

# Early Year Milk Price and Child Stunting

## :Evidence from Zambia

Chungmann Kim(presenter)<sup>1</sup>   Kathy Baylis<sup>2</sup>

<sup>1</sup>Ag. and Consumer Economics Dept., University of Illinois Urbana Champaign

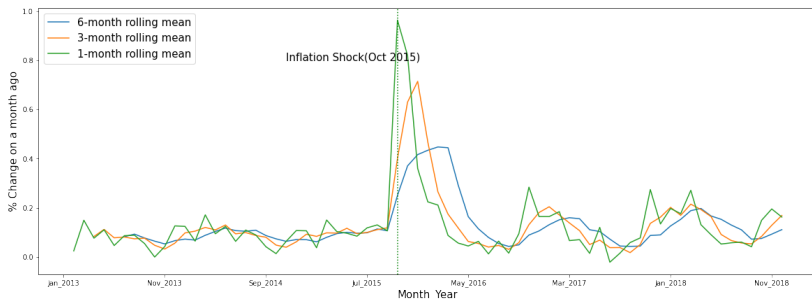
<sup>2</sup>Geography Dept., University of California Santa Barbara

31, Aug, 2023

# Introduction

## Price Shock in Zambia

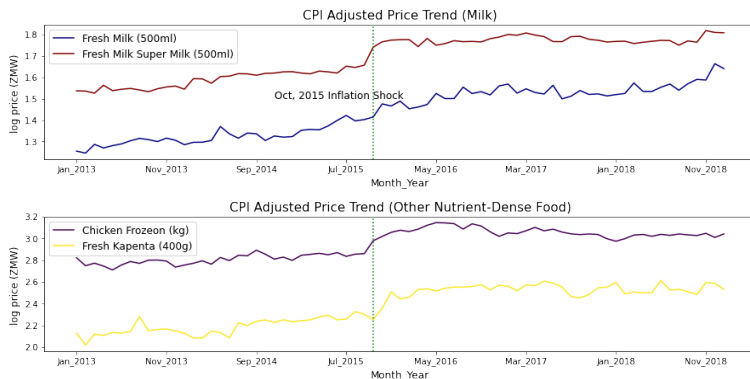
Zambia Currency (Kwacha) lost nearly 40% of its value against USD in late 2015 (Bertelsmann Stiftung, 2018)



Source : IMF CPI - Food data

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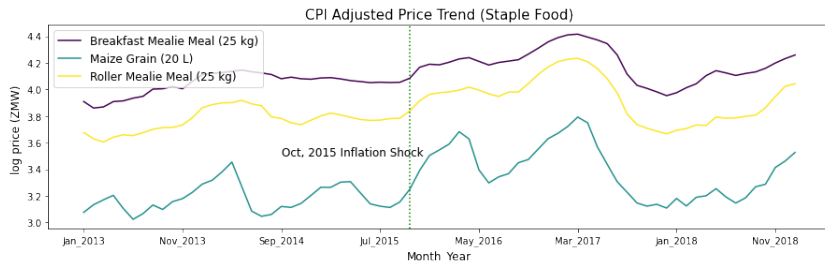
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Note : Author's calculation using Zambia Price Data (2013-2018)

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- Early-life chronic malnutrition carries long-lasting negative effects: diminished education, cognitive impairment, reduced earnings, and increased poverty risk. (Headey, Hirvonen, and Hoddinott, 2018).

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- Early-life chronic malnutrition carries long-lasting negative effects: diminished education, cognitive impairment, reduced earnings, and increased poverty risk. (Headey, Hirvonen, and Hoddinott, 2018).
- Few studies accounts for the biological period of exposure to economic shock (Woldemichael, Kidane, and Shimeles, 2022)



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- Does **timing of exposure** to price shock matter?
- **Which** socio-economic group are vulnerable to the price increase?

# Data

## Child Nutrition & Demographics

- Zambia Demographic Health Surveys (2018)
  - Outcome Variable : **Stunting** (=1 if Height for Age Z-Score < 2 SD)
  - Covariates : Maternal Education, Health Environment, and other demographics (Child level & HH level)

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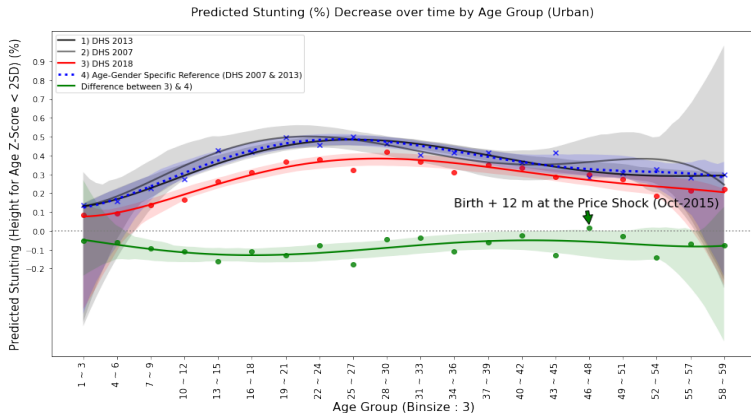
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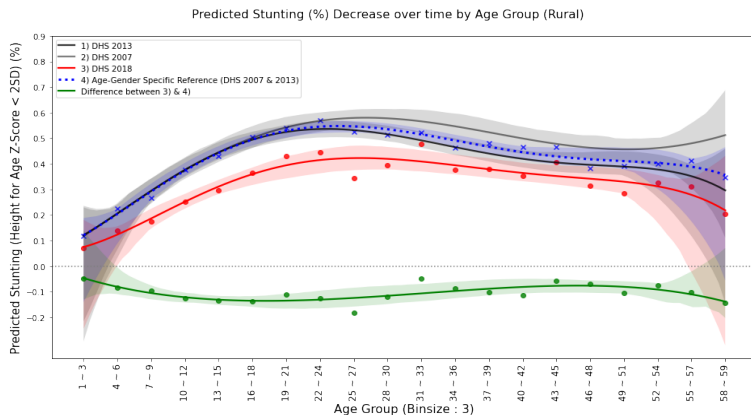
## Prevalence of Stunting over Time by Age Group (Urban VS Rural )



Note : Author's calculation using Zambia DHS 2007, 2013, 2018

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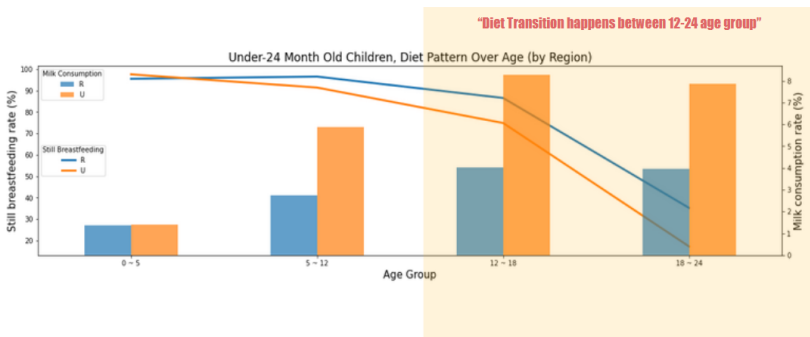
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# Context

## Diet Transition - from Breastfeeding to Milk Complement (Urban VS Rural)

- Urban children consume more milk, quit breastfeeding and are fed milk earlier than rural children.
- This transition largely occurs in the 2<sup>nd</sup> year (12m-24m).



Note : Author's calculation using Zambia DHS (2018) 24 hours Diet Module

# Model Specification & Estimation Strategy

Price Measures in the Key Developmental Period (12 - 24m)  
(Woldemichael, Kidane, and Shimeles, 2022)

$$\text{Early Year Average Price}_{b,d(t)}^C = (N - n)^{-1} \cdot \sum_{t=n}^N \cdot \text{Price}_{b,d(t)}^C$$

$d$  : district,  $b$  : year-month of birth for each child

$n$  and  $N$  : month indicators that define both start and end of key developmental period, and superscript

$C$  : commodity(e.g. fresh milk or mealie meal.)

# Model Specification & Estimation Strategy

## Model Specification

### Reduced Form Nutrition Production Function (Behrman and Deolalikar, 1988)

$$\begin{aligned}
 y_i = & \alpha + \delta_1 * (N - n)^{-1} \cdot \sum_{t=n}^N \log price_{b,d(t)}^{Fresh\ Milk} \\
 & + \delta_2 * (N - n)^{-1} \sum_{t=n}^N \log price_{b,d(t)}^{Mealie\ Meal} \\
 & + \gamma * HH_h + \zeta * Mother_m + \eta * Child_i + \theta_{FE} + \epsilon_i
 \end{aligned}$$

$y_i = Stunting_{2018} (= 1 \text{ if } HAZ \leq 2SD)$ ,  $d$  : district,  $b$ : birth cohort

$HH$  = wealth group dummy, number of under 5 children, head of sex,  $Mother$  = mother's education level, birth at first birth,  $Child$  = age by month, sex, birth order,  $FE \in \{\emptyset, cluster, household\}$

# Model Specification & Estimation Strategy

## Estimation Strategy

- Linear Probability Model (LPM)

LPM with fixed effects simplifies the interpretation of the coefficients, but it only provides the linear approximation to average partial effects.

- Correlated Random Effects (CRE) Probit Model

We model the unobserved cluster/household effect as a function of the data (Wooldridge, 2013) (Wooldridge, 2013). We add means of time-varying covariates (e.g. price measures) to the model and estimate them with random effects probit.

# Descriptive Statistics

We focus on children aged **24 - 60 months**, who completed the first two years of life.  
**No systematic differences** in nutritional status at birth between groups.

Table 1 : Balance At Birth (Cluster Level) - Age > 24m

	Age <12 month at Price Shock			Age >= 12 month at Price Shock			Diff
	N	Mean	SD	N	Mean	SD	
Birth Weight (kg)	1284	3.313	0.631	2615	3.227	0.602	-0.038
Birth Size							
Average	1719	0.618	0.485	3385	0.616	0.486	-0.001
Larger than Average	1719	0.192	0.394	3385	0.199	0.399	0.006
Smaller than Average	1719	0.079	0.269	3385	0.084	0.277	0.005
Very Large	1719	0.052	0.222	3385	0.052	0.222	0
Very Small	1719	0.023	0.152	3385	0.02	0.141	-0.003
First 3 days, Given anything other than breast milk	670	0.055	0.229	2324	0.059	0.237	0.005

# Estimation Result by Region (Urban vs Rural)

Milk price increase is statistically associated with a rise in stunting risk in **only in urban areas**.

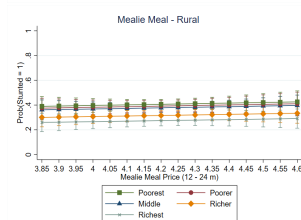
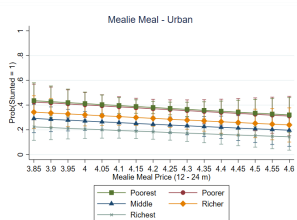
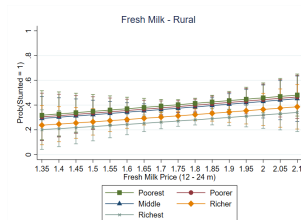
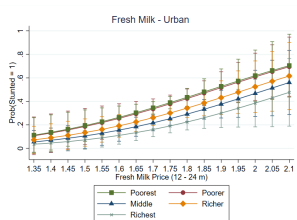
Table 1. LPM-FE vs CRE Probit Model and Average Marginal Effects

	Urban					Rural				
	(1) Base	(2) Cluster FE	(3) CRE Probit (Cluster)	(4) HH FE	(5) CRE Probit (HH)	(6) Base	(7) Cluster FE	(8) CRE Probit (Cluster)	(9) HH FE	(10) CRE Probit (HH)
<b>log Fresh Milk Price</b>	<b>0.340**</b>	<b>0.961**</b>	<b>0.906**</b>	<b>2.121**</b>	<b>1.711***</b>	-0.014	0.176	0.242	-0.378	-0.362
	(0.154)	(0.431)	(0.365)	(0.978)	(0.691)	(0.084)	(0.242)	(0.244)	(0.515)	(0.378)
<b>log Mealie Meal Price</b>	-0.089	-0.203	-0.163	-0.195	0.012	0.090	0.062	0.047	0.205	0.207**
	(0.116)	(0.172)	(0.168)	(0.470)	(0.171)	(0.075)	(0.097)	(0.098)	(0.230)077	(0.328)
Observations (Number of Groups)	1431	1431 (198)	1431 (198)	1431 (1328)	1431 (1328)	3453	3453 (347)	3453 (347)	3453 (3020)	3453 (3020)
Controls										
Household Wealth (1-5)	Yes	Yes	Yes	No	No	Yes	Yes	Yes	No	No
Child Characteristics	No	Yes	Yes	Yes	Yes	No	Yes	Yes	Yes	Yes
Household Characteristics	No	Yes	Yes	No	No	No	Yes	Yes	No	No
Mother's Characteristics	No	Yes	Yes	Yes	Yes	No	Yes	Yes	Yes	Yes
Cluster Fixed Effects	No	Yes	No	No	No	No	Yes	No	No	No
Household Fixed Effects	No	No	Yes	Yes	Yes	No	No	Yes	Yes	Yes
Cluster / HH Mean Values Included	No	No	Yes	No	Yes	No	No	Yes	No	Yes

Notes : Coefficients and standard errors are from separate models. Cluster/HH mean values are included for time varying price measures. CRE probit standard errors are clustered bootstrap estimates, while LPM standard errors are corrected for each clustering level. Each model includes different set of covariates (see bottom of each column), but results are not shown. Asterisks \*\*\*, \*\*, and \* indicate significance at the 1%, 5%, and 10% levels, respectively.

# Estimation Result : Heterogeneity by Wealth Group

Children from **poorer households** in **urban area** are more vulnerable to **milk price hikes**.



Note : Figures are from model (3) and (8) from Table 1

# Robustness Test 1

## Adding Weather Extreme Events (e.g. Drought and Extreme Hot)

Models are robust to **weather extreme events**.

Table 2. Robustness Test

	CRE Probit							
	Urban				Rural			
	Price Only (Table 1) Cluster	HH	Price & Weather Events Cluster	HH	Price Only (Table 2) Cluster	HH	Price & Weather Events Cluster	HH
log Fresh Milk Price	0.906** (0.261)	1.711** (0.691)	0.685* (0.381)	1.613*** (0.500)	0.242 (0.244)	-0.362 (0.378)	0.232 (0.176)	-0.471 (0.372)
log Mealie Meal Price	-0.163 (0.168)	0.012 (0.171)	-0.082 (0.173)	-0.762 (0.136)	0.047 (0.098)	0.207** (0.682)	0.014 (0.097)	0.188*** (0.072)
Observations (Number of Groups)	1431 (198)	1431 (1328)	1423 (198)	1423 (1303)	3453 (347)	3453 (3020)	3147 (347)	3147 (2768)

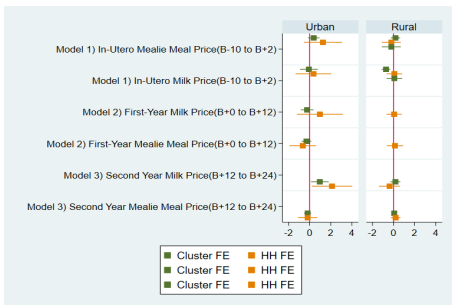
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## Robustness Test 2

### Models using Price Measures with Different Time Boundaries

Milk prices are only positive and significant when exposure occurs between **12 - 24m** in **urban area** both at cluster & household level.



Note : Marginal effects from models with different time boundaries

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Thank you for your attention

Contact:  
ck24@illinois.edu

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