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Sri Lankan Households a Decade after the Indian Ocean Tsunami

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Abstract

We estimate the causal effect of the Indian Ocean tsunami in Sri Lanka on household income and consumption eight years after the event, using a quasi-experimental method. A strong association between area-wide tsunami disaster shock and increases in household income and consumption in the long-term emerged from our empirical investigation. Deviating from the common observation on short-term impacts, these results are suggestive of an optimistic potential for some long-lasting potentially successful recovery scenarios. Still, Sri Lanka received a very large amount of external transfers post-tsunami, much larger than is typical for disaster events and one which may not be replicable in other cases. Our findings suggest a more nuanced picture with respect to household consumption impacts. We observe a reduction of food consumption and only find an increase in non-food consumption. The increase in non-food consumption is much smaller than the observed increase in income. We also find that households in high-income regions experienced much better recovery from the disaster.

JEL Q54, R11

Keywords: Sri Lanka, Tsunami, disaster, household survey, long-run impact

1. Introduction

The 2004 Indian Ocean earthquake elevated the ocean floor by at least three meters generating a very powerful tsunami that killed 226,000 and displaced more than 2 million people in about a dozen countries. A large number of casualties and property damage associated with high-intensity disasters like tsunamis obviously have short term impacts by reducing economic activity, but relatively less is known about the long-term economic losses.¹ Here, we are interested in households' vulnerability to the long-term impacts of the disaster; specifically, we aim to identify the impact of the tsunami on Sri Lankan households. In this case, the event was completely unexpected and thus undoubtedly exogenous, but its impact is not. Households' socio-economic characteristics, their exposure and vulnerability to the hazard itself, their resilience and access to tools and mechanisms to manage the disaster's aftermath, their preferences, their decisions when the circumstances around them change, and their choices during the post-event reconstruction all eventually determined the disaster's long-term consequences (Hallegatte *et al*, 2014; Hallegatte and Przyluski, 2010; McCarthy and Smith, 2009; Mechler, Bayer, & Peppiatt, 2006; World Bank, 2013).

Sri Lanka, an island country in the Indian Ocean, is densely populated, with a 2015 population of 20.7 million (Central Bank, 2015). The population is 74.9% Sinhalese, 11.2% Sri Lankan Tamil, 4.1% Indian Tamil, and 9.3% Sri Lankan Moor. 20.5%, 26.3% and 45% of population work in the agriculture, industry and service sectors, respectively (Department of Census and Statistics, 2012). Administratively, Sri Lanka is divided into 9 Provinces, and these are divided further into 25 Districts. Each District is divided into Divisional Secretariats (DS). Each DS consists of several Grama Niladhari (GN) Divisions, the lowest administrative unit. Currently, there are 324 DS Divisions and 14,009 GN Divisions in the country (Department of Census and Statistics, 2015). In the World Bank's classification, Sri Lanka is a lower-middle income country, though social indicators suggest a higher standard of living when compared to other countries in South Asia. From the early 1980s, the country was impacted by an armed uprising; seven districts out of 25 were intensely affected by the

¹ Economic losses – sometime referred to as induced or indirect losses – are considered in terms of changes to flows of goods, services and business or even government revenues. These can emerge in various spatial scales, in different economic sectors, and can change rapidly or evolve slowly over longer periods of time (Rose, 2009; Hallegate & Przyluski, 2010; Cochrane, 2004).

conflict with the Liberation Tigers of Tamil Eelam (LTTE). The armed conflict ended in 2010. Prior to the 2004 tsunami, approximately 25 percent of the population lived in the coastal region, while 70 percent of tourist hotels and 62 percent of industrial units and almost all fisheries were also located there.

The tsunami reached 13 Districts out of 14 coastal districts, the death toll reached almost 35,500, and the disaster affected more than one million people. Infrastructure was severely damaged and the overall economic losses totalled USD 1.5 billion, approximately 5% of the country's GDP (Department of Census and Statistics, 2005) – a figure proportionally similar to the damage experienced by Japan in the 2011 earthquake/tsunami disaster. Tourism and fisheries were the two sectors most seriously affected. 150,000 people lost their main income source, 50% of them in the fisheries sector. Before 2004, tourism was one of the top income earners in Sri Lanka with the gross earnings amounting to USD 416 million during 2004 from peak tourist arrivals of 566,200.

A decade after this catastrophic disaster, this paper evaluates the long-term household-level consequences of this event in Sri Lanka. The average effect of the tsunami on household income and consumption residign in the seven affected districts is examined in quasi-experimental (diff-in-diff) setting using cross sectional household data. Since the intensity of damages—death, displacement and property damag caused by tsunami—vary across districts, the analysis also sheds light on the spatial dynamics of the disaster recovery.

2. Related Literature on Disaster Losses

The literature on the losses associated with disasters examines consequences at both the micro- and macro-level. At the macro level, the available empirical evidence does not reach any consensus view, but much of the variance in results is a function of the different foci in each paper—the spatial scale, type of disaster, time horizon, etc. Two recent papers provide regression meta-analysis of this literature (Klomp and Valckx, 2014; Lazzaroni and van Bergeijk, 2014). This literature, however, provides few details about the causal channels that lead from the trigger event itself to the macroeconomic aggregate impacts that are identified (be they on aggregate employment or unemployment, production, or fiscal and trade deficits). The micro-econometric literature is better placed to provide some impetus for the important endeavour of unveiling these causal chains.

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In the short term, households manage the financial risk they are exposed to because of natural hazards through several risk transfer and risk management tools. Explicit insurance contracts and credit are the primary market-based arrangements available to manage financial risks, but implicit insurance arrangements (from kin, from governments, or from the international community) also play a role in transferring risk away from affected households.² Similarly for firms, the availability of resources for post-disaster appears to play a central role in post-disaster recovery.³

The literature, however, suggests that low-income households are credit constrained and their ability to self-insure is limited. This results in variation in expenditures associated with realized risks (Mogues, 2011; Baez and Mason, 2008; Carter et al., 2007; Baez, 2006; Jansen and Carter, 2013). Because of inadequate risk transfer, poor households use costly ways to confront risk, ways that may have long-lasting adverse consequences. These consequences can be especially severe for the poorest households (Dercon and Christiaensen, 2011; Dercon et al., 2007).

Households accumulate saving to cushion the welfare loss. Evidence suggests that household saving is higher in countries with greater risk exposure and frequency of intense disasters (Aizenman and Noy, 2015; Skidmore, 2001). When productive assets are lost due to disaster, households may need to reduce consumption to protect their remaining assets (Barrett *et al.*, 2007; Kazianga and Udry, 2006; Little et al., 2006). Either way, they can fall into "poverty traps" from which recovery is difficult without external assistance. Carter and Barrett (2006), for example, provide evidence for such 'asset poverty traps' among pastoralists in northern Kenya.

Reductions in household consumption can have significant adverse consequences to wellbeing, including even reduced height and body mass index for children experiencing these events (Alderman, Haddinot and Kinsey, 2006; Haddinot, 2006). In the cross-country analysis, Cuaresma (2009) revealed a strong negative correlation between disasters and secondary school enrolment. Further, Bhalotra, Sanhueza and Wu (2011) revealed evidence

² Sawada and Shimizutani (2007), for example, observe that the households with collateral and free from binding borrowing constraints were better able to cope with income loss following the 1995 Kobe earthquake. With limited access to explicit risk transfer financial instruments, the poor respond to shocks by pooling risk through social mechanisms such as credit cooperatives and mutual assistance pools (Baez, 2006; Little *et al*, 2006; World Bank, 2013).

³ Following the tsunami, Sri Lankan firms with access to loans or grants performed better (De Mel et al., 2011).

from Chile and Cruso and Miller (2015) from Peru that those exposed to massive earthquakes (in 1960 and 1970, respectively) still show a relative deficit in educational and labour outcomes many decades later. More worrying, the latter project also shows that children whose mothers were affected at birth also manage to complete fewer years of schooling.

When formal insurance or other safety nets are not available, the poor diversify their income sources to lower risk exposure. Typically, this may be done either through labour sector diversification locally or through emigration. Lynham, Noy and Page (2012) observe people migrating away from 1960 Tsunami in Hawaii, for example, but Brata, Groot and Rietveld (2014) identify only a temporary change in population in response to the 2004 tsunami and the Nias earthquake in 2005 in Indonesia.⁴ And indeed, research shows that remittances play an important role in the households reactions to disasters (Deshingkar and Aheeyarse 2006; Harvey and Group, 2007; Mohapatra et al., 2009; Savage et al., 2007). Suleri and Savage, 2006) observe that households receiving remittances are able to recover earlier from the 2005 Afghanistan earthquake. Gröger and Zylberberg (2015) observe rural household members in Vietnam migrating to urban areas in the aftermath of the 2009 catastrophic typhoon, and households with settled migrants receiving more remittances in the typhoon's aftermath. Le de, Guillard and Friesen (2014) observe an unequal access to remittances, with the poor having less access and therefore struggling more to recover from the 2012 cyclone in Samoa.

3. The Aftermath of the 2004 Tsunami in Sri Lanka

In the immediate aftermath of the 2004 catastrophe, the government established the Centre for National Operations (CNO) with special powers to oversee the coordination of agencies involved in rescue and relief. Three task-forces; (1) The Task Force for Rescue and relief (TAFRER); (2) The task Force for Logistics, Law and Order (TAFLOL) and (3) the Task Force for Rebuilding the Nation (TAFREN) were set up to address specific aspects of the relief effort (Government of Sri Lanka, 2005).

⁴ This temporary effect may be due to the strong positive influence of the disaster on the peace deal between the Government of Indonesia and the Free Aceh movement in 2005. Halliday (2012) reveals people migrating away after an earthquake in El Salvador.

The Ministry of Finance set up a coordination system with national agencies, donor agencies, and international non-governmental organisations, to support 16 primary interventions or activities identified as priorities. The coordination and facilitation roles implementing reconstruction were eventually transferred to a newly formed Reconstruction and Development Agency (RADA). Reconstruction activities were coordinated at District and Division level by the Government Agent at each level (Post Tsunami Reconstruction and Recovery, Joint Report of GOSL and Development Partners, 2005).

Initially, the government declared a development-free buffer zone along the coast restricting reconstruction, but later in 2005 this policy was abandoned due to strong opposition from the public (Ingram et al., 2006). New houses were provided based on proven ownership claims for destroyed houses or, alternatively for those unable to claim land ownership, through a donor-driven program. For completely damaged houses, the government provided land and cash grants and additional donor assistance to rebuild houses. For partially damaged houses, cash grants were provided. Financing for tsunami reconstruction in Sri Lanka relied mainly on foreign grants. A total of 3.9 billion USD was committed; of which 2.0 billion came from foreign grants and 0.9 billion each from foreign loans (mostly multilateral) and the government of Sri Lanka. Almost 1.4 billion were spent on reconstruction already by the end of 2006 – two years after the tsunami (Jayasuriya and McCawley, 2010). Later, the tsunami reconstruction was managed within the mainstream development program. Despite very large flows of relief and reconstruction funds, reconstruction in Sri Lanka faced problems of coordination, escalation of construction costs, and other similarly common difficulties in post-disaster recoveries (Athukorala, 2012; Jayasuriya and McCawley, 2010; Munasinghe *et al.*, 2007).

4. Empirical Analysis of Household Survey

This study isolates the causal effect of the 2004 tsunami in a quasi-experimental analysis using pooled cross-sections of information from household surveys. We include observations of 84,303 households obtained in five Household Income and Expenditure Survey waves conducted in 1995, 2002, 2006, 2009 and 2012 by the Sri Lanka Department of Census and Statistics. The longitudinal nature and richness of household level information

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covering nearly a decade before and a decade after the tsunami make these surveys well suited to investigate the tsunami's long-term impact at the household level.

The surveys collected detailed information on household and individual demographics, employment, consumption, income and other related data corresponding to the preceeding calendar year. The demographic characteristics include variables such as sex, age, marital status, ethnicity, religion, level of education, employment status, house ownership; household food consumption (including in-kind consumption), household non-food expenditure (household expenditure on housing, education, health, transport, recreation, household expenditure on durables and on insurance and savings; the income of the household (from paid employments, non-agricultural and agricultural activities); cash receipts (pension, disability and relief, food stamp, property rent, dividends); and remittances (from abroad and from within the country).

Due to the civil conflict that ended in 2009, the data for eight districts out of total 25 districts are not available before the end of the conflict. Thus, survey data for 17 Districts from 1995 to 2013 were used in the analysis. Details about the data available in each wave of the household survey and their definitions are available in Appendix Table 1. The summary statistics are available in Table 1. The composition of ethnic and religious groups, and composition of households according to their location (urban, rural, estate) in the survey sample are all comparable approximately to the national figures available from the census, and the sample was collected, in principle, to be nationally representative.

To isolate the causal effect of the disaster precisely, randomised experimentation is ideal, and a (time-series) panel of households would be preferable. Since randomised experimentation is impossible and panel data are unavailable, we instead use crosssectional data pre- and post-disaster in a quasi-experimental analysis. The identification strategy relies on the standard common trend assumption in difference-in-difference estimations, and we show that this assumption appears appropriate in this case.

To identify the treatment group, we exploit the spatial variation of tsunami damage using reported deaths, displacements, homelessness and infrastructure damage (replacement cost) due to tsunami across districts. Our aim is to reveal the average causal effect from the area-wide tsunami shock. Out of 25 Districts, 13 Districts were affected by the tsunami (Appendix Table 2) and reported mortality rates, the number of people displaced, the

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number of people that became homeless, and the damage to public infrastructure. Out of the 13 affected districts, the HH survey covered only seven districts both before and after the tsunami, as the other six districts were directly involved in the conflict. Our treatment group comprises the households in the affected seven districts for which we have data.

In the absence of household level reported tsunami damage, we first estimate the average effect for all households in the affected districts. However, the exposure of the districts and the households in those districts to the tsunami varied considerably. The regions on the eastern and southern coasts were directly exposed to the tsunami waves coming from east-south-east and higher damages were reported in those districts facing in that direction. We also exploit the damage information available across districts to examine how recovery vary when accounting for the level of damage. The district level reported deaths, displacements, homelessness, and the value of public infrastructure damaged are used as the proxies for exposure to the tsunami to identify the relative intensity of the shock across districts. We define a shock index by dividing the impact figures by the total number of households in the respective districts. We thus estimate a second model using a different treatment group: households in the three most highly affected regions (for which we have data): Galle, Matara, and Hambantota.

The previous literature argued that the long-term impacts vary with the vulnerability of the affected households (duPont and Noy, 2016). To examine the impact of vulnerability on recovery, we estimate two other models. One (Model 3) is estimated for a different treatment group excluding the two richest districts (Colombo and Gampaha) and the other (Model 4) for a treatment group including only the two richest districts excluded in Model 3. To further examine the sensitivity of our results, we investigate whether the causal effect depends on the type of damage. We estimate models using three disaster index variables: displacement, homelessness, and deaths.⁵

Our identification uses difference-in-difference (DID) method. The standard empirical specification is:

$$Y_{idt} = \beta_1 + \beta_2 Post_i + \beta_3 T_d + \beta_4 Post_i T_d + U_{idt}$$
⁽¹⁾

⁵ We exclude the infrastructure index, as it is highly correlated with displacement and homelessness.

Modifying the standard model, our empirical specification takes the following.

$$Y_{idt} = \beta_1 + \beta_2 Post_i T_d + \beta_3 \,\delta_t + \beta_4 X_{idt} + \beta_5 \gamma_d + U_{idt} \tag{2}$$

 Y_{idt} is the outcomes of interest- household monthly consumption and household monthly income- and; the unit observed is household *i*, in district *d* and time *t*. T_d is the treatment dummy defining membership in treatment cross section (affected=1, not affected=0) and U_{idt} are the unobserved affects. *Post*_i is a dummy variable to distinguish sample by pre and post treatment (post-tsunami=1, pre-tsunami=0). β_2 is the treatment effect of interest.

Other than the common trend assumption, treatment exogeneity is a key for unbiased estimation. Treatment effects are naturally heterogeneous across households depending on household characteristics and the community level characteristics that they live in. By adequately controlling for such heterogeneity, the unobserved variation ideally should account for the average causal effect. Household demographic and socio-economic covariates X_{idt} are incorporated into the model to control for household heterogeneity. Gender (male=1, female=0), age (years), years of education, head of household employment status (binary variables for employed, unemployed and employed in paid occupation), household's ethnic group (binary indicators), and house ownership (binary) are all used in the model estimation. Pre shock differences of the treatment and control group that could be due to different time trends are controlled by year fixed effects (δ_t). The geographic differences are controlled by district fixed effects (γ_d); these therefore control for the *Post_i* and T_d in eq. 1, respectively. Finally, the other differential effects and the mean of the error term U_{idt} are assumed zero. We estimate model 1-4 using this specification.

In a different set of regressions, instead of dummy variable for treatment as in the previous model (eq. 2), we introduce treatment indices (T_d^{ind}) into the model. Separate treatment index variables are used for death, displacement and homelessness.

$$Y_{idt} = \beta_1 + \beta_2 T_d^{ina} + \delta_t + X_{idt} + \gamma_d + U_{idt}$$
(3)

The models are estimated using OLS. Since heteroscedasticity and serial correlation may be present in the data, we use robust and district clustered standard errors for inferences.

5. Results

We rely on several waves of the national household survey conducted by the Sri Lankan government. Table 1 provides some descriptive statistics of this data. In total, 64% of the observations are of households in the years following the tsunami, while 51% of the households live in the districts affected by the tsunami – 33% are of households in the tsunami areas observed in the aftermath. In some of the estimations, we include covariates that are also correlated with income or consumption. The details about these, and the regression results for the two dependent variables, income, and consumption are provided in Appendix Tables 6 and 7, respectively.

5.1 Income

Table 2 presents our benchmark results for household income. We observe the average treatment (tsunami) effect (ATE) on household income by examining the coefficient (β_2 in equation 2) for the interaction of the treatment indicator (tsunami-affected district), and post-tsunami year. Columns I-V present several estimations based on different assumptions regarding the standard errors and the inclusion of district fixed effects and district year trends. Column II controls for household heterogeneity; and district fixed-effects are included in column III. The inclusion of district fixed-effects reduces the size of the estimated ATE coefficients, suggesting that some of the observed ATE in column II is an over estimation associated with the higher income in the coastal districts. Column IV includes district-year interactions, but these over-fit the model, and remove much of the impact we associate with the tsunami itself (as our identification of treatment relies on households residing in a tsunami-affected district). Column V is our preferred specification as it includes district clustered standard errors, though results there are not dramatically different from the other specifications.⁶

To further examine the validity of our findings, we estimated the log-normal model using the log transformation of the dependent variables (income and consumption). The kernel density plot of residuals with normal density overlaid on the plot (in the Appendix 14) show

⁶ The one exception is column IV that includes district-specfic time interactions. Since we only have a few observations post disaster, and our identification relies on cross-district differences, the inclusion of this term incorporates much of the identified tsunami impact.

slightly skewed error distribution. The inter-quartile range test in appendix 5 shows the presence of 3-4% outliers in the non-log model. Appendix 15 shows that the log-normal model rectifies this issue (these estimates are shown in table 6 column I).⁷

Table 2, column V (with robust and clustered standard errors) shows the causal effect (ATE) of the tsunami impact on household income. The results indicate a positive average effect on household income in affected districts. Income increased due to the tsunami in the year 2006 by Rs. 7067 (47 US\$), slightly reduced in the year 2009 (Rs. 5872, 39.14 US\$) and increased significantly in the year 2012 by Rs. 15003 (100 US\$). The effect is clearly seen in the normalised income (the predicted residuals regressing income against household covariates and district fixed effects and collapsed by mean and survey years for both control and treatment groups) in figure 1.

In figure 1, the two parallel lines for treatment and control group before tsunami clearly indicates the absence of different income trend between treatment and control group pretsunami and appears to confirm that our common-trend assumption is valid. The divergence of the two lines in the figure post-tsunami starkly demonstrates our finding: households residing in tsunami affected districts experienced an increase in income in post-tsunami years.

The same model as in column V of table 1, when estimated for the treatment group of the most-affected regions (column VI), reveals a much more moderate income increase shortly after the tsunami - Rs. 2045 in 2006. This effect, compared to an increase of more than three times as much for all affected regions, suggests that the districts that benefited the most were those that were perceived as tsunami affected (and therefore received assistance) but were less heavily damaged. In the longer-term, in the 2009 and 2012 observations, the estimated impact of the tsunami on household income for heavily affected regions becomes negative (but statistically insignificant); compared to large and

⁷ Log tranformations create other issues. Income data has considerable zero and negative observations and log transformation drops those observations (column I of table 6). Another model is estimated using log-modulus transformation (John and Draper, 1980) to preserve the sign and zero observations (column II of table 6). Nevertheless, the log transformation drastically changes the distribution of of the dependent variable. Extreme values are obviously possible in our data since we use time series data with 15 years time gap. This is evident in the scatter plots of log and non-log income and consumption in Appendix 17. Given that the change in distribution can significantly affect the average income and consumption and ATE in each year, which our main variable of interest in identifying the causal effect.

significant positive impact on all affected districts. Again, this suggests the role that assistance played in recovery, with less affected districts benefiting the most.

Column VII estimated the same model but excluded the two richest regions (one of them is the capital, Colombo). Maybe not surprisingly, in that case as well, we see a much more moderate increase in the positive impact of the tsunami recovery on household income. Again, this difference is most likely associated with the increased access to post-tsunami assistance for the two richest (and urban) regions that were now excluded. Confirmation of that is found in column VIII, where the model estimated focuses on the ATE for the richest two regions. The identified ATE is 3-5 times as large for the richest as for the poorest affected regions (column VII).

In table 3, we describe the ATE on the various sources of household income. In particular, we separately examine the ATE for paid income, agricultural income, remittances, transfers, dividends, and rents (and other income). We estimate these using the same specification as in table 2 column V – our preferred specification. In all the different income sources, the identified ATE for the three post-tsunami years is positive, though the estimated coefficient is often not statistically significant. The increase in monthly income from non-agricultural activities is the largest, and also the most statistically significant: in 2006 by Rs. 4559 (31.4 USD), in 2009 by Rs. 3911 (27.0 USD) and 2012 by Rs. 6246 (43.1 USD). As in the table 2, we observe more moderate increases in income in the immediate aftermath, and a larger impact observed in the 2012 survey. In most of the households' income streams, the increase in the 2006 and 2009 is not statistically different from zero (no treatment effect) but it is statistically significant and positive for 2012. We also observe that not only do we identify the biggest ATE for non-agricultural income, the model's goodness-of-fit is best for that income source. Our ability to determine the cross-household differences in other income sources is much more limited, and the observed tsunami impact (ATE) is in any case much smaller for these. However, the impacts described in table 3 are the ATE averaged over all affected regions, and the districts were exposed to the tsunami in varying degrees. Results by district (Appendix 8) show that household in all affected regions except Puttlam experienced increase of income and the positive impacts persisted into the 2012 survey. Figure 2 illustrates the variation of impact among the affected districts.⁸ We note that the

⁸ Conversion of Sri Lankan Rupees to USD is based on the August 2016 exchange rate of 1USD=145 Rs.

2006 survey was conducted less than two years after the tsunami disaster and 2009 marked the most intense fighting between the government and the LTTE, just before the conflict ended in 2010. These observations suggest that our finding that income increased more in 2012 may be quite intuitive. When examining districts separately, we find that Colombo (CMB) and Gampaha(GMP) have significantly higher incomes; while most other districts either have lower income post-tsunami or no observed change.

5.2 Consumption

Tables 4 and 5 include the detailed estimation of the impact of the tsunami on household consumption and its components, respectively. As discussed above with respect to income, our preferred specification in column V of table 4 includes the household covariates and the district fixed-effects, and is estimated with district clustered standard errors. In column V, the estimated ATE for consumption is positive in the three years post tsunami for which we have a measurement, but is statistically significant only for the immediate aftermath, and for the longer term (in 2012). As was the case for income, the impact in the longer term (2012) is larger than the impact in the immediate aftermath. Importantly, however, the positive observed increase in consumption is much smaller than the increase we identified in household income. It appears that the increase in income does not translate very well into increases in welfare (consumption).

As we observed that the increase in income was much smaller for the most heavily damaged districts (column VI in table 2), we observe that the same districts did not experience an increase in income but rather appeared to have experienced a decrease in consumption (though the decrease is not statistically significant – see column VI in table 4). The difference between the two richest regions and the rest of the affected regions is less stark for consumption than it was for income, but we still observe a bigger increase in consumption in the richest districts than we observe for the other affected districts in the longer term (columns VII and VIII in table 4).

The normalised income is presented in figure 3 (predicted residuals when regressing consumption against household covariates and district fixed effects are collapsed by mean and survey years for both control and treatment groups). The figure depicts the parallel

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consumption trends before the 2004 tsunami, and the steeper increase in consumption observed for the treated (impacted) districts. At the district level (results in the appendix 9), we find temporary gains for Matara and Gall; persistent gains for Kalutara, Gampaha, Hambantota, Colombo and Puttlam. Figure 4 illustrates the variation of impact across the impacted districts.

The surveys we use include details questions about expenditure patterns, so we use the same diff-and-diff methodology to identify ATE for each component of expenditure (table 5). When distinguishing between food and non-food expenditures, we find that, worryingly, food consumption actually decreased in the affected districts as a consequence of the tsunami, while non-food consumption increased (and that accounts for the overall increase identified in table 4). Especially notable is the decrease in food consumption observed in 2009 (column I table 5), after the inflow of external assistance has ebbed five years after the tsunami. When we examine the separate components of non-food consumption we find, overall, increases in all of the items, with the largest identified increase in, not surprisingly, housing – an increase that is probably related to the cost of reconstruction and the increase in construction costs that is typical of most disaster recoveries. ⁹

6. Robustness Analyses

In tables 2-5, we found that the tsunami and the ensuing inflow of assistance significantly increase household income, but that consumption gains were much more limited. What explains these lack of 'pass-through' from income to consumption is not immediately obvious. Some expenditure and income components are missing in our data – most relevant seem investment in property. Due to inconsistent survey questions across surveys, all income and expenditure components are not available in our complete time series dataset. Therefore, the average household consumption is approximately Rs. 3000 higher than the average household income reported in our summary statistics (Table 1). In order to provide more information on the 'missing income,' we exclude the year 2002 survey (the survey wave with most missing components) and repeated the analysis while including ad-hoc income. These regressions are available from the authors, but they clearly, and similarly,

⁹ Non-consumption expenditures (column X) include the expenses on household security and payment of liabilities (savings, payment of tax, thrift societies, debt payment etc.).

show a positive impact on income and consumption. Still, we again observe income increases to be substantially bigger than the identified consumption increases.

In table 6, we reestimate the key equations using log income (column I and II) and log consumption (column III). Results are similar to the results we previously obtained in the benchmark estimations in column V in table 2 and column V in table 4 for income and consumption, respectively. We see an increase in consumption that is significantly higher than the observed increase in consumption (which in even negative in column III).

Subsequently, in table 7, we aim to account for the relative intensity of tsunami damages in each district. The tsunami's impact is measured separately by deaths, displacements, and homelessness. Comparing the regressions modelling consumption (columns I-III) and those modelling income (columns IV-VI) reveal, again, that there is little pass-through from increases in post-tsunami income to post-tsunami consumption. What we observe is that the intensity of damages is correlated with more income (probably because there are higher assistance flows/aid) but with less consumption.

In addition, we estimated the diff-in-diff model with "affected households and pre-disaster 2002-year interaction variable" as a type of placebo test. When 'pretending' that the tsunami occurred in 2002, we verify that the income and consumption of treatment and control group were not significantly different before the disaster event. We find ((results available upon request) non-significant ATE for both outcome variables (income and consumption). The treatment and control groups are not different in their income and consumption before the tsunami; providing additional evidence to support the validity of our results.

The causal effect we find could also be biased if the tsunami caused affected households to migrate, thus changing the sample dramatically. Given the reconstruction policy process in Sri Lanka whereby land was allocated nearby to the areas of destruction, we argue that migration after the disaster should have been negligible. In order to verify this claim, we examined the net migration of affected districts between 1982 and 2012 (Appendix 20). We indeed find that people migrated away from the affected regions of Matara, Galle, and Hambantota districts and migrated into tsunami affected Colombo, Gampaha, Kaluthara and Puttlam districts. Considering the overall migration pattern in Sri Lanka, most people migrated into the tsunami affected coastal and urban areas. It is most likely that migrants

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out from Matara, Galle and Hambantota settled in tsunami-affected Colombo, Gampaha or Kaluthara districts.

7. Conclusion

Few research projects attempt to identify and quantify the long-term impact of a catastrophic disaster on household wellbeing. Most research attention is usually directed to an event in its immediate aftermath, and interest eventually wanes. From a macroeconomic perspective, the majority of the evidence points to no-effect at the national level and more adverse local long-term effect. Yet, little is really known about the impacts at the micro/household level in the longer term. Here, we estimated the causal effect of the Indian Ocean tsunami in Sri Lanka on household income and consumption eight years after the event, using a quasi-experimental method.

A strong association between area-wide tsunami disaster shock and increases in household income and consumption in the long-term emerged from our empirical investigation. Deviating from the common observation in the literature on short-term impacts, these results are suggestive of an optimistic potential for long-lasting positive consequences. Nevertheless, our analysis does not provide any evidence to evaluate whether the observed positive effect is due to so-called 'creative destruction' – the replacement of old with new capital and technology. Sri Lanka received a very large amount of external transfers post-tsunami, much larger than is typical for disaster events (Becerra et al., 2014 and 2015). The increases in consumption and income can thus be associated not with a 'build-back-better' reconstruction, but purely due to the infusion of unusually massive amounts of external resources for rebuilding. This infusion is not replicable in other cases, and only an accounting for the amount of assistance received, per district, could possibly start to allow us to differentiate between these two hypotheses.

Our findings suggest a more nuanced picture with respect to household consumption impacts. We observe a reduction of food consumption and only find an increase in non-food consumption. The increase in non-food consumption is much smaller than the increase in income. As previously observed in the literature, it is clearly evident that households in highincome regions experience better recovery from disaster. The positive impact on income is

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largely 'enjoyed' by the higher income regions. Household consumption follows a similar pattern. As was the case more generally, it is impossible to know, at this point, whether the most dramatic increases in incomes and consumption in the wealthier regions was due to their better resilience (and reduced vulnerability) or a function of their improved ability to capture the inflows of aid into the affected regions. After all, both of these possibilities have been documented in research projects on short-term post-disaster dynamics.

Further improvements to the analysis remain. More rigorous analysis can be obtained by incorporating more variation of the shock using smaller geographical units. This may address possible errors generated by the method used to identify the shock in the present approach. An evaluation of the plausibility of "creative destruction" hypothesis can also be undertaken by controlling for external financing; which we were unable to do with the available data in this case. Understanding and identifying the distributional impact of the tsunami is also a policy-relevant question we plan to pursue.

Table1: Summary statistics

	OPE	МЕАМ	STD DEV	Мим	MAY
	84202		42		1
SEX (HOUSEHOLD HEAD)(DOMMY)	84303	./9	.42	10	1
AGE (HOUSEHOLD HEAD)	84303	51	14.04	10	99
EDUCATION(YR) (HOUSEHOLD	84303	1	2.94	0	18
HEAD)		~-		-	
ETHNIC GROUP-	84303	.85	.35	0	1
SINHALESE(DUMMY)				_	_
ETHNIC GROUP-TAMIL(DUMMY)	84303	.14	.35	0	1
OTHER ETHNIC (DUMMY)	84303	.00	.06	0	1
Religion-Buddhist(Dummy)	84303	.79	.40	0	1
Religion-Hindu(Dummy)	84303	.07	.25	0	1
Religion-Muslim(Dummy)	84303	.06	.24	0	1
Religion-Catholic(Dummy)	84303	.07	.26	0	1
OTHER RELIGION	84303	.00	.01	0	1
MARRIED (HOUSEHOLD HEAD)	84303	.80	.40	0	1
WIDOWED DEVOICED (HOUSEHOLD	84303	.18	.38	0	1
HEAD)					
NOT MARRIED (HOUSEHOLD HEAD)	84303	.02	.15	0	1
HOUSEHOLD SIZE	84303	4	1.88	1	20
House ownership(Dummy)	84303	.79	.40	0	1
TOTAL SEASONAL AND NON-	84303	102	759.31	0	128003
SEASONAL CROP CULTIVATED LAND					
(PERCH)					
HOUSEHOLD IN RURAL	84303	.81	.39	0	1
SECTOR(DUMMY)					
HOUSEHOLD IN URBAN	84303	.13	.34	0	1
SECTOR(DUMMY)					
HOUSEHOLD IN ESTATE	84303	.05	.22	0	1
SECTOR(DUMMY)				-	•
	84303	.57	.49	0	1
	0.000		140	•	•
	84303	03	16	0	1
	04000	.00	.10	U	•
	84303	34	17	0	1
	04000	.54		U	•
	84303	28	45	0	1
TRANSFER DAYMENTS(DUMMY)	04000	.20	.40	U	1
HOUSEHOLDS DESERVING - DENSION	84202	10	20	0	4
AND DISABILITY DAYMENTS	84303	.10	.30	0	1
	84202	07	26	0	4
RECEIVING	84303	.07	.20	0	1
REMITTANCE-LOCAL (DUMMY)	04202	07	26	•	
HOUSEHOLD RECEIVING	84303	.07	.20	0	1
REMITTANCE-FOREIGN (DUMMY)			077.00	•	
HOUSEHOLDS RECEIVING	84303	142	377.06	0	30000
TRANSFER PAYMENTS (RS)				-	
HOUSEHOLDS RECEIVING -PENSION	84303	959	4418.38	0	156000
AND DISABILITY PAYMENTS(RS)				_	
HOUSEHOLD RECEIVING	84303	4439	33519.53	0	2400000
REMITTANCE-LOCAL(RS.)					
HOUSEHOLD RECEIVING	84303	7790	52799.3	0	2050000
REMITTANCE-FOREIGN(RS.)					
HOUSEHOLDS-AFTER	84303	.64	.48	0	1
TSUNAMI(DUMMY)					
HOUSEHOLDS AFFECTED BY	84303	.51	.50	0	1
TSNAMI(DUMMY)					
AFFECTED HOUSEHOLDS OBSERVED	84303	.33	.47	0	1
AFTER TSNAMI					
INCOME (RS./MONTH)	84303	8536.23	12497	-3595	98732
CONSUMPTION	84303	11692.86	10584	297	87436
EXPENDITURE(RS./MONTH)					

Table 2: Impact of tsunami on Household Income

MODELS		AVERAGE EF	FECT OF AF	ECTED REGIONS		REGIONS WITH	WITHOUT	RICHEST
						HIGH INTENSITY	RICHEST	REGIONS
							REGIONS	
INDEPENDENT	(1)	(11)	(111)	(IV)	(V)	(VI)	(VII)	(VIII)
VARIABLES	ROBUST S.	ROBUST S. E	ROBUST	ROBUST S. E	DISTRICT	DISTRICT	DISTRICT	DISTRICT
	E		S.E		CLUSTERED	CLUSTERED	CLUSTERED	CLUSTERED
					ROBUST SE	ROBUST SE	ROBUST SE	ROBUST SE
YEAR_2002	No	-700	-361	774	-361	261	-389	-590
		(95) ***	(92) ***	(334) **	(484)	(214)	(262)	(586)
YEAR_2006	No	-1879	-105	-818	-105	2791	-127	-239
		(121)***	(100)	(70) ***	(1245)	(461) ***	(1241)	(1270)
YEAR_2009	No	-3185	-664	1648	-664	2198	-690	-804
		(94) ***	(80) ***	(381)***	(861)	(160) ***	(848)	(886)
YEAR_2012	No	-1319	1476	-717	1476	11334	1418	1101
		(164)***	(137)***	(100)***	(1791)	(543) ***	(1800)	(1854)
TREATMENT*2006	6878	10972	7067	4985	7067	2045	2711	11766
	(214)***	(212)***	(195)	(373) ***	(2421)***	(420) ***	(1573) *	(1845) ***
TREATMENT*2009	5057	10440	5872	3832	5872	-412	2882	9259
	(186)	(168)	(165)	(578) 🛄	(1674)	(326)	(1245)	(1095)
TREATMENT*2012	16005	18892	15003	-2239	15003	-418	8068	22981
	(248) ົົ	(252)	(224)	(178)	(3997) 🗰	(888)	(3172) ^^	(2917)
TREATMENT	4400							
	(89)							
PRE TSUNAMI	-528	No	No	NO	No	No	No	NO
	(67)	¥		N/		N	¥	¥
HOUSEHOLD	NO	YES	YES	YES	YES	YES	YES	YES
COVARIATES	Na	Na	N	Vee	Vee	¥=-	Vee	V=a
DISTRICT FIXED	NO	NO	TES	TES	YES	TES	TES	TES
	No	No	No	Vrc	No	No	No	No
TREND	NO	NO	NO	165	NO	NO	NO	NO
CONSTANT	3448	4883	-896	936	-896	352	26	-542
CONSTANT	(56)***	(818)***	(782)***	(758)	(1082)	(999.02)	(939)	(1103)
F	(00)	808.47	1802 72	775.21	(1002)	(353.02)	(303)	(1100)
P-SOUARED		0 29	0.41	0.53	0.49	0.24	0 3946	0.55
		84303	8/303	84303	8/303	55921	65921	58827
ORSERVATIONS		84505	84303	64303	04303	55921	05921	50027
OBSERVATIONS								

INCOME SOURCES	PAID	Agricultural Income	Non- agricultural Income	REMITTANCE	TRANSFERS	DIVIDENDS	RENTS AND OTHER INCOME
INDEPENDENT							
VARIABLES							
YEAR_2002	638	280	-1599	45	36	26	305
	(312) ^{**}	(317)	(634) ^{**}	(35)	(58)	(11) ^{**}	(113) ^{***}
YEAR_2006	599	-96	-1083	171	13	10	298
	(1054)	(171)	(333) ***	(94) [*]	(31)	(10)	(113) ^{***}
YEAR_2009	67	-34	-1073	147	11	6	232
	(735)	(171)	(331) ***	(77) [*]	(52)	(5)	(85) ^{***}
YEAR_2012	1760	59	-990	338	137	6	185
	(1518)	(196)	(345) ***	(162) ^{**}	(77) [*]	(7)	(77) ^{**}
TREATMENT*2006	1786	231	4559	93	175	12	177
	(1098)	(286)	(1750) ***	(110)	(55.53) ^{***}	(22)	(142)
TREATMENT*2009	940	325	3911	136	62	33	428
	(765)	(220)	(1470)***	(93)	(52)	(15) ^{**}	(178) ^{**}
TREATMENT*2012	5673	717	6246	947	828	110	447
	(1869) ^{***}	(326) ^{**}	(2202) ^{**}	(257) ^{***}	(168) ^{***}	(42) ^{***}	(122) ***
HOUSEHOLD COVARIATES AND DISTRICT FIXED EFFECTS	YES	YES	YES	YES	YES	YES	YES
CONSTANT	-906	-29	952	-124	-9	-14	-91
	(723)	(216)	(371)***	(81)	(88)	(15)	(94)
R-SQUARED NUMBER OF OBSERVATIONS	0.33 84303	0.12 84303	0.24 84303	0.05 84303	0.08 84303	0.01 84303	0.03 84303

Table 3: Impact of tsunami on household Income by source of income

Notes: Robust standard errors in bracket. ***, **, *, stand for significance at 1%, 5% and 10% respectively. No district year trends

Table 4: Impact of tsunami on Household consumption

MODELS		Average ef	FECT OF AFFEC	TED REGIONS		REGIONS WITH HIGH	WITHOUT	RICHEST REGIONS
INDEPENDENT VARIABLES	(I) Robust S. E	(II) Robust S. E	(III) Robust S. E	(IV) Robust S. E	(V) District Clustered Robust SE	(VI) DISTRICT CLUSTERED ROBUST SE	(VII) DISTRICT CLUSTERED ROBUST SE	(VII) DISTRICT CLUSTERED ROBUST SE
YEAR_2002	Νο	-2618 (154) ^{***}	452 (72) ^{***}	1232 (197)	452 (205) ^{***}	261 (214)	388 (243)	581 (158) ***
YEAR_2006	No	-1335 (136) ^{***}	2829 (131) ^{***}	2846 (566)	2829 (445) ^{***}	2791 (461) ^{**}	2849 (452) ^{***}	2897 (449) ^{***}
YEAR_2009	Νο	-2548 (159) ^{***}	2236 (100) ^{***}	2735 (330)	2236 (175) ^{***}	2198 (160) ^{**}	2255 (170) ^{***}	2304 (180) ^{***}
YEAR_2012	Νο	6696 (213) ^{***}	11140 (185) ^{****}	10686 (519)	11140 (567) ^{***}	11334 (543) ^{**}	11424 (548) ^{***}	11041 (600) ^{***}
TREATMENT*2006	-1902 (172)	4144 (158) ***	1269 (192) ^{***}	1555 (696)	1269 (507) ^{***}	2045 (420) ^{**}	1686 (457) ^{***}	877 (703)
TREATMENT*2009	-3588 (164) ^{***}	2964 (118) ^{***}	207 (161)	1070 (454)	207 (275)	-412 (326)	-58 (318)	492. (247) ^{**}
TREATMENT*2012	8309 (211) ^{***}	5697 (240) ^{***}	2786 (239) ^{***}	2167 (796)	2786 (1103) ^{***}	-418 (888)	1241 (1237)	4560 (725) ^{***}
TREATMENT	3284 (75) ^{***}	No	No	Νο	Νο	Νο	Νο	No
PRE TSUNAMI	-5577 (88) ^{***}	No	No	No	No	No	Νο	No
HOUSEHOLD COVARIATES	No	YES	YES	YES	YES	YES	YES	YES
DISTRICT FIXED EFFECTS	No	Νο	YES	YES	YES	YES	YES	YES
DISTRICT YEAR	Νο	Νο	Νο	YES	No	No	Νο	Νο
CONSTANT	11771 (80)	3735 (744) ^{***}	-428 (744)	869 (743)	-428 (1083)	352 (999)	3380 (612) ^{***}	2458 (1449) [*]
F	2894.63	771.33	554.05	231.36				
R-SQUARED	0.19	0.25	0.28	0.29	0.28	0.24	0.1623	0.30
NUMBER OF OBSERVATIONS	84303	84303	84303	84303	84303	55921	66246	58827

Notes: Robust standard errors in bracket. ***, **, *, stand for significance at 1%, 5% and 10% respectively.

DEPENDENT	FOOD	Non-	Housing	CLOTHING	PERSONAL	Сомм. &	EDUCATION	HEALTH	HH NON-	Non-cons.
VAR.	CONS	FOOD			CARE	TRANSP.			DURABLES	EXPENDITURE
INDEPENDENT		CONS.								
VAR.	I	п	ш	IV	v	VI	VII	VIII	IX	x
XEAD 2002		1001	267	207	17	107	60	60	10	0.49
YEAR_2002	-942	1391	207	-237	17	(21)	(12)	(12)	-18	948
	(184)	(197)	(30)	(33)	(3)	(31)	(12)	(12)	(8)	(101)
YEAR_2006	-253	3085	637	806	40	604	180	91	-10	1192
	(174)	(401)***	(70)	(28)	(7)	(46)	(19)	(24)	(9)	(205)
YEAR 2009	-592	2831	396	66	38	657	160	73	-11	1211
-		(140)***	(27)	(23)	(4)	(52)	(11)	(14)	(6)	(85)
	(135)									
YEAR 2012	2728	8414	2178	385	219	1717	531	377	81	2444
	(214)	(453) ***	(168)	(29)	(8)	(109)	(36)	(50)	(8)	(193)
	***	()								
TREATMENT*200	-274	1552	285	18	8	303	61	103	4	742
6	(155)	(439) ***	(116)	(23)	(8)	(86)	(22)	(34)	(9)	(211)
	*									
TREATMENT*200	-538	761	273	-67	3	99	125	109	-14	329
9	(120)	(274) ***	(160)	(22)	(8)	(72)	(40)	(29)	(7)	(164)
	* * *									
TREATMENT*201	-141	2897	985	З	45	374	256	243	-12	974
2	(252)	((928) ***	(558)	(23)	(26)	(245)	(108)	(57)	(9)	(468)
	VEC	VEC	VEC	VEC	VEC	VEC	Vec	VEC	VEC	VEC
HOUSEHOLD	YES	YES	YES	YES	YES	YES	YES	YES	YES	YES
DISTRICT FIXED										
EFFECTS										
DISTRICT YEAR	No	No	No	No	No	No	No	No	No	No
TREND										
CONSTANT	3025	-3432	-823	97	-5	-770	-262	-167	40.3979	-1325
	(222)	(771)	(392)	(14)	(9)	(185)	(64)	(76)	6	(153)
B COULDED	0.07	0.00	0.20	0.16	0.37	0.12	0.06	0.03	(4)	0.07
R-SQUARED	0.27	0.23	0.29	0.16	0.27	0.12	0.06	0.03	0.12	0.07
NUMBER OF	8430	84303	84303	84303	84303	84303	84303	84303	84303	84303
OBSERVATION	3									

Table 5: Impact of tsunami on Household consumption by type of consumption

Notes: Robust standard errors in bracket. ***, **, *, stand for significance at 1%, 5% and 10% respectively.

DEPENDENT VARIABLES	LOG OF	LOG OF	LOG OF MONTHLY CONSUMPTION
INDEPENDENT Variables	I	II	III
YEAR_2002	31 (.25)	-1.76 (.45) ^{***}	.048 (.026) [*]
YEAR_2006	87 (.43) ^{**}	-3.3 (1.39) ^{**}	.32 (.05) ***
YEAR_2009	-2.01 (.84) ^{**}	-1.17 (.84)	.27 (.02) ***
YEAR_2012	-1.47 (.86)	73 (.84)	.98 (.03) ^{***}
TREATMENT*200 6	1.32 (.54) ^{**}	3.20 (1.48) ^{**}	.03 (.06)
TREATMENT*200	2.27	1.23	05
9	(.96) **	(.96)	(.02) **
TREATMENT*201	2.4	1.45	08
2	(.97) **	(.95)	(.04) **
HOUSEHOLD COVARIATES	YES	YES	YES
DISTRICT FIXED EFFECTS	YES	YES	YES
DISTRICT YEAR TREND	Νο	No	Νο
CONSTANT	5.68	5.68	7.97
	(.59)***	(.59) ***	(.045) ***
R-SQUARED	0.78	0.74	0.35
N. OF OBS.	71234	84303	84303

Table 6: Impact on consumption and income using log of income and consumption (log normal)

Notes: Column I dropped all observations for whom the dependent variable ≤0.While column two includes them (by adding a small constant. Robust standard errors in the bracket. ***, **, *, stand for significance at 1%, 5% and 10% respectively.

DEPENDENT	CONSUMPTION			INCOME		
VARIABLES						
INDEPENDENT	DEATH	DISPLACEMENT	HOMELESSNESS	DEATH	DISPLACEMENT	HOMELESSNESS
VARIABLES						
YEAR_2002	465.	463	462	-264	-260	-261
	(208)	(209)	(209)	(510)	(513)	(519)
YEAR_2006	3276.	3231	3047	3664	3276	2924
	(282.)	(260)	(280)	(2018)	(1953)	(1992)
YEAR_2009	2388	2429	2431	2275	2011	1686
	(204)	(193)	(191)	(1523)	(1477)	(1510)
YEAR_2012	12738	12752	12675	8917	8375	7462
	(649) ***	(625)	(624)	(3627)	(3459)	(3494)
TREATMENT*2006	84313	4642	12250	-51074	5190	18628
	(23367) ***	(1383)-	(2971)	(103401)	(6525)	(16751)
TREATMENT*2009	-21425	-1874	-2976	3145	5530 (5485)	18367
	(24075)	(553)	(2114)	(76281)		(13663)
TREATMENT*2012	-71482	-3615	-3178	111721	16042	52080
	(73629)	(3341)	(10143)	(180436)	(14300)	(34625)
HOUSEHOLD	YES	Yes	Yes	YES	YES	Yes
COVARIATES						
DISTRICT FIXED	YES	YES	YES	YES	Yes	Yes
EFFECTS						
DISTRICT YEAR	No	No	No	No	No	No
TREND						
CONSTANT	-851	-883	-839	-3419	-3125	-2770
	(1084)	(1065)	(1043)	(1534.30)	(1495)	(1538)
R-squared	0.28	0.28	0.28	0.44	0.44	0.44
NUMBER OF	84303	84303	84303	84303	84303	84303
OBSERVATIONS						

Table 7: Consumption and income consequences by type of damage

Notes: Robust standard errors in bracket. ***, **, *, stand for significance at 1%, 5% and 10% respectively.



Figure 2: Normalised income of affected households by districts





Figure 3: Normalised consumption of affected and non-affected households

Figure 4: Normalised consumption of affected households by District



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Appendix 1: Components of household income and consumption

Income components	Description	1995, 2006,	2002
	-	2009 and	
		2012	
Paid income	Income from paid employments	V	V
	(wage/salaries, commissions, bonus,		
	arrears)		
Agricultural Income	Income from engage in agricultural	V	V
	activities including value of amount		
	consumed by the household		
Non-agricultural Income	Income from engage in non-agricultural	V	V
	activities including value of amount		
-	consumed by the household		
Remittance	Local and foreign remittance	v	V
Transfers	Receipts of government transfer payment,	v	v
Dividende	disability and relief payments	-1	-1
Dividends Dente and athen income	Dividends and interests	V	V
Adhea income	Property rents and other cash receipts	V	V
Adnoc income	Loans taken, sales of assets, withdrawal of	N	v
	sociatios renovment of loops given	v	^
	insurance compensations lottery and		
	other adhoc gains		
Income components	Description	I	I
Food	Value of consumed food (include freely reco	eived or home g	rown) of the
	main house hold excluding boarders and ser	rvants	,
Non food	Household expenditure on housing fuel and	light	
	Personal care		
	Health expenses		
	Transport and communication expenses		
	Education expenses		
	Expenditure in recreation and cultural activi	ties	
	Expenditure on nondurable household good	ls	
	Household services (laundry, grinding etc.)		
	Clothing textiles and foot wear		
	Non consumption expanses Souings no	umont of Inclu	anco dobt
	income tay contributions to trade unions	s thrift societies	ance, uebt,
	security navments (provident fund) ever	penses on socie	al activities
	donations, loans given		
Servants	Expenses on servants food and non-food co	onsumption	
Servants	Durable household goods Non consumption expenses: Savings, pa income tax, contributions to trade unions security payments (provident fund), exp donations, loans given. Expenses on servants food and non-food co	yment of Insur s, thrift societie penses on socia ponsumption	rance, debt, s and social al activities,

		displaced	population that	public infrastructure
District	Deaths	population	become homeless	damage (Rs. Million)
Jaffana	2640	39607	20734	1716.4
Mulativu	3000	22657	22831	2166.1
Trincomalee	1078	81643	36326	3446
Batticaloa	2840	61912	70282	3208.4
Ampara	10436	75172	67707	3959.2
Hambantota	4500	17723	8955	1296.5
Matara	1342	13206	28860	2216.9
Galle	4214	128077	53440	4289.9
Kaluthara	256	27713	24855	1009.4
Colombo	79	31239	24457	235.1
Gampaha	6	1449	4401	348.1
Puttlam	4	66	228	16.9
Kilinochchi	0	1603	1186	232.3
Mannar	0	0	0	11

Appendix 2: Damage caused by Tsunami

Source: Department of census and Statistics (2006)

Variable	Num.	Mean	Std. Dev.	Min	Max
	Observations				
Total income	67498	11278.19	15241.21	-3500	98732.24
Total	67498	12003.76	10705.96	297	87436.11
consumption					

Appendix 3: Average income before and after Tsunami excluding year 2002 data

Appendix 4: Correlation of disaster damage intensities

Damage Variables	Deaths	Infrastructure	Displacements	Homelessness
Deaths	1.0000			
Infrastructure	0.6006	1.0000		
Displacements	0.7696	0.9476	1.0000	
Homelessness	0.6406	0.8792	0.8755	1.0000

Appendix 5: IQR test for severe outliers

Explanatory Variable	Income	Log of Income	Consumption	Log of consumption
Low values	0.19%	0.49%	0.00%	0.1%
High values	3.83%	0.11%	2.97%	0.00%

Appendix 6: Determinants of Income

VARIABLES	COEFFICIEN	ROBUST	VARIABLES	COEFFICIEN	Robust
	т	STD. ERR.		т	STD. ERR.
		(CLUSTERED			(CLUSTERED
))
YEAR_2002	-1234	23***	CMB(COLOMBO)	4347	91***
YEAR_2006	-908	39***	GMP(Gамрана)	1435	44***
YEAR_2009	-2196	23***	KTR(KALUTARA)	-472	42***
YEAR_2012	8287	134***	KDN(KANDY)	527	22***
SEX	-154	45***	NEL(NUWARA ELIYA)	-1014	46***
Age	1	2	GLL(GALLE)	-417	43***
LEVEL OF	21	11*	MTR(MATARA)	-925	46***
WIDOWED	-103	102	ΗΒΤ(ΗΑΜΒΑΝΤΟΤΑ)	-3815	321***
AND					
DIVORCED					
HH SIZE	22	23	KRN(KURUNEGALA)	-3676	323***
OWNERSHIP	134	74 [*]	PTLM(PUTTLAM)	-5042	293***
OF HOUSE					
TOTAL AREA	.12	.07	PLNR(POLONNARUWA	-5500	331***
OF CROP)		
CULTIVATED			-		
SINHALESE	-80	126	ANUR (ANURADHAPUR	-6322	330***
			A)		
OTHER	-387	612	BDL(BADULLA)	-7099	320***
ETHNIC					
TRANSFER	-2581	349***	MONR(MONERAGALA)	-6453	334***
PAYMENT					
PENSION &	2578	364***	RTN(RATNAPURA)	-7018	333***
DISABILITY					
PAYMENT					
REMITTANCE	324	127***	KGL(KEGALLE)	-7196	329***
-LOCAL					
REMITTANCE	2025	181***			
-FOREIGN					
EMPLOYED	-156	96			
UNEMPLOYE	-350	305			
D					
NON PAID	-94	77			
EMPLOYED					
RURAL	-232	93***			
SECTOR					
ESTATE	-484	174***			
SECTOR					
CONSTENT	7977	523***			

Notes: Robust standard errors in bracket. ***, **, * stand for significance at 1%, 5% and 10% respectively. This regression controlled for the district year trends

VARIABLE	COEFFICIENT	ROBUST STD.	VARIABLE	COEFFICIENT	ROBUST STD.
		ERR.			ERR.
		(CLUSTERED)			(CLUSTERED)
YEAR_2002	-405	26	CMB(COLOMBO)	6280	387
YEAR_2006	-334	37	GMP(Gамрана)	4648	243
YEAR_2009	3227	244	KTR(KALUTARA)	2241	182
YEAR_2012	11889	149	KDN(KANDY)	1340	146
SEX	137	108	MTL(MATALE)	948	107
Age	13	6	NEL(NUWARA ELIYA)	2552	340
LEVEL OF	264	91	GLL(GALLE)	2658	97
EDUCATION					
WIDOWED AND	-511	130	MTR(MATARA)	2498	107
DIVORCED					
HH SIZE	286	51	ΗΒΤ(ΗΑΜΒΑΝΤΟΤΑ)	3558	30
OWNERSHIP OF	1641	248	KRN(KURUNEGALA)	1360	55-
HOUSE					
TOTAL AREA OF	.04	.04	PTLM(PUTTLAM)	1603	126
CROP					
CULTIVATED					
SINHALESE	274	196	PLNR(POLONNARUWA)	2587	89
OTHER ETHNIC	-179	549	ANUR (ANURADHAPURA)	1225	96
TRANSFER	08	.05	BDL(BADULLA)	501	96
PAYMENT					
PENSION &	.07	.01	RTN(RATNAPURA)	1136	63
DISABILITY					
PAYMENT					
REMITTANCE-	00	.001	KGL(KEGALE)	1198	8 6
LOCAL					
REMITTANCE-	.001	.0004			
FOREIGN					
EMPLOYED	-106	159			
UNEMPLOYED	37	452			
NON PAID	94	108			
EMPLOYED					
RURAL SECTOR	-1852	185			
ESTATE SECTOR	-1819	582			
CONSTENT	1141	858			

Appendix 7: Determinants of consumption

Notes: Robust standard errors in bracket. ***, **, *, stand for significance at 1%, 5% and 10% respectively. This regression controlled for the district year trends

DISTRICTS	COLOMBO	GAMPAH	KALUTAR	GALLE	MATARA	HAMBANT	PUTTLAM
INDEPENDENT		Α	Α			ΟΤΑ	
VARIABLES							
YEAR_2002	-979	-65	-329	-318	-551	-458	-683
	(473)	(417)	(244)	(253)	(207)	(202)	(281)
YEAR_2006	-440	41	-78	-80	-198	-156	-268
	(1263)	(1273)	(1266)	(1266)	(1258)	(1261)	(1258)
YEAR_2009	-1003	-525	-639	-641	-759	-718	-828.
	(868)	(879)	(864)	(865)	(854)	(856)	(854)
YEAR_2012	892	1493	1368	1449	1299.	1359	1240
	(1868)	(1841)	(1839)	(1843)	(1841)	(1842)	(1842)
TREATMENT*2006	13579	9742	5223	3526	2799	1541	-1362
	(1183)	(1194)	(1201)	(1207)	(1205)	(1216)	(1193)
TREATMENT*2009	10177	8146	5440	3223	2577	2493	-775
	(794)	(805)	(813)	(816)	(817)	(823)	(808)
TREATMENT*2012	26057	19780	14622	8529	7563.	9823.	-722
	(1779)	(1787)	(1783)	(1762)	(1766)	(1762)	(1758)
HOUSEHOLD	YES	YES	YES	YES	YES	YES	YES
COVARIATES AND							
DISTRICT FIXED							
EFFECTS							
CONSTANT	-1001	-418	-106	-371	-1660	-1691	-86
	(1222)	(1288)	(993)	(963)	(1526)	(1370)	(995)
R-SQUARED	0.58	0.53	0.44	0.39	0.3982	0.4041	0.4078
NUMBER OF	50998	48274	46702	46785	45683	44343	44188
OBSERVATIONA							

Appendix 8: Impact of tsunami on household Income by Districts

Notes: Robust standard errors in bracket. ***, **, *, stand for significance at 1%, 5% and 10% respectively. No district year trend

Appendix 9: Impact of tsunami on household consumption by Districts

DISTRICTS	COLOMBO	G амрана	KALUTARA	GALLE	MATARA	Ηαμβαντότα	PUTTLAM
INDEPENDENT							
VARIABLES							
YEAR_2002	477	665	734	448	479	374	483
	(140)	(167)	(231)	(174)	(162)	(237)	(164)
YEAR_2006	2835	3004	3032	2889	2910	2856	2908
	(460)	(434)	(441)	(454)	(450)	(465)	(451)
YEAR_2009	2242	2409	2438	2296	2317	2262	2314
	(178)	(152)	(166)	(151)	(146)	(170)	(148)
YEAR_2012	11122	11306	11599	11460	11440	11406	11428
_	(626)	(567)	(533.01)	(532)	(533)	(542)	(539)
TREATMENT*2006	1268	183	1365	1973	2018	2182	1059
	(593)	(432)	(425)	(417)	(429)	(411)	(435)
TREATMENT*2009	303	367	612	-937	-238	325	95
	(377)	(110)	(105)	(117)	(119)	(132)	(107)
TREATMENT*2012	5152	3817	4266	-1535	-369	1538	2072
	(524)	(515)	(485)	(551)-	(556)	(545)	(551)
HOUSEHOLD	YES	YES	YES	YES	YES	YES	YES
COVARIATES AND							
DISTRICT FIXED							
EFFECTS							
DISTRICT YEAR	No	No	No	No	No	No	No
TREND							
CONSTANT	1896	4001	394	1021	-176	2651	3706
	(1237.093)	(825)	(694)	(503.77)	(1258)	(971.66)	(487)
R-SQUARED	0.31	0.27	0.26	0.24	0.24	0.25	0.25
NUMBER OF	50998	48274	46702	46785	45683	44343	44188
OBSERVATIONS							

Notes: Robust standard errors in bracket. ***, **, *, stand for significance at 1%, 5% and 10% respectively.

Appendix 10: Impact on Income and Consumption excluding year 2002 data

DEPENDENT	MONTHLY	MONTHLY	Аднос
VARIABLES	INCOME (RS.)	HOUSEHOLD	INCOME
	(EXCLUDE YEAR	CONSUMPTION	(EXCLUDING
INDEPENDENT	2002 DATA)	(Rs.)	VEAD 2002
VARIABLES		(EXCLUDE YEAR	
		2002 DATA)	DATA)
YEAR_2006	-2547	2941	-2303
	(799) ***	(457) ***	(1266) ***
YEAR_2009	-3024	2354	-2229
	(1069) ***	(182)***	(1352) ***
YEAR_2012	145	11323	-1098
	(1326)	(618)***	(717)
TREATMENT*2006	3345	1077	-3859
	(1538) **	(524) **	(2677)
TREATMENT*2009	1535	12	-4463
	(1958)	(417)	(2803)
TREATMENT*2012	11994	2704	-2904
	(2507) ***	(1264)**	(1946)
HOUSEHOLD	YES	YES	YES
COVARIATES			
DISTRICT FIXED	YES	YES	No
EFFECTS			
DISTRICT YEAR	No	No	No
TREND			
CONSTANT	870	3616	887
	(729)	(960)***	(795)
R-SQUARED	0.52	0.27	0.19
NUMBER OF	67498	67498	67498
OBSERVATIONS			

Notes: Robust standard errors in bracket. ***, **, *, stand for significance at 1%, 5% and 10% respectively.

Appendix 11: Economic growth rate



Source: International Monetary Fund, World Economic Outlook Database, April 2015



Appendix 12: Tsunami affected Districts in Sri Lanka

Source: Department of census and statistics (2005)



Appendix 13: Damage (Calculated per household) across Districts

Source: Author calculated using Census and Statics data (2005)





Distribution of Residuals income



Leverages against squared residuals income



Distribution of Residuals consumption



Leverages against squared residuals consumption

Appendix 15: Distribution of Residuals using Log





Distribution of Residuals income

Leverages against squared residuals Income



Distribution of Residuals consumption



Leverages against squared residuals Consumption





Income

Log of Income















Income

Log of income





Consumption

Log of consumption





Appendix 19: Distribution of consumption



Appendix 20: District level net migration



Source: Department of Census and Statistics (2013)

WORKING PAPERS IN ECONOMICS AND FINANCE



TE WHARE WĀNANGA O TE ŪPOKO O TE IKA A MĀUI

0

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