

Looking upstream: enhancers of child nutritional status in post-flood rural settings

Jose Manuel Rodriguez-Llanes, Shishir Ranjan-Dash, Alok Mukhopadhyay, Debarati Guha-Sapir

Background. Child undernutrition and flooding are highly prevalent public health issues in many developing countries. Yet we have little understanding of preventive strategies for effective coping in these circumstances. Education has been recently highlighted as key to reduce the societal impacts of extreme weather events under climate change, but there is a lack of studies assessing to what extent parental education may prevent post-flood child undernutrition. **Methods and materials.** One year after large floods in 2008, we conducted a two-stage cluster population-based survey of 6–59 months children inhabiting flooded and non-flooded communities of Jagatsinghpur district, Odisha (India), and collected anthropometric measurements on children along with child, parental and household level variables through face-to-face interviews. Using multivariate logistic regression models, we examined separately the effect of maternal and paternal education and other risk factors (mainly income, socio-demographic, and child and mother variables) on stunting and wasting in children from households inhabiting recurrently flooded communities (2006 and 2008; $n = 299$). As a comparison, separate analyses on children in non-flooded communities were carried out ($n = 385$). All analyses were adjusted by income as additional robustness check. **Results.** Overall, fathers with at least completed middle education (up to 14 years of age and compulsory in India) had an advantage in protecting their children from child wasting and stunting. For child stunting, the clearest result was a 100 to 200% lower prevalence associated with at least paternal secondary schooling (compared to no schooling) in flooded-areas. Again, only in flooded communities, an increase in per capita annual household income of 1,000 rupees was associated to a 4.7–4.9% lower prevalence of child stunting. For child wasting and only in flooded areas, delayed motherhood was associated to better nutritional outcomes (3.4% lower prevalence per year). In flooded communities, households dedicated to activities other than agriculture, a 50 to 51% lower prevalence of child wasting was estimated, suggesting farmers and fishermen as the most vulnerable livelihoods under flooding. In flooded areas, lower rank castes were at higher odds of both child wasting and stunting. **Conclusions.** In the short-term, protracted nutritional response in the aftermath of floods should be urgently implemented and target agricultural livelihoods and low-rank castes. Education

promotion and schooling up to 14 years should have positive impacts on improving children nutritional health in the long run, especially under flooding. Policies effectively helping sustainable livelihood economic development and delayed motherhood are also recommended.

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2 **post-flood rural settings**

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24 Abstract

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29 parental education may prevent post-flood child undernutrition.

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31 population-based survey of 6–59 months children inhabiting flooded and non-flooded communities of
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33 with child, parental and household level variables through face-to-face interviews. Using multivariate
34 logistic regression models, we examined separately the effect of maternal and paternal education and
35 other risk factors (mainly income, socio-demographic, and child and mother variables) on stunting and
36 wasting in children from households inhabiting recurrently flooded communities (2006 and 2008; $n =$
37 299). As a comparison, separate analyses on children in non-flooded communities were carried out ($n =$
38 385). All analyses were adjusted by income as additional robustness check.

39 **Results.** Overall, fathers with at least completed middle education (up to 14 years of age and
40 compulsory in India) had an advantage in protecting their children from child wasting and stunting. For
41 child stunting, the clearest result was a 100 to 200% lower prevalence associated with at least paternal
42 secondary schooling (compared to no schooling) in flooded-areas. Again, only in flooded communities,
43 an increase in per capita annual household income of 1,000 rupees was associated to a 4.7–4.9% lower
44 prevalence of child stunting. For child wasting and only in flooded areas, delayed motherhood was
45 associated to better nutritional outcomes (3.4% lower prevalence per year). In flooded communities,
46 households dedicated to activities other than agriculture, a 50 to 51% lower prevalence of child wasting
47 was estimated, suggesting farmers and fishermen as the most vulnerable livelihoods under flooding. In
48 flooded areas, lower rank castes were at higher odds of both child wasting and stunting.

49 **Conclusions.** In the short-term, protracted nutritional response in the aftermath of floods should be
50 urgently implemented and target agricultural livelihoods and low-rank castes. Education promotion and
51 schooling up to 14 years should have positive impacts on improving children nutritional health in the
52 long run, especially under flooding. Policies effectively helping sustainable livelihood economic
53 development and delayed motherhood are also recommended.

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56 **Introduction**

57 Among all the disaster risks associated to a warming climate, flooding has become the most frequent
58 and since the nineties it has been affecting *grosso modo* 100 million people a year. This is more than any
59 other disaster type worldwide, climate-related or not (EMDAT, 2015). Yet and relative to their
60 importance, the health consequences of flooding have been rarely investigated (Ahern et al., 2005;
61 Alderman et al., 2012). Even if, at the time of writing, an equivalent 37.8% (ie, 2.8 billion people) of the
62 current world population was affected by floods in the past 25 years, little serious actions are being
63 taken.

64 Education has been recently stressed as one key to reduce the societal impacts of climate change
65 (Striessnig, Lutz & Patt, 2013). The study of educational attainment as a promotor of disaster resilience
66 recently received substantive attention in a collection of 11 papers (Muttarak & Lutz, 2014). The results
67 of this special issue are a convincing step forward. These studies were undertaken in different countries
68 and regions, and targeted relevant outcomes in different phases of the disaster cycle in populations
69 exposed to diverse climate-related hazards. Overall, they found that higher-educated groups avoided
70 high-risk areas to settle, were better prepared, reacted more efficiently to early warnings and had lessen
71 impacts on health and social variables, which indicated according to the authors better coping and
72 recovery in these groups (Muttarak & Lutz, 2014). For the particular case of flooding, the evidence on
73 the role of education as a promotor of positive health outcomes is scant and contentious (Lowe, Ebi &
74 Forsberg, 2013). But what are the plausible pathways linking more education to health improvements?
75 Indeed, formal education is crucial for acquisition and processing of information (eg, literacy),
76 improvement of cognitive abilities, decision making and long-term planning, and generally it leads to
77 securing skilled jobs and ultimately higher income and better health (briefly reviewed in Striessnig, Lutz
78 & Patt, 2013 and Lindeboom, Nozal & Klaauw, 2009).

79 The transgenerational effects of education on health have received substantial attention. Although the
80 effect of maternal education on child stunting and general child health has been studied (Milman et al.,
81 2005; Lindeboom, Nozal & Klaauw, 2009), the effect of paternal education on child health has been
82 more rarely examined (Moestue & Huttly, 2008; Semba et al., 2008). Assessing the extent to which
83 father education affects child's health is an important step to understand the relative contribution of
84 fathers to the family wellbeing. Considering father's education is also a premise to further study the

85 synergies between father and mother education levels in the promotion of child's health (Semba et al.,
86 2008).

87 Adequate child nutrition is a key indicator of wellbeing and development of particular importance in
88 developing countries (Black et al., 2013). Nevertheless, studies connecting flood-exposure to the
89 nutritional status of children have been rare and none of them considered the role of father education
90 in preventing the health impacts of floods among children (Phalkey et al., 2015). To the best of our
91 knowledge, no study to date has investigated particularly the role of parental education on child's
92 nutritional status in post-flood settings. We only found one study which investigated the sole effect of
93 mother education, not father education, on malaria parasitemia in Sub-Saharan children (Siri, 2014).
94 Importantly, no study has looked at the association of education on child wasting, let alone in the
95 context of floods. The investigation of child wasting is relevant as evidence suggest that stunting and
96 wasting represent different processes of undernutrition (Ricci & Becker, 1996). Recent evidence
97 suggests that exposure to floods can be associated to increases in both child stunting (Rodriguez-Llanes
98 et al., 2011) and child wasting (Rodriguez-Llanes et al., 2016) and thus both deserve attention in post-
99 flood settings.

100 The large floods occurring in rural Odisha, India in September 2008 produced massive damage to
101 agriculture, water and sanitation, communication networks and severe disruption to the normal
102 functioning of the entire rural society at large (Government of Orissa, 2008). As part of an integrated
103 project to investigate the health, social, and economic impacts of disasters, we carefully planned and
104 conducted in September 2009 a representative survey of children affected and non-affected by these
105 floods in 265 communities. To get further insight into prevention strategies for the health impacts of
106 floods, which are absent in the literature (Bouزيد, Hooper & Hunter, 2013) we examined in this study the
107 effect of maternal and paternal education and other risk factors on stunting and wasting in children
108 from families living in flooded and non-flooded communities.

109

110 **Materials & Methods**

111 **Design and sample**

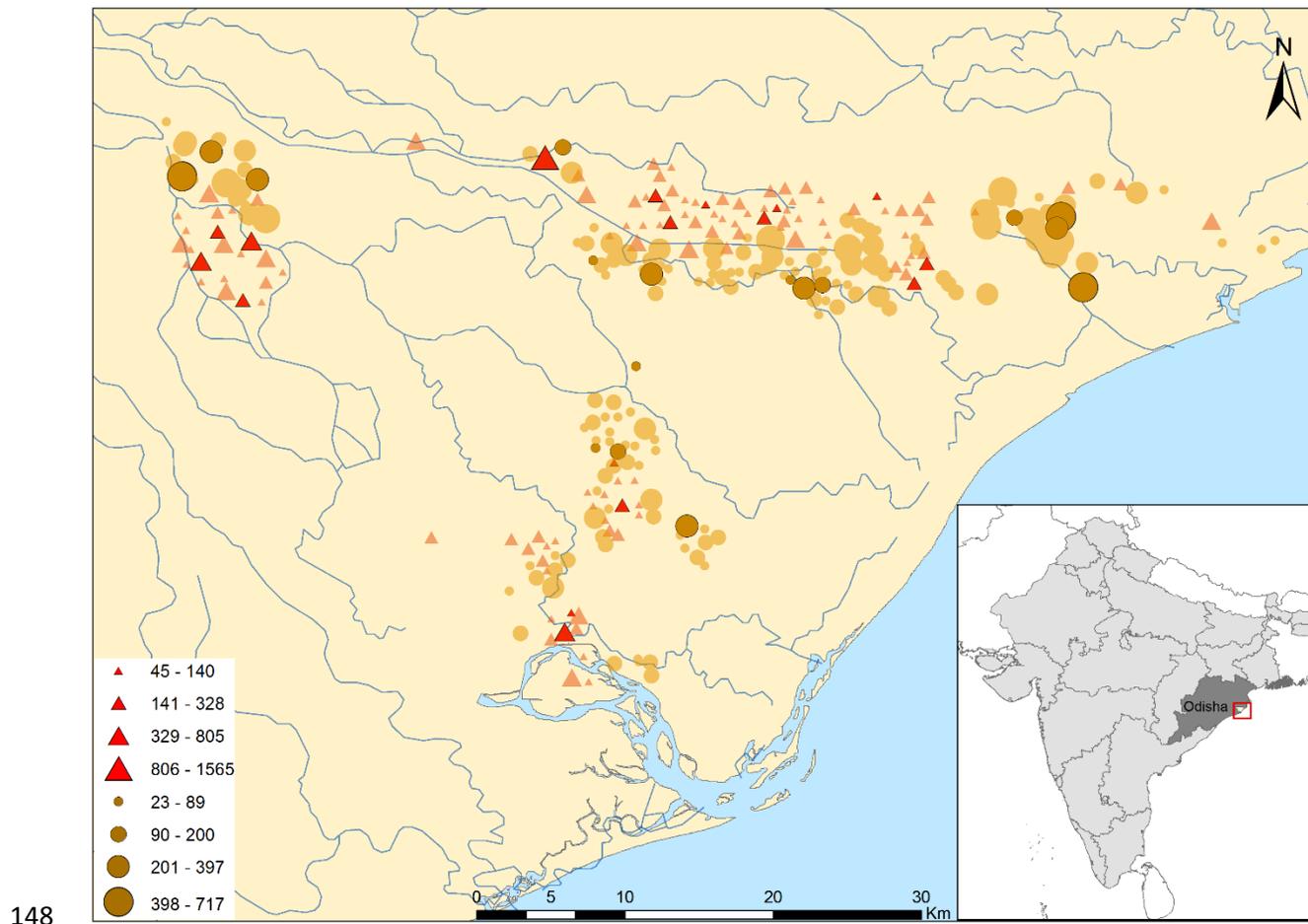
112 Jagatsinghpur is a coastal district of the state of Odisha, located within the Bay of Bengal, India. The
113 district's population is around one million, 90% of which inhabit rural areas (Census of India, 2011). The
114 region is located in a large and fertile flood plain, crossed by large rivers, which makes it very attractive

115 for fishing activities, livestock rearing and crop farming. However, this land is also regularly affected by
116 heavy monsoons which often produce flooding. In the last decade, Jagatsinghpur has been hardly hit by
117 five major floods, including coastal flooding associated to cyclone Paradip (05B) in 1999, followed by
118 heavy rain floods in 2001, 2003, 2006 and the latest flood occurring prior our investigation, which took
119 place in mid-September 2008 (Government of Orissa, 2008).

120 We conducted a population-based survey one year after the 2008 flooding in rural Jagatsinghpur
121 district, Odisha, India. A two-stage cluster survey was necessary to obtain a probability sample of
122 children 6 to 59 months of age in 265 villages from four severely flood-affected blocks of the district
123 (Kujang, Biridi, Balikuda and Tirtol). Overall, 122 of these were flooded and 143 non-flooded in
124 September 2008. The percentage of households flooded in each village was obtained for 2006 and 2008
125 events from OSDMA data (OSDMA, 2009; Fig. 1).

126 We used a two-stage cluster design to select our sample as our population of interest was clustered in
127 villages and the information on the population was scant (Groves et al., 2009). In this study, the Primary
128 Sampling Units (PSUs) were the villages and the children were the Secondary Sampling Unit (SSU). A 30
129 (PSU) by 30 (SSU) design, which should provide a probability sample of 900 children, was fixed. We
130 initially enumerated the 265 villages along with each village population size projected for 2009 from
131 census data (Census of India, 2001), and subsequently 30 clusters from 29 villages were selected. This
132 first selection was done using Probability Proportionate to Size (PPS) sampling with replacement (ie, one
133 large village was selected twice). This method picks up randomly villages, but the chances of selection
134 are proportional to the size of the village, with selection probabilities favoring larger villages (Groves et
135 al., 2009). At the second stage a fixed number of children (ie, 30) are selected by cluster, independently
136 of village size. The result is that chances are compensated with equal probabilities of selection of any
137 eligible child listed in the sampling frame (Groves et al., 2009). Importantly, an updated list of eligible
138 children was obtained from the ICDS centers (1 ICDS center for every 1,000 inhabitants) and validated
139 with the ward members of each village. A total 3,671 eligible children were listed from 29 villages within
140 a month prior training and piloting of the survey instrument. Once the lists compiled, it was detected
141 that three flooded villages (Korana, Jamphar and Raghunathpur) and two non-flooded (Muthapada,
142 Sureilo) did not have enough eligible children (ie, less than 30). We created 5 new clusters by merging
143 the list of children in each of these 5 villages ($n = 280$) with those of the closest non-selected flooded or
144 non-flooded village, respectively, of our list (Fig. 1). Finally, thirty children per cluster were randomly
145 selected.

146 **Figure 1: Study site, eligible villages and original sample of flooded and non-flooded villages in**
 147 **Jagatsinghpur district, Odisha, India.**



149 Triangles represent flooded villages; circles those non-flooded. Size of polygons is proportionate to village size as
 150 measured by number of households (see map legend). Polygons overimpressed identified villages selected.

151

152 **Ethical approval**

153 This study was approved by the Community Health Ethics Committee, Voluntary Health Association of
 154 India, New Delhi. Persons eligible to participate in the study were not offered any monetary incentive.
 155 Written informed consent was obtained for every head of household visited. In case the respondent was
 156 an illiterate, we asked a literate person from the community to read out the consent form and explain it
 157 to the head of the family. Then we obtained the thumb impression of the respondent. In those case, the
 158 person who readout the consent form also signed as a witness. Research procedures were consistent
 159 with the Declaration of Helsinki (1997). Interviews were administered after obtaining informed consent.

160 The protocol was reviewed by a small group of scientists who had experience working with survivors of
161 natural disasters and amended based on their recommendations.

162 **Data collection: instruments and measures**

163 Our survey instrument was adapted from the core one developed and approved by a multidisciplinary
164 consortium of researchers from the MICRODIS project. The instrument development was based on
165 interim literature reviews, and follows the UNICEF conceptual framework on child malnutrition (UNICEF,
166 1997). The questionnaire was validated prior to our study in research sites in India, Indonesia, The
167 Philippines, Vietnam and the UK. We collected background information at the household level, and
168 more specifically from mothers and fathers, covering basic socio-demographic characteristics, wealth,
169 child caring practices, healthcare access, maternal and paternal education, income and credit practices,
170 water and sanitation, food consumption patterns; demographics, nutrition and health status data at the
171 child level.

172 To assess nutritional status using anthropometric indicators, weight and height/length were recorded.
173 Children were weighed without clothes. Weight was measured to the nearest 100 grams by trained
174 research assistants using a beam balance (<10 kg; Raman Surgical Co., Delhi-33, India) and an electronic
175 balance for those children heavier than 10 kg. For children younger than 2 years of age, length
176 measurements were taken to the nearest millimeter in recumbent position with an infantometer
177 (Narang Medical, Delhi-110 028, India). If children were older than 2 years, they were measured
178 standing up with an adjustable board calibrated in millimeters. Each research assistant measured and
179 weighted each child twice to minimize measurement errors and use the average value of both
180 measurements to gain precision. These instruments were calibrated daily.

181 Study questionnaires were administered by twelve experienced research assistants (Rodriguez-Llanes et
182 al., 2011) from the Voluntary Health Association of India (VHAI) who received specific training on
183 anthropometry and interview procedures for this study in late August 2009. The questionnaire was
184 piloted in 12 households (6 in flooded villages and 6 in non-flooded) and improved based on the inputs
185 of the pilot exercise. The study questionnaire was translated to the local language in Odisha (Oriya) and
186 subsequently back translated into English by different professional translators and a researcher checked
187 the level of agreement between both versions. Duration of interviews ranged from 45 to 60 minutes and
188 all field work was completed between 6 and 24 September 2009.

189 **Study variables**

190 As our aim was to estimate the association effect of formal maternal and paternal education on child
191 undernutrition we excluded variables in our dataset which may have mediated this effect (Schisterman,
192 Cole & Platt, 2009). We supported our choice by examining these variables (eg, caregiving practices,
193 food security, and access to health care, water and sanitation) framed within the UNICEF framework for
194 child malnutrition (UNICEF, 1997). Overall 51 variables were assessed in a previous study (Rodriguez-
195 Llanes et al., 2016) but for the purpose of this study we only analyzed distal determinants of child
196 undernutrition.

197 Two outcomes were used in our study: stunting (height-for-age) and wasting (weight-for-height).
198 Stunting is an indicator of chronic malnutrition, whereas wasting often evaluates acute nutritional stress
199 at individual and population levels. The new WHO standard, was used to calculate the z scores for these
200 indicators. Malnutrition was a binary variable indicating whether a children is malnourished, z score <-2
201 (1) or not (0) at the time of the interview.

202 Overall, 17 variables were examined as potential predictors. The two fundamental variables of this study
203 were the level of formal education attained by both children's parents, mother and father. Formal
204 education was the only assessed in this study, as we excluded parents with technical training or
205 professional studies (Fig. 2). We studied education using the same categories as recorded in the original
206 questionnaire, which follow benchmarks in the Indian educational system: never attended school (0
207 years of schooling), completed elementary school (5 years of schooling), middle school (8 years), high
208 school (12 years) and completed university studies (15 years or more). For mothers, age at marriage, age
209 at first delivery and age at birth of the selected child were reported and analyzed in years; the later
210 calculated by subtracting the age of each child (converted to years) to the mother's age. Father's age at
211 birth of the selected child was obtained by same calculations. Child's sex was a binary variable. The age
212 of each child in months was obtained from birth certificates and vaccination cards. If these were not
213 available, local calendars were used. The birthweight of each child was recorded in grams from birth
214 certificates and for analytical purpose we expressed birthweight per 100 grams. The count of children
215 younger than 5 years living in a household were used in our analyses. The principal means of livelihood
216 were dichotomized as agriculture (taken as reference), grouping households dedicated to fishing,
217 livestock rearing and crop farming versus non-agricultural for any other reported activity. Two religions
218 were present in the study area: Hinduism, taken as the reference group, and Islam. The caste of the
219 household was based on the household head and was grouped as a scheduled caste, other backward
220 class or general class (reference category). The general class is the higher caste status. The scheduled

221 caste is the social group historically subject to the higher deprivation levels in the country. Owned land
222 was originally collected in acres but was expressed in hectares and analyzed as a continuous variable.
223 Annual household income was recalculated per capita using data on household size and expressed per
224 1,000 Indian rupees (INR) and modeled as a continuous predictor. A household owning any livestock,
225 including chicken, were modelled against those households not owning any. Finally, we recorded the
226 exact number of persons residing in each household (household size) but were dichotomized as more
227 than 4 or otherwise to account for overcrowding as done by a previous study (Semba et al., 2008). The
228 levels of occupation (employed, unemployed or working as housewife) of mother and father were
229 examined to better comprehend the level of gender empowerment within the household.

230 **Statistical methods**

231 Two expert data managers entered the data. An external researcher (JMR) run exploratory data analysis
232 to identify errors in data entry and other implausible values (Day, Fayers & Harvey, 1998). ENA for
233 SMART software (version November 2008) was used to calculate nutritional indicators with the 2006
234 World Health Organization Standard (ENA, 2008). ENA software is built up with specific functions for
235 detection of outliers and impossible values. These were used to identify further problematic values
236 among the calculated nutritional indicators. Each potential error discovered was discussed and
237 investigated to determine where they originated and subsequently corrected.

238 The main predictor in this study was mother and father education. The original sample size was
239 determined based on requirements of a previous study (Rodriguez-Llanes et al., 2016). Regarding the
240 main research question addressed by this report, overall, available sample sizes were sufficient for
241 subgroup analyses to detect prevalence ratios of 2 or more with a 80% power (Sullivan, Dean & Soe,
242 2009).

243 We examined the relationship between parental education and child nutritional status separately in
244 repeatedly flooded and non-flooded cohorts. A first consideration was not to model maternal and
245 paternal education jointly, but in separate analyses (Fig. 2). Intuitively, individuals with similar education
246 tend to get married more often and this was reflected by our data: variance inflation factor (VIF) did
247 show these two variables being highly collinear. As such, 8 models were fitted on this data (2 datasets x
248 2 outcomes x 2 predictors each). These models are summarized in Fig. 2.

249 Bivariate and multiple adjusted logistic regression models with a quasi-binomial distribution to control
250 for overdispersion were fitted (Lumley, 2010). For multivariate models, we included only variables with
251 $p < 0.1$ in bivariate associations with undernutrition. We first run a full model with all variables having a
252 p -value lower than 0.1, a bit more conservative than some authors recommend (Vittinghoff et al., 2005).
253 On each full model, the VIF on each predictor was calculated and predictors with the highest VIF were
254 sequentially eliminated until all remaining predictors had a VIF lower than 4. Backward selection was
255 then applied to the remaining variables. At each step the non-significant variable with the largest p -
256 value was excluded to obtain a final model including only significant variables at alpha level of 5% (ie, p -
257 value < 0.05). As proposed in recent literature (Siri, 2014; Muttarak & Lutz, 2014), we examined the
258 impact of household annual per capita income on the effects on paternal education to ensure that
259 education and income are independent contributors of child nutrition outcomes. We did that in all 8
260 models.

261 Results were provided as prevalence ratios, crude [PR] or adjusted [aPR] with 95% confidence intervals.
262 Given the limits of our sample size, interactions were not examined. All tests were two-tailed with $\alpha =$
263 0.05. All analyses were weighted. Weights were calculated as the inverse of the selection probabilities.
264 Statistical analyses were conducted in R (version 3.0.2)(R Development Core Team, 2008) with the
265 survey package (Lumley, 2010).

266 **Results**

267 **Study sample**

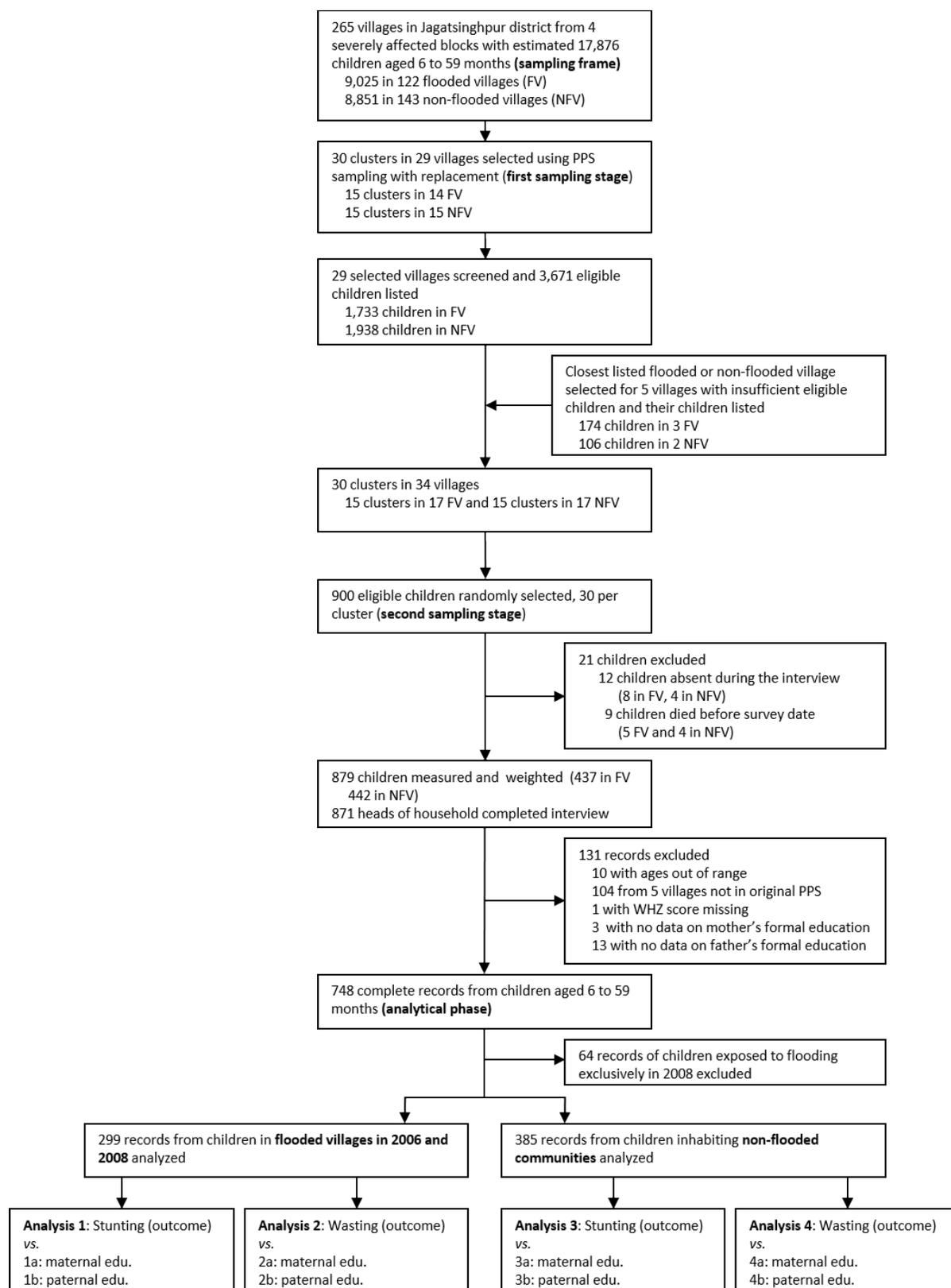
268 A sample representative of the children population 6 to 59 months of age was obtained in 265 villages of
269 Jagatsinghpur district. Our analyses excluded children with missing observations on relevant variables
270 and from villages inundated in September 2008 but not in 2006. Fig. 2 provides further details on each
271 stage of the sampling and exclusions before statistical analysis.

272 Table 1 presents descriptive information for the variables analyzed in recurrently flooded communities
273 ($n = 299$) and those non-flooded ($n = 385$). Mother's illiteracy was less prevalent in flooded areas
274 compared to non-flooded. Whereas more mothers completed middle school in flooded communities,
275 more completed university studies in non-flooded ones. The percentage of fathers completing high
276 school was 10% higher in flooded communities compared to non-flooded, while those having university
277 degrees was similar. In general the flooded population was more educated relative to the non-flooded,

278 and fathers were more educated than mothers. Almost 75% of the mothers attained at least middle
279 school amongst the flooded villages relative to 67% in the non-flooded. For a similar comparison
280 amongst fathers, the percentages were higher, 89% and 81.2%, respectively.

281 Flooded households showed higher reliance on agriculture activities as per their higher percentages of
282 land and livestock owned, which corresponded to higher proportions dedicated to agricultural activities
283 (41.8% in flooded vs 35.8%% in non-flooded). Households of the highest cast were also more common in
284 the sample of flooded villages whereas schedule and other backward castes were found more often in
285 non-flooded communities. Unemployment rates among fathers were 2.1% in non-flooded compared to
286 2.3% in recurrently-flooded populations. As many as 97.7% of all interviewed mothers worked as
287 housewives in both flooded and non-flooded communities.

288 **Figure 2: Flow diagram of sampling procedure, sample obtained and analyses undertaken.**



289

290

291 **Table 1: Prevalence of stunting, wasting and baseline characteristics in rural flooded and non-flooded**
 292 **communities of Odisha, India.**

293

Variables	Flooded (<i>n</i> = 299)		Non-flooded (<i>n</i> = 385)	
	<i>n</i>	% or mean (SE)	<i>n</i>	% or mean (SE)
Stunting	299	30.4 (0.03)	385	29.0 (0.03)
Wasting	299	51.5 (0.03)	385	20.3 (0.03)
Maternal education				
None	17	5.4 (0.01)	37	9.5 (0.02)
Primary school	59	19.9 (0.03)	96	23.5 (0.02)
Middle school	106	36.0 (0.03)	105	26.1 (0.02)
High school	85	29.3 (0.03)	108	28.7 (0.03)
College or more	32	9.3 (0.02)	39	12.2 (0.02)
Paternal education				
None	6	2.5 (0.01)	15	4.1 (0.01)
Primary school	29	8.5 (0.02)	55	14.2 (0.02)
Middle school	87	25.4 (0.03)	116	29.6 (0.03)
High school	116	41.0 (0.03)	123	31.0 (0.03)
College or more	61	22.6 (0.03)	76	21.2 (0.03)
Mean age at marriage of the mother, years	299	21.8 (0.18)	385	21.5 (0.17)
Mean age of the mother at first delivery, years	299	23.7 (0.17)	385	23.4 (0.17)
Mean age of the mother at birth of selected child, years	299	26.1 (0.24)	385	25.9 (0.21)
Mean age of the father at birth of selected child, years	299	31.3 (0.27)	385	31.1 (0.27)
Mean annual income per household capita, 1,000 rupees	299	7.5 (0.60)	385	7.3 (0.63)
Means of livelihood linked to agriculture	299	41.8 (0.03)	385	35.8 (0.03)
Mean land owned, hectares	299	0.6 (0.05)	385	0.4 (0.04)
Any livestock owned	184	64.7 (0.03)	209	55.2 (0)
Religion				
Hindu	248	89.9 (0.01)	357	92.7 (0.01)
Muslim	51	10.1 (0.01)	28	7.3 (0.01)
Caste				
General	83	33.2 (0.03)	68	20.5 (0.03)
Other backward	100	32.4 (0.03)	183	41.9 (0.03)
Scheduled caste	66	24.6 (0.03)	106	30.3 (0.03)
No caste	50	9.8 (0.01)	28	7.3 (0.01)
No. of individuals eating from same kitchen > 4	240	78.7 (0.03)	287	76.5 (0.02)
Mean no. of children under five eating from same kitchen	299	1.4 (0.04)	385	1.5 (0.05)
Child female	140	45.0 (0.03)	175	44.4 (0.03)
Mean birthweight, 100 grams	299	27.7 (0.25)	385	27.4 (0.21)
Mean child age, months	299	33.4 (1.08)	385	32.5 (0.97)

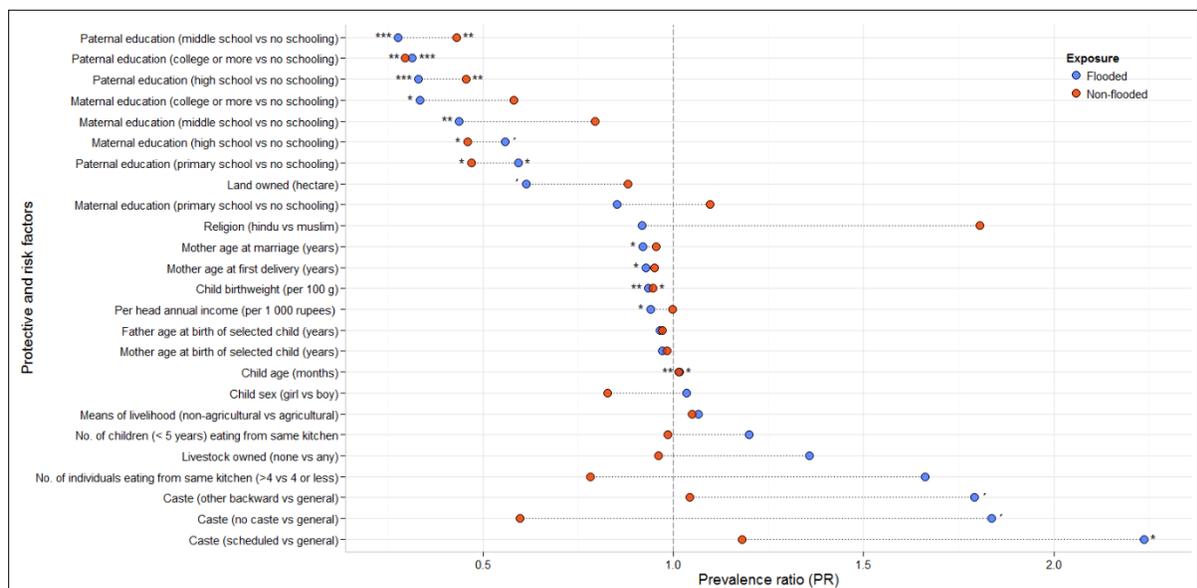
294

295 **Correlates of child stunting**

296 In univariate analyses, paternal and maternal education played an important role in reducing child
 297 stunting (Fig. 3). Overall, the protective effect of education on child stunting were substantial for most
 298 educated groups, amongst flooded communities and in men. Compared to mothers never attending
 299 school, a 3-fold (200%) lower prevalence of stunting was observed in most educated ones (ie, those
 300 completing university studies). But generally the effects observed on paternal education were larger
 301 and stronger compared to maternal education (Fig. 3).

302

303 **Figure 3: Factors associated to child stunting in repeatedly flooded and non-flooded communities of**
 304 **rural coastal Odisha, India.**



305

306 Blue dots show the prevalence ratios in repeatedly flooded communities. Red dots in non-flooded. Dotted lines
 307 show the relative difference in the univariate effects between flooded and unflooded communities. *** p<0.001;
 308 ** p<0.01; * p<0.05, ' p<0.1.

309

310 An additional hectare of land owned in flooded communities had a substantial protective effect,
 311 although non-significant ($p = 0.07$; detailed results in Table S1), but the effect was lower in non-flooded.
 312 A per capita household annual income improvement of 1,000 rupees was also associated with a
 313 decrease of about 6% in child stunting, but again, only in flooded households (Fig. 3, Table S1).
 314 Belonging to a backward caste was detrimental to child stunting but only in flooded communities, and
 315 especially deleterious in the most deprived caste in India, the scheduled caste, with more than a 2-fold
 316 associated increase in the prevalence of child stunting.

317

318 Multivariate analyses on child stunting

319 In adjusted analyses, the positive effect of maternal and paternal education on child nutrition weakened
320 and remained statistically significant only for models on paternal education in flooded communities; and
321 marginally significant in non-flooded (Table 2). In repeatedly-flooded communities, fathers with
322 completed middle, high school or university studies were consistently associated with 2.5-, 2.1- and 2.7-
323 fold lower prevalence of child stunting, respectively. In those non-flooded, completed middle or high
324 school by fathers conferred lower protection to their children, with associated 1.9- ($p = 0.055$) and 1.8-
325 fold ($p = 0.085$) lower prevalence of stunting but none reached statistical significance. Having completed
326 college was comparably as protective in non-flooded areas as in flooded (2.8-fold reduction), and the
327 effect statistically significant ($p = 0.015$).

328 Per capita annual income was a consistent effect in flooded studied areas and across models on
329 maternal and paternal education with 4.7% to 4.9% lower prevalence of stunting for each yearly
330 increase in 1,000 rupees per capita at household level. Schedule caste remained a relevant factor in
331 multivariate analysis modeling the effect of maternal education but only in flooded communities (Table
332 2). The two final models for non-flooded-communities were also adjusted for the effect of income but
333 no substantial change in the coefficients were noted.

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347 **Table 2: Multivariate logistic regression models for maternal and paternal education and further risk**
 348 **factors associated to stunting in flooded and non-flooded children populations in rural Odisha, India.**

	Flooded (n = 299)		Non-flooded (n = 385)	
	aPR (95% CI)	p-value	aPR (95% CI)	p-value
Models for maternal education				
Maternal education
No schooling	1	..	1	..
Primary school	0.895 (0.558, 1.434)	0.645	1.275 (0.657, 2.475)	0.473
Middle school	0.572 (0.335, 0.978)	0.042	0.927 (0.461, 1.866)	0.832
High school	0.839 (0.468, 1.504)	0.556	0.523 (0.242, 1.127)	0.099
College or more	0.598 (0.228, 1.567)	0.296	0.717 (0.263, 1.954)	0.516
Child age (months)	1.015 (1.003, 1.027)	0.014	1.015 (1.004, 1.026)	0.008
Caste
General	1	..	NA	NA
Other backward	1.654 (0.864, 3.164)	0.130	NA	NA
Scheduled caste	2.007 (1.068, 3.770)	0.031	NA	NA
No caste	1.560 (0.772, 3.151)	0.216	NA	NA
Per head annual income (per 1,000 rupees)	0.955 (0.913, 0.998)	0.043	NA	NA
Models for paternal education				
Paternal education
No schooling	1	..	1	..
Primary school	0.779 (0.478, 1.268)	0.315	0.523 (0.260, 1.052)	0.070
Middle school	0.404 (0.247, 0.661)	<0.001	0.530 (0.278, 1.012)	0.055
High school	0.487 (0.300, 0.624)	0.004	0.563 (0.293, 1.081)	0.085
College or more	0.372 (0.205, 0.791)	0.047	0.360 (0.159, 0.816)	0.015
Child age (months)	1.013 (1.001, 1.024)	0.033	1.012 (1.001, 1.023)	0.035
Child birthweight (per 100 grams)	NA	NA	0.950 (0.905, 0.997)	0.037
Per head annual income (per 1,000 rupees)	0.953 (0.911, 0.996)	0.032	NA	NA

349 NA, results not available if variable not retained in multivariate model.

350 **Correlates of child wasting**

351 Crude associations showed an important role of parental education to reduce child wasting, with a
 352 preponderance of paternal over maternal effects of formal education and the largest effects observed in
 353 non-flooded communities (Fig. 4). An association between means of livelihood (non-agriculture vs
 354 agriculture) and child wasting favoring households not relying on agriculture activities was observed,
 355 with a 47% lower prevalence of child wasting compared to those dedicated to agricultural activities (Fig.
 356 4, detailed results in Table S2).

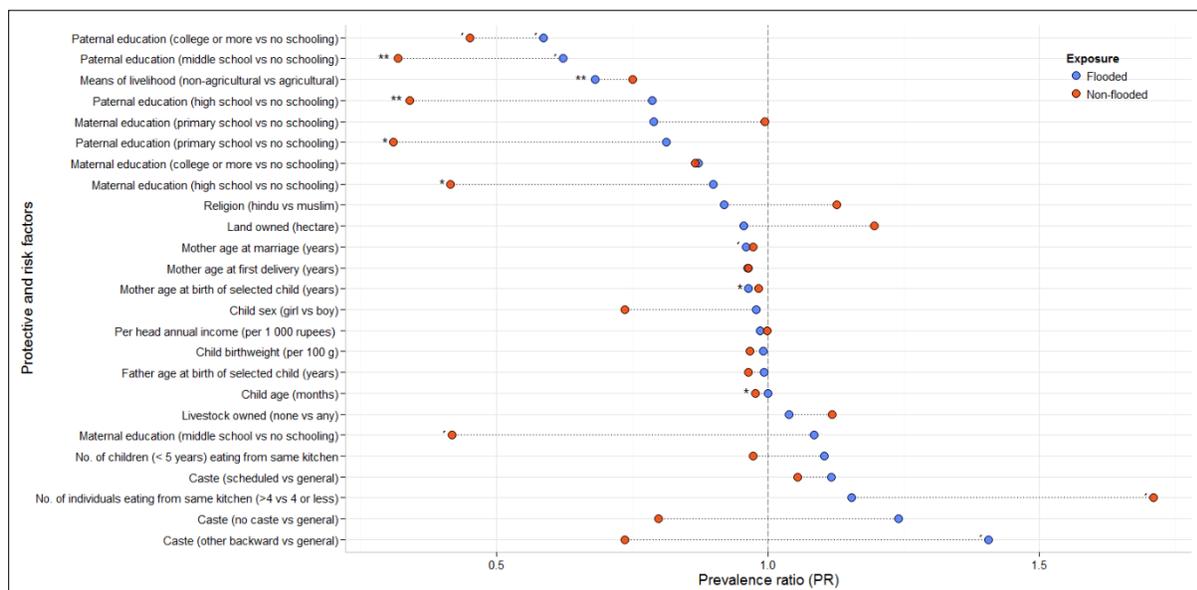
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358 **Multivariate analyses on child wasting**

359 The results for paternal education were different in adjusted models compared to those showed by
 360 bivariate associations (Table 3, Table S2, Fig. 4).

361

362 **Figure 4: Factors associated to child wasting in repeatedly flooded and non-flooded communities of**
 363 **rural coastal Odisha, India.**



364

365 Blue dots show the prevalence ratios in repeatedly flooded communities. Red dots in non-flooded. Dotted lines
 366 show the relative difference in the univariate effects between flooded and unflooded communities. ** $p < 0.01$; *
 367 $p < 0.05$, ' $p < 0.1$.

368

369 In flooded communities, the protective effects of different levels of paternal education were all
 370 significant ($p < 0.001$) except for that of primary education ($p = 0.117$). However, larger effects were
 371 observed in the non-flooded communities surveyed in our study starting from primary education (2.9-
 372 fold lower prevalence, $p = 0.019$), middle (2.9 times lower, $p = 0.012$) or high school (2.7-fold difference,
 373 $p = 0.008$) up to those with university degrees who were associated with 2 times lower prevalence of
 374 wasting among their children, though the association remained insignificant ($p = 0.081$). In flooded
 375 households dedicated to activities other than agriculture, a 50 to 51% lower prevalence of child wasting
 376 was estimated (Table 3).

377 In modeling the effect of maternal education in flooded communities, we found that mother's age at
 378 birth of the studied child was associated with an improved nutritional outcome of their children. Each
 379 additional year was associated with a 3.4% lower prevalence of wasting ($p = 0.005$; Table 3). In flooded
 380 communities only and for the model on maternal education, belonging to other backward castes
 381 showed to be deleterious on child wasting (Table 3). All four final models in this section were also
 382 adjusted for annual per head income but none of those inclusions impacted our results.

383

384 **Table 3: Multivariate logistic regression models for maternal and paternal education and further risk**
 385 **factors associated to wasting in flooded and non-flooded children populations in rural Odisha, India.**

		Flooded ($n = 299$)		Non-flooded ($n = 385$)	
		aPR (95% CI)	p-value	aPR (95% CI)	p-value
Models for maternal education					
Maternal education	
	No schooling	1	..	1	..
	Primary school	0.817 (0.531, 1.257)	0.359	0.819 (0.435, 1.542)	0.536
	Middle school	1.013 (0.730, 1.407)	0.939	0.338 (0.157, 0.728)	0.006
	High school	0.858 (0.598, 1.231)	0.406	0.360 (0.178, 0.726)	0.005
	College or more	0.975 (0.609, 1.563)	0.917	0.637 (0.247, 1.643)	0.351
Mother age at birth of selected child (years)		0.967 (0.944, 0.989)	0.005	NA	NA
Means of livelihood (non-agricultural vs agricultural)		0.660 (0.526, 0.828)	<0.001	NA	NA
Child age (months)		NA	NA	0.975 (0.956, 0.994)	0.009
Caste	
	General	1	..	NA	NA
	Other backward	1.396 (1.016, 1.917)	0.040	NA	NA
	Scheduled caste	1.272 (0.887, 1.826)	0.191	NA	NA
	No caste	1.393 (0.913, 2.130)	0.125	NA	NA
Number of individuals eating from same kitchen	
	2, 4	NA	NA	1	..
	>4	NA	NA	1.870 (1.059, 3.301)	0.032
Models for paternal education					
Paternal education	
	No schooling	1	..	1	..
	Primary school	0.777 (0.567, 1.064)	0.117	0.348 (0.144, 0.840)	0.019
	Middle school	0.571 (0.432, 0.754)	<0.001	0.349 (0.155, 0.787)	0.012
	High school	0.699 (0.579, 0.844)	<0.001	0.366 (0.175, 0.763)	0.008
	College or more	0.521 (0.363, 0.750)	<0.001	0.490 (0.220, 1.090)	0.081
Means of livelihood (non-agricultural vs agricultural)		0.667 (0.526, 0.846)	<0.001	NA	NA
Child age (months)		NA	NA	0.977 (0.958, 0.997)	0.023

386 NA, results not available if variable not retained in multivariate model.

387

388 Discussion

389 This study identifies education as one key investment in reducing the health impacts of extreme events
390 under climate change risks while promoting sustainable human development. The most striking finding
391 is that paternal, and not maternal education, appeared to be the strongest predictor of lower child
392 stunting and wasting in these communities. In non-flood settings, large samples from Indonesia and
393 Bangladesh studied the effect of parental education on child stunting, and comparable positive effects
394 were found for maternal and paternal education (Semba et al., 2008). The effect of paternal education
395 on child wasting and stunting was found to be comparable to that of mothers' also in India and Vietnam
396 (Moestue & Huttly, 2008). In flood settings, however, these effects were not as clear as in available
397 studies. Hossain & Kolsteren (2003) found no effect of mother education on child wasting recovery four
398 months after large floods in Bangladesh. Also in Bangladesh, Del Ninno & Lundberg (2005) found no
399 effect of maternal education on child growth between three waves of anthropometric data collected
400 within 15 months after the floods. None of these studies considered the effect of paternal education.

401

402 In our specific setting, we hypothesize that the observed differences in effects favoring paternal
403 education could be partly explained by the low economic independence of women in our sample: nearly
404 98% of mothers in both exposure groups were housewives and we assumed had no revenues. We
405 acknowledge that this topic is complex and more studies on this specific interaction should be
406 conducted on datasets having larger samples than ours. However, in India evidence exist pointing to low
407 maternal economic and physical autonomy as a contributor to child stunting (Shroff et al., 2009; Imai et
408 al., 2014).

409

410 A second observation is that the effect of paternal education on child stunting were larger and
411 statistically significant in flooded communities while the same estimated effects in non-flooded
412 communities were of smaller magnitude. It looks as if the already positive effect of education might be
413 boosted in a post-flood situation. Several studies have shown that education might substantially
414 improve coping after disasters, through avoiding income loss after a disaster (Garbero & Muttarak,
415 2013), diversification of economic activites (van der Land & Hummel, 2013) or choosing sustainable
416 mechanisms for coping prevail over short-term views (Wamsler, Brink & Rantala, 2012; Helgeson, Dietz
417 & Hochrainer-Stigler, 2013). In a disaster situation, decisions on allocation of the constraint resources
418 available can be crucial. More educated parents might be able to invest in coping strategies with longer

419 term benefits, have more savings or simply work in professions which are less affected by floods. Our
420 results on child wasting showing that non-agricultural livelihoods withstood better the shock caused by
421 flooding reinforce this point. Similar results on a higher likelihood of post-disaster migration in low
422 educated farmers are consistent with our views (van der Land & Hummel 2013). Crop farming, livestock
423 rearing and fishing were the most affected activities according to reports from the Government of
424 Odisha: around half the production of the 0.44 Million Kharif crops flooded were lost, more than 2.3
425 Million livestock flood-affected, and above 6,300 fishing boats with their nets and other equipment
426 damaged (Government of Orissa, 2008). Our findings are consistent with this pattern of affectedness
427 and reveal their mid-term consequences on child nutritional health. Young children of agricultural
428 livelihoods were the most impacted regarding the prevalence of wasting one year after the floods,
429 which was around 50% higher than in other livelihoods. Importantly, no effect was observed for similar
430 analyses conducted in non-flooded communities, whether univariate or multivariate. This reinforces the
431 hypothesis that crop destruction and overall food insecurity is a very likely pathway to child
432 undernutrition among the flooded populations (Leaning & Guha-Sapir 2013). There is much to do in
433 terms of mitigation strategies to reduce the initial impact of floods on the livelihoods of these vulnerable
434 populations. At the same time, there is an urgent need to reform and expand the relief response, which
435 was not commensurate to the magnitude and duration of the problem. The basic needs of the affected
436 were only covered by the government on the first 15 days following the floods. We need to ensure that
437 these livelihoods are satisfied on around a year after the floods and that we target most affected
438 livelihoods. Crops take months to be harvested since they are planted. Animal stocks need time to
439 recover fully, same for repairing boats or buying new ones. Without more dedicated resources, the risk
440 is on perpetuating poverty cycles.

441

442 Another consistent result was the positive effect of income on child nutrition in flood-affected
443 communities. For example, an increase in 5,000 rupees per capita in yearly income within households
444 should be associated with 25% lower prevalence of child stunting. Sustainable livelihood economic
445 development is then a plausible strategy to reduce the nutritional burden of floods according to our
446 data. In contrast, and again in flood-affected livelihoods only, lower caste was associated with worst
447 nutritional status for both wasting and stunting variables, suggesting that caste is still in association with
448 lower opportunities. Our study findings show that these social determinants of different impacts get
449 visible in an extreme situation such as a disaster but we were unable to offer an explanation on the
450 mechanism explaining this pattern in the data. Qualitative research might plausibly be very useful to

451 investigate the reasons further. Finally, an important result was that mothers giving birth later had a
452 lower likelihood of having a wasted children. Our model suggest that by delaying 5 years the birth of a
453 given child, the associated prevalence of child wasting in case of floods would be on average 17% lower.
454 It is noteworthy to point that this result was independent from the effect of mother education, adjusted
455 for in this model.

456

457 Taking all the above together, education promotion in rural areas, especially among women, would be
458 extremely efficient. In our study area about 30% of the mothers did not completed middle school
459 education, which is mandatory in India. There is also an education gap with men in our study area, for
460 whom this percentage was less than 20%, a 10% difference with women. It is worth mentioning that
461 looking across all models in our study, primary education (up to 11 years of age) was not a significant
462 contributor to better child nutrition in 7 out of 8 models reported. However, providing 3 additional years
463 of schooling (up to 14 years) had a significant positive impact on child nutrition. Moreover, the
464 magnitude of effects reported for middle, high and college education were quite comparable, suggesting
465 that at least in these rural communities, middle education is sufficient. Our results are in agreement
466 with the Indian policy on education. More efforts are then needed to ensure that every Indian children
467 receives the 8 years of compulsory formal education. By promoting more education and adequate family
468 planning (Van Braeckel et al., 2012), women might naturally delay motherhood, with potential benefits
469 to children's nutritional health. To ensure full benefits, we hypothesized that mother education might
470 lead to increased employment rates amongst women, as this would also contribute to additional higher
471 household income (with associated benefits shown by our study). Notably, it has been demonstrated
472 that income allocation by mothers is more efficient to improve family's health, including child nutritional
473 status, compared to fathers (Thomas, 1990).

474

475 **Strengths and limitations of the study**

476 There a few methodological considerations, strengths and limitations that need to be addressed here.
477 Firstly, all the reported results were adjusted, as an additional robustness check, by annual per capita
478 household income. As such all effects reported were independent of income, and thus our results can be
479 methodologically comparable to those recently published (Muttarak & Lutz, 2014). Second, we often
480 found in the literature maternal and paternal education jointly modelled (Semba et al., 2008). However,
481 we found that these two variables were highly correlated in our data and enough evidence using
482 variance inflation factor analysis, which supported that they should be modelled separately. Future

483 studies should be aware of this. Third, father education was highly correlated to caste and mother's age
484 at birth of a child, and this is the reason why models modeling father education do not include these
485 variables. Strengths of this study include carefully consideration of income, as well as having analyzed
486 data on a non-flooded group which allowed us to observe that some effects such as mother's age at
487 birth, income or caste become important determinants only in the extreme circumstances of flooding.
488 Additional positive features of this study were its population-based design, the consideration of
489 recurrently flooded communities only, coupled with the analysis of mid-term and more long-term child
490 undernutrition. On the minus side, our study was not based on a very large sample size, which could
491 have been missed the detection of smaller effects and provided tighter confidence intervals. Similarly a
492 larger sample size might have allowed us to analyze most severe forms of malnutrition, such as severe
493 wasting and stunting. In the aftermath of flooding in Bangladesh, Choudhury & Bhuiya (1993) have
494 shown maternal education to play a role in increasing severe forms of underweight. Also, as the findings
495 presented are based on cross-sectional data, care is needed to not attribute causal relationships to the
496 associations showcased here. Importantly, it is difficult to generalize our results to other settings such as
497 urban slums, as vulnerabilities might be different there. However, we are confident that there might be
498 similarities between our results and other rural areas around the world in which subsistence farming is
499 the norm. With its limitations, this is one of the first studies to look at child undernutrition in flood
500 settings and study potential preventive factors in short and long-term. More studies are needed to
501 solidify the evidence base for global action.

502

503 **Conclusions**

504 We recommend strengthening the response to floods and targeting most vulnerable agricultural-
505 dependent livelihoods, which showed an associated 50% higher prevalence of child wasting. Policies for
506 relief should be reviewed to ensure a longer period of support and proper take up by development
507 programs, which should give enough time and support to these households relying on agricultural
508 activities to fully recover.

509 Education promotion in general and the strict fulfillment of schooling until the compulsory 14 years of
510 age is strongly recommended to protect child's health in the face of future flooding. Policies effectively
511 helping sustainable livelihood economic development and diversification of economic activities, delayed
512 motherhood, which certainly include education promotion, are vividly praised.

513 Given that global climate changes are set to increase flooding both in frequency and severity which will
514 further aggravate this situation, we recommend urgent action to be taken (UNICEF, 2014).

515

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