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Application of Quantile Regression to examine changes in the distribution of stunting of Indian children aged 0-36 months using four rounds of NFHS data.

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Abstract

Background: In India, it has been observed that the prevalence of stunting among under-five children decreased, but the prevalence is still alarmingly high. In previous studies, traditional (linear and logistic) regression analyses were used, and these analyses were limited to encapsulated cross-distribution variations. Our study's objective was to examine how the different determinants are heterogeneous in various percentiles.

Methods: This article examined the change in the stunting distribution of children and examined the relationships between the key covariate's trends and patterns in stunting among children aged <3 years over a period of 24 years. Four successive rounds of the National Family Health Survey data 1992-93, 1998-99, 2005-06, and 2015-16 were used for analysis. The final study included 206579 children aged <3 years (N= 106136 male, 100443 female). To explain and analyze differences in the stunting distribution, the lambda-mu-sigma (LMS) method was used. Trends in stunting distribution over time were analysed using separate sex-stratified quantile regression (QR). The selected socioeconomic, demographic and other predictors considered for this analysis.

Results: The quantile regressions have clearly indicated that mothers who have higher than primary level education were beneficial to decrease child malnutrition at the lower end of the distribution. The age, birth order, mother's body-mass-index (BMI) and wealth, among others, were some more determining factors for HAZ. Results of selected quantile regression estimated at 5th, 10th, 25th, 50th, 75th, 90th, and 95th quantiles. The wealth index was a highly negative association with lower quantiles compared to upper quantiles in stunting

However, in the age classification, as the age increases, there was a negative association in the upper quantiles of stunting. Small size at birth was having a negative association in all the quantiles of stunting.

Conclusions: The outcome of various covariates working differently across the stunting distribution was suggested by quantile regression. The major discrepancies in different aspects were underlined by socioeconomic and demographic aspects of India. The heterogeneity of this effect was shown using quantile regression.

Keywords: Body Mass Index (BMI); Quantile regression; Stunting; Trend; Socioeconomic; Demographic

Background

In India, malnutrition was a significant public health problem among under-five children. The survival and early development of children were severely affected by malnutrition. Also, it vigorously affects the health of pregnant and nursing mothers. The overall resistance-to diseases and future performance in school and at work was also determined by Malnutrition [1, 2]. In the world, among under-five children, about 3.5 million die each year due to child malnutrition. In developing countries, more than one-third of all child deaths under the age of five were mainly caused by child undernutrition [3]. Therefore, nutrition was a significant factor in the health sector, especially in developing countries [4].

The term malnutrition includes undernutrition and overnutrition, and a significant factor causing child mortality worldwide [5]. In child nutritional status, the three broadly known pointers to describe the physical development of children were stunting (short height for age), wasting (low weight for height/length) and underweight (low weight for age) [6 - 8]. Among the children aged 0-59 months, there was a 38%, 36%, and 21% prevalence of stunted children, underweight, and wasting respectively, according to National Family Health Survey (NFHS-4) [9]. Around the world, 165 million children under five years of age were affected by stunting [10]. Stunting and other forms of undernutrition reduce a child's chance of survival. It was characterized as the percentage of children aged 0 to 59 months whose height for age was below minus two standard deviations (moderate stunting) and minus three standard deviations (severe stunting) from the median of the WHO Child Growth Standards [11, 12]. To measure child growth, the height for age Z-score (HAZ) was used.

In early childhood, particularly stunting increases the chance of susceptibility to sickness. Since stunting was associated with suboptimal brain development, it has practically permanent effects on physical, mental development and was probably going to have long-lasting adverse ramifications for intellectual capacity, school performance, and future benefits [10]. The aforementioned effects on development and uniformity caused by child malnutrition were most likely to keep the children in the limbo of poverty, which affects the development of nations [13, 14]. Thus, improving children's nutritional status and lessening the prevalence of malnutrition to advance their physical and mental improvement was extraordinarily significant [15]. Better wellbeing and nourishment have resulted in financial development just as value concerns [16, 17]. According to the WHO, stunting was a reliable measure of overall social deprivation [18]. Hence, the World Health Assembly has set a global target to reduce the number of stunted under-five children to 40 percent by 2025 [10, 13].

Child health was predominantly noteworthy due to its connection to child poverty and the build-up of adult human capital. The improvement of needy children's health and nutrition has been viewed as an effective method of improving school participation and upgrade financial development since learning converts into gains in long-run productivity [19, 20].

There was considerable literature analysing the determinants of nourishing child status in a broad scope of nations and utilizing distinctive econometric techniques. The studies that were conducted earlier have used standard multiple linear or logistic regressions. The more significant part of those past investigations was that they concentrated on finding the determinants at the mean and odds using logistic regression [21 - 27]. The utilization of these methods may prompt inaccurate strategy intercession measures if the relationship between child nutritional status and the specific socioeconomic and demographic determinants is heterogeneous at the various percentiles of the nutritional distribution. Moreover, just a couple of studies focused on the distinction at various points of the conditional nutritional dispersion [23, 28, 29].

The paper's limitation can be resolved by analysing demographic and socioeconomic factors across varied points of the conditional stunting (HAZ) distribution in India. Moreover, the changes in the stunting distribution of children's and connections between the key covariates patterns and trends in stunting were investigated by using the quantile regression model [30, 31].

The objective of this paper was to find the association between demographic, socioeconomic and health factors of child nutritional status from 4 rounds of National Family Health Survey (NFHS) data of children aged <3 years in India over 24 years by using quantile regression. Adequate research has not been done in this area. This analysis was expected to improve the structure of successful intercession measures designed to tackle child malnutrition and improve child health.

Methods

Data

Data for this analysis were taken from four consecutive rounds of the NFHS of India conducted during 1992–93, 1998–99, 2005–06, and 2015–2016. The NFHS is an expansive scope household study led over the states and union territories of India. The International Institute for Population Sciences (IIPS), Mumbai, India, directed the numerous rounds of the survey with

community-oriented assistance from a few national and international associations [32]. The surveys conducted led to more information on reliable estimates of fertility, Infant and children mortality, nutritional status of children, better use of MCH services at a national, state-level and across the urban and rural residence.

The initial three rounds of the NFHS were intended to give state-level information. Nonetheless, the fourth round of study, which includes a more prominent sample size, yields assessments of most factors for all 640 districts in India [9]. Each of the four rounds of the study incorporated multi-stage sampling design, two-stage sampling design in rural areas, and three stages in urban areas [33]. The NFHS gathered information utilizing distinctive interview schedules, household schedules and qualified women/individual schedules, and for the fourth round, the Biomarker Questionnaire was incorporated [9]. In all the rounds, the substance of the schedule remains the same.

The household response rate of the first-round was 96% and 98 % for the remaining three surveys. The individual response rate for the first and second rounds was 96%, 94% for the third round, while it was 97% in the fourth round of the survey. All four rounds of the NFHS furnish data on anthropometric pointers with a different age group of children. For instance, NFHS-1 gathered data from children aged below four years and NFHS-2 gathered data from children below three years, while NFHS-3 and 4 gathered data from children below five years. Hence, to make the assessments equivalent, the examination was limited to the children aged <3 years old. For this study, in total, only arthrometric indicators of unit-level data of 206579 children aged <3 years (N = 106136 male and 100443 female) over 24 years period were considered.

Dependent variable

In this analysis, only stunting as a dependent variable to find the independent variable's distribution on different quantiles was used. To estimate the stunting marker, the reference population of the WHO 2006[34] was considered. As per the WHO, the child was classified as stunted if the height-for-age Z-score of a child is <2 standard deviations (SDs), then the child was classified as stunted. If the child's Z score value is <3, it was classified as severely stunted. The HAZ score value considered from -6 to +6 remaining cases were excluded/flagged.

Independent variable

Five quantiles of the socioeconomic variable wealth index have been taken into account in this study. The quantiles are: (poorest, poorer, middle, richer, and richest). Demographic variables such as caste - Scheduled Caste (SC), Scheduled Tribe (ST), Other Backward Class (OBC), and Other castes were taken into consideration. Child characteristics such as child's age in

months (0-6, 7-12, 13-24, and 25-59), birth order (1,2,3,4, and above), and size at the time of birth (small, average and large) were taken into consideration. The maternal factors include maternal age of 15-49 years, BMI of mothers (underweight: Body mass index <18.5, normal: 18.5<BMI<25, and overweight BMI>25.0), and mother's education (No education, <5 years of schooling, 5-7, 8-9, 10-11, 12 years or more) were considered. The wealth index variable was computed only for the NFHS 3 and 4 due to the NFHS 1 and 2 survey data unavailability for analysis. Moreover, the Body Mass Index (maternal) variable was computed for the NFHS 2, 3, and 4 due to the unavailability of the data in NFHS 1.

Statistical Analysis

Quantile Regression

Koenker and Bassett (1978) first introduced the key idea of quantile regression [13, 35, 36]. This procedure has an advantage over the conventional common least-squares method. It does not accept a steady effect of the independent factors over the whole distribution of the dependent variable [31, 37]. This methodology was utilized to consider a heterogeneous impact of every determinant alongside various percentiles of the conditional distribution of the dependent variable [30, 31].

Koenker and Bassett (1978) show that the empirical quantile function is the solution of the minimization problem defined by [37]:

$$\begin{aligned}\hat{\beta}_\tau &= \underset{\beta_\tau \in R^K}{\operatorname{argmin}} \left\{ \sum_{i: y_i \geq x_i' \beta_\tau} \tau |y_i - x_i' \beta_\tau| + \sum_{i: y_i < x_i' \beta_\tau} (1 - \tau) |y_i - x_i' \beta_\tau| \right\} \\ &= \underset{\beta_\tau \in R^K}{\operatorname{argmin}} \sum_i \rho_\tau |y_i - x_i' \beta_\tau|\end{aligned}$$

With $\rho_\tau(z)$ can be defined as:

$$\rho_\tau(z) = \begin{cases} \tau(z) & \text{if } z \geq 0 \\ (\tau - 1)z & \text{if } z < 0 \end{cases} = (\tau - I(z < 0))z$$

Let x_i where $i= 1, \dots, n$ a sample, a $K \times 1$ vector of regressors, $y_i = x_i' \beta_\tau + \varepsilon_{\tau i}$, $0 < \tau < 1$, $\rho_\tau(z)$ is the check function, and $I(\cdot)$ usual indicator function.

In this study, quantile regression analysis was performed to identify the independent variables related to child anthropometric Z score over the seven (5th, 10th, 25th, 50th, 75th, 90th, and 95th) percentiles. This analysis was performed using the SAS University Edition software.

Lambda Mu Sigma (LMS) Method

Using quantile regression, we investigated longitudinal changes in the stunting distribution overtime for the separate sex-stratified. In addition, we used the LMS method to determine the age-specific secular trends in the child nutritional status measures, thereby allowing us to examine the temporal trends in selected seven percentile points of the stunting.

The LMS method summarizes the shifting distribution by three curves representing the median, coefficient of variation and skewness; the latter expressed as a Box-Cox power [38]. For a given value of covariate, to transform the response to standard normality LMS technique applies a Box-Cox transformation; to get the quantiles, an opposite Box-Cox transformation was applied to the quantiles of the standard normal distribution [39, 40]. A large sample size was required to gauge the percentiles in each age group with proper accuracy. The division may lose data from nearby groups. To maintain a strategic distance from division, Cole and Green (1992) built up a Box-Cox transformation based on the semiparametric method from the LMS technique presented by Cole in 1988 [41].

LMS technique in the GAMLSS package R version 3.4.3 (R Development Core Team, Vienna, Austria) to get figures of stunting distribution for 1992 and 2016, along with the lines showing the adjustments in the stunting measures [42, 43] were utilized. Additionally, the LMS method to decide the age-specific patterns in the stunting measures for both genders, along with the lines permitting us to analyse the worldly patterns in explicit percentile purposes of the stunting, were utilized in the analysis.

Results

Table 1. presents descriptive statistics for the individual demographic, socioeconomic, and health-related variables for each gender.

Tables 2 show the calculated coefficients for the selected seven percentiles (5th, 10th, 25th, 50th, 75th, 90th, and 95th) of male and female, HAZ. Table 2 also illustrates that both male and female child age (25-36, 13-24, 7-12 age in months), Mothers education (No education, < 5, 5-7, 8-9 years of schooling), child-size at birth (small, average), birth order (4 or more, three) and type of caste (SC) were negatively associated when compared to their reference category in all percentiles and also found statistically significant. Factors such as child's age in months, child-size at birth, type of caste, and birth order are highly negative associated with higher quantiles than the lower quantile. The factor mother's education (No education and < 5 years) is high at (10th, 25th and 50th) percentile when compared to other percentiles. Type of caste (ST) had a negative association in (5th, 10th) percentile but a positive association in the higher percentiles (25th, 50th, 75th, 90th, and 95th).

Tables 3 shows the calculated coefficients at the specific percentiles for both males and females for the NFHS-2 data. Child age (25-36, 13-24, 7-12 age in months), Mothers education (No education, < 5, 5-7, 8-9 years of schooling), child-size at birth (small, average), birth order (three), and type of caste (SC, OBC) factors were negatively associated and also statistically significant in all quantiles. The factor mother's education has high coefficient values in lower percentiles and low coefficient values in higher percentiles. The factor birth order (4 or more and three) is negatively associated in all except 90th and 95th percentile.

In females, Type of caste (ST) had a negative association in 5th quantile, but a positive association in the higher percentiles (10th, 25th, 50th, 75th, 90th, and 95th), and significant association was observed in (25th, 75th, 90th, and 95th). The factor mothers BMI (<18.5) had a negative association in all percentiles, and it has high coefficient values from 25th to 95th percentile compared to the 5th and 10th percentile. Moreover, a significant association was observed in (25th to 95th) percentile. Similarly, mothers BMI (>25) had a positive association in all quantiles when compared to the reference category (BMI 18.5 to 25), Indicating that mothers' BMI is could be an effective measure to combat child stunting.

Table 4 presents the results of quantile regression estimates at the selected percentile levels for the NFHS -3 data. Child's age in months had a statistically significant negative association with HAZ when compared to the reference age group (0-6 months) for both genders, children whose age (25-36, 13-24, 7-12 months) had a high negative association in the higher quantiles when compared to the lower quantiles.

In the maternal education variable, when compared to the reference was taken as mothers had >12 years education, mothers education category (<5, 5-7, 8-9, 10-11 years) had a statistically significant negative association (except 95% percentile) with HAZ for both genders. In females, Children whose mother's education had a high negative association at the lower quantiles were switched to low at the higher quantiles (no significant association observed in the 95% percentile). In male children whose mother's education had a high negative association at the middle quantiles and it is switched to low at the higher quantiles.

For the variable child's size at birth (reference category: Large), child's -size at birth at small and average has a negative association and significant was observed in the male and female children. In male children, it is highly negative associated with the higher quantiles compared to the lower quantiles. In female children, size at birth was less negatively associated in the lower quantiles than the higher quantiles. The type of caste (children who belong to SC and OBC) in male children has negatively statistically significant (10th percentile to 90th

percentile) associated with the lower quantiles, and it is positively associated in the 95th percentile.

For the female children, type of caste has negatively statistically significant (10th percentile to 90th percentile) associated in the lower quantiles, and it is positively associated in the 95th percentile

For the variable birth order (4 or more, three and second) is negatively associated in all except 90th and 95th percentile when compared to the reference category

The quantile regression estimates showed a positive association between BMI>25 and HAZ. However, in the lower percentiles, it shows the statistical significance for both genders and in the 95th percentile, no significance was observed. It indicates that Mother's BMI could be an effective measure to combat child stunting.

Quantile regression results presented a statistically significant negative association between household economic status and HAZ. There have been significantly higher HAZ scores of wealth index for children who belong to the poorest, poorer, middle and richer households than the richest household. For example, in 5th percentile, children from the poorest, poorer, middle and richer household's wealth index had a -0.73, -0.52, -0.40, and -0.46 disadvantage in terms of height over children from the richest households.

Table 5 presents the results of quantile regression estimates at the selected percentile levels for the NFHS -4 data. Child age in months had a statistically significant (except female at 5th percentile for the age 7-12 months) negative association with HAZ when compared to the reference age group for both genders, children whose age (25-36, 13-24, 7-12 months) had a high negative association in the higher quantiles when compared to the lower quantiles.

Mother's education category (<5, 5-7, 8-9, 10-11 years) had a statistically significant negative association (except female at 95th percentile) with HAZ for both genders.

For the variable child size at birth, child-size at birth at small and average (except 5th percentile) has a negative association and statistically significant (except both genders at 5th and 10th percentile) was observed in the male and female children. Type of caste (children who belong to SC and OBC) in male children it has negatively statistically significant (except 10th percentile of OBC) associated.

For the variable birth order (4 or more, three and second), a negatively statistically significant association was observed in all except 90th and 95th percentile compared to the reference category. Also, the higher birth order has a significantly negative association in the lower quantiles of stunting.

The quantile regression estimates showed a positive association between BMI>25 and HAZ. However, in the lower percentiles, it shows the statistical significance for both genders and in the 95th percentile, no significance was observed. It indicates that mother's BMI could be an effective measure to combat child stunting.

Quantile regression results showed a negative association between household economic status and HAZ. Children from the poorest, poorer and middle household's wealth index had higher HAZ scores compared with those from the richest household's wealth index. In particular, children from the poorest, poorer and middle household's wealth index had a -0.62, -0.48, and -0.31 disadvantage in terms of height over children from the richest households.

Figure 1 presents the smoothed distribution curve using the LMS method for the differences in the stunting over time in males and females aged 0-36 months in selected years. A decreasing trend is seen in all of the percentile curves from 1992 to 2016, in which the levels are decreased in the higher percentile levels among both genders and each age group. Differentiation of the plots noted that females' stunting percentile curves were marginally stable than those of male children. For both genders, when age increases, it automatically indicates that there are more prone to be stunted, but when compared to the NFHS 1 with NFHS, 4 curves were less decreased.

Discussion

Using 24 years period of data from the four rounds of NFHS, the determinants associated with stunting and trend in the stunting distribution among children aged 0-36 months of data using quantile regression were identified. The LMS method was used to construct the curves of the 5th, 10th, 25th, 50th, 75th, 90th, and 95th percentiles. There were significant works investigated the determinants of child nutritional status in a wide scope of nations and utilizing distinctive econometric techniques. Most of the past investigations have generally utilized standard linear regression and logistic regressions. The utilization of these regression estimation techniques may prompt in wrong approach to policy intervention measures, if the relationship between child nutritional status and the different determinants are heterogenous at the various percentiles [44]. One of the most important objectives in developing countries was to decrease child malnutrition. This paper embodied a quantile regression approach to investigate the effects of demographic, socioeconomic and health-related factors on stunting at different percentiles [13].

The analysis identified that the child's age in months is one of the strongest determinants of child nutritional status. Results showed a negative increment in coefficients from lower percentiles to upper percentiles compared to the first six months of life. The analysis also

reveals that stunting is least in the first six months of life, and it is increased with increasing age [23, 45, 46].

The present study showed a negative association between the mother's educational status and HAZ. Children of mothers with no education, <5 years. 5-7 years of mother's education had higher negative HAZ scores than the children whose mother's education had >12 years. It was observed that the children of mothers with less education had worse HAZ scores than children of mothers with high education. This analysis has proven that a mother's educational status plays a crucial in child nutrition. Educational accomplishments highly influence the nutritional information, awareness and risks associated with insufficient food. Assessments conducted in the past have revealed that mothers who were educated fed their children better since they have the higher capability and could benefit effectively from health care providers regarding medicinal information and nutrition-rich foods [47- 49].

The present study showed a negative association and also statistically significant between household economic status and HAZ. Children from the poorest, poorer, and middle household economic status had higher negative HAZ scores than the richest households. It is observed that the children of low-income families had worse HAZ scores than children from high-income households. The study illustrates that factors like child age, birth order of the child, mother's education and household economic status were more important than aggregate economic conditions [50]. For the factor size at birth, male children whose weight at birth was small had a nutritional drawback in height compared with their best peers.

The mother's nutritional status (BMI) was also associated, and it was statistically significant, with children's nutrition. The results showed a substantial effect on a mother's BMI on a child's nutritional status. Children of mothers with lower BMI are more likely to be stunted when compared with their counterparts.

Strength and limitations of the study

The study showed trends in child nutritional status from 1992 to 2016 using NFHS four rounds of data covering most of India. A more effective approach using quantile regression and LMS method to identify the heterogeneous effect of the child nutritional status with household, socioeconomic and health-related determinants.

Various intervention measures are required to consider the different effects of child nutritional status determinants with various percentiles of the HAZ distribution. Child malnutrition can

only be tackled with a multifaceted approach that includes targeted intervention, and it cannot be combated with a one-size-fits-all policy [44].

LMS curve with different percentiles for 1992 and 2016 quantified the trends in the performance of genders concerning child nutritional status. This study investigates how factors associated with child nutritional status on the stunting distribution on the selected percentiles. Limitations of the analysis based on secondary data also apply to this study. Cross-section data was used for this study.

Conclusion

This quantile regression suggests that the effects of different covariates worked differently across the stunting distribution. Predictors of stunting viz: child age in months, birth order, size at birth, and wealth index and their associations vary across different quantiles of stunting in India. Since India's socioeconomic and demographic characteristics have underlined the significant disparities in many aspects, national strategies to tackle stunting should be tailored appropriately for various segments. It would be adequate to carefully integrate applicable interventions according to the objective and target population for individuals' wellbeing and the country's development. Various measures could combat child nutrition among lower socioeconomic strata: increase in the family awareness regarding right feeding practices through mass media or other community-based programs, subsidizing the cost of nutrition-rich food and execute universal health care coverage.

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Availability of data and materials

The secondary data “NFHS” used in this study are publicly available at <https://dhsprogram.com/what-we-do/survey/survey-display-355.cfm>

Authors' contributions

Thirupathi Reddy M contributed to the data collection, analysis, and manuscript preparation. Vishnu Vardhana Rao M developed the study protocol, secured funds, supervised the study, and guided in manuscript preparation.

Ethics approval and consent to participate

This study is based on a publicly accessible survey dataset; therefore, this request is not applicable.

Competing interests

The authors declare that they have no competing interests.

Table 1 Descriptive Statistics and prevalence of stunting children aged 0-36 months by factors at different categories based on (NFHS- 1, 2, 3, 4) data.

Variable (n and %) in each category)	Category	NFHS-1		NFHS-2		NFHS-3		NFHS-4	
		Female	Male	Female	Male	Female	Male	Female	Male
Child age in months	25-36	3175(29.8%)	3343(30.6%)	3581(30.3%)	3934(30.5%)	4041(33.5%)	4456(34.5%)	22001(33.4%)	23430(33.3%)
	13-24	3637(34.1%)	3730(34.1%)	3948(33.4%)	4222(32.8%)	3945(32.7%)	4302(33.3%)	21911(33.2%)	23273(33.0%)
	7-12	1663(15.6%)	1715(15.7%)	1925(16.3%)	2118(16.4%)	2019(16.8%)	2150(16.6%)	10881(16.5%)	12007(17.1%)
	0-6	2181(20.5%)	2153(19.7%)	2373(20.1)	2611(20.3%)	2048(17%)	2008(15.5%)	11114(16.9%)	11684(16.6%)
	Missing	0(0%)	0(0%)	0(0%)	0(0%)	0(0%)	0(0%)	0(0%)	0(0%)
Type of caste	SC	1313(12.3%)	1445(13.2%)	2156(18.2%)	2382(18.5)	2080(17.3%)	2292(17.7%)	12527(19%)	13201(18.8%)
	ST	1268(11.9%)	1262(11.5%)	1783(15.1%)	1822(14.1%)	2018(16.7%)	1935(15%)	13292(20.2%)	13493(19.2%)
	OBC	-	-	3419(28.9%)	3660(28.4%)	3930(32.6%)	4262(33%)	25747(39.1%)	28179(40%)
	Other caste	8075(75.8%)	8234(75.3)	4387(37.1)	4958(38.5%)	3513(29.1%)	3855(29.8%)	11362(17.2%)	12478(17.7%)
	Missing	0(0%)	0(0%)	82(0.7%)	63(0.5%)	512(4.3%)	572(4.4%)	2979(4.5%)	3043(4.3%)
Mothers Education	No education	5927(55.6%)	6143(56.1%)	5784(48.9%)	6140(47.7%)	4731(39.3%)	4916(38.1%)	19260(29.2%)	20041(28.5%)
	<5 years	739(6.9%)	703(6.4%)	1071(9.1%)	1090(8.5%)	867(7.2%)	911(7.1%)	4062(6.2%)	4111(5.8%)
	5-7 years	1462(13.7%)	1409(12.9%)	1831(15.5%)	1856(14.4%)	1932(16%)	1946(15.1%)	10321(15.7%)	11294(16%)
	8-9 years	938(8.8%)	1068(9.8%)	1210(10.2%)	1486(11.5%)	1812(15%)	1965(15.2%)	12082(18.3%)	13279(18.9%)
	10-11 years	848(8.0%)	862(7.9%)	997(8.4%)	1185(9.2%)	1095(9.1%)	1383(10.7%)	7638(11.6%)	8075(11.5%)
	>12 years	703(6.6%)	71(6.6%)	928(7.8)	1123(8.7%)	1616(13.4%)	1795(13.9%)	12544(19%)	13594(19.3%)
	Missing	39(0.37%)	39(0.4%)	6(0.05%)	5(0.04%)	0(0%)	0(0%)	0(0%)	0(0%)
		(Small)	2369(22.2%)	2129(19.5%)	3144(26.6%)	2970(23.1%)	2668(22.1%)	2511(19.4%)	8333(12.6%)
Size at birth	Average	6921(64.9%)	7177(65.6%)	7071(59.8%)	7858(61%)	6612(54.9%)	7214(55.9%)	45296(68.7%)	48796(69.3%)
	Large	1266(11.9%)	1564(14.3%)	1604(13.6%)	2051(15.9%)	2605(21.6%)	3021(23.4%)	10991(16.7%)	12349(17.5%)
	Missing	100(0.9%)	71(0.6%)	8(0.07%)	6(0.05%)	168(1.4%)	170(1.3%)	1287(2%)	1264(1.8%)
		4 or more	3130(29.4%)	3270(29.9%)	3090(26.1)	3395(26.3%)	2712(22.5%)	2893(22.4%)	10122(15.4%)
Birth order	Three	1918(18%)	1951(17.8%)	2061(17.4%)	2286(17.7%)	1911(15.9%)	2147(16.6%)	10127(15.4%)	11487(16.3%)
	Second	2669(25%)	2668(24.4%)	3126(26.4%)	3467(26.9%)	3445(28.6%)	3762(29.1)	20912(31.7%)	22162(31.5%)
	First	2939(27.6%)	3052(27.9%)	3550(30%)	3737(29%)	3985(33.1%)	4114(31.9%)	24746(37.5%)	25945(36.9%)
	Missing	0(0%)	0(0%)	0(0%)	0(0%)	0(0%)	0(0%)	0(0%)	0(0%)
		Poorest	-	-	-	-	2165(18%)	2171(16.8%)	16886(25.6%)
Wealth Index	Poorer	-	-	-	-	2288(19%)	2256(17.5%)	15704(23.8%)	16535(23.5%)
	Middle	-	-	-	-	2487(20.6%)	2692(20.8%)	13318(20.2%)	14374(20.4%)
	Richer	-	-	-	-	2689(22.3%)	2948(22.8%)	11254(17.1%)	11886(16.9%)
	Richest	-	-	-	-	2424(20.1%)	2849(22.1%)	8745(13.3)	9954(14.1%)
	Missing	-	-	-	-	0(0%)	0(0%)	0(0%)	0(0%)
		<18.5	-	-	4152(35.1%)	4661(36.2%)	3970(32.9%)	4276(33%)	16004(24.3%)
Mothers BMI	BMI>25	-	-	668(5.6%)	699(5.4%)	1044(8.7%)	1208(9.4%)	8377(12.7%)	8972(12.7%)
	BMI 18.5-25	-	-	6957(58.8%)	7456(57.9%)	6959(57.7)	7371(57)	41113(62.4%)	43421(61.7%)
	Missing	-	-	50(0.4%)	69(0.5%)	80(0.7%)	61(0.5%)	413(0.6%)	405(0.6%)

(-) No Data available

Table 2: Quantile Regression Estimations for the different quantiles of the female (10517) and male (10830) children's, Dependent variable HAZ (Stunting) NFHS-1

Parameter	0.05		0.1		0.25		0.5		0.75		0.9		0.95	
	Female	Male	Female	Male	Female	Male	Female	Male	Female	Male	Female	Male	Female	Male
Intercept	-3.31 ^a	-3.82 ^a	-2.12 ^a	-2.45 ^a	-0.73 ^a	-1.11 ^a	0.30 ^a	0.018 ^a	1.18 ^a	1.30 ^a	2.40 ^a	2.38 ^a	3.25 ^a	2.98 ^a
Child age in Months														
25-36in Months	-0.97 ^a	-0.63 ^a	-1.38 ^a	-0.94 ^a	-1.7 ^a	-1.28 ^a	-1.75 ^a	-1.56 ^a	-1.80 ^a	-1.8 ^a	-1.89 ^a	-1.93 ^a	-2.00 ^a	-1.80 ^a
13-24 months	-0.68 ^a	-0.52 ^a	-1.01 ^a	-0.80 ^a	-1.32 ^a	-1.08 ^a	-1.40 ^a	-1.35 ^a	-1.40 ^a	-1.52 ^a	-1.55 ^a	-1.63 ^a	-1.57 ^a	-1.48 ^a
7-12 months	-0.32 ^a	-0.25 ^b	-0.50 ^a	-0.33 ^a	-0.56 ^a	-0.40 ^a	-0.64 ^a	-0.50 ^a	-0.71 ^a	-0.67 ^a	-0.79 ^a	-0.78 ^a	-0.95 ^a	-0.67 ^a
0-6 months	Reference	Reference	Reference	Reference	Reference	Reference	Reference	Reference						
Mothers Education														
No education	-0.68 ^a	-0.73 ^a	-1.03 ^a	-1.13 ^a	-1.21 ^a	-1.26 ^a	-1.05 ^a	-1.11 ^a	-0.76 ^a	-1.09 ^a	-0.74 ^a	-0.82 ^a	-0.46 ^b	-0.59 ^a
<5 years	-0.33 ^a	-0.21	-0.55 ^a	-0.49 ^a	-0.82 ^a	-0.77 ^a	-0.81 ^a	-0.80 ^a	-0.67 ^a	-0.77 ^a	-0.70 ^a	-0.70 ^a	-0.60 ^b	-0.48
5-7 years	-0.14	-0.27 ^c	-0.35 ^a	-0.60 ^a	-0.59 ^a	-0.70 ^a	-0.61 ^a	-0.62 ^a	-0.44 ^a	-0.73 ^a	-0.60 ^a	-0.57 ^a	-0.43 ^c	-0.39
8-9 years	-0.02	-0.03	-0.36 ^a	-0.26 ^c	-0.45 ^a	-0.49 ^a	-0.47 ^a	-0.44 ^a	-0.29 ^a	-0.49 ^a	-0.55 ^a	-0.67 ^a	-0.68 ^a	-0.63 ^b
10-11 years	-0.05	0.09	-0.006	-0.05	-0.23 ^b	-0.25	-0.16 ^c	-0.25 ^a	-0.02	-0.29 ^a	-0.21	-0.33 ^c	0.25	-0.17
>12 years	Reference	Reference	Reference	Reference	Reference	Reference	Reference	Reference						
Size at birth														
Small	-0.35 ^a	-0.21 ^c	-0.34 ^a	-0.50 ^a	-0.45 ^a	-0.48 ^a	-0.53a ^a	-0.58 ^a	-0.57 ^a	-0.63 ^a	-0.69 ^a	-0.62 ^a	-0.94 ^a	-0.69 ^a
Average	-0.22 ^b	-0.04	-0.24 ^a	-0.23 ^a	-0.23 ^a	-0.17 ^a	-0.29 ^a	-0.28 ^a	-0.32 ^a	-0.28 ^a	-0.39 ^a	-0.31 ^a	-0.39 ^b	-0.39 ^b
Large	Reference	Reference	Reference	Reference	Reference	Reference	Reference	Reference						
Type of caste														
(SC)	-0.05	-0.06	-0.11	-0.17 ^b	-0.15 ^b	-0.18 ^a	-0.13 ^b	-0.15 ^a	-0.23 ^a	-0.12 ^c	-0.18 ^b	-0.06	-0.17	-0.13
ST	-0.33 ^a	-0.21 ^c	-0.12	-0.16 ^c	0.12 ^c	-0.07	0.17 ^a	0.09 ^c	0.24 ^a	0.31 ^a	0.53 ^a	0.58 ^a	0.93 ^a	1.03 ^a
Other caste	Reference	Reference	Reference	Reference	Reference	Reference	Reference	Reference						
Birth order														
4 or more	-0.13	-0.19 ^b	-0.14 ^b	-0.12	-0.15 ^b	-0.08	-0.13 ^a	-0.08 ^c	-0.13 ^b	-0.04	-0.001	-0.11	-0.09	-0.15
Three	-0.07	-0.03	-0.03	-0.06	0.04	0.004	-0.02	-0.003	-0.11 ^c	-0.08	-0.11	-0.1	-0.33 ^b	-0.04
Second	-0.03	0.02	0.01	0.04	0.05	0.04	0	0.01	-0.02	0.04	-0.01	0.06	-0.11	0.1
First	Reference	Reference	Reference	Reference	Reference	Reference	Reference	Reference						

a: 1% Significance; b: 5% Significance; c: 10% Significance

Table 3: Quantile Regression Estimations for the different quantiles of the female (11582) and Male (12742) children's, Dependent variable HAZ (Stunting) NFHS-2

Parameter	0.05		0.1		0.25		0.50		0.75		0.9		0.95	
	Female	Male												
Intercept	-1.99 ^a	-2.29 ^a	-1.48	-1.50 ^a	-0.57 ^a	-0.59 ^a	0.53 ^a	0.41 ^a	1.38 ^a	1.41 ^a	2.41 ^a	2.69 ^a	2.98 ^a	3.29 ^a
Child age														
25-36in Months	-1.75 ^a	-1.32 ^a	-1.81 ^a	-1.48 ^a	-1.75 ^a	-1.51 ^a	-1.87 ^a	-1.63 ^a	-2.02 ^a	-1.89 ^a	-2.20 ^a	-2.24 ^a	-2.17 ^a	-2.28 ^a
13-24 months	-1.52 ^a	-1.20 ^a	-1.45 ^a	-1.33 ^a	-1.41 ^a	-1.41 ^a	-1.51 ^a	-1.53 ^a	-1.72 ^a	-1.64 ^a	-1.87 ^a	-1.95 ^a	-1.93 ^a	-1.93 ^a
7-12 months	-0.48 ^a	-0.54 ^a	-0.51 ^a	-0.59 ^a	-0.62 ^a	-0.61 ^a	-0.69 ^a	-0.67 ^a	-0.84 ^a	-0.78 ^a	-1.01 ^a	-1.03 ^a	-1.01 ^a	-0.96 ^a
0-6 months	Reference													
Mothers Education														
No education	-1.22 ^a	-1.15 ^a	-1.21 ^a	-1.18 ^a	-1.11 ^a	-1.20 ^a	-1.04 ^a	-0.98 ^a	-0.82 ^a	-0.86 ^a	-0.66 ^a	-0.76 ^a	-0.38 ^b	-0.61 ^a
<5 years	-0.95 ^a	-0.79 ^a	-0.87 ^a	-0.74 ^a	-0.69 ^a	-0.81 ^a	-0.69 ^a	-0.65 ^a	-0.51 ^a	-0.64 ^a	-0.43 ^a	-0.77 ^a	-0.31	-0.55 ^a
5-7 years	-0.52 ^a	-0.73 ^a	-0.54 ^a	-0.81 ^a	-0.55 ^a	-0.71 ^a	-0.57 ^a	-0.67 ^a	-0.50 ^a	-0.55 ^a	-0.41 ^a	-0.60 ^a	-0.23	-0.57 ^a
8-9 years	-0.64 ^a	-0.52 ^a	-0.52 ^a	-0.49 ^a	-0.40 ^a	-0.49 ^a	-0.48 ^a	-0.37 ^a	-0.33 ^a	-0.33 ^a	-0.30 ^b	-0.42 ^a	-0.1	-0.42 ^b
10-11 years	-0.50 ^a	-0.46 ^a	-0.28 ^b	-0.35 ^a	-0.30 ^a	-0.35 ^a	-0.33 ^a	-0.25 ^a	-0.17 ^c	-0.2	-0.15	-0.19	-0.1	-0.30 ^c
>12 years	Reference													
Size at birth														
Small	-0.42 ^a	-0.53 ^a	-0.45 ^a	-0.56 ^a	-0.55 ^a	-0.57 ^a	-0.60 ^a	-0.59 ^a	-0.52 ^a	-0.56 ^a	-0.50 ^a	-0.61 ^a	-0.49 ^a	-0.55 ^a
Average	-0.16 ^c	-0.27 ^a	-0.13 ^c	-0.33 ^a	-0.24 ^a	-0.34 ^a	-0.28 ^a	-0.32 ^a	-0.22 ^a	-0.32 ^a	-0.26 ^a	-0.42 ^a	-0.23 ^c	-0.36 ^a
Large	Reference													
Type of cast														
SC	-0.21 ^a	-0.18 ^b	-0.16 ^b	-0.27 ^a	-0.19 ^a	-0.23 ^a	-0.19 ^a	-0.29 ^a	-0.18 ^a	-0.21 ^a	-0.20 ^a	-0.26 ^a	-0.27 ^b	-0.26 ^b
ST	-0.06	-0.18 ^c	0.01	-0.25 ^a	0.12 ^b	-0.14 ^b	0.056	-0.13 ^a	0.15 ^b	-0.04 ^c	0.16 ^b	0.14 ^c	0.30 ^b	0.43 ^a
OBC	-0.07	-0.22 ^a	-0.14 ^b	-0.25 ^a	-0.05	-0.11 ^b	-0.05	-0.08 ^c	-0.09 ^c	-0.04	-0.13 ^c	-0.07	-0.20 ^c	-0.03
Other caste	Reference													
Birth order														
4 or more	-0.29 ^a	-0.22 ^a	-0.31 ^a	-0.25 ^a	-0.33 ^a	-0.23 ^a	-0.29 ^a	-0.23 ^a	-0.17 ^a	-0.13	0.03	0.05	0.29 ^b	0.17
Three	-0.12	-0.04	-0.08	-0.11 ^c	-0.13 ^b	-0.08	-0.18 ^a	-0.09 ^c	-0.61	-0.11 ^c	-0.02	-0.14 ^c	-0.06	-0.20 ^c
Second	-0.1	-0.0575	-0.1	-0.067	-0.08 ^c	-0.04	-0.11 ^b	-0.06	-0.07	-0.05	0.03	-0.03	-0.03	0.08
First	Reference													
Mothers BMI														
BMI<18.5	-0.01	-0.07	-0.05	-0.07	-0.08 ^b	-0.09 ^a	-0.09 ^a	-0.17 ^a	-0.17 ^a	-0.24 ^a	-0.28 ^a	-0.28 ^a	-0.44 ^a	-0.50 ^a
BMI>25	0.40 ^a	0.2	0.37 ^a	0.27 ^a	0.51 ^a	0.29 ^a	0.428 ^a	0.24 ^a	0.40 ^a	0.28 ^b	0.36 ^a	0.28 ^b	0.19	0.23
BMI 18.5-25	Reference													

a: 1% Significance; b: 5% Significance; c: 10% Significance

Table 4: Quantile Regression Estimations for the different quantiles of the female (11315) and male (12133) children's, Dependent variable HAZ (Stunting) NFHS-3

Parameter	0.05		0.1		0.25		0.5		0.75		0.9		0.95	
	Female	Male												
Intercept	-2.18 ^a	-2.43 ^a	-1.51 ^a	-1.68 ^a	-0.50 ^a	-0.54 ^a	0.47 ^a	0.58 ^a	1.40 ^a	1.55 ^a	2.19 ^a	2.44 ^a	2.95 ^a	2.96 ^a
Child age in months														
25-36in Months	-0.86 ^a	-0.89 ^a	-1.05 ^a	-1.02 ^a	-1.36 ^a	-1.28 ^a	-1.51 ^a	-1.47 ^a	-1.68 ^a	-1.67 ^a	-1.67 ^a	-1.72 ^a	-1.88 ^a	-1.56 ^a
13-24 months	-0.86 ^a	-0.87 ^a	-0.98 ^a	-1.01 ^a	-1.26 ^a	-1.28 ^a	-1.42 ^a	-1.45 ^a	-1.58 ^a	-1.58 ^a	-1.59 ^a	-1.59 ^a	-1.74 ^a	-1.41 ^a
7-12 months	-0.16	-0.26 ^b	-0.35 ^a	-0.41 ^a	-0.46 ^a	-0.56 ^a	-0.54 ^a	-0.63 ^a	-0.57 ^a	-0.68 ^a	-0.56 ^a	-0.70 ^a	-0.61 ^a	-0.46 ^a
0-6 months	Reference													
Mothers Education														
No education	-0.79 ^a	-0.50 ^a	-0.64 ^a	-0.55 ^a	-0.51 ^a	-0.58 ^a	-0.55 ^a	-0.63 ^a	-0.42 ^a	-0.51 ^a	-0.26 ^b	-0.52 ^a	-0.04	-0.57 ^a
<5 years	-0.90 ^a	-0.19	-0.59 ^a	-0.14	-0.47 ^a	-0.28 ^a	-0.58 ^a	-0.37 ^a	-0.42 ^a	-0.31 ^a	-0.2	-0.26 ^c	-0.03	-0.12
5-7 years	-0.31 ^b	-0.30 ^b	-0.24 ^a	-0.36 ^a	-0.24 ^a	-0.40 ^a	-0.32 ^a	-0.50 ^a	-0.30 ^a	-0.41 ^a	-0.29 ^b	-0.42 ^a	-0.16	-0.39 ^b
8-9 years	-0.31 ^b	-0.15	-0.21 ^b	-0.14	-0.18 ^b	-0.24 ^a	-0.24 ^a	-0.39 ^a	-0.21 ^a	-0.34 ^a	-0.19	-0.33 ^a	0.06	-0.11
10-11 years	-0.12	-0.21 ^c	0.05	-0.13	-0.07	-0.15 ^c	-0.17 ^a	-0.33 ^a	-0.22 ^a	-0.27 ^a	-0.21	-0.29 ^b	0.07	-0.33 ^b
>12 years	Reference													
Size at birth														
(Small)	-0.32 ^a	-0.31 ^a	-0.35 ^a	-0.39 ^a	-0.35 ^a	-0.40 ^a	-0.27 ^a	-0.45 ^a	-0.19 ^a	-0.44 ^a	-0.19 ^b	-0.42 ^a	-0.13	-0.49 ^a
Average	-0.11	-0.03	-0.10 ^c	-0.03	-0.12 ^a	-0.10 ^b	-0.06	-0.15 ^a	-0.03	-0.19 ^a	-0.08	-0.16 ^b	0.01	-0.18 ^c
Large	Reference													
Type of caste														
SC	-0.1	-0.11	-0.16 ^b	-0.23 ^a	-0.24 ^a	-0.24 ^a	-0.24 ^a	-0.23 ^a	-0.25 ^a	-0.23 ^a	-0.09	-0.19 ^b	0.04	0.01
ST	0.04	-0.04	0.01	-0.09	0.05	-0.06	0.09 ^c	-0.04	0.16 ^a	-0.07	0.34 ^a	0.14	0.54 ^a	0.41 ^a
OBC	-0.14 ^c	-0.1	-0.14 ^a	-0.17 ^a	-0.13 ^a	-0.15 ^a	-0.12 ^a	-0.15 ^a	-0.12 ^b	-0.14 ^a	0.01	-0.05	0.05	0.12
Other caste	Reference													
Birth order														
(4 or more)	-0.25 ^a	-0.51 ^a	-0.29 ^a	-0.37 ^a	-0.25 ^a	-0.25 ^a	-0.18 ^a	-0.18 ^a	-0.16 ^a	-0.08	-0.07	0.14 ^c	0.09	0.34 ^a
Three	-0.15	-0.24 ^a	-0.18 ^a	-0.17 ^b	-0.11 ^b	-0.16 ^a	-0.05	-0.15 ^a	-0.03	-0.12 ^b	0.04	-0.05	0.05	0.23 ^c
Second	-0.04	-0.22 ^a	-0.08	-0.07	-0.11 ^b	-0.05	-0.04	-0.09 ^b	-0.03	-0.01	0.02	0.04	0.01	-0.03
First	Reference													
Mothers nutritional status														
BMI<18.5	-0.07	-0.14 ^b	-0.08 ^c	-0.14 ^b	-0.16 ^a	-0.17 ^a	-0.21 ^a	-0.18 ^a	-0.18 ^a	-0.18 ^a	-0.29 ^a	-0.18	-0.41 ^a	-0.24 ^a
BMI>25	0.11	0.08	0.16 ^b	0.07	0.18 ^a	0.09	0.16 ^a	0.12 ^b	0.25 ^a	0.11	0.27 ^b	0.01	0.01	-0.07
BMI 18.5-25	Reference													
Wealth Index														
(Poorest)	-0.73 ^a	-0.72 ^a	-0.79 ^a	-0.82 ^a	-0.73 ^a	-0.80 ^a	-0.64 ^a	-0.66 ^a	-0.64 ^a	-0.61 ^a	-0.54 ^a	-0.53 ^a	-0.67 ^a	-0.46 ^a
Poorer	-0.52 ^a	-0.73 ^a	-0.52 ^a	-0.71 ^a	-0.52 ^a	-0.60 ^a	-0.48 ^a	-0.57 ^a	-0.46 ^a	-0.50 ^a	-0.44 ^a	-0.47 ^a	-0.47 ^a	-0.43 ^a
Middle	-0.40 ^a	-0.50 ^a	-0.49 ^a	-0.53 ^a	-0.42 ^a	-0.48 ^a	-0.39 ^a	-0.31 ^a	-0.38 ^a	-0.37 ^a	-0.31 ^a	-0.44 ^a	-0.43 ^b	-0.52 ^a
Richer	-0.46 ^a	-0.26 ^b	-0.40 ^a	-0.30 ^a	-0.26 ^a	-0.30 ^a	-0.24 ^a	-0.20 ^a	-0.25 ^a	-0.22 ^a	-0.16	-0.22 ^b	-0.38 ^b	-0.40 ^a
Richest	Reference													

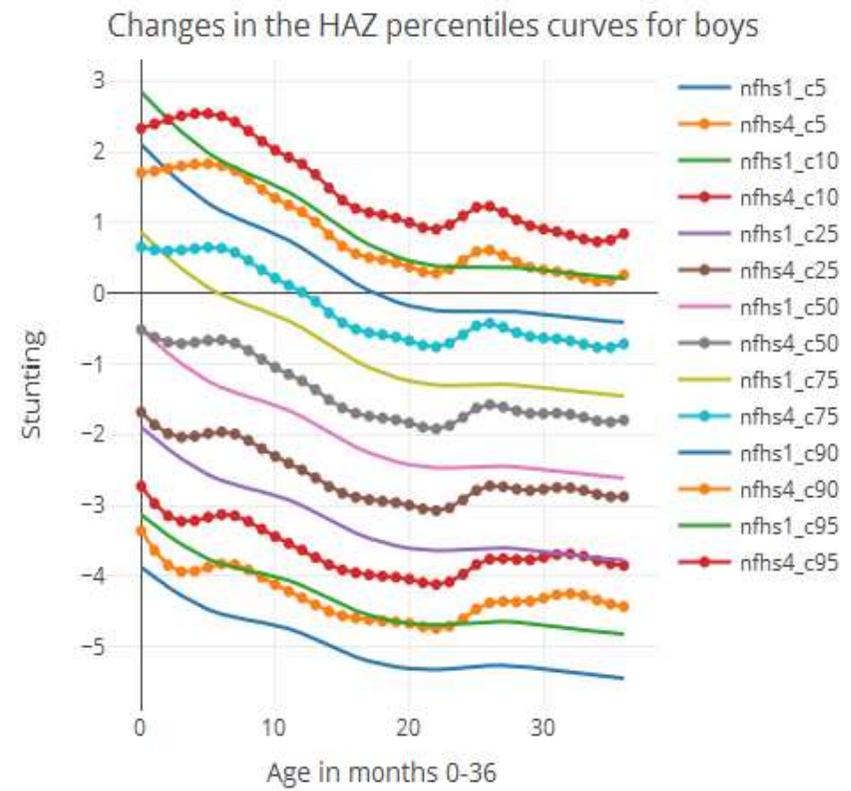
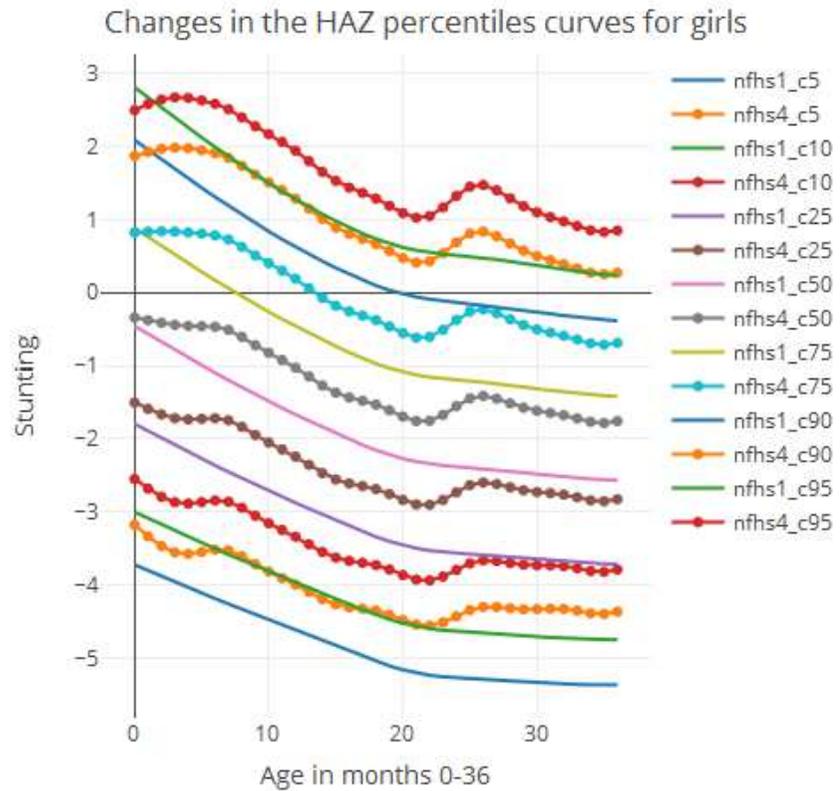
a: 1% Significance; b: 5% Significance; c: 10% Significance

Table 5: Quantile Regression Estimations for the different quantiles of the Female (61582) and Male (66018) children's, Dependent variable HAZ (Stunting). NFHS-4

Parameter	0.05		0.1		0.25		0.5		0.75		0.9		0.95	
	Female	Male												
Intercept	-2.72 ^a	-3.28 ^a	-1.87 ^a	-2.36 ^a	-0.66 ^a	-0.95 ^a	0.41 ^a	0.26 ^a	1.54 ^a	1.37 ^a	2.73 ^a	2.49 ^a	3.56 ^a	3.29 ^a
Child age														
25-36 in Months	-0.50 ^a	-0.12 ^a	-0.67 ^a	-0.36 ^a	-1.02 ^a	-0.82 ^a	-1.23 ^a	-1.15 ^a	-1.41 ^a	-1.36 ^a	-1.42 ^a	-1.41 ^a	-1.37 ^a	-1.37 ^a
13-24 months	-0.52 ^a	-0.41 ^a	-0.68 ^a	-0.59 ^a	-1.00 ^a	-1.00 ^a	-1.15 ^a	-1.21 ^a	-1.27 ^a	-1.28 ^a	-1.20 ^a	-1.26 ^a	-1.19 ^a	-1.15 ^a
7-12 months	-0.07	-0.11 ^b	-0.13 ^a	-0.11 ^a	-0.28 ^a	-0.29 ^a	-0.35 ^a	-0.35 ^a	-0.38 ^a	-0.37 ^a	-0.32 ^a	-0.30 ^a	-0.27 ^a	-0.28 ^a
0-6 months	Reference													
Mothers Education														
No education	-0.50 ^a	-0.46 ^a	-0.53 ^a	-0.46 ^a	-0.48 ^a	-0.47 ^a	-0.43 ^a	-0.46 ^a	-0.34 ^a	-0.42 ^a	-0.27 ^a	-0.31 ^a	-0.15	-0.31 ^a
<5 years	-0.34 ^a	-0.27 ^a	-0.36 ^a	-0.26 ^a	-0.37 ^a	-0.31 ^a	-0.32 ^a	-0.35 ^a	-0.32 ^a	-0.38 ^a	-0.26 ^a	-0.40 ^a	-0.30 ^b	-0.38 ^a
5-7 years	-0.30 ^a	-0.24 ^a	-0.30 ^a	-0.29 ^a	-0.30 ^a	-0.32 ^a	-0.31 ^a	-0.31 ^a	-0.29 ^a	-0.33 ^a	-0.24 ^a	-0.30 ^a	-0.17 ^c	-0.29 ^a
8-9 years	-0.09 ^c	-0.07	-0.17 ^a	-0.12 ^a	-0.21 ^a	-0.17 ^a	-0.25 ^a	-0.22 ^a	-0.24 ^a	-0.24 ^a	-0.24 ^a	-0.27 ^a	-0.20 ^b	-0.30 ^a
10-11 years	-0.01	-0.08	-0.10 ^b	-0.08 ^c	-0.11 ^a	-0.09 ^a	-0.11 ^a	-0.12 ^a	-0.12 ^a	-0.10 ^a	-0.10 ^c	-0.07	-0.08	-0.1
>12 years	Reference													
Size at birth														
(Small)	-0.35 ^a	-0.22 ^a	-0.36 ^a	-0.27 ^a	-0.36 ^a	-0.41 ^a	-0.41 ^a	-0.45 ^a	-0.48 ^a	-0.47 ^a	-0.61 ^a	-0.54 ^a	-0.64 ^a	-0.62 ^a
Average	0.02	-0.02	-0.002	-0.04	-0.05 ^b	-0.11 ^a	-0.08 ^a	-0.13 ^a	-0.13 ^a	-0.17 ^a	-0.2 ^a	-0.24 ^a	-0.19 ^b	-0.29 ^a
Large	Reference													
Type of cast														
(SC)	-0.10 ^b	-0.19 ^a	-0.17 ^a	-0.24 ^a	-0.20 ^a	-0.23 ^a	-0.24 ^a	-0.22 ^a	-0.24 ^a	-0.22 ^a	-0.21 ^a	-0.19 ^a	-0.31 ^a	-0.13 ^c
ST	0.01	-0.20 ^a	-0.01	-0.18 ^a	0.001	-0.09 ^a	0.02	-0.07 ^a	0.05 ^c	-0.07 ^a	0.14 ^a	0.05	0.29 ^a	0.18 ^b
OBC	-0.06	-0.12 ^a	-0.12 ^a	-0.14 ^a	-0.15 ^a	-0.16 ^a	-0.16 ^a	-0.16 ^a	-0.18 ^a	-0.20 ^a	-0.16 ^a	-0.22 ^a	-0.17 ^b	-0.23 ^a
Other caste	Reference													
Birth order														
(4 or more)	-0.30 ^a	-0.12 ^b	-0.25 ^a	-0.14 ^a	-0.23 ^a	-0.16 ^a	-0.21 ^a	-0.12 ^a	-0.18 ^a	-0.11 ^a	-0.13 ^a	-0.02	-0.1	0.07
Three	-0.17 ^a	-0.10 ^b	-0.18 ^a	-0.11 ^a	-0.15 ^a	-0.10 ^a	-0.14 ^a	-0.09 ^a	-0.11 ^a	-0.07 ^a	-0.06	-0.04	0.03	0.02
Second	-0.08 ^b	0.02	-0.10 ^a	-0.03	-0.07 ^a	-0.06 ^a	-0.07 ^a	-0.04 ^b	-0.05 ^b	-0.03	-0.08 ^b	0.03	-0.01	0.10 ^c
First	Reference													
Mothers BMI														
BMI<18.5	-0.11 ^a	-0.06 ^c	-0.13 ^a	-0.10 ^a	-0.16 ^a	-0.15 ^a	-0.20 ^a	-0.19 ^a	-0.27 ^a	-0.21 ^a	-0.42 ^a	-0.34 ^a	-0.57 ^a	-0.46 ^a
BMI>25	0.28 ^a	0.26 ^a	0.24 ^a	0.25 ^a	0.20 ^a	0.19 ^a	0.15 ^a	0.13 ^a	0.09 ^a	0.12 ^a	0.02	0.08 ^c	-0.11	0.03
BMI 18.5-25	Reference													
Wealth Index														
(Poorest)	-0.62 ^a	-0.56 ^a	-0.67 ^a	-0.62 ^a	-0.68 ^a	-0.63 ^a	-0.60 ^a	-0.58 ^a	-0.57 ^a	-0.51 ^a	-0.54 ^a	-0.47 ^a	-0.53 ^a	-0.46 ^a
Poorer	-0.48 ^a	-0.31 ^a	-0.47 ^a	-0.40 ^a	-0.46 ^a	-0.39 ^a	-0.45 ^a	-0.38 ^a	-0.46 ^a	-0.39 ^a	-0.47 ^a	-0.42 ^a	-0.46 ^a	-0.39 ^a
Middle	-0.31 ^a	-0.20 ^a	-0.33 ^a	-0.17 ^a	-0.32 ^a	-0.23 ^a	-0.32 ^a	-0.25 ^a	-0.34 ^a	-0.24 ^a	-0.38 ^a	-0.25 ^a	-0.34 ^a	-0.18 ^b
Richer	-0.20 ^a	-0.08	-0.18 ^a	-0.09 ^b	-0.19 ^a	-0.10 ^a	-0.17 ^a	-0.11 ^a	-0.20 ^a	-0.12 ^a	-0.30 ^a	-0.11 ^b	-0.23 ^b	-0.07
Richest	Reference													

a: 1% Significance; b: 5% Significance; c: 10% Significance

Figure 1: LMS curves for child aged 0-36 months in NFHS 1 (1992) and NFHS 4 (2016)



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Figures

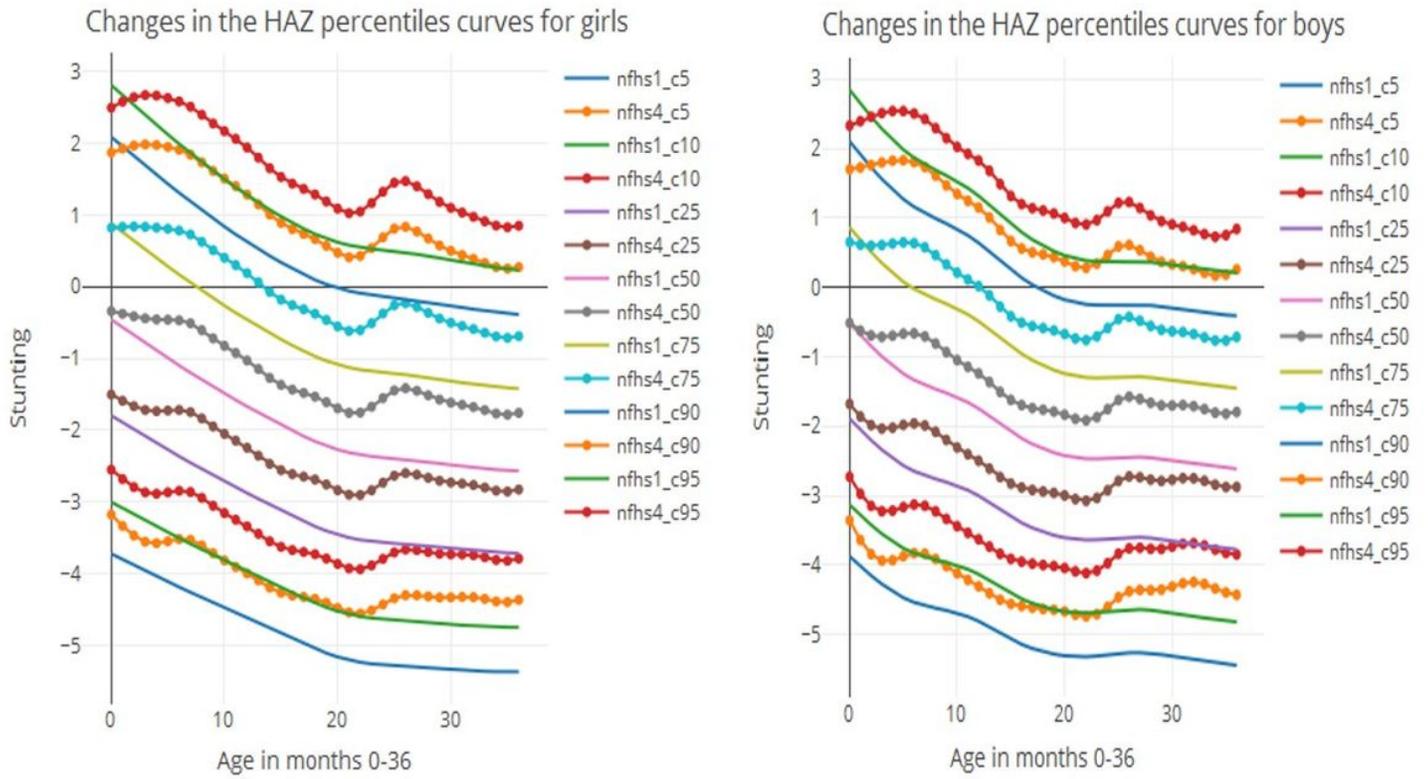


Figure 1

LMS curves for child aged 0-36 months in NFHS 1 (1992) and NFHS 4 (2016)