

Food, Fuel, and Facts: Distributional Effects of Global Price Shocks

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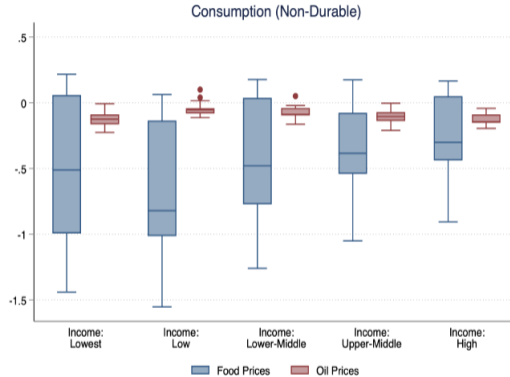
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ASSA 2024 Annual Meeting, San Antonio

A quick summary

- How do commodity price shocks affect **households** in emerging markets?
 - Important question given recent large global **commodity price shocks**
- Analyze empirically in Indian context exploiting
 1. **rich household survey** data on income and consumption
 2. time-series variation in global commodity prices
- Shed light on **distributional effects** and **transmission mechanisms**

Main takeaway



- Poorer households **more affected** by food price shocks
- Oil price shocks affect households more homogeneously
- Heterogeneous consumption responses mirror heterogeneity in **earnings responses**

Overview of my comments

1. Identification & more on macro effects
2. Econometric odds and ends
3. Mechanism & story

Commodity price shocks and their macro impact

- The authors study food and oil price shocks. Use as instruments:
 1. Baumeister and Hamilton (2019) oil supply shock
 2. Self-constructed food price shock
 - residualize global food price index from a common demand and food-specific demand factor estimated in a DFM of non-energy commodity prices
- Before studying distributional impacts, I would like to see **macro impact**
 - Sense-check how well shocks are identified
 - Helps putting distributional effects into perspective. How is overall Indian economy affected?
 - Compare aggregate consumption response and average household consumption response to assess representativeness of survey

Empirical specification

Panel LP-IV:

$$\begin{aligned} c_{i,t+h} - c_{i,t-1} = & \beta_{0, \text{food}}^{g,h} \text{ext}_t^{\text{food}} \times \mathbb{1}_{i \in g(t)} + \beta_{0, \text{oil}}^{g,h} \text{ext}_t^{\text{oil}} \times \mathbb{1}_{i \in g(t)} + \sum_{j=1}^J \alpha^h (c_{i,t-j} - c_{i,t-j-1}) \\ & + \sum_{k=1}^K \beta_{k, \text{food}}^h \text{ext}_{t-k}^{\text{food}} + \sum_{k=1}^K \beta_{k, \text{oil}}^h \text{ext}_{t-k}^{\text{oil}} + \sum_{d=0}^D \delta^h D_{t-d} + \gamma^{g,h} X_t \times \mathbb{1}_{i \in g(t)} \\ & + \delta_{c,t} + \delta_{1,t} + \delta_{e,t} + \delta_{\text{city},t} + \delta_{\text{age},t} + \mathbb{1}_{i \in s} \times \mathbb{1}_{\text{year}} + \mathbb{1}_{i \in s} \times \mathbb{1}_{\text{month}} + \epsilon_{i,t+h} \end{aligned}$$

- Baseline specification uses global commodity price shocks as instruments
- But only second stage is reported. How does the **first stage** look? Are the commodity price shocks strong instruments?
 - Important for inference

Empirical specification

Panel LP-IV:

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- Baseline specification estimates effects of food and oil price shocks **jointly**
- Controlling contemporaneously for food prices when looking at oil price shock nets out any potential effects of oil price shocks through higher food prices
 - Less of an issue in IV specification, problematic in OLS
 - My suggestion would be to run two **separate** LP regressions

Empirical specification

Panel LP-IV:

$$\begin{aligned} c_{i,t+h} - c_{i,t-1} = & \beta_{0, \text{food}}^{g,h} \text{ext}_t^{\text{food}} \times \mathbb{1}_{i \in g(t)} + \beta_{0, \text{oil}}^{g,h} \text{ext}_t^{\text{oil}} \times \mathbb{1}_{i \in g(t)} + \sum_{j=1}^J \alpha^h (c_{i,t-j} - c_{i,t-j-1}) \\ & + \sum_{k=1}^K \beta_{k, \text{food}}^h \text{ext}_{t-k}^{\text{food}} + \sum_{k=1}^K \beta_{k, \text{oil}}^h \text{ext}_{t-k}^{\text{oil}} + \sum_{d=0}^D \delta^h D_{t-d} + \gamma^{g,h} X_t \times \mathbb{1}_{i \in g(t)} \\ & + \delta_{c,t} + \delta_{1,t} + \delta_{e,t} + \delta_{\text{city},t} + \delta_{\text{age},t} + \mathbb{1}_{i \in s} \times \mathbb{1}_{\text{year}} + \mathbb{1}_{i \in s} \times \mathbb{1}_{\text{month}} + \epsilon_{i,t+h} \end{aligned}$$

- Baseline specification uses many **fixed effects**, e.g. to account for state-specific trends
- If shocks are truly exogenous, these controls are not necessary. How sensitive are results to inclusion of fixed effects?
 - Important to understand where identifying variation is coming from

- Why are distributional effects of oil and food price shocks **so different**?
- It would be useful to discuss the **food and fuel consumption shares** more prominently
 - Food share is over 50% even for the top decile of households. Is that realistic? How is food defined?
- Earnings heterogeneity seems to play an important role. But **what is driving** earnings heterogeneity?
 - Do you have information on occupation? Can you look at households with occupations in agricultural sector?

- What about heterogeneity in **financial income**?
 - You show some aggregate responses but don't you also have information on this in the survey?
 - Could this help explain the relatively stronger response of richer households to oil price shocks?

- Robustness to commodity price shock instruments:
 1. Oil supply shocks: Caldara, Cavallo, and Iacoviello (2019), Känzig (2021)
 2. Food price shocks: De Winne and Peersman (2016)
- Use Brent or more relevant benchmark for Indian oil prices
- Be more transparent about the size of the shock. One sd shock? By how much do prices change?
- Grouping households by permanent income or education
estimate permanent income using Mincerian-type regression

To sum up

- Great paper on a very timely and important topic!
- Impressive data effort
- Convincing new evidence for emerging markets