

Anemia in Senegal: The Effect of the 2008 World Food Price Crisis

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Abstract

I explore the effect of the 2008 world food price crisis on the health of Senegalese children who were in-utero at the time. While previous work has looked at the effect of the 2008 crisis on child health, mine is the first to use a fetal origins approach, which allows data collected several years after the crisis to be used in analysis. I find suggestive evidence that the 2008 crisis had a negative impact on maternal health in urban areas, and that its effect was diminishing with distance from Dakar. I also find that the effect of exposure in-utero is apparent in 2010 but not in 2012, suggesting dissipation over time. Because iron is a critical nutrient in early-life development, this dissipation masks an important negative shock to health.

Introduction

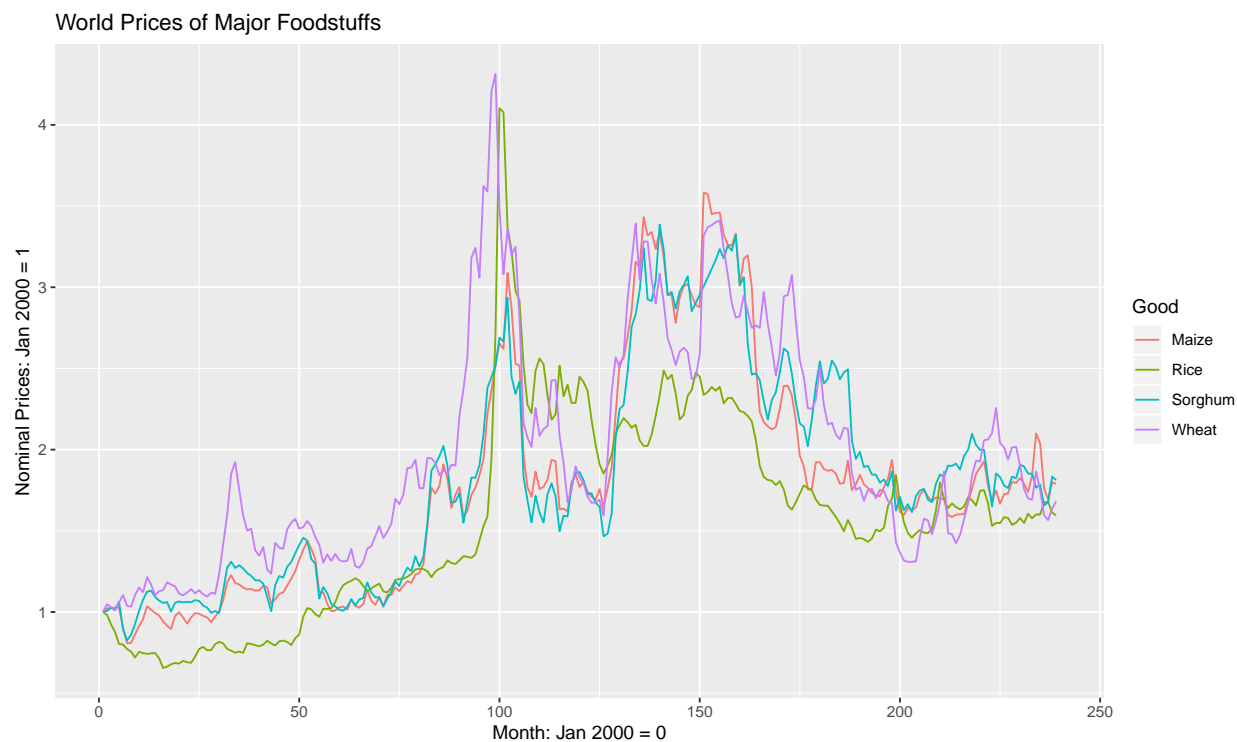
This project explores the effect of the 2008 world food price crisis on infant anemia in Senegal. Anemia is characterized by low levels of hemoglobin in the blood and can be caused by deficiencies in iron, folate and other micro-nutrients. By WHO estimates, 1.62 billion people around the world suffer from anemia. Anemia disproportionately affects women and the poor. Most cases of anemia are caused by iron-deficiency due to malnutrition. In children, anemia impairs cognitive and motor development. In adults, anemia causes fatigue and for pregnant women increases the probability of death during childbirth. In 2004, the World Bank estimated that anemia was responsible nearly one million deaths per year.

Maternal anemia has a direct effect on infant anemia. Unless remedied through increased iron consumption, anemia will remain as the infant grows into a child. During menarche, iron requirements increase for girls, and many become anemic. Thus an anemic mother begets anemic child who becomes an anemic woman who begets an anemic child. This inter-generational aspect of anemia is well appreciated by the WHO and is recognized as an important issue at the nexus of public health, nutrition and female empowerment (Camaschella 2015).

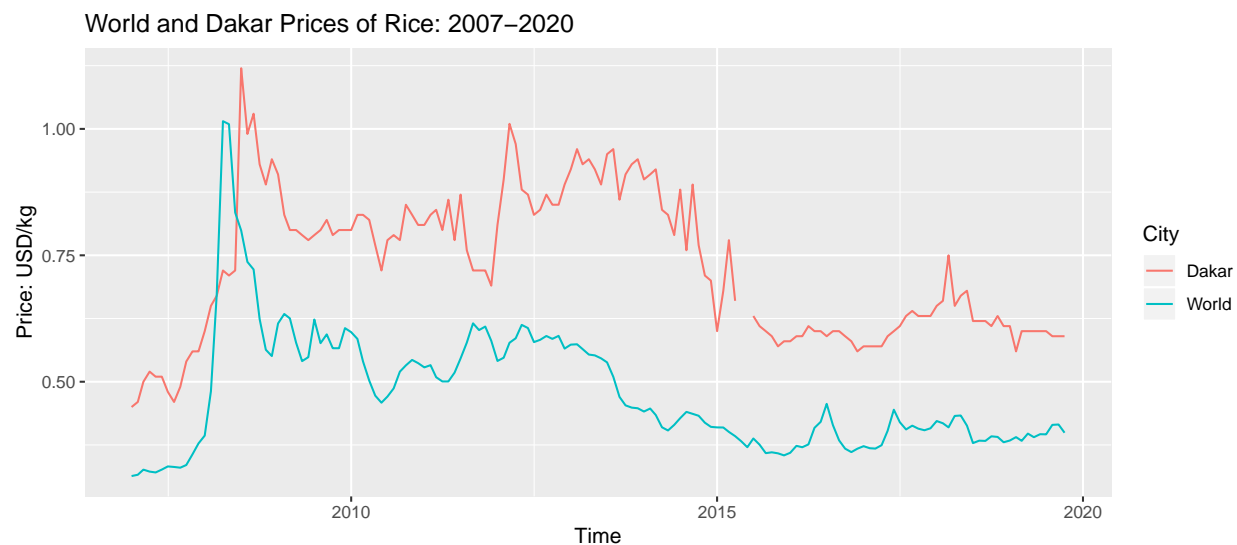
Anemia is a major public health concern. I use the unforeseen and intense increase of world food prices in 2008 as an instrument to analyze the intergenerational effects of maternal anemia. By looking at childhood hemoglobin data in Senegal 2-4 years after the crisis, I am able to both quantify the effect of the 2008 crisis and examine the effect of in-utero shocks on anemia in childhood. Both the causes of anemia and the consequences of the 2008 crisis have been explored extensively by researchers. I am, to my knowledge, the first to connect the two.

2008 World Food Price Crisis

After an extended period of low and stable food prices, world food prices increased dramatically from February to April, 2008. The causes are debated, but include American agriculture policy as well as financial speculation (Rosset 2008). The prices of both rice and wheat doubled in fewer than 6 months, while the prices of other staples increased as well. This increase led to riots and unrest in low to medium-income countries around the world, including Senegal.



The world price spike is apparent in retail prices in Senegal. Here I plot the retail price of rice in Dakar. Rice prices follow a similar pattern to the world price, with a delay, shifted upwards by transportation costs and retail markup.



The 2008 crisis drew attention from the fields of peasant studies, nutrition, and development, as well as from Science and Nature.¹ The crisis was viewed as evidence of the risk poor countries take on by specializing

¹See Cohen and Garrett (2010); Hadley et al. (2011); Martin-Prevel et al. (2012); Minot (2010); Ruel et al. (2009); Wodon and Zaman (2008); Von Braun (2008); Barrett (2010).

in cash-crops for export and relying on imported food. Much of the rhetoric in these fields advocates for reduced integration into the world market, and more self-sufficiency in food production (Rosset 2008).

As emphasized by Sen (1980), food security depends on the combination of level and volatility of food entitlements. From the simplest model of comparative advantage, we know that trade will induce specialization. This specialization will increase average incomes, but the effect of specialization and trade on volatility is unresolved (Giovanni and Levchenko 2009; Burgess and Donaldson 2010). The 2008 crisis can be thought of as an extreme case of market volatility. My contribution is to quantify the effect of this unforeseen shock on maternal health in Senegal.

Existing studies in economics explore the effect of the 2008 crisis on health and welfare (Arndt et al. 2016; Dimova and Gbakou 2013). Dimova et. al. estimate welfare elasticity with respect to price in Cote d'Ivoire during the crisis and argue that the majority of rural dwellers actually benefited from the shock, while urban dwellers suffered. Arndt et. al. take a more similar approach to mine, looking at the effect of high food prices on child weight-for-height. The authors find that weight-for-height was negatively correlated with food price increases, but study is limited by its dataset, which relies on direct observation during and after the crisis.

Using a fetal origins approach opens a wider range of evidence, as data from 2-4 years post-crisis can be used. Having a broader sample allows me to disentangle the effect of the crisis from seasonal fluctuations, unlike Arndt et. al. (2016). By looking at hemoglobin, I gain improved sensitivity relative to weight-for-height, as hemoglobin is sensitive to both quality and quantity of food. Relative to Dimova et. al., my approach measures the direct effect of the 2008 crisis on maternal and fetal health, rather than a synthetic measure of welfare.

Moreover, the increased variation from the fetal origins approach allows me to test for spatial heterogeneity in the effect of the crisis. This will allow me to provide econometric evidence towards the discussion being held in adjacent fields about market integration and food sovereignty. Previous work has emphasized a rural-urban divide, with the urban poor suffering from increased food prices, and the rural poor minimally affected (Dimova and Gbakou 2013; Barrett 2010; Ruel et al. 2009). While it is true that rural-dwellers are more likely to be net food-sellers, this hypothesis overlooks the broad class of rural poor who are net food-consumers. I will be able to test both this rural-urban hypothesis, and my own distance hypothesis.

The Economy of Senegal: Food and Trade

Senegal is a small, poor economy on the west coast of Africa, with a population of 16 million and a GDP per capita of \$3,675 (PPP). The climate is Sahelian, with a rainy season. Dakar is the port and economic hub. The coastal areas practice horticulture and fishing, while the hinterlands close to Dakar engage in commercial groundnut production for export. The river basins in the north and south are farmed for rice, while millet, sorghum and maize are the staples in drier areas. See the FEWS Livelihood Zone Map below:

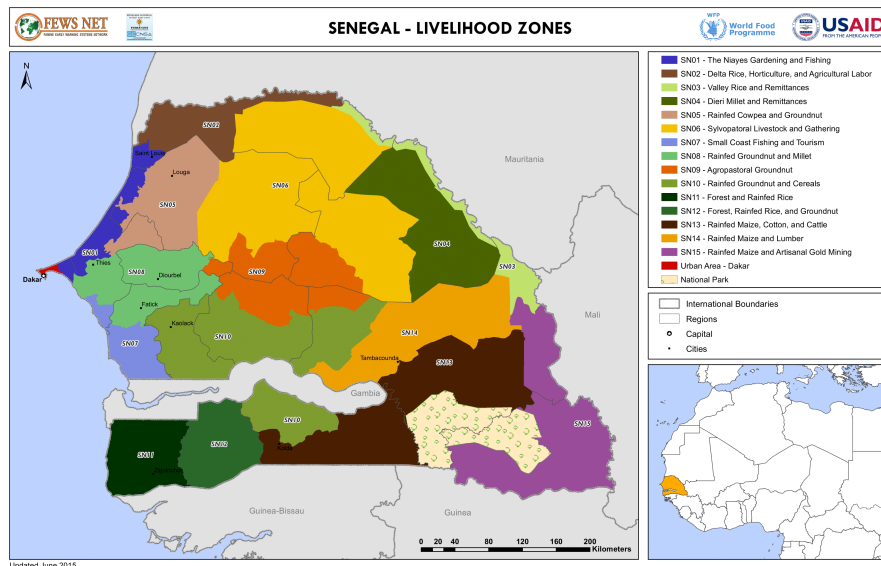
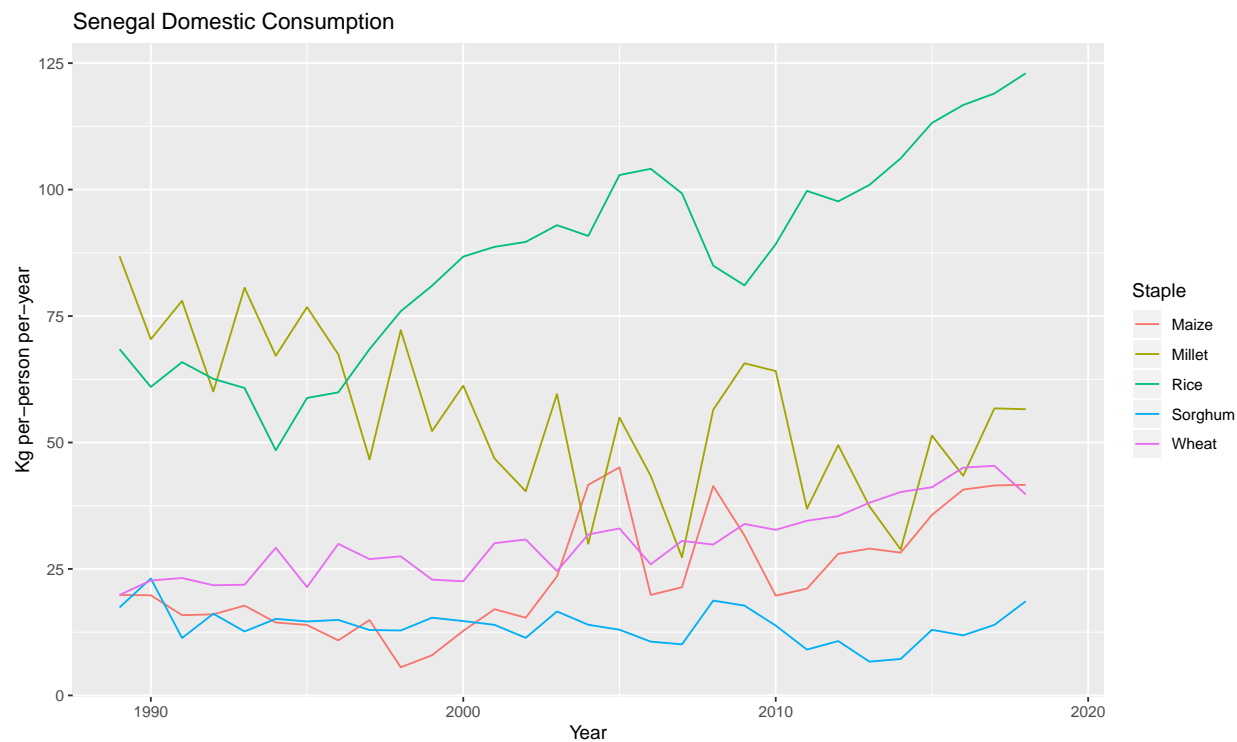
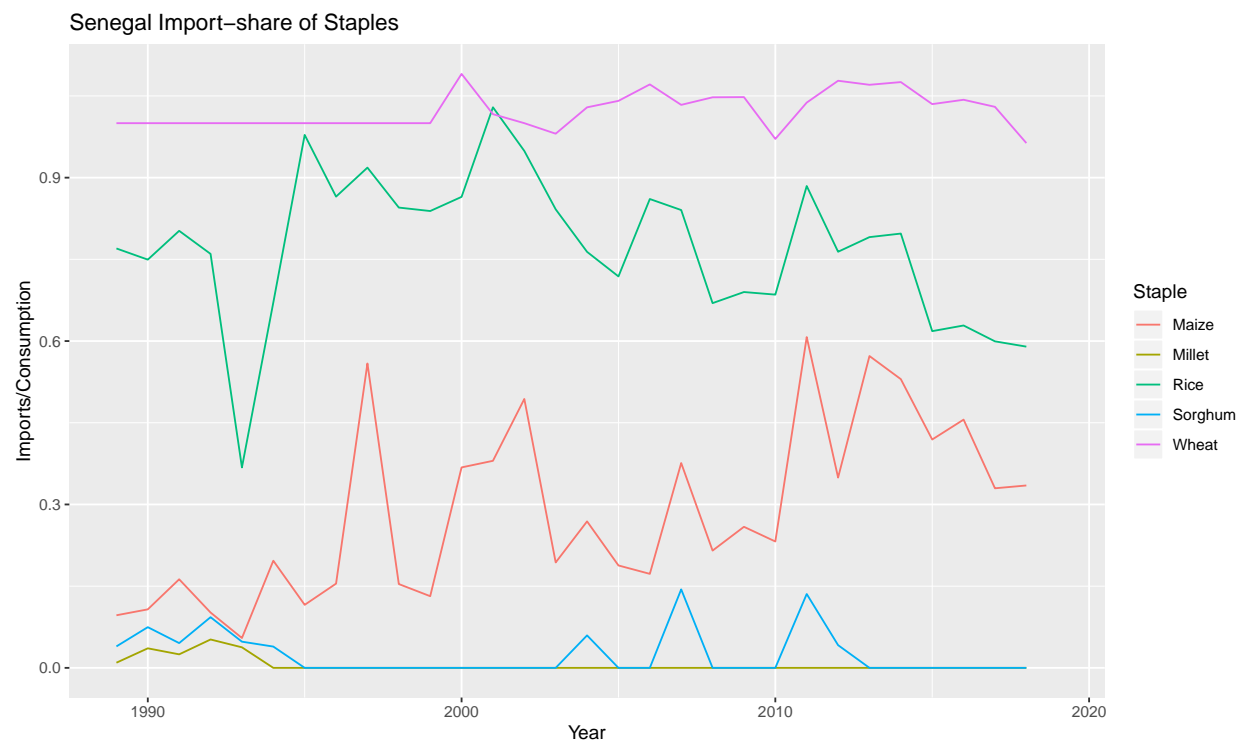
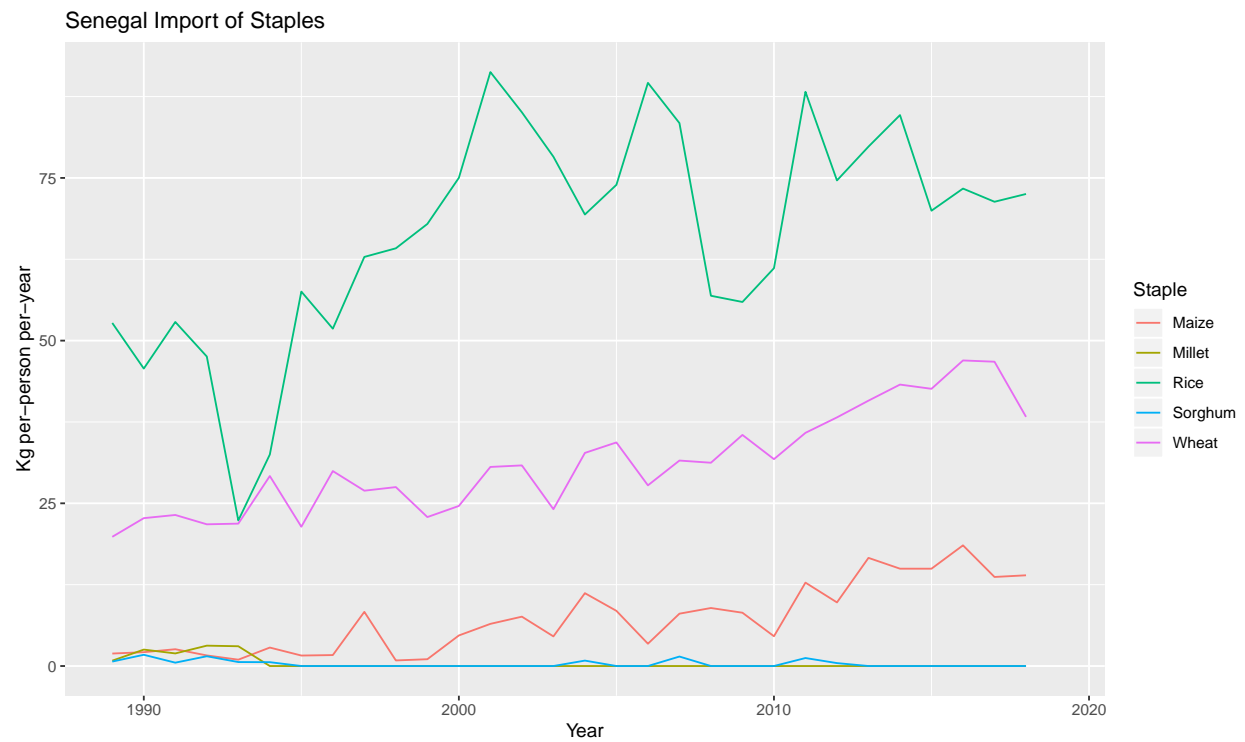


Figure 1: FEWs Livelihood Maps

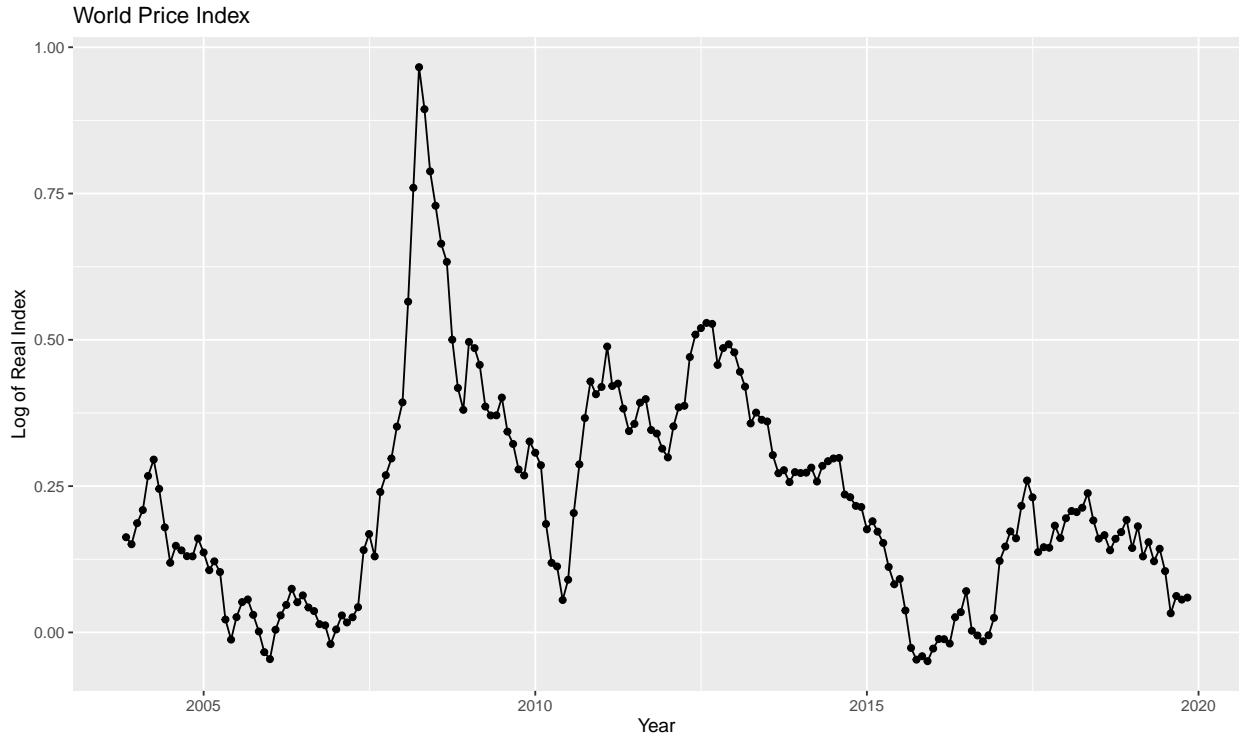
Although agriculture is the primary means of livelihood for 69% of the population, imports make up a large part of the food supply. The main staples are rice, maize, sorghum, wheat and millet. As seen in the graph below, foods which are mainly imported, like wheat and rice, have increased in importance in Senegal over the past thirty years, while locally produced staples millet and sorghum have declined in per-capita consumption.





We see that in 2008, when prices increased dramatically, consumption of millet and sorghum increased while that of rice decreased. This might reflect some substitution away from high-priced, imported goods. It would indicate some reserve capacity for local food production.

To construct a world price index that is relevant for anemia in Senegal, I use two weights. One, the amount of consumption of a particular staple in Senegal in 2008, and two, the iron per 100 grams of that staple.² I took the weighted average of the commodity prices in real USD, and multiplied by the CFA-USD exchange rate in order to get prices in terms of local currency. I plot the index in log-terms below, so that proportional changes can be easily read.



Mechanisms

To put it simply, this paper explores the relationship of world food prices while a child was in-utero to that child's hemoglobin level age 1-5. The chain of reasoning goes as follows:

1. An increase in the world price of food leads to an increase in local prices in Senegal;
2. which causes food consumption to decrease in quality and/or quantity;
3. which causes iron intake to fall for pregnant women;
4. which reduces fetal iron stores;
5. leading to reduced hemoglobin as an infant and child.

Each of these mechanisms has specific caveats and mediators. I go through each of these in order to think about what the regression should be.

1. World Price, Local Price and Transportation Costs

The effect of world price on local price will depend on transportation costs and local supply (Minot 2010). Consider a simple model. P_a is the price in autarky and is determined via local supply and demand. Because

²The final weights were 0.39 0.13 0.18 0.30 for rice, maize, sorghum and wheat respectively. I exclude millet from the study because it has not been imported or exported since the early 1990's.

Senegal is an small-open-economy (SOE), trade provides an infinitely elastic supply curve at price $P_w + t$, the world price plus transportation costs. Given an increase in the world price of food, there are three cases for how the local price will respond:

1. Food was not imported before the price increase, and it is not be imported after the increase. Price will be determined by local supply and demand, $\frac{\Delta P_L}{\Delta P_w} = 0$.
2. Food was imported before the price increase, but is not imported afterwards. In this case, the local supply and demand curves meet at a lower price than P_w . The economy moves up the local supply curve and $0 < \frac{\Delta P_L}{\Delta P_w} < 1$.
3. Food was imported before the price increase, and continues to be imported afterwards. In this case, the new price is still $P_w + t$ and $\frac{\Delta P_L}{\Delta P_w} = 1$.

Transportation costs also change the relative price of commodities faced by producers, affecting investment. Transportation costs make commercial, export-oriented production less profitable relative to local, subsistence production. Thus, we should also expect food-producing capacity to increase with transportation costs. Greater food-producing capacity would shift the local supply curve outward, making the world price less likely to bind.

2. Price and Consumption

As price increases there are substitution and income effects which lead to lower consumption. Given that the 2008 price increase affected all of the major, imported staples, I would expect the substitution effect to be relatively small.³ As poor families spend a greater share of income on food than rich families, the income effect, and therefore total impact of the spike, should decrease with household wealth.

Iron, unlike calories, is a function of both food quality and quantity. As prices increase, the household must either reduce quantity or change the type of food they are consuming. In Burkina Faso, which neighbors Senegal, the 2008 price shock lead to decreased amounts and diversity of food (Martin-Prevel et al. 2012). As I am particularly interested in female iron consumption, gender inequality looms large. While there are a host of issues surrounding intra-household allocation and gender, I will not treat those in this paper. Instead, I will assume that female iron consumption decreases with prices in a similar way as household iron consumption.

Given a price shock, households will try to smooth consumption across time, attenuating the effect of price on consumption. However, most households, especially the poor, are credit constrained. Thus, we would expect a nonlinear effect of price on consumption, with large increases in price having disproportionately large effect on consumption. As savings, and therefore ability to smooth in a credit-constrained world, are a stock, prices in previous months will also affect the ability to smooth consumption. If prices have been high for several months, then a household is less likely to be able to smooth consumption given another month of high prices.

3-5. Iron and Anemia in Mother and Child

Hemoglobin is typically limited by iron stores in the body, which depends disease environment and the quality and quantity of food intake. In a healthy person, there are iron-stores about equal to the amount of iron in use in the person's blood. There is a small, constant loss of iron. The bio-availability of iron of an individual is a stock, while iron intake is a flow, meaning that past iron consumption will affect current iron-stores and hemoglobin levels (Camaschella 2015).

Iron is passed from the mother to fetus, allowing the development of fetal hemoglobin, which differs from adult hemoglobin. Over the first four months of life post-birth, the infant breaks down the fetal hemoglobin

³One possibility is that the increase in the price of imported goods lead to increased consumption of local crops, like millet. Millet is relatively high in iron compared to rice, so this would attenuate any results I find.

and synthesizes adult hemoglobin. Little iron can be transmitted through breast-milk, so the fetus must develop sufficient stores of iron in order to make the transition from fetal to adult hemoglobin as they grow and their blood volume increases. The additional iron required for pregnancy is around 1,000 mg— nearly half of the total iron store in a healthy, non-pregnant woman (Lactation 1990).

Infants, on the other hand, need relatively little iron, given that they develop sufficient stores in-utero. Breast-milk contains very little iron, around 0.1 mg per gram (USDA). As illustration, consider that a typical infant would consume less than 700 grams of breast milk a day in the first six months (Neville et al. 1988), meaning a total iron intake of around 54 mg over the first six months. This is trivial compared to the 1,000 mg needed during pregnancy. Moreover, as long as the woman is breastfeeding, the absence of menstruation will reduce her iron loss rate, increasing her stores. Thus, during the first six months of life hemoglobin in infants should be relatively insensitive to maternal iron consumption.⁴ Here I list the CDC’s recommended daily intake of iron for mother and infant across different periods.

| Period | Maternal Iron Intake | Period | Infant Iron Intake |
|--------------|----------------------|-----------|--------------------|
| Pregnancy | 27 mg | | |
| Lactation | 9 mg | 0-6 mo | .27 mg |
| Menstruation | 18 mg | 7-12 mo | 11 mg |
| | | Childhood | 7 mg |

We see that iron availability is critical during the fetal period, but not during the breastfeeding period. This discontinuity in the response of infant hemoglobin to nutritional environment (very sensitive up until birth, and then insensitive) offers a fetal-origins identification strategy. Given a specific event, we can compare those who had already been born to those still in-utero.

Data

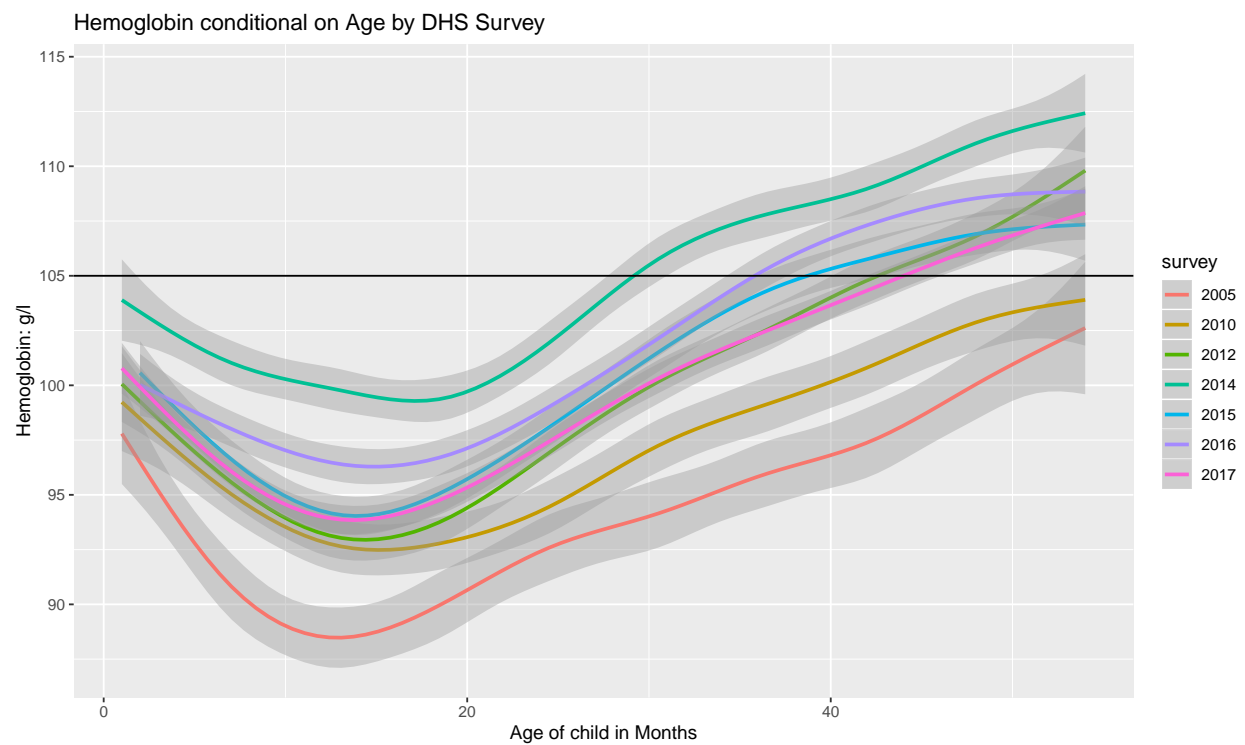
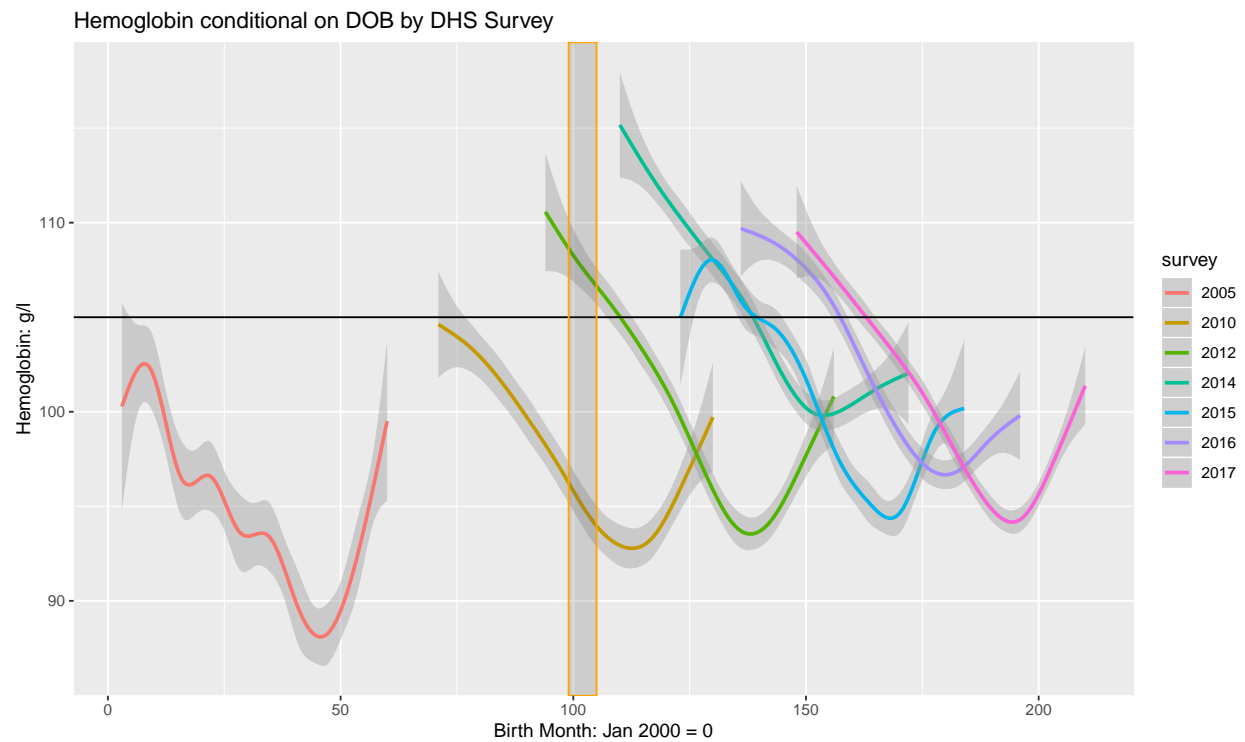
The DHS measures the hemoglobin of infants and children age 6 months to 5 years. There are two DHS surveys of Senegal, 2010 and 2012, which include children in-utero during the 2008 crisis. The DHS data includes location and households characteristics. The variables which I will use for this study are hemoglobin, child age and month of birth, household wealth, whether a household lives in an urban *commune* and driving-distance to Dakar.

| Urban | Survey | Wealth | Distance ⁵ | N |
|-------|--------|---------|-----------------------|------|
| Rural | 2010 | -0.0656 | 3.14 | 2252 |
| Rural | 2012 | -0.0680 | 3.01 | 3492 |
| Urban | 2010 | 0.0928 | 2.25 | 974 |
| Urban | 2012 | 0.0658 | 2.59 | 1573 |

Hemoglobin

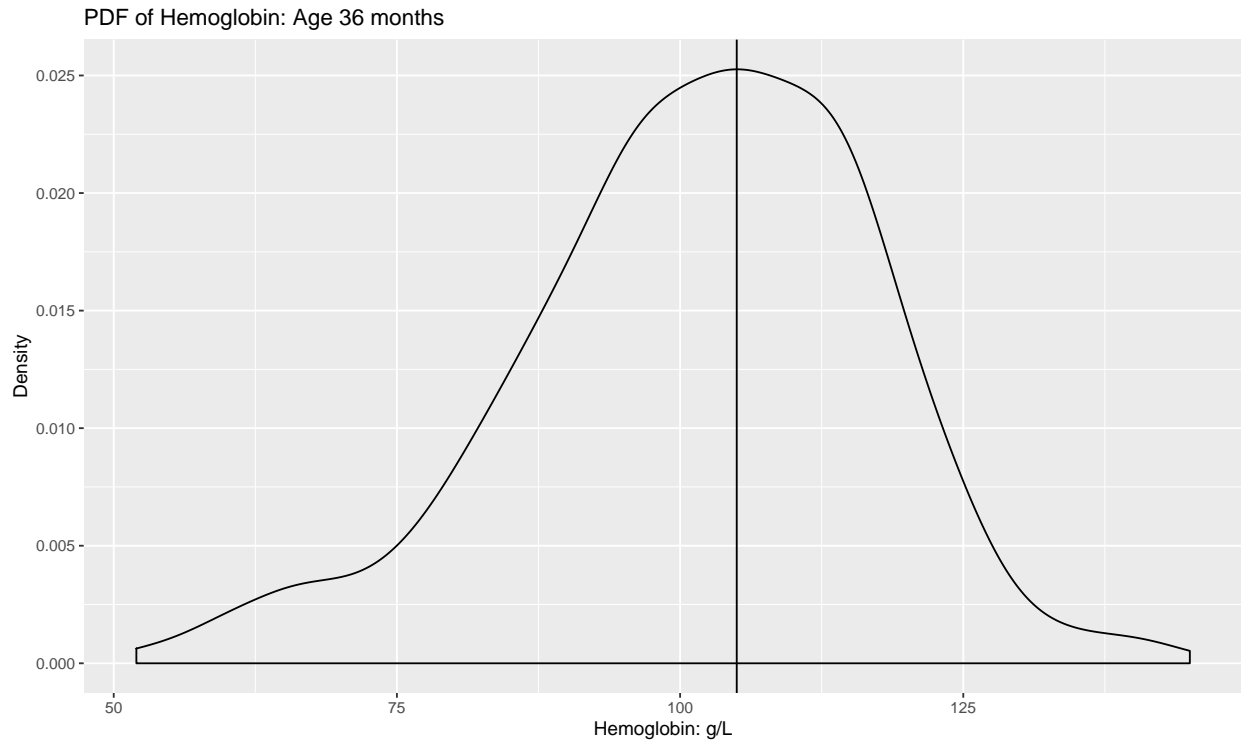
At birth, the infant has a large amount of fetal hemoglobin which must be broken down and replaced by adult hemoglobin. In general, hemoglobin levels decrease from birth until 6 months to 2 years, then they begin to increase again. We see such a pattern in the DHS data, which I plot below. The yellow window indicates the 2008 crisis.

⁴Note that this all relies on women breastfeeding their children. In Senegal, breastfeeding is overwhelmingly common during the first six months, with very little cessation before the child reaches a year old (Mané et al. 2006).

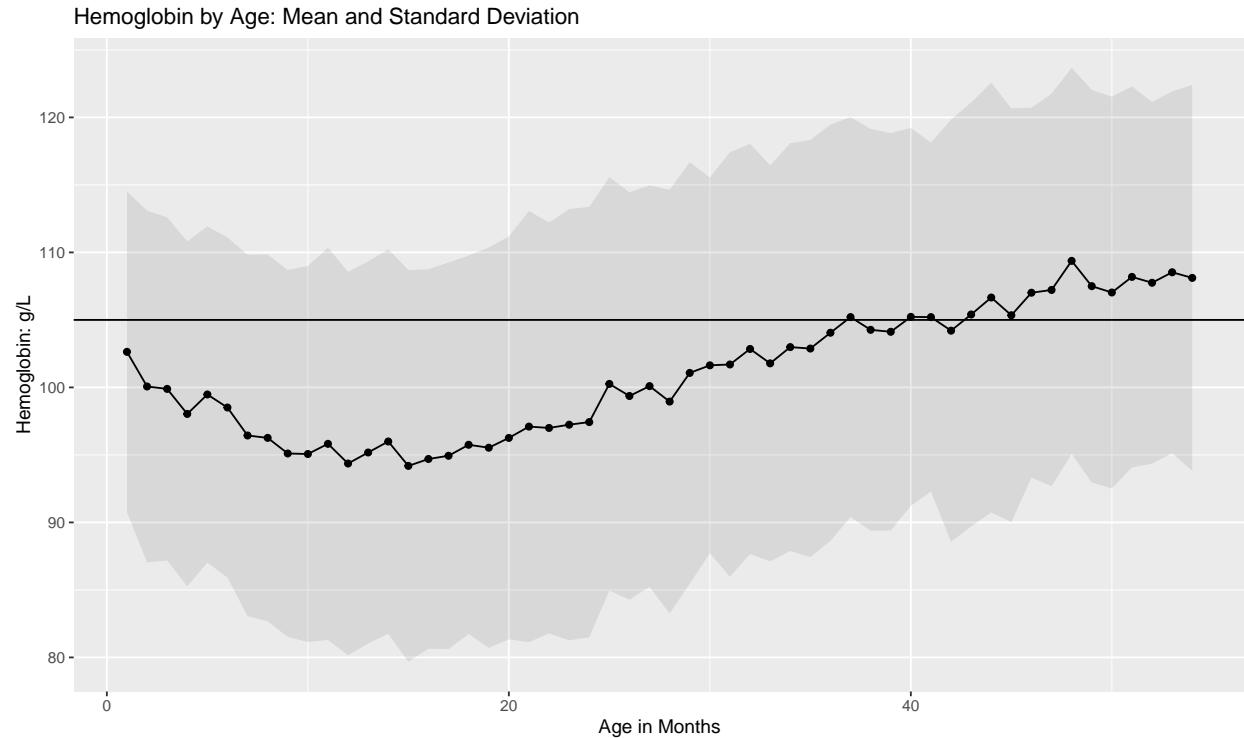


Anemia is relatively rare in children, compared to in infants or adults. Anemia resurges in girls upon menarche. The cutoff for anemia in infants and children, 105 grams per liter of blood is shown as a black, horizontal line.

Within an age group, hemoglobin follows a symmetric, uni-modal distribution. For reference, below I provide a the distribution of hemoglobin for children of age 36 months:



In order to account for age, I will use a z-score approach analogous to how child height and weight is treated by the DHS. We are interested in the 2008 price shock. The only surveys which provide information on the shock are 2010 and 2012, so I use the other 5 surveys (2005, 2014-17) to make z-scores for the surveys of interest. Taking the mean and standard deviation by age, I then normalize hemoglobin by age.

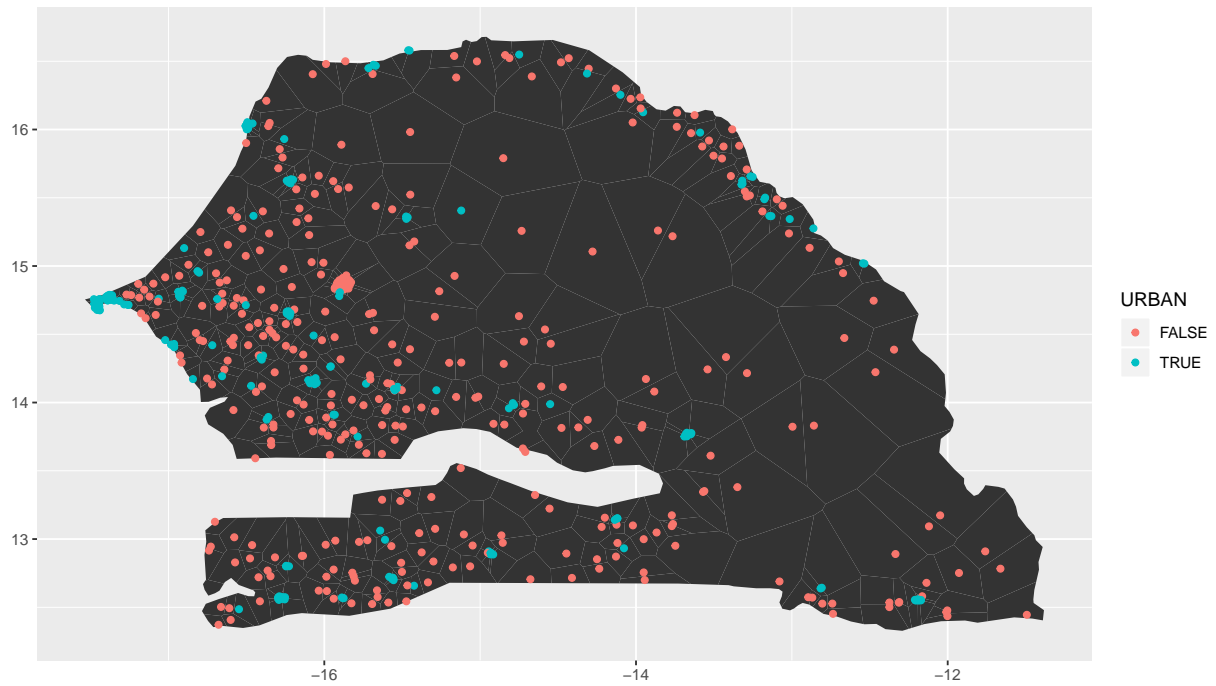


We see that most of the population is anemic during early childhood, suggesting that maternal iron is generally insufficient for the fetus to develop the iron-stores it requires. The standard deviation of hemoglobin by age is around 14-15 g/L. This will be important later in interpreting results, which are reported in standard deviations.

Other Variables of Interest

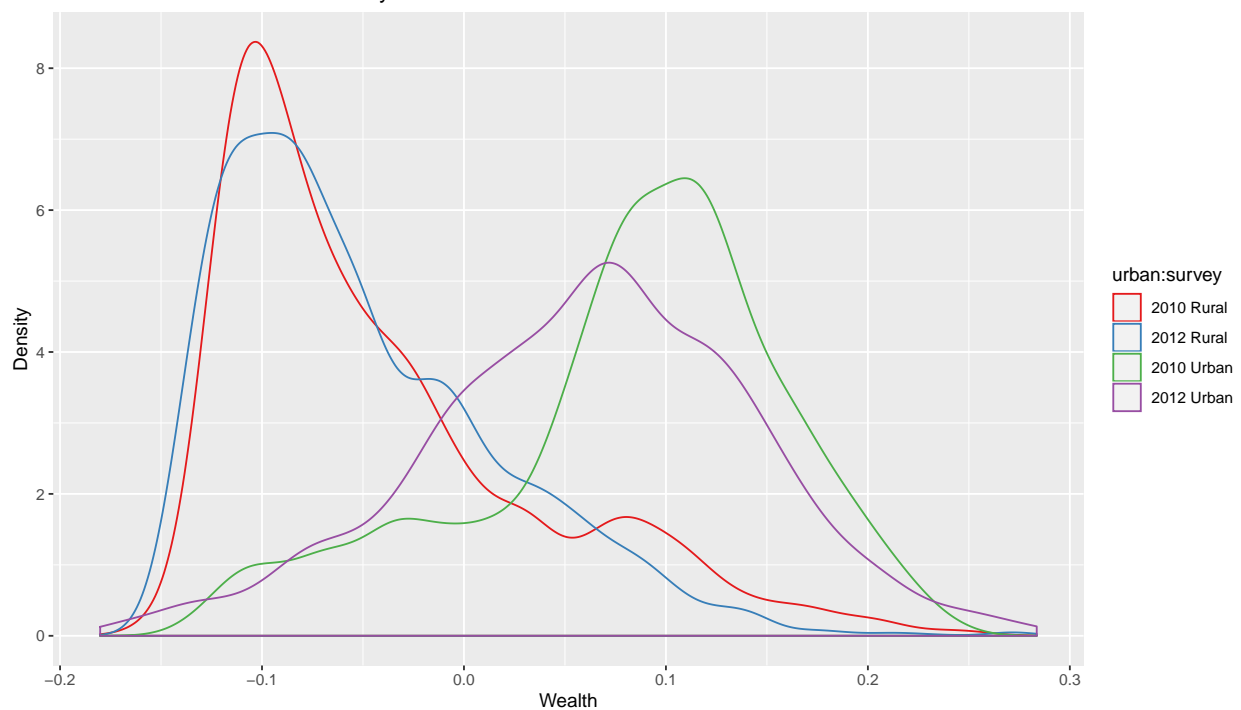
The DHS does not capture income, but it does measure wealth as a principle-components analysis of observed household possessions. Each DHS observation is tied to a “cluster”—generally a village or neighborhood. These clusters are geo-located with displacement, creating neutral measurement error. I use the coordinates of each cluster to measure the driving-distance to Dakar through the OpenStreetMaps API. The DHS also records whether a household is urban or rural. This urban-rural distinction is inherited directly from the census demarcation of the country being studied. In Senegal, urban is defined on the level of the *commune* (roughly translated as township). Below I plot the distribution of DHS clusters across Senegal, colored by whether they are urban or rural.

Spatial Distribution of Data

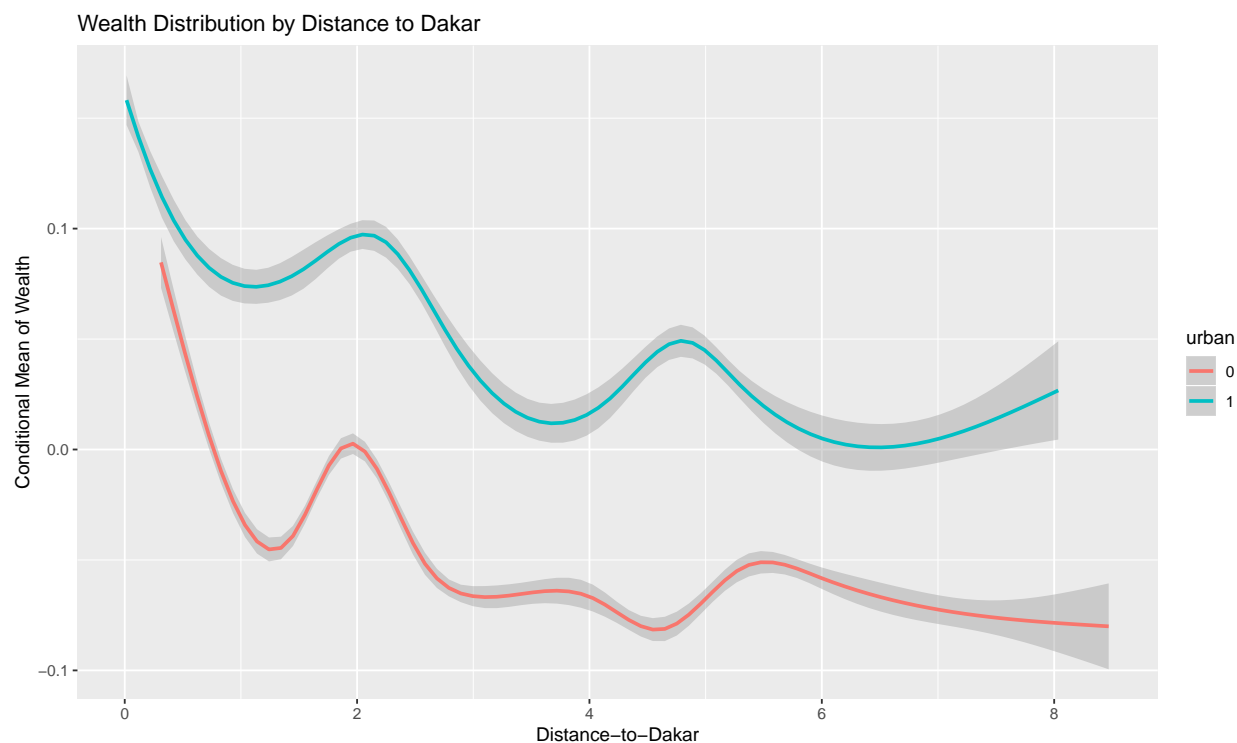


We will see below that the variable of interest, “exposure”, is balanced across observables. However, I will also be interacting exposure with other variables. These are not balanced. Consider the distribution of wealth across urban and rural areas:

Wealth Distribution across Surveys

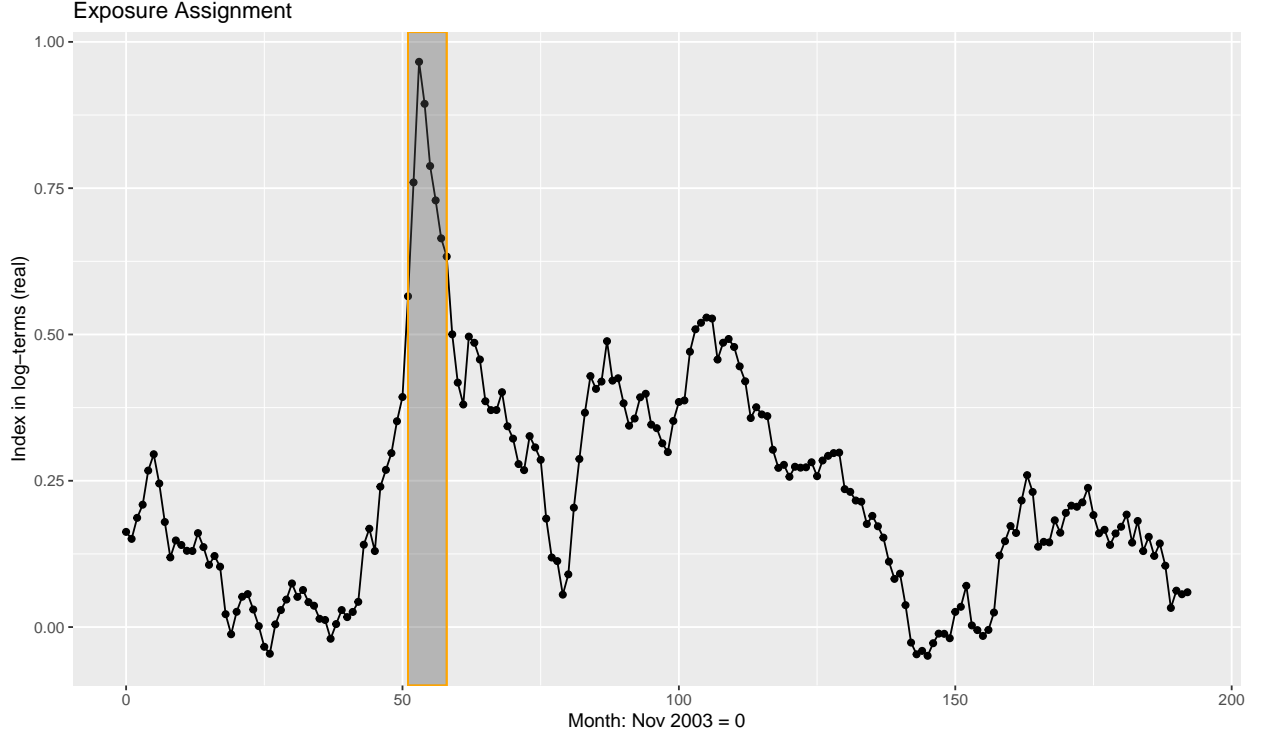


Urban areas are wealthier than rural areas. We also see that for both the rural and urban populations, mean wealth decreases as we move away from Dakar.



Methodology

The 2008 crisis forms a contiguous 8-month period in which the price index is greater than at any other point in the 20 year span. I will take all those in-utero during this period, February to September 2008, to be exposed. The degree of exposure will be the proportion of months of gestation which fall during the crisis period, weighted by the price index. For anyone born before March 2008, or conceived after September 2008, the in-utero exposure will be zero. For someone born in November 2008, the exposure measure would be $\frac{8}{9}$, as they were in-utero for the entirety of the crisis period.



Identification relies on exposure being as good as randomly assigned. Obviously exposure is collinear with month of birth, but it should be orthogonal to any other variables. The 2008 crisis was caused by the developments in large, rich economies and can be taken as exogenous to Senegal. Exposure, being as good as random, should then be balanced across other observables.

| | Distance-to-Dakar | Wealth | Urban |
|---------------------|---------------------|---------------------|--------------------|
| Distance to Dakar | -0.0009 (0.0005) | | |
| Wealth | | 0.0200* (0.0084) | |
| Urban | | | 0.0010 (0.0017) |
| R ² | 0.0001 | 0.0002 | 0.0000 |
| Adj. R ² | 0.0001 | 0.0001 | -0.0000 |
| Num. obs. | 24944 | 34677 | 34677 |
| RMSE | 0.1695 | 0.1459 | 0.1459 |

*** $p < 0.001$, ** $p < 0.01$, * $p < 0.05$

Table 3: Balance Test

While exposure is unrelated to distance to Dakar and being urban, it is correlated with wealth. Because wealth is positively related to hemoglobin, there will be omitted variable bias in the estimation of the effect of exposure. While we can control for wealth, the correlation is still disturbing, as exposure ought to be orthogonal to household characteristics by definition. The correlation could be due to selective mortality which disproportionately affected poor households. Such selection would place downward bias on my estimate of the effect of the crisis.⁶

For results, I present two kinds of evidence:

⁶I test for selection in the appendix, and find no evidence to support this hypothesis. It is difficult to rule it out completely.

1. The deviation from a quadratic time trend of the exposed cohorts, estimated separately by survey; and,
2. The median hemoglobin-for-age by birth-quarter.

I use median-regression, rather than least-squares, as it is less sensitive to heterogeneous response across the distribution.⁷

$$H_i = \alpha_0 + \beta Exp_i + \delta Exp_i * Dist_i + \alpha_1 MOB + \alpha_2 MOB^2$$

where H_i is hemoglobin-for-age, Exp_i is the number of months in-utero during the crisis, $Dist_i$ is the driving time to Dakar for the cluster of individual i . I will also estimate a more complex relationship, in order to test the distance hypothesis next to the urban-rural and wealth hypotheses discussed above, in order to speak to the broader literature:

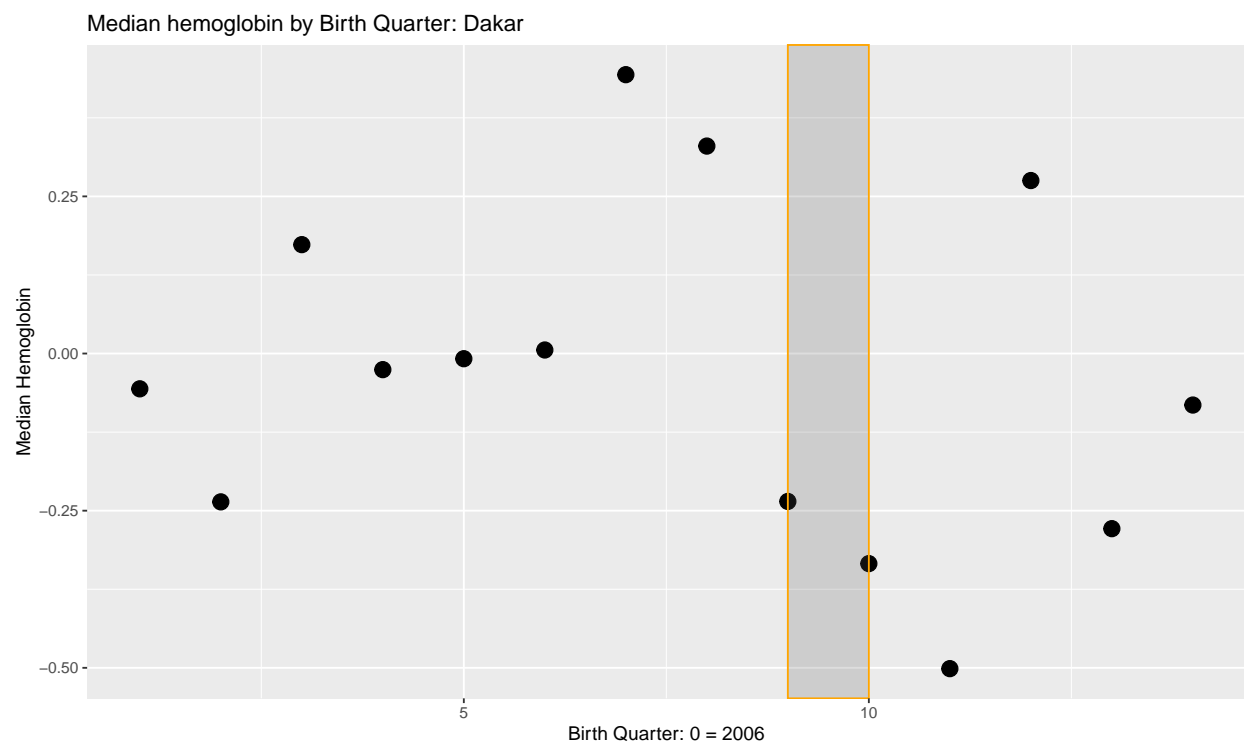
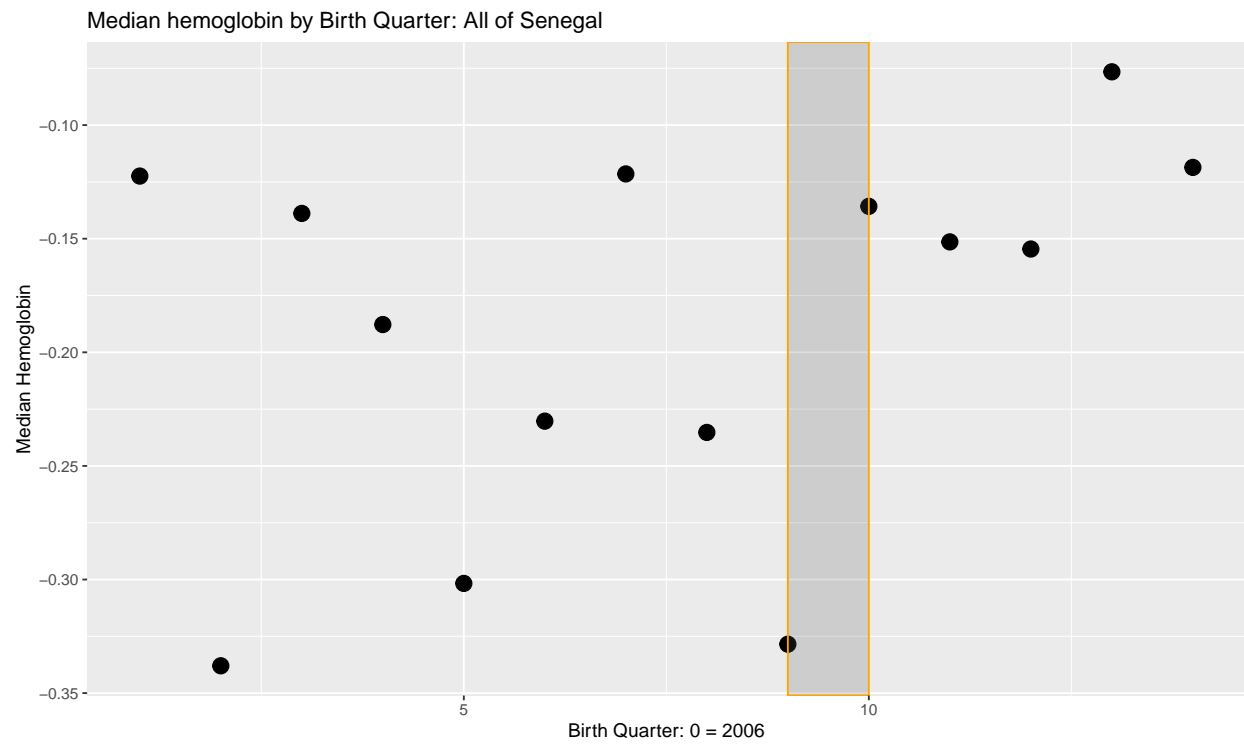
$$H_i = \alpha_0 + \beta Exp_i + \delta Exp_i * Dist_i + \alpha_1 MOB + \alpha_2 MOB^2 + \rho Exp_i * Wealth + \alpha_3 Dist_i + \alpha_4 Wealth$$

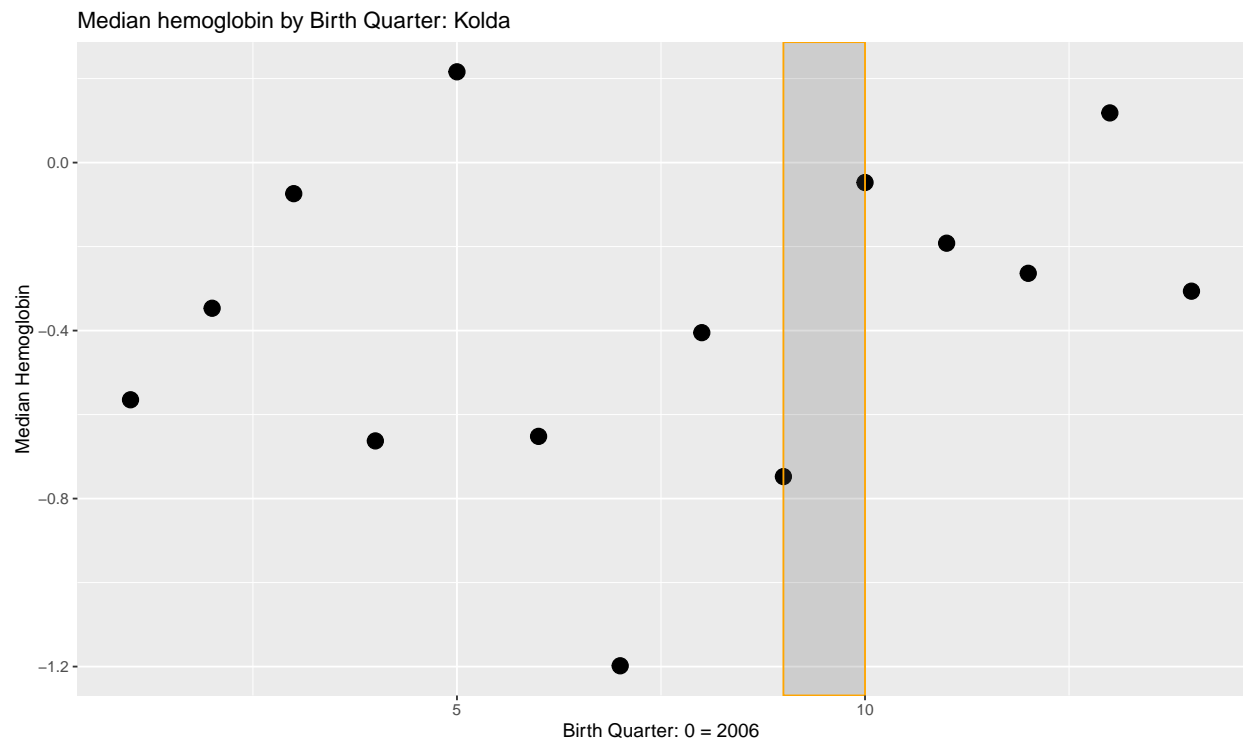
This regression is run separately for each survey year, and for urban and rural households. My hypotheses are that $\delta > 0$ and $\rho > 0$. In the first regression, I expect that $\beta < 0$, because distance-to-Dakar must be non-negative. However, in the second specification, The sign of β will depend on the underlying distributions of wealth, which can be negative.

Results

Here I plot the median hemoglobin by birth-quarter. I highlight the two quarters which are exposed to the 2008 crisis in-utero. First, I present the data for all of Senegal. If one squints, one may discern a dip. Second, I plot the data for Dakar. According to my predictions, this should be where the dip is most obvious. While certainly not obvious, there is a more apparent dip than for all of Senegal. Third, I present the data for Kolda, a region to the southeast of Dakar, relatively isolated and with significant food production. I would not expect to see much of a dip here, and no dip is evident.

⁷We would expect little to no response of hemoglobin to external shocks at the high and low ends of the distribution. At the high end, those in good health are presumably insulated from price shocks by SES. At the low end, a certain level of hemoglobin is required by the body, and it cannot diminish beyond some point without causing death.





This evidence is unclear. The effect of the 2008 crisis could not have been not uniformly large across the country, or else it would be apparent.

Next I turn to regression, in order to test the hypotheses stated above. I use quantile regression evaluated at the median, bootstrapped standard errors, and all regressions include a quadratic time trend.⁸

| | 2010 | 2012 | 2010 | 2012 |
|------------------------------|------------------------|--------------------|----------------------|---------------------|
| Intercept | -0.1737*** (0.0312) | 0.0060 (0.0186) | -0.0705 (0.0550) | 0.0239 (0.0385) |
| Exposure | -0.0200 (0.1047) | 0.0131 (0.0858) | -0.2175 (0.1551) | 0.0869 (0.1246) |
| Distance to Dakar | | | -0.0326* (0.0150) | -0.0058 (0.0120) |
| Distance to Dakar * Exposure | | | 0.0611 (0.0352) | -0.0310 (0.0355) |
| Num. obs. | 3226 | 5065 | 3226 | 5065 |
| Percentile | 0.5000 | 0.5000 | 0.5000 | 0.5000 |

*** $p < 0.001$, ** $p < 0.01$, * $p < 0.05$

Table 4: Parsimonious Regression Results

As expected from the graphs above, exposure alone does not have an impact on hemoglobin. However, once we allow the effect to depend on distance-to-Dakar, the 2008 crisis appears to have a negative effect on hemoglobin in 2010 but not in 2012. The result is not significant, but it is suggestive. We see below that once we break up the sample into urban and rural groups, and allow for the effect to differ by wealth, the effect of the crisis on urban-dwellers is large and significant. Thus, the small effect we see above is the average of the large effect on the urban-poor with the small (or even positive) effect on the rich or rural

⁸The results are stable to different polynomials in time.

part of the sample. Exposure has no effect on hemoglobin in 2012, which is consistent with the shock to hemoglobin dissipating over time.

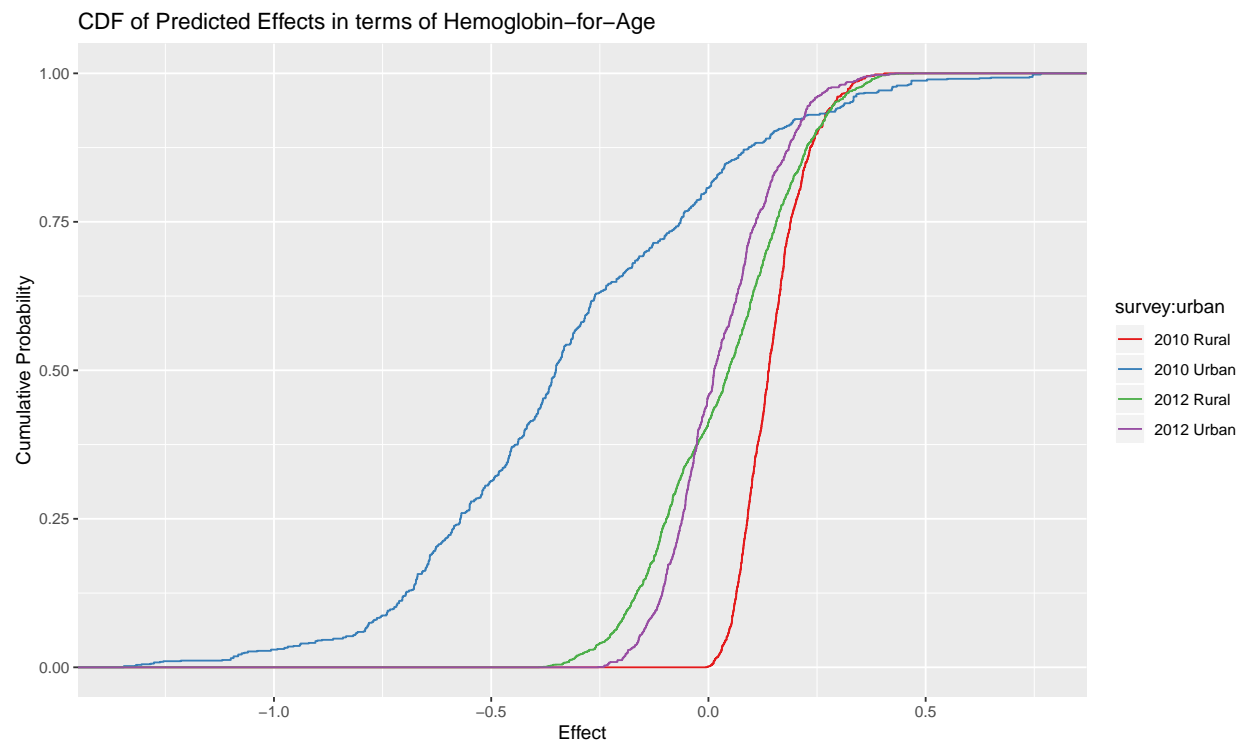
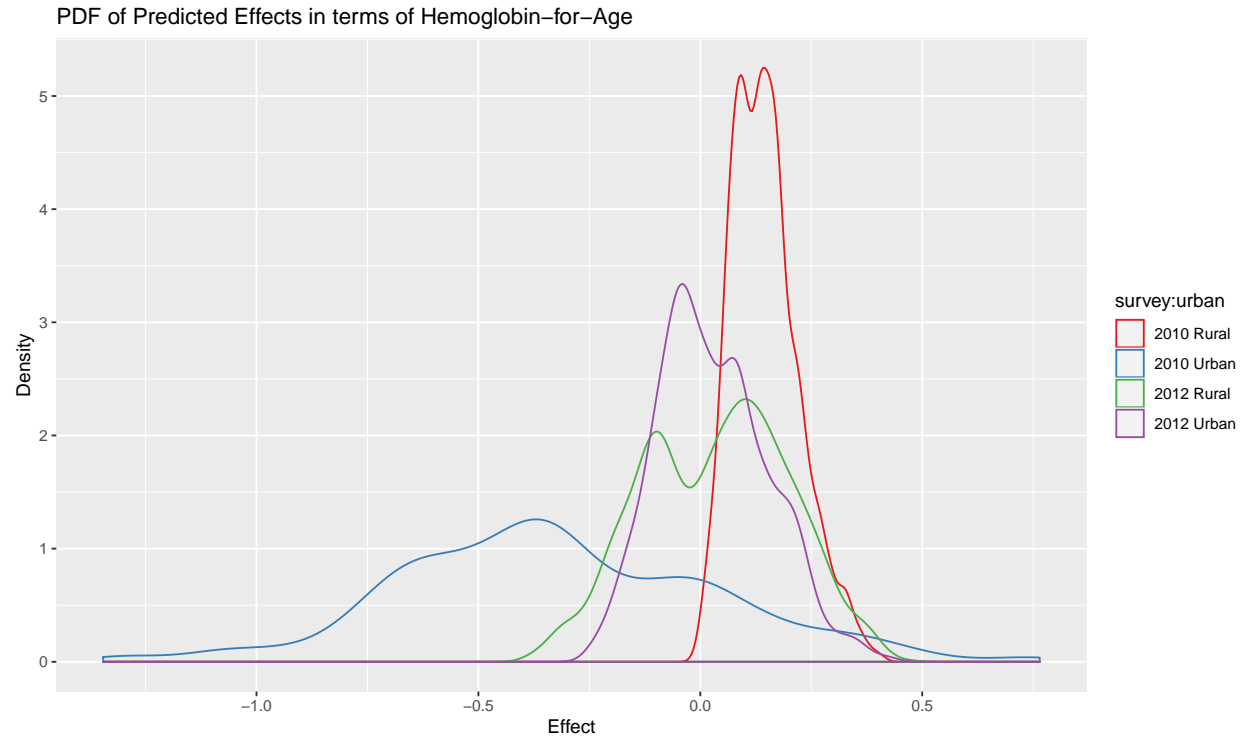
| | 2010: Urban | 2010: Rural | 2012: Urban | 2012: Rural |
|------------------------------|-----------------------|------------------------|---------------------|-----------------------|
| Intercept | −0.1367 (0.1104) | −0.2328*** (0.0705) | −0.0144 (0.0864) | −0.0342 (0.0513) |
| Exposure | −1.1218** (0.4034) | 0.0794 (0.1940) | −0.2327 (0.2875) | 0.3460 (0.1908) |
| Distance to Dakar | 0.0186 (0.0275) | 0.0056 (0.0185) | 0.0354 (0.0194) | 0.0139 (0.0139) |
| Wealth | 1.8144** (0.6263) | 1.8563*** (0.3841) | 1.2966* (0.5501) | 2.1454*** (0.3809) |
| Distance to Dakar * Exposure | 0.1804* (0.0806) | 0.0300 (0.0509) | 0.0684 (0.0664) | −0.0774 (0.0421) |
| Wealth * Exposure | 3.8185 (2.2887) | 0.9274 (1.5318) | 1.0001 (1.7194) | 0.8146 (1.1945) |
| Num. obs. | 974 | 2252 | 1573 | 3492 |
| Percentile | 0.5000 | 0.5000 | 0.5000 | 0.5000 |

*** $p < 0.001$, ** $p < 0.01$, * $p < 0.05$

Table 5: Regression Results by Survey and Urban-Rural

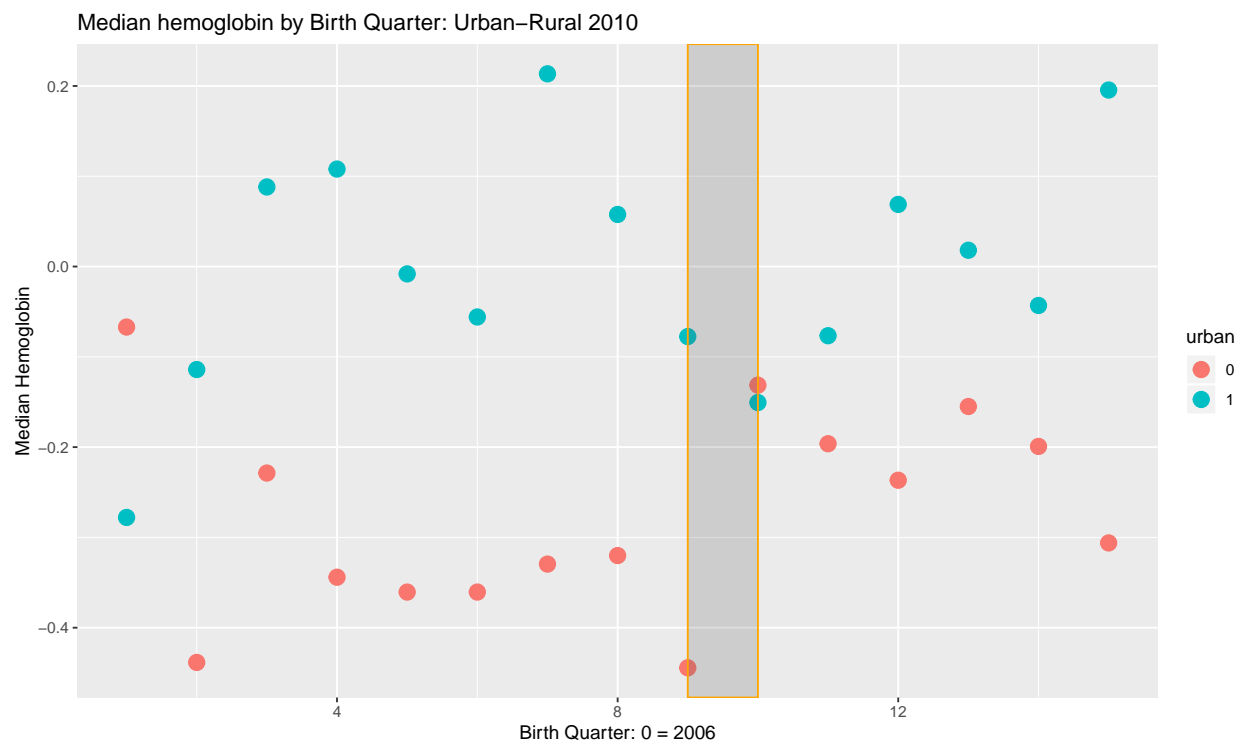
Here we see that exposure is negatively related to hemoglobin for urban-dwellers. The effect of exposure is greater for the poor across all specifications. For the rural population, the effect of the crisis on 2010 hemoglobin is noisy but slightly positive. In 2012, the results are attenuated, and the urban rural difference has become almost unnoticeable. The attenuation is expected because hemoglobin reflects iron stocks. The stocks will be affected by various other shocks over time, masking the signal from the 2008 shock.

With many interaction terms, interpretation depends on the underlying distribution of the covariates. The estimate on exposure above is for someone with *wealth* = 0 and living in Dakar. As Dakar is the wealthiest part of the country, only a very small fraction of the sample will be expected to have such an effect. For interpretation, I calculate the predicted effect of the 2008 crisis given the coefficients estimated above, given each individual's values of wealth and distance-to-Dakar. As we move to the right along the x-axis, the individuals are wealthier and live further from Dakar. Those at the far left of the distribution are the poorest and closest to Dakar.



The implied effects for the urban 2010 population are large. For the urban-poor, the effect ranges from -1 to -.5 standard deviations in 2010. The standard deviation of hemoglobin by age in Senegal is around 14-15 g/L. To understand the magnitude of this shocks, consider a 10 g/L shock to hemoglobin. If every child in Senegal was exposed to such a shock, rates of anemia would be expected to increase from ~60% to ~85%. As

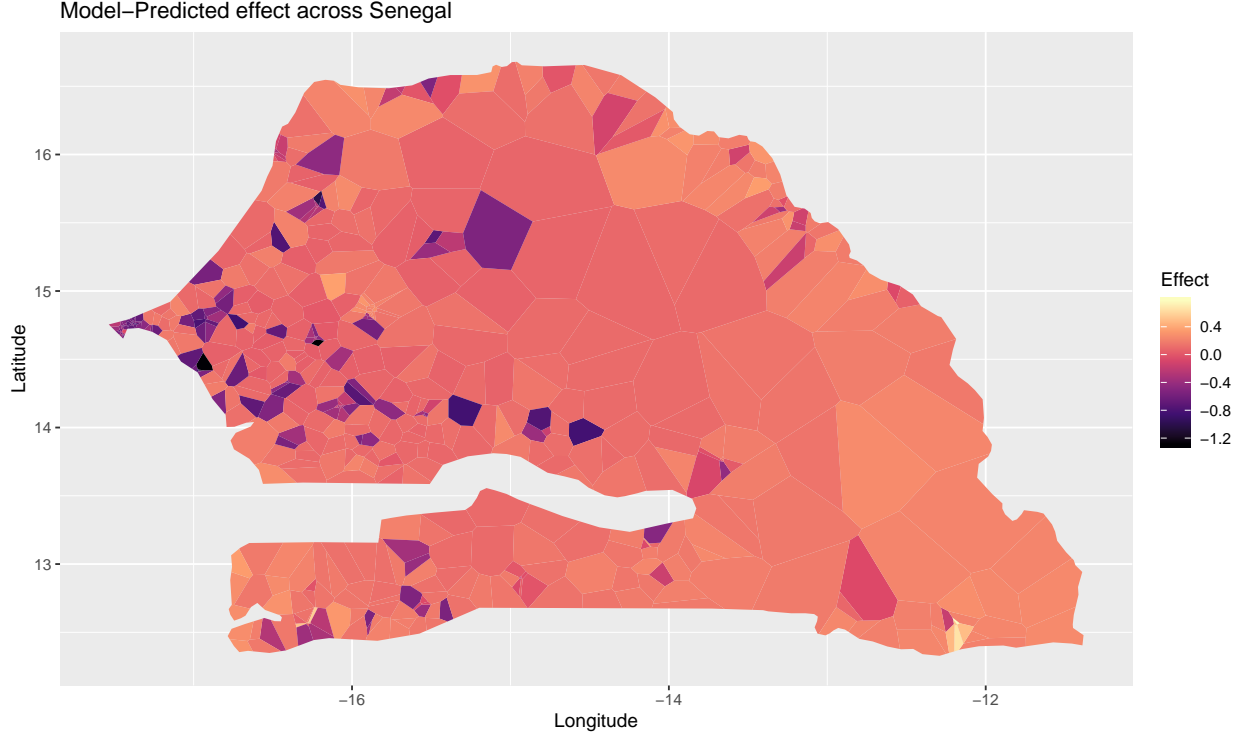
a further illustration, consider a typical poor household in Dakar (10th percentile of the wealth distribution within Dakar). The expected effect would be ~ 14 g/L of hemoglobin. For a child not to be driven into anemia after such a shock, they would have to have initial hemoglobin of 120 g/L, the 90th percentile of the overall distribution of hemoglobin in Senegal. The effect of the crisis is noticeable in the raw data. Below I plot the median hemoglobin by birth quarter, just as I did above, but this time split into rural and urban groups and only including observations from the 2010 survey.



I did not include this in the first set of such graphs because it was not created until after seeing the regression results above. Therefore, it should not be interpreted as *separate* evidence than the regression results above. Rather, it is an illustration of those results using the raw data. In contrast, the first three birth-quarter graphs I show were created *before* running regressions.

Spatial Distribution of Effects

To explore the spatial distribution of effects, I produce maps of the effect of the 2008 crisis across regions of Senegal. First, I use the coefficients from the 2010 estimates above to predict the effect of exposure by DHS cluster. Distance-to-Dakar is defined on the cluster level, and then I use the median wealth in each cluster to estimate the effect. I then plot the clusters, colored by the predicted effect of exposure.



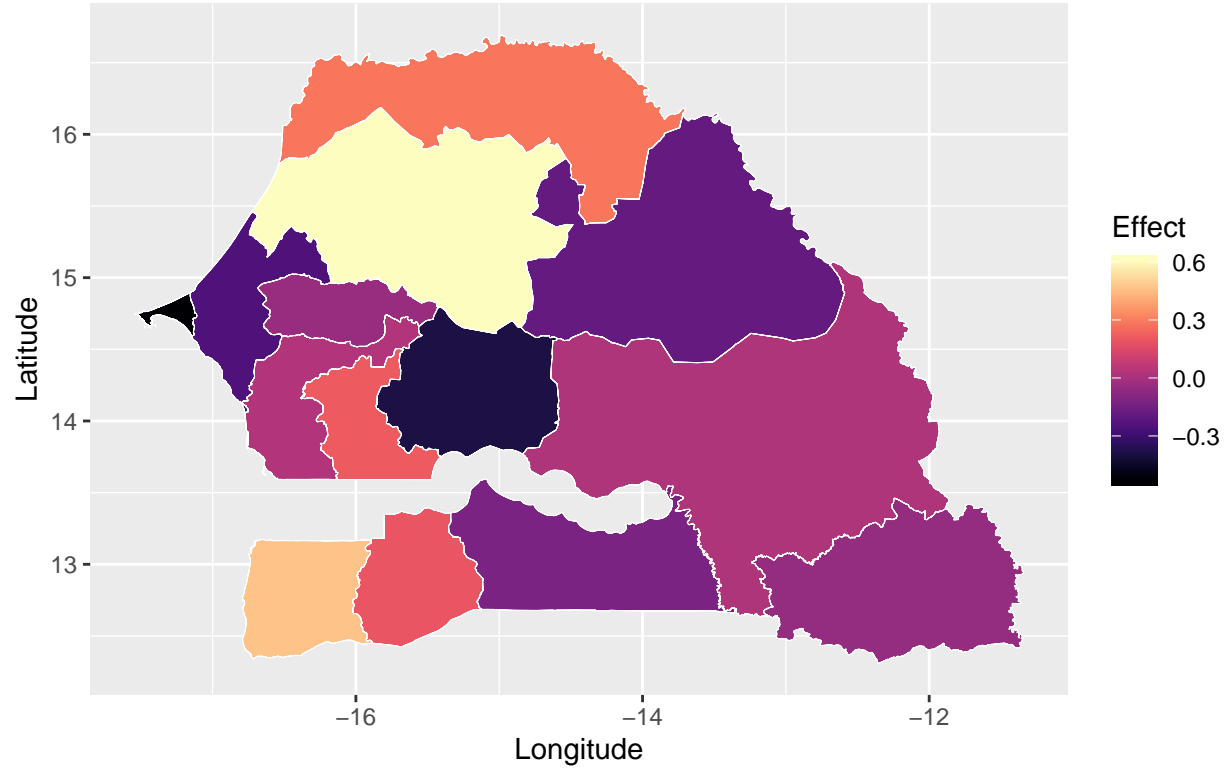
We see obvious islands of effect— these are urban clusters. In general, the urban-rural effect is far greater than the distance-to-Dakar effect, as even rural clusters close to Dakar have relatively small effects. Dakar itself does not exhibit a large effect. This is because the median wealth in Dakar is the relatively great. Moreover, the specification forces the effect of distance-to-Dakar to be linear.

The regression above does not allow for spatial heterogeneity of response, except through distance-to-Dakar. Therefore, I run a more flexible regression, which allows the effect of exposure to vary by region:

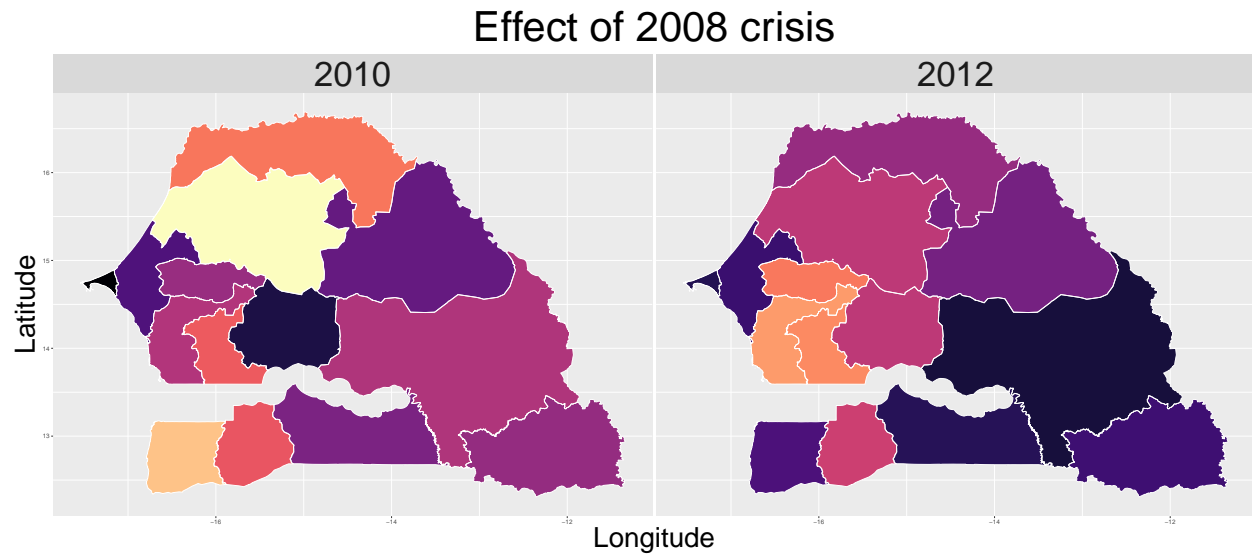
$$H_i = \alpha_0 + \beta Exp_i + \delta Exp_i * Region + \alpha_1 MOB + \alpha_2 MOB^2$$

I report the predicted effect of exposure in each region.

Effect of 2008 crisis: 2010



I do not control for urban-rural or for wealth, as I want to capture the regional effects, inclusive of differences in wealth and the proportion who are urban. We see that Dakar and its neighbor Thiès are most strongly affected. These are highly urban regions, but also are relatively wealthy. Outside of these two regions, remoteness seems to have no effect, indicating a nonlinear relationship of distance-to-Dakar and effect of the 2008 crisis.



Here I plot the same map, but include the results for 2012 in order to compare magnitudes. We see that for many regions the effect is inconsistent across years, suggesting that the crisis had no consistent effect. For Dakar and Thiès, on the other hand, the effect of the crisis is consistent across years and persists into 2012.

Discussion

The Effect of Exposure

The results above provide support for the following hypotheses:

1. The 2008 crisis had a negative impact on the maternal nutrition of urban-dwellers, but not rural-dwellers;
2. The effect of the 2008 crisis on nutrition was greater for the poor than the rich;
3. The effect of the 2008 crisis on child hemoglobin is discernible in 2010, but not in 2012, suggesting a dissipation of signal; and,
4. For urban dwellers, the effect of the 2008 crisis was greatest for those closest to Dakar.

For those affected– the urban poor– the estimated effect of being exposed in-utero to the 2008 crisis is between 10-14 g/L of hemoglobin, where the cutoff for anemia is defined at 105 g/L. I would argue that these effects are best understood as lower bounds. Because with the fetal origins approach we implicitly compare those in-utero and those already born, the estimated effect of exposure is net any negative impact on children already born. Also, given the persistence in iron stocks and budget constraints, those conceived after the

crisis also likely felt some negative effects. I run a more narrow regression in the Appendix, which discards around half the data and focuses on those exposed and those born just before exposure. The estimated effect of the crisis is larger than in my main specification, further supporting the argument that my main results form a lower bound.

There is substantial attenuation of the effect of the 2008 crisis between 2010 and 2012. This should not be interpreted as recovery of health, but rather as the masking of an informative signal. Iron during the first 1000 days of life (including gestation) is extremely important, and iron deficiency impairs cognitive and motor development in ways that can never be recovered (Lozoff 2007). Therefore, looking at the effect of in-utero shocks too late in life can mask important differences. Two children who seem the same in 2012 in terms of hemoglobin, actually had very different hemoglobin in 2010, and therefore are on very different biological trajectories.

Interpretation of Variables

Although exposure was defined with respect to prices, the crisis was more than just an increase in prices. During the crisis there were protests, unrest and a change in the type of agricultural production, with an increase in staples like millet and sorghum. Thus, we cannot rule out that the estimated effect of the crisis has more to do with the general characteristics of the period than with prices *per se*. For example, the strong effect in Dakar might come from social unrest, as there was rioting in Dakar during the period.

Similarly, distance-to-Dakar should be considered as a general proxy for integration to the world market. Distance captures a broad set of socio-economic conditions. We would expect areas further from Dakar to be poorer, have less education and have more subsistence agriculture relative to commercial agriculture. Thus, the estimated effect will not be of transportation costs alone, but rather the combination general set of characteristics associated with being further from Dakar. Likewise, “wealth” in the DHS is a synthetic index, rather than an actual measure of household assets. Thus, wealth in this study ought to be thought of as a continuous measure of socio-economic status, which is correlated with household assets and income.

Connecting to a Broader Literature

My results support the existing view of the food policy literature (Cohen and Garrett 2010; Dimova and Gbakou 2013) that increases in food prices negatively affect urban households while having little effect on the rural populace. While the results are too noisy to reject the null hypothesis that the crisis had a null effect on rural households, we also cannot rule out that the nutrition rural households improved during the crisis. This might due to the increased income from sale of food, as well as substitution toward locally produced, high-iron crops like millet and sorghum.

All of this discussion and analysis has been specific to Senegal. I would like to stress that one would expect very different effects of the crisis in different countries. As shown above, millet production actually increased during 2008. This may reflect a reserve capacity for food production, or a fortunate coincidence of good yields coinciding with high price of imports. In either case, it suggests significant value in having substantial domestic food production capacity. Given a sudden increase in food prices, households may adapt by shifting production towards food. Even if adaptation isn’t possible, the independence of world price and local weather conditions imply that having some domestic production and some imported food supply would shrink the variance of food consumption. A country which completely abandons local food production in order to specialize in cash crops, would be in a significantly more vulnerable position than Senegal was in 2008. Thus, my work suggests some empirical support to the arguments for “food-sovereignty” in the food-policy literature (Rosset 2008).

Conclusion

I find suggestive evidence that the 2008 world food crisis negatively affected the health of women, as shown by the health of their children, in urban Senegal. The effect is strongest and most consistent around Dakar. Within the context of the food-policy literature, my results are direct evidence to the hypothesis that food price shocks negatively affect urban households, without having a large effect on rural households. My results suggest that the effect of a negative shock to maternal nutrition may have a large effect on child hemoglobin, outweighing the effects on children. This negative shock is apparent 2 years ex-post, but has dissipated by 4 years ex-post. As Iron is a critical micronutrient for fetal and infant neuro-development, although the shock has dissipated, its effect on long-run health and welfare will remain.

These results are evidence that the health of poor people in poor countries depends on the movement of international markets. The 2008 food price crisis is a specific episode which had a strong, negative impact on the urban-poor of Senegal. Since 2008 urbanization, and food imports, have continued to increase in Senegal. Incomes have also risen. Whether the country is more or less vulnerable today to a price shock like that of 2008 is difficult to answer.

We expect urban populations to be less vulnerable to drought and other natural disasters. There would seem to be a trade-off then, between vulnerability to market shocks and vulnerability to shocks from nature. Urban, well-connected populations are more vulnerable to the market, while rural, remote populations are vulnerable to nature. Here we must bring in that universal feature of social analysis: class. In either case, it is the poor who are vulnerable. Any policy which thinks of mitigating the effect of these shocks, whether they come from the market or from nature, must keep this in mind.

This is a fetal origins paper. In general, the fetal-origins literature tends to concentrate on the fetus. I would like to draw attention to the mother. Poor fetal nutrition is one and the same as poor maternal nutrition. Any discussion of infant health must consider gender inequality (Osmani and Sen 2003). The negative consequences of the 2008 crisis on child hemoglobin are direct evidence to the malnutrition of poor, urban women of Senegal. In order to ensure the health of a population, the health of women of childbearing age must be the focus.

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Appendix

Here I run several alternative specification to those reported above. Each of these was ran after running the regressions above, in response to a specific concern or question. Because I did not pre-commit to these specifications, they should be interpreted weakly. However, they do generally support my conclusions. Of particular importance in the narrow-window regression, which finds even stronger effects than my main specification.

Narrow Window

In the main specification, we are comparing infants exposed to the 2008 crisis in-utero to two control groups:

1. Those who are exposed in childhood; and,
2. Those who are conceived after the crisis.

For reasons discussed above, including breastfeeding and delayed transmission of shocks, we would expect both of these groups to also experience a negative effect of the crisis. Thus, the ideal control group is to narrow the regression to the window just around those who had just been born and those who were still in-utero. I did not do this as my main specification because it would involve throwing out much of the data set, and would reduce my power. Next I report a more narrow specification, where I have thrown out over half the data-set. The data are restricted to those who were born between April 2007 and July 2009. The while the standard errors increase, the point estimates increase in magnitude, as expected if the effect is concentrated in these groups.

| | 2010 | 2012 | 2010 | 2012 |
|------------------------------|------------------------|--------------------|----------------------|---------------------|
| Intercept | -0.1737*** (0.0313) | 0.0060 (0.0203) | -0.0705 (0.0556) | 0.0239 (0.0382) |
| Exposure | -0.0200 (0.0991) | 0.0131 (0.0918) | -0.2175 (0.1622) | 0.0869 (0.1244) |
| Distance to Dakar | | | -0.0326* (0.0148) | -0.0058 (0.0115) |
| Distance to Dakar * Exposure | | | 0.0611 (0.0375) | -0.0310 (0.0341) |
| Num. obs. | 3226 | 5065 | 3226 | 5065 |
| Percentile | 0.5000 | 0.5000 | 0.5000 | 0.5000 |

*** $p < 0.001$, ** $p < 0.01$, * $p < 0.05$

Table 6: Test for Lagged Effects

| | 2010: Urban | 2010: Rural | 2012: Urban | 2012: Rural |
|------------------------------|-----------------------|------------------------|---------------------|-----------------------|
| Intercept | 0.2438 (0.2191) | -0.5061*** (0.1369) | 0.0663 (0.3562) | -0.3522 (0.2670) |
| Exposure | -1.6559** (0.5606) | 0.3994 (0.2752) | -0.3887 (0.6498) | 0.9637* (0.4783) |
| Distance to Dakar | -0.0528 (0.0410) | 0.0666 (0.0416) | 0.0670 (0.0664) | 0.0999* (0.0512) |
| Wealth | 0.9826 (1.3104) | 2.7116*** (0.7879) | 0.5571 (1.9582) | 1.3552 (1.0269) |
| Distance to Dakar * Exposure | 0.2762** (0.0903) | -0.0577 (0.0792) | 0.0145 (0.1158) | -0.1982** (0.0828) |
| Wealth * Exposure | 4.8522 (2.8088) | -0.3558 (2.0317) | 2.0310 (3.2243) | 2.2534 (1.9023) |
| Num. obs. | 446 | 1017 | 424 | 888 |
| Percentile | 0.5000 | 0.5000 | 0.5000 | 0.5000 |

*** $p < 0.001$, ** $p < 0.01$, * $p < 0.05$

Table 7: Test for Lagged Effects

I take this as convincing evidence of the effect of the 2008 crisis. However, it was not the specification to which I pre-committed, so I have included it in the appendix as supporting evidence, rather than as my main specification.

Lead-lag

As noted in the prices section, it can take several months for world prices to diffuse to cities in Senegal. Thus one might expect the exposure variable to be mis-assigned, as those who were really exposed to the shock were born several months later. Therefore, I take the twice lagged values of exposure, and regress them in the same way as before. Interestingly, the effect of the crisis is now negative for both urban and rural dwellers. This suggests that the price effect takes longer to be transmitted to rural areas.

| | 2010 | 2012 | 2010 | 2012 |
|------------------------------|------------------------|---------------------|---------------------|---------------------|
| Intercept | -0.1821*** (0.0314) | 0.0097 (0.0196) | -0.0777 (0.0564) | 0.0270 (0.0362) |
| Exposure | 0.0185 (0.0810) | -0.0133 (0.0664) | -0.1453 (0.1462) | 0.1018 (0.1331) |
| Distance to Dakar | | | -0.0307 (0.0157) | -0.0056 (0.0115) |
| Distance to Dakar * Exposure | | | 0.0446 (0.0355) | -0.0469 (0.0368) |
| Num. obs. | 3224 | 5065 | 3224 | 5065 |
| Percentile | 0.5000 | 0.5000 | 0.5000 | 0.5000 |

*** $p < 0.001$, ** $p < 0.01$, * $p < 0.05$

Table 8: Test for Lagged Effects

| | 2010: Urban | 2010: Rural | 2012: Urban | 2012: Rural |
|------------------------------|----------------------|-----------------------|----------------------|-----------------------|
| Intercept | -0.2224 (0.1137) | -0.1912* (0.0778) | -0.0655 (0.0829) | -0.0395 (0.0489) |
| Exposure | 0.1055 (0.3839) | -0.1193 (0.1802) | 0.0657 (0.2593) | 0.3799* (0.1794) |
| Distance to Dakar | 0.0227 (0.0253) | 0.0026 (0.0205) | 0.0589** (0.0188) | 0.0131 (0.0142) |
| Wealth | 1.7467** (0.6561) | 1.8736*** (0.4241) | 1.5815** (0.5530) | 2.1349*** (0.3576) |
| Distance to Dakar * Exposure | 0.0480 (0.0740) | 0.0281 (0.0459) | -0.0633 (0.0646) | -0.0900 (0.0570) |
| Wealth * Exposure | 0.0500 (1.9689) | 0.0467 (1.3039) | -1.0935 (1.9637) | 0.0833 (1.1413) |
| Num. obs. | 974 | 2250 | 1573 | 3492 |
| Percentile | 0.5000 | 0.5000 | 0.5000 | 0.5000 |

*** $p < 0.001$, ** $p < 0.01$, * $p < 0.05$

Table 9: Test for Lagged Effects

Selection

Next I test for fetal selection. Female fetuses are more robust than males, so if the 2008 crisis caused selection due to in-utero stress, we should see an increase in the number of girls. I find no such result.

| | Girl |
|---------------------|-----------------------|
| (Intercept) | 1.4863*** (0.0061) |
| exposure | -0.0007 (0.0203) |
| R ² | 0.0000 |
| Adj. R ² | -0.0001 |
| Num. obs. | 8358 |
| RMSE | 0.4999 |

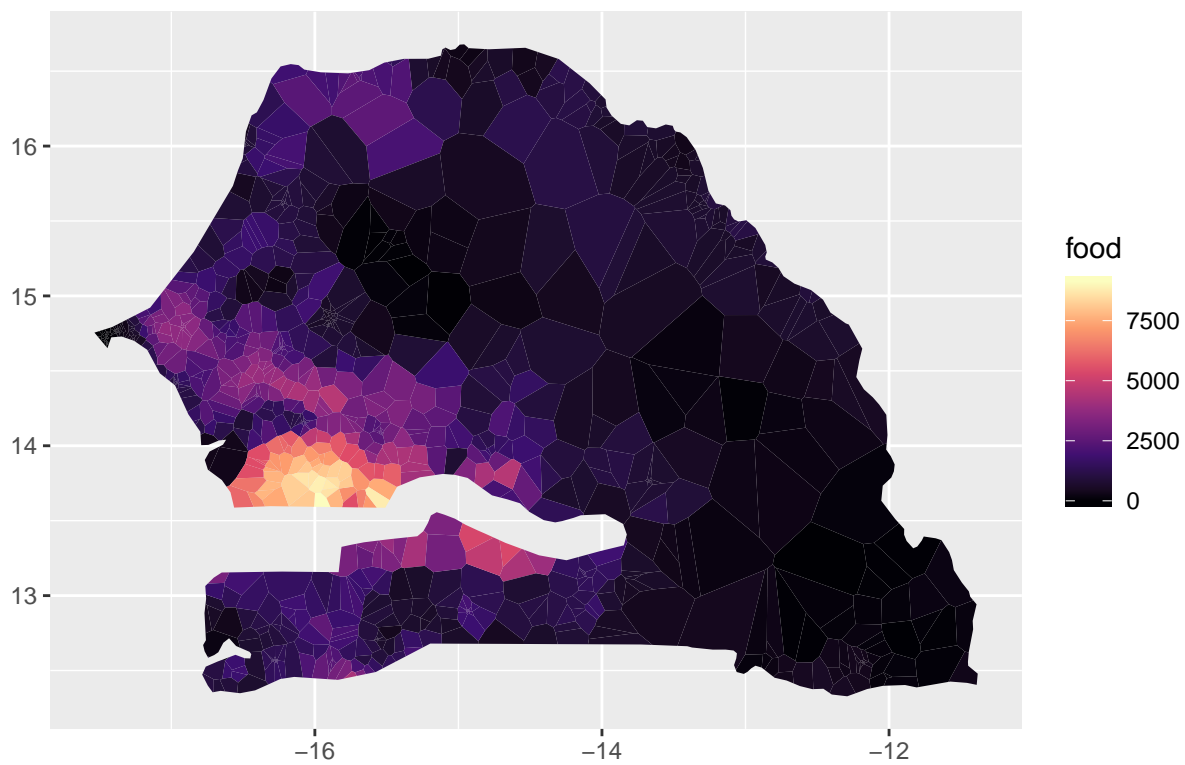
*** $p < 0.001$, ** $p < 0.01$, * $p < 0.05$

Table 10: Statistical models

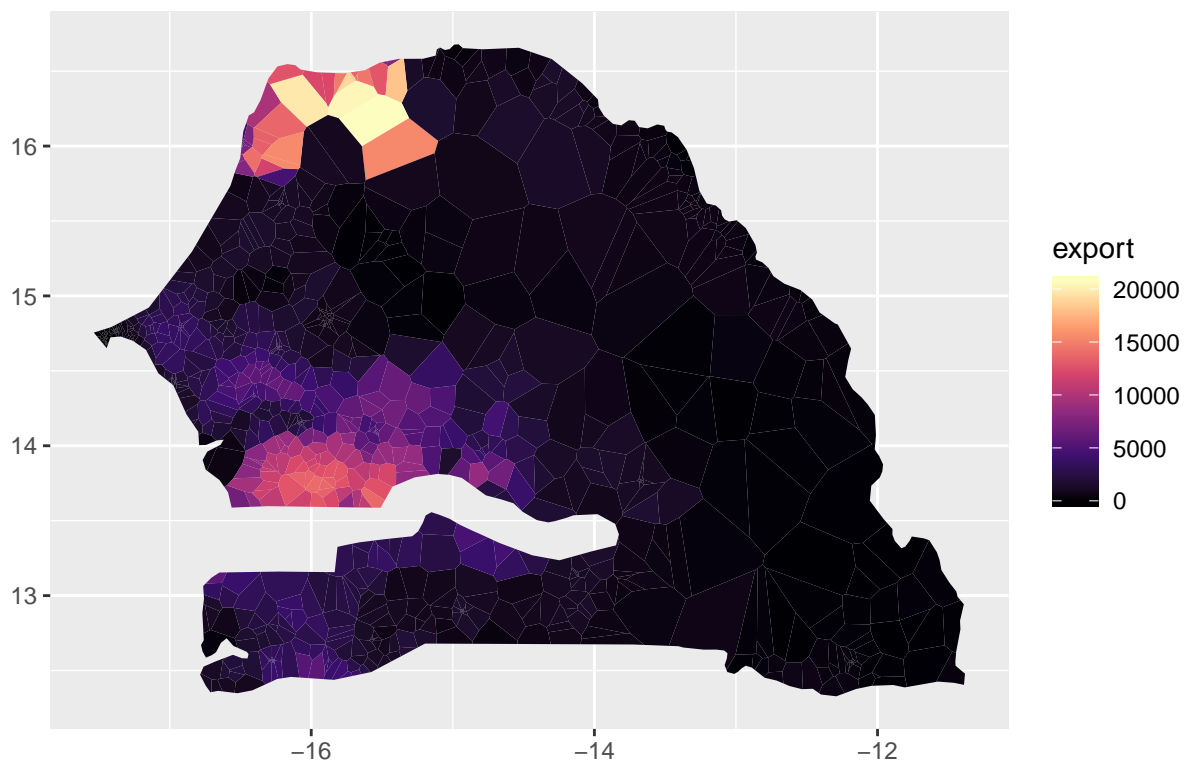
Appendix 2: More information on the spatial distribution of variables in Senegal

In case the reader is interested in the distribution of agriculture and wealth across Senegal (as I was), I have included some heat-maps which depict variables at the cluster level. Below I report food production, as well as commercial agriculture (sugarcane, palm nuts and cotton).

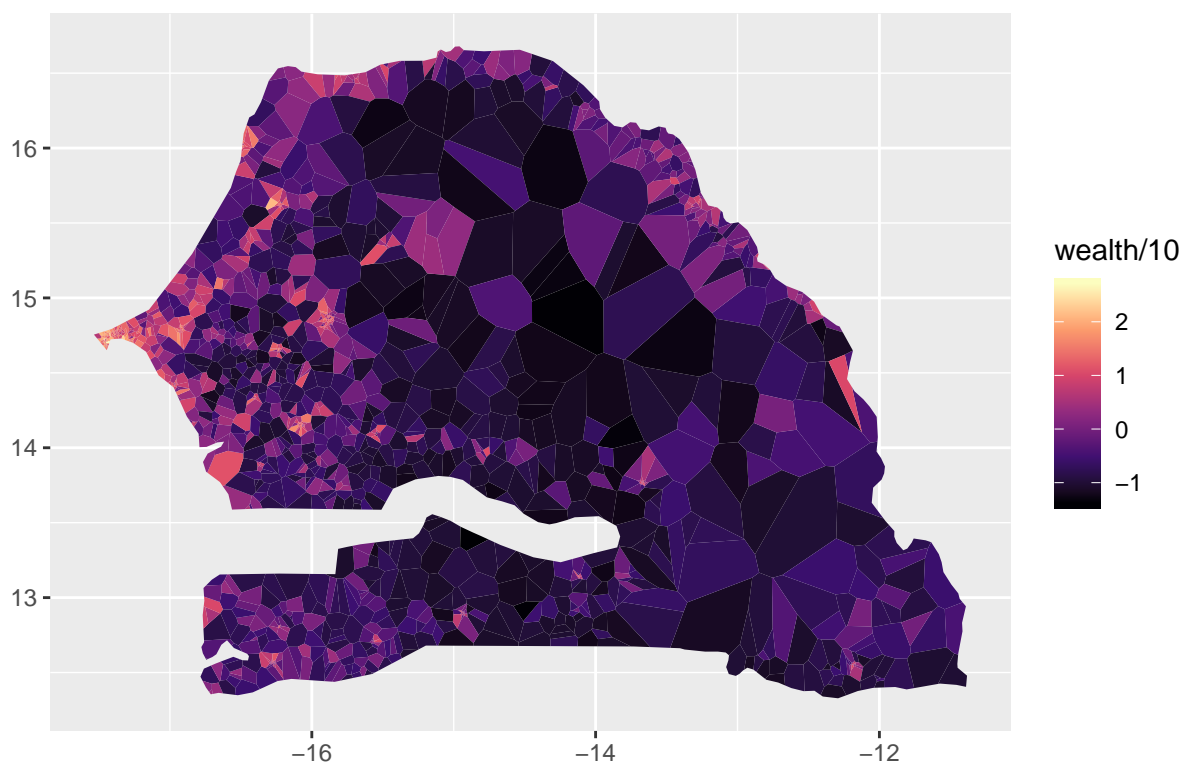
Food Production across Senegal



Commercial Agriculture across Senegal



Wealth across Senegal



Land Ownership across Senegal

