million tonnes in demand for food and supply of food would result in food prices increasing by 0.3–1.3 per cent annually.

Turning to other drivers, we find that each is positive and statistically significant when introduced one at a time. The results suggest that an increase in global prices, MSPs, fiscal deficit (percentage of GDP) and
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<th>III</th>
<th>IV</th>
<th>V</th>
<th>VI</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Constant</strong></td>
<td>5.047***</td>
<td>3.578***</td>
<td>0.557</td>
<td>4.634***</td>
<td>4.603***</td>
<td>3.731***</td>
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<tr>
<td></td>
<td>[133.856]</td>
<td>[13.057]</td>
<td>[1.321]</td>
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<td>0.013***</td>
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<td></td>
<td>[2.311]</td>
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<td><strong>Agricultural Wages</strong></td>
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<td>0.031***</td>
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**Table 4. Effect of Demand–Supply Gap on Food Prices**

**Source:** Authors’ calculations.

**Notes:** Robust t-statistic in brackets. ***, ** and * imply significance at 1%, 5% and 10%, respectively.
faster wage growth, has a positive and significant impact of food prices. However, when introduced simultaneously, it is only wage growth, which has a significant impact on food prices, apart from demand-supply gap.

The channels through which fiscal deficit impacted inflation in India merit some elaboration with the combined fiscal deficit of the state and the central governments averaging 7.7 per cent during the period 2008–2009 to 2013–2014, compared to an average of 5.7 per cent in the previous four years. As discussed above, higher fiscal deficit can lead to an uptick in inflation either if the banking system is financing the fiscal deficit or if the resultant higher demand cannot be immediately met with an increase in supply. Evidence suggests that in India, the second channel played a more dominant role than the first one. As pointed out in Khundrakpam and Pattanaik (2010), the increased borrowing programme of the government did not result in a high money growth, since the growth in demand for credit from the private sector exhibited significant deceleration. Money supply growth during 2008–2009 to 2013–2014 averaged 14.8 per cent well below 19 per cent growth averaged between 2004–2005 and 2007–2008, while bank credit growth to the commercial sector nearly halved from 26.3 per cent to 15.5 per cent.

The risk to inflation emerged from the fact that the fiscal stimulus measures were primarily aimed at bolstering consumption instead of investment. The measures included waiver of loans to the farmer, liberal increase in the salaries of the government employees and expansion of a scheme that guaranteed 100 days of employment to a household across entire India. Consequently, the growth in government consumption jumped from 6.6 per cent during 2004–2005 and 2007–2008 to 9.2 per cent between 2008–2009 and 2011–2012 with the food subsidy bill inching up from 0.65 per cent of GDP in 2006–2007 to over 1 per cent in 2012–2013. These measures boosted aggregate demand, including the demand for food items, which accounts for nearly 46 per cent of the consumption basket. The increase in food consumption as a result of additional fiscal outlay is over and above the growth in consumption due to change in dietary preferences and per capita income growth.

In addition, the central bank also attempted to boost aggregate demand by reducing key policy variables such as the repo rate, statutory liquidity ratio and cash reserve ratio, introducing several refinance windows to provide liquidity and relaxing prudential norms with respect to provisioning and risk weights. For a list of fiscal and monetary measures introduced to boost aggregate demand, see Dasgupta and Sen Gupta (2011).
Conclusion

The main objective of this article was to empirically examine the often-cited hypothesis that rising per capita income in emerging markets has shifted the dietary habit towards protein-rich commodities, resulting in high food prices. Using household survey data to estimate expenditure elasticity, we find the elasticity to be highest in case of fruits, followed by milk and milk products, vegetables, and meat and fish. This has meant that in recent years, demand has exceeded supply of all these products, barring fruits. Empirical estimates show that the demand–supply gap is an important driver of rise in food prices, along with other factors such as MSPs, global prices, fiscal deficit and agricultural wages.

Going forward, high positive expenditure elasticity for protein- and vitamin-rich products implies that a rise in per capita income will result in a rise in aggregate demand for these commodities. This will further widen the demand–supply gap in coming years, unless steps are taken to significantly raise productivity of Indian agriculture. Hence, apart from augmenting the supply of various food products, to keep the demand–supply gap in check, measures will have to be taken to keep in check other drivers of food inflation.

Indeed, this is what has been observed during the last two years, when in spite of further rise in the demand–supply gap for certain commodities, as per capita income increased, food inflation moderated due to improvements in most of the other drivers of food prices. For example, global food prices contracted by an average of 4.7 per cent in 2013–2014 and 2014–2015, compared to an increase of 10.8 per cent between 2006–2007 and 2012–2013. Similarly, consolidated government deficit (as a percentage of GDP) also improved from around 8 per cent to below 6.7 per cent. At the same time, the growth in MSPs has significantly reduced from an average annual rate of 12.7 per cent during 2006–2007 and 2012–2013 to an average of 3.7 per cent in 2013–2014 and 2014–2015. Finally, rural wages, which were growing at an average of 14.6 per cent during 2006–2007 and 2012–2013, saw its growth decline to sub-10 per cent in recent years.

Acknowledgement

We would like to thank Laurence Ball, Charan Singh and conference participants of International Conference on Food Price Volatility: Causes and Challenges, Rabat, Morocco, for their very helpful comments. We also express thanks to an anonymous referee for insightful comments. The views expressed in this article are ours and do not necessarily reflect the views or policies of the Asian
Development Bank (ADB), or its Board of Governors, or the governments they represent.

Appendix

Table A1. Summary Statistics of the Main Variables

<table>
<thead>
<tr>
<th>Variable</th>
<th>Number of Observations</th>
<th>Mean</th>
<th>Standard Deviation</th>
<th>Minimum</th>
<th>Maximum</th>
</tr>
</thead>
<tbody>
<tr>
<td>Domestic Food Inflation (%)</td>
<td>60</td>
<td>9.708</td>
<td>8.495</td>
<td>–5.512</td>
<td>40.171</td>
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<tr>
<td>Demand Supply Gap (Million Tonnes)</td>
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<td>–2.787</td>
<td>19.589</td>
<td>–35.429</td>
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<td>MSP Inflation (%)</td>
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<td>2.850</td>
<td>9.132</td>
<td>0.528</td>
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<td>Agricultural Wage Inflation (%)</td>
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<td>11.205</td>
<td>6.626</td>
<td>1.680</td>
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<tr>
<td>Global Food Inflation (%)</td>
<td>57</td>
<td>9.411</td>
<td>15.601</td>
<td>–33.376</td>
<td>69.000</td>
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<td>Fiscal Deficit (% of GDP)</td>
<td>60</td>
<td>7.310</td>
<td>1.552</td>
<td>3.990</td>
<td>9.330</td>
</tr>
</tbody>
</table>

Source: Author’s calculations.

References


