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Does the Quality of Electricity Matter? Evidence from Rural India

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Abstract

This paper estimates the returns to household income due to improved access to electricity in rural India. We examine the effect of connecting a household to the grid and the quality of electricity, defined as hours of daily supply. The analysis is based on two rounds of a representative panel of more than 10,000 households. We use the district-level density of transmission cables as instrument for the electrification status of the household. We find that a grid connection increases non-agricultural incomes of rural households by about 9 percent during the study period (1994-2005). However, a grid connection and a higher quality of electricity (in terms of fewer outages and more hours per day) increases non-agricultural incomes by about 28.6 percent in the same period.

JEL classification: O12, O18, Q48

Keywords: Electricity Supply, Quality, India, Energy and Development, Infrastructure

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1 Introduction

In 2009, 288 million people, corresponding to about a fourth of India's population, had no access to electricity (World Energy Outlook 2011). In 2005, 364 million people did not have access, while another 300 million received intermittent supply (Balachandra, 2011). The massive grid failure of July 2012 affected 670 million individuals, almost 10 percent of the world's population and made world headlines (New York Times, 2012). According to the World Bank, unreliable electricity supply has been a major obstacle to Indian economic development, limiting its comparative advantage in labor-intensive products (World Bank, 2010; Rud, 2012). Connecting all households to the grid and providing them assured electricity is likely to have a major impact on the Indian economy.

The aim of this paper is to estimate the economic returns to rural households from a grid connection as well as from the quality of electricity supply. We study the effect on rural household income - we include all income from wages and business income, except income from agricultural assets for which we have insufficient data.

Rural electrification may affect incomes of households in several ways (Oda and Tsujita, 2011; Modi, 2005). It may free members of the household from domestic chores, or let them perform these tasks in the evening. The resulting increase in the labor supply may drive down wages, especially for females. Electrification may also increase the productivity of some activities. The productivity of agricultural labor may improve due to technologies such as sprinkler or drip irrigation, which is likely to have an upward effect on wages. Assured supply of electricity creates opportunities for entrepreneurial activities which can take place within the household, such as rice-milling and production of oil from oilseeds. Other ancillary industries, such as the repair and welding of agriculture implements such as ploughs and tractors, may now be possible. These activities are likely to increase farm incomes, which may result in a general increase in labor supply and wages. Finally, electrification could also affect labor supply by children. The need to collect different kinds of fuels, including animal and agricultural waste and fuelwood may diminish and therefore their labor supply to market activities may increase, or they may allocate their time to other pursuits such as education (Modi, 2005; Oda and Tsujita, 2011).

There are many studies that qualitatively describe the Indian electricity sector (Balachandra, 2011; Khandker et al., 2010; Bhide and Monroy, 2011). Most papers identifying the effect of rural electrification using econometric techniques such as propensity score matching, difference-in-difference or instrumental variables usually deal with countries other than India. Bensch et al. (2010) focus on the case of Rwanda, while Khandker

¹The "quality" is defined in terms of the average hours of daily electricity received by the household.

²70 percent of India's population lives in rural areas, the focus of our analysis.

et al. (2009a and 2009b) study Bangladesh and Vietnam, specifically rural electrification initiatives undertaken by the World Bank. Two recent papers focus on the impact of connecting a household to the electricity grid. Dinkelman (2011) studied the labor market effects of an electrification project in South Africa using an instrumental variable approach. Lipscomb et al. (2013) examine the case of Brazil and, using instrumental variables, find a positive impact of electrification on various development indicators. The main contribution of our paper is the focus on the quality of electricity, which has not been done previously.

A variety of factors may lie behind the electrification of certain areas and the non-electrification of others. Governments usually aim infrastructure investment to already growing areas. Other economic trends may affect the investment decision. For instance, a rich village is probably more likely to be electrified than a poor village. The likelihood of being connected to the grid may also depend on the proximity to a big city, or on the population density of the region. For all these reasons, disentangling the impact of infrastructure investments such as electricity on development outcomes has been discussed extensively in the literature (Duflo and Pande, 2007; Roller and Waverman, 2001; Aschauer, 1989; Garcia-Mila and McGuire, 1992; Holtz-Eakin, 1993).

We tackle this endogeneity issue regarding electrification by using an instrumental variable approach. We construct an instrument which measures the difference in the density of transmission cables in each district from the national average. The argument is straightforward. If a household is located in a district served by a higher density of transmission cables the probability of being connected to the network, and of receiving a better quality power supply, is higher than when it is located in a district with a lower cable density (Brown and Sedano, 2004). We then interact the variation of this measure with the initial state of electrification in order to capture the possible heterogeneity in the impact of an improvement of the transmission network within a district. The data on the transmission network is constructed by measuring the length of the transmission cables within each district using ArcGIS.

Decisions regarding investments on the Indian transmission network lie with the federal government, while state governments are charged with their implementation (Modi, 2005; Balachandra, 2011). Transmission lines are a major infrastructure investment and require a high voltage pole at their destination. The primary purpose of establishing transmission lines is to electrify urban and industrial areas (Brown and Sedano, 2004). Once the federal government decides to build a transmission line between two points, the planner's problem is to find the shortest and the least costly route between these two points, and the objective of bringing electricity to remote villages is not considered during this stage.³ For this reason, the objective of rural electrification in India may

 $^{^3}$ The per mile cost of setting up a new transmission network may range from several hundred thousand

have played a small role on how the transmission lines were installed and located. Modi states that federal legislation in India did not explicitly mention rural electrification until 2003 (Modi, 2005).⁴ However, state-level elected officials have a large incentive to improve rural electrification, but they only control investment in distribution networks, not in transmission (Balachandra, 2011). While distribution networks are much less costly, they can only be installed if there is a transmission line within a feasible radius. For these reasons, this paper uses the transmission network, rather than the distribution network, as an instrument for the probability of electrification in rural areas.

Our results suggest that the 16% increase in grid connections between 1994 and 2005 increased non-agricultural incomes of rural households by about 9 percent. However, higher quality of electricity (a 32% increase, where quality is measured in terms of fewer outages and more hours per day) increased non-agricultural incomes by about 28.6 percent. This improvement in electrification rates translates to Rupees 574 per person, while the improvements in the quality of power supply translates to Rupees 1,852 per person in a household that experienced an average improvement in electrification probability and power quality, respectively. The results regarding a new grid connection are comparable in magnitude to those of Dinkelman (2011) and Lipscomb et al. (2013). The former finds an impact of electrification of roughly 16% on male earnings and no impact on female earning, while the latter finds that an increase in the electrification rate of 10% increases income by 9%.

The remainder of the paper is organized as follows. Section 2 discusses electrification policy in India. Section 3 describes the data used in our analysis and presents some stylized facts. Section 4 outlines the methodology used and presents results. Section 5 discusses potential channels and section 6 concludes the paper.

2 Background

Electricity supply in India is inefficient at all stages of production, transmission and distribution (Modi, 2005). At subsidized prices, supply lags demand, especially during peak periods, leading to frequent outages and voltage fluctuations. In 2007, average per capita consumption was 543 kWh, lower than Sub-Saharan Africa (578 kWh) while the average for OECD countries was about 8,500 kWh. Average per capita consumption in India is about a fourth of Chinese consumption (2,346 kWh).⁵ These low figures are partly because India is home to a third of the world's population that is not connected to an electric grid. Cost recovery is low and has actually declined over time (69 percent in

dollars to several million (Brown and Sedano, 2004).

⁴In the federal planning budget, only about 10 percent of the electricity outlay is spent on distribution, the rest goes towards power generation and transmission.

⁵International Energy Agency, 2009.

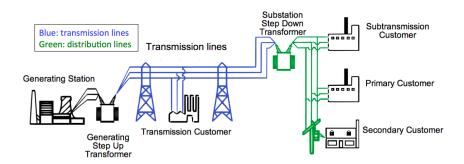
2001-02). These low rates are mainly due to losses during transmission and distribution, which rose from 25% in 1997-98 (Modi, 2005) to about 39% in 2000-01 (Oda and Tsujita, 2011), while the average for neighboring countries is about 10%. Estimates suggest that only 55% of the power supplied is billed and only 41% is paid for (Modi, 2005). Infrastructure theft has led to further declines in coverage (Balachandra, 2011).

Figure A.1 and Figure A.2 plot the transmission network with the rail and road networks, respectively. Note that both rail and road networks have a national coverage while the electricity transmission network is quite unevenly distributed, with higher coverage in the northern industrial states. Large parts of the country have no transmission lines hence no electricity. Regions with poor transmission network connectivity include eastern Karnataka, western Andhra Pradesh, Jammu & Kashmir, Arunachal Pradesh, Assam, Aizawl, Imphal, Nagaland, Gujarat (outside of the Ahmedabad metro region), western Rajasthan and eastern Madhya Pradesh. However, there has been significant improvement in average consumption and production of electricity. The generation capacity grew from 1,362 MW in 1947 to nearly 74,699 MW in 1991. Over the same period, per capita consumption increased from 15.55 kWh to 252.7 kWh (Modi, 2005). Currently, capacity exceeds 170 GW.

According to Brown and Sedano (2004), the cost of setting up a transmission network ranges from several hundred thousand dollars if there are no geographical obstacles such as hills or mountains, to over a million dollars per km. Figure 1 shows the main components of an electricity network. The transmission network carries electricity from production sites to high demand locations. This network is characterized by a high voltage (High-Voltage Direct-Current of 765, 500, 345 and 230 kV), as this reduces transport losses. Because of the high voltage, it is not possible to directly connect a house or a firm to the line, and a step-down transformer must be installed to decrease the voltage and then channel power through a subtransmission network, which is characterized by lower voltage, usually 69, 115 or 138 kV. At the end of the sub-transmission network there is another electrical substation which moves electricity into the distribution network. This last section of the grid is characterized by lower voltages (less than 50 kV) and travels very short distances, between 50 and 100 km. The electricity transported by the distribution network is at a voltage that can be consumed by the retail consumer.

Setting up new segments of the transmission network is costly and differs from the investment in the distribution network. The cost depends on its length, but also on other factors related to the landscape it must traverse. For example, installing cables over rolling hills is costlier than over flat land because the line will need a higher number of supporting structures. If the line has to bypass a mountain, the cost is likely to increase, as more structures are needed. Because of the high cost of setting up a transmission network, and because it is mainly used to transfer electricity to high de-

Figure 1: Schematic of the Power Grid



Source: United States Department of Energy.

mand centers, its route is not designed to serve retail customers between the two points (Brown and Sedano, 2004). The engineer's problem at this stage is to find the route that is shortest and the most cost effective. Once the transmission cables are installed, the state government can withdraw power through a step down transformer and build a new distribution network. In our data sample, over 80% of the households live in villages with populations less than 5,000 and out of these, about a quarter live in villages with less than 1,000 inhabitants. These small villages are likely to have only a marginal impact on the location of transmission lines, an observation critical to our identification strategy.

We differentiate between connecting to the power grid and the quality of power supply (a measure of outages and average daily hours of supply which are generally correlated).⁶ The density of transmission lines thus determines the quality of power supply, as well as the probability of being connected to a grid. A household situated in a district characterized by a high density of transmission cables is more likely to be connected to the grid, and conditional on being connected, it is more likely to receive higher quality power.

 $^{^6}$ We use the term "quality" of power rather than "reliability" which in the engineering literature has a precise meaning relating to interruptions in power supply.

3 Data and Stylized Facts

The panel data contains two survey waves conducted in 1994 and 2005. The first wave is part of the Human Development Profile of India (HDPI) and covers 33,230 rural households. A share of these households was then re-interviewed for the India Human Development Survey (IHDS) in 2005, which covered 41,554 rural and urban households. The households to be re-visited in 2005 were chosen by first randomly ranking villages, and then fulfilling a variety of conditions needed for the survey to be representative by starting in the first village and then moving down the list of villages. The survey is thus representative at the country level, but not necessarily at smaller geographical units such as a district. The final data consists of a representative panel comprising 9,791 rural households which were interviewed both in 1994 and 2005. Households are asked whether or not they receive electricity, and if they do, information is obtained regarding the quality of the electricity supply. While the definition of power quality varies across rounds, we carry out various robustness tests to ensure that the impact of power quality is correctly identified. The panel also contains a wide variety of information at the individual, household and village level.

Our main variables are constructed as follows. House is a dummy variable which takes a value of 1 if the household owns the house in which the family is currently living. Children, Teen and Adult represent the percentage composition of the household in terms of children (between 0 and 14 years old), teens (between 15 and 21 years old) and adults (older than 22 years old), respectively. Size represents the total number of individuals in the household. Livestock is a dummy variable which takes the value 1 if the household owns farm animals. Hindu is a dummy variable taking a value of 1 if it is a Hindu household and 0 if it belongs to another religious denomination.

Finally, household income is measured as non-agricultural income in per-adult equivalent terms. It is computed by using the OECD definition of equivalence scale, which is given by $1 + 0.7(N_{Adults} - 1) + 0.5N_{Children}$ (OECD, 1982). This expression gives us the number of adult equivalent members of the household. Unfortunately, there is some discrepancy in variable definitions between the two surveys. For example, the question-naire used in 2005 was much more detailed on agricultural income than the one used in 1994. The latter questionnaire also contains questions about losses, while in the former, agricultural income was not allowed to be negative. Therefore, for the sake of our analysis, data on agricultural income could not be included in household total income. This being said, all of our results should be considered as lower bounds of the actual effect. One of the main effects of electrification is on agriculture through the installation of irrigation pumps.

Table 1 presents the household characteristics considered in this paper. Per adult

Table 1: Household Characteristics

	1	994	2	2005	-	
Variable	Mean	$Std.\ Dev.$	Mean	Std. Dev.	$log\Delta$	Obs.
Income per Adult Equivalent	6,472.4	7,639.8	8,722.5	10,895.0	0.130	9791
Household Size	6.577	3.322	5.546	2.759	-0.074	9791
Share of Children	0.341	0.21	0.3	0.222	-0.056	9791
Share of Teenager	0.146	0.166	0.136	0.173	-0.031	9791
Share of Adults	0.513	0.187	0.564	0.216	0.041	9791
Hindu	0.854	0.353	0.848	0.359	-0.003	9791
Home Ownership	0.968	0.175	0.98	0.139	0.005	9791
Livestock Ownership	0.704	0.456	0.845	0.361	0.079	9791

Notes: The table reports descriptive statistics of the household controls used in the estimation. The average of the indicator variables Hindu, Home and Livestock. Child, Teen and Adult represent percentages. The column Δ shows the change between 1994 and 2005.

equivalent income increased significantly over the period studied, and so did its volatility. On average, there were 6.6 individuals per household, while this number declined to 5.5 by 2004. About one third of the individuals within households were children, 14 percent were teenagers and slightly more than half of the household population were adults. The share of adults has increased over time, and the share of children as well as teenagers has decreased. Approximately 85 percent of the households in the sample are Hindus. With respect to asset ownership, almost the entire sample owned the dwelling they were living in. In 1994, 70 percent of the sample owned livestock while this number increased to 85 percent in 2005.

3.1 Household Electrification

Table 2 reports household grid connection rates for the 18 states surveyed in this study. The electrification rate at the national level increased during the study period from 53 percent to 69 percent. All states surveyed exhibit an increase in connection rates. States with relatively high income levels, such as Andhra Pradesh, Tamil Nadu and Kerala have done better in terms of grid connections. Poorer states, such as Bihar, Orissa and Uttar Pradesh have under-performed.⁷

We next present the 1994 district-level electrification rates in Figure 2. The districts that are not in our sample are presented in white, and the variation in electrification rates across surveyed districts are presented with different shades. As can be seen from the map, the higher rural electrification rates are observed around Delhi and Gujarat, as well as the southern regions of India such as Andhra Pradesh and Tamil Nadu. While the survey used in this paper focuses on rural areas and isolated small villages, the

⁷This relationship between income and electrification rates is sharper at the state level. Figure A.3 shows that there is a strong and positive correlation between state-level income and electrification rates in 2005. The same relationship holds for 1994, not presented here for brevity.

higher electrification in the relatively industrialized areas indicates that village electrification benefits from industrial activities, most likely due to better infrastructure. The electrification rates in 2005, presented in 3 also show a similar structure.

In Figure 4 we present changes in the electrification rate between 1994 and 2005 that is used in the empirical strategy. Surprisingly, a few districts experienced reductions in electrification rates. This may be due to theft or deterioration of infrastructure, which is quite common in India (Balachandra, 2011). The increase in electrification rates mainly occurred in the Eastern states such as Jharkhand, rather than the industrial states that had high initial electrification rates. While controlling for all district-level factors through a fixed effects specification in a first difference model, we also run a series of robustness checks to ensure that the exclusion restrictions are not violated by the differential changes in electrification rates across regions.

Table 2: Household Electrification Rate by State (%)

	Elec	trificati	on	Qualit	y of Suj	pply
	1994	2005	Δ	1994	2005	Δ
Andhra Pradesh	54	86	32	31	78	47
Bihar	10	21	11	7	27	20
Chhattisgarh	36	68	32	27	57	30
Gujarat	71	87	16	44	80	36
Haryana	82	90	8	45	60	15
Himachal Pradesh	91	97	6	69	99	30
Jharkhand	20	64	44	17	71	54
Kerala	76	89	13	44	96	52
Maharashtra	67	78	11	40	80	40
Madhya Pradesh	62	71	9	41	45	4
Orissa	17	34	17	21	94	73
Punjab	77	95	18	45	64	19
Rajasthan	49	54	5	36	49	13
Tamil Nadu	66	88	22	50	98	48
Uttar Pradesh	19	39	20	19	50	31
Uttarakhand	33	66	33	19	61	42
West Bengal	13	36	23	15	86	71
India	53	69	16	39	71	32

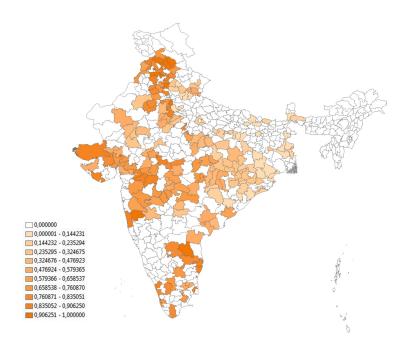
<u>Notes</u>: The table reports the percentage of households that reported a grid connection in 1994 and 2005. Assam is not reported due to low number observations in this state. Only three household were surveyed in this state.

The quality of power supply at the household level is constructed as follows. The 1994 wave asked if household power supply was continuous, if the household experienced on average one or two or more than two power outages per week. The electricity supply is characterized as high quality if the household reported no power outages. We assume

the threshold for high quality to be 18 hours per day.⁸ The 2005 wave asked how many hours of electricity per day does the household receive on average. The quality of the power supply variable is then defined as a variable that takes the value of zero when the household is not connected to the grid, 0.5 when it is connected but receives a low quality power supply, and 1 when it is connected and receives high quality power supply.

The second panel (Table 2) shows the average quality index across states. At the national level, the quality index increased by 32 percentage points between 1994 and 2005, which is approximately an 80% increase. There is a correlation between the increase in grid connection rates and the quality index, with a correlation coefficient of 0.48. For example, in Orissa, the grid connection rate doubles while the quality index increased by 73 percentage points. In Madhya Pradesh, grid connection rate increased only by 9 percentage points and the quality index by 4 points.

Figure 2: Electrification rate in the survey districts in 1994.



 $\underline{\text{Notes}}$: The information on electrification rates is obtained from the Human Development Profile of India. Districts left white are not part of the survey sample. Source: ESRI ArcGIS World Package and Geocommons.

⁸The model is re-estimated with threshold values of 16, 17, 19 and 20 hours. The main results hold with slight changes in the magnitudes of the effects. 18 hours is preferred in the paper as it reflects the productive hours of an agricultural household.

0,000000 0,000001 - 0,144231 0,144232 - 0,235294 0,235295 - 0,324675 0,324676 - 0,476923 0,476924 - 0,579365 0,579366 - 0,658537 0,658538 - 0,760870 0,760871 - 0,835051 0,835052 - 0,906250 0,906251 - 1,000000

Figure 3: Electrification rate in the survey districts in 2005.

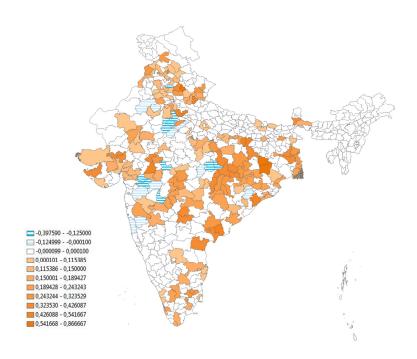
Notes: The information on electrification rates is obtained from the India Human Development Survey. Districts left white are not part of the survey sample. Source: ESRI ArcGIS World Package and Geocommons.

3.2 The Transmission Network

In order to construct the variable for the length of the transmission cables within each district, maps of the Indian transmission network published by the Indian Ministry of Power are used. These maps are then superimposed on the map of Indian districts from Census 2001 using ArcGIS (Government of India, 2001). Next, transmission cables are split along the borders of the districts in order to measure their exact length within each district. The density of the transmission network is then computed by taking the total length of all the line segments within each district, and dividing it by the district surface area. The deviation of this measure from the national average is then defined as the normalized transmission cable density.

Figure 5 presents the map of this normalized cable density measure for 1994, where the darker shades represent higher cable density. The districts marked white are not part of the transmission network (they have no transmission cables). If a household is located in a district characterized by a positive density, then the probability that the household is connected to the grid and receives high quality electricity is higher, while

Figure 4: Variation in electrification rate in the survey districts between 1994 and 2005.



Notes: The information on electrification rates is obtained from the Human Development profile for 1994 and India Human Development Survey for 2005. The figure shows the change in the electrification rate in each district. Districts left white are not part of the survey sample. Source: ESRI ArcGIS World Package and Geocommons.

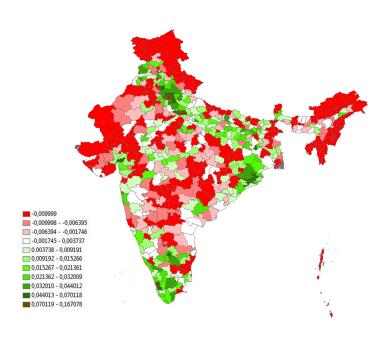
the opposite is true for negative values (Brown and Sedano, 2004). Similarly, Figure 6 shows the normalized cable density for 2005.

In order to understand how the new infrastructure was distributed across the country during the 11 years of the study period, the difference in the normalized cable density measures is presented in Figure 7. Note that this does not reflect the increase in the overall increase cable density as it is normalized for both years. The negative values indicate that the gap between the district and the national average of transmission cable density has increased, therefore the district did not receive a equiproportionate share of transmission investment given its area. The positive values indicate that the district received more than its equiproportionate share of infrastructure investment. The figure shows that the positive values are clustered in the eastern region, areas close to Delhi and in the state of Gujarat. This implies that political connections may be an important factor in attracting infrastructure investment, which we discuss later in the paper.

⁹The distribution of cables prior to normalization and the cable network are presented in Figure A.4.

Table 3 summarizes information on the density of transmission cables by state. During the study period, the transmission network expanded by about 23% (more than 7,000 km). However, there is considerable heterogeneity across states. For example, West Bengal saw a decline of 3.3% while the network size in Gujarat increased by 96.7%. The states which saw their network size increase (Gujarat, Maharashtra, Punjab, Rajasthan and Uttar Pradesh) also experienced large gains in mean household income. In order to compare the infrastructure with other countries, the total cable length as well as total length divided by area are presented for China and the U.S. in the last two lines. Compared to India, the total length is approximately 4 times higher in China and 6.4 times higher in the U.S. When we take the total area of the countries into account, India has 12 km of transmission cable per $1000 \ km^2$, while this number is 16 km in China and 26 km in the United States.

Figure 5: Normalized density of transmission cables per km^2 by district, 1994



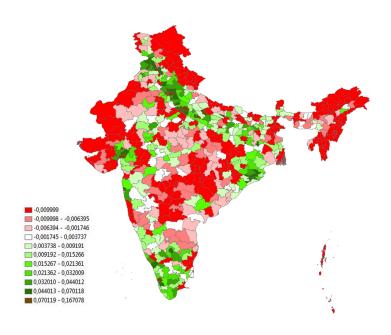
Notes: Data obtained from the Indian Ministry of Power. Cable length within districts is measured using ESRI ArcGIS World Package and Geocommons. The figure represents the cable length divided by the area of the district, normalized with respect to the national mean.

Table 3: Density of transmission cables

	tal # of district ace surveyed
	ın Total surface
	Real mean income
2005	Length/km ²
	$Total\\ length~(km)$
	Real mean income
1994	$Length/km^2$
	$Total \\ length \; (km)$

Notes: Data for transmission cables are obtained from the Indian Ministry of Power. Length within the states is measured using ArcGIS. Real mean income is calculated using the households included in the sample. ¹Yang (2006). ²Brown and Sedano (2004)

Figure 6: Normalized density of transmission cables per km^2 by district, 2005

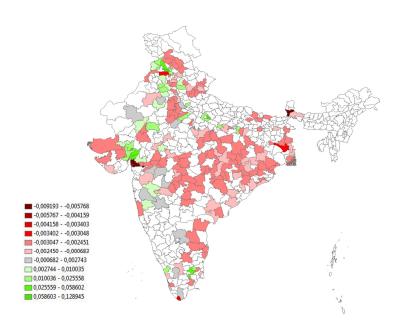


Notes: Data is obtained from the Indian Ministry of Power. Cable length within districts is measured using ESRI ArcGIS World Package and Geocommons. The figure represents the cable length divided by the area of the district, normalized with respect to the national mean.

3.3 Description of data

In order to provide a descriptive background for the impact of electrification on household income, we take advantage of the panel aspect of the survey and compare households that changed their electrification status from 1994 to 2005. In particular, we compare households that change their status from "not connected" to "connected" and households for which the status remained the same. Further, those with electricity connection are divided between low and high quality electricity, where quality is measured by the number of hours per day as described above. By following households through changes in electrification status, we test whether or not the change in their real income was significant. The results presented in Table 4 show that among 4,613 households in our sample in 1994 who were not connected to the grid, 2,447 remained unconnected, 1,118 were connected but received low quality electricity supply and 978 were connected but

Figure 7: Difference in the normalized density of transmission cables per km^2 between 1994 and 2005.



Notes: Data is obtained from the Indian Ministry of Power. Cable length within districts is measured using ESRI ArcGIS World Package and Geocommons. The figure represents the cable length divided by the area of the district and normalized with respect to the national mean. Districts left white are not part of the survey sample.

received a high quality supply.¹⁰ While the real incomes of the households who remained unconnected significantly increased by 1.9%, this increase was higher at 2.8% for households that received low quality electricity, and even higher at 3.4% for households that received high quality electricity.¹¹ This suggest that the incomes of connected households increased more than the unconnected households, without establishing a causal relationship or ruling out other factors that can affect household income.

The households that were connected, but received a low quality supply in 1994 also experience a differential change. Those that continued to receive lower quality electricity had a 2.1% increase, and those who experienced improvements in electrification and moved to a high quality electricity supply had a higher increase in their income, at 3.6%. Households that had a high quality supply in 1994 and continued to receive

¹⁰Approximately three quarters of the households report data on the hours of electricity they received. The full sample is used in the analysis wherever possible.

¹¹These differences are statistically significant.

Table 4: Household Electrification and Quality of Supply

	Not Connected (2005)	Connected Low Quality Supply (2005)	Connected High Quality Supply (2005)	Total
Not Connected (1994)				
Number of Households	2,447	1,188	978	4,613
Change in Income (%)	1.9***	2.8***	3.4***	,
Connected, Low Quality Supply (1994)				
Number of Households	532	1,762	1,848	4,142
Change in Income (%)	0.09	2.1***	3.6***	,
Connected, High Quality Supply (1994)				
Number of Households	83	299	653	1.035
Change in Income (%)	2.3	2.1**	4.6***	,
Total	3,062	3,249	3,479	9,790

Notes: The table reports the absolute number of households moving from one state to the other between 1994 and 2005 and the variation in the household total income, not including income from agriculture. Asterisks report statistics for a test of two means between 1994 income and 2005 income. *** p<0.01, ** p<0.05, * p<0.1. High Quality Supply is defined as more than 18 hours of electricity per day on average in 2005; in 1994, households were asked about the number of outages experienced per week.

a high quality supply in 2005 had the highest increase in their real incomes at 4.6%. However, only 653 out of 9,790 households are in this group, representing 6.6% of the sample.

The data also show that some households experienced deterioration in their electricity connection. Some 532 households with low quality electricity and 83 households with high quality electricity in 1994 lost their electricity connection. These households did not experience a significant increase in their income. The reduction in electrification coverage in some states were shown in Figure 4. Finally, 299 households switched from high quality to low quality electricity. While their incomes increase over the time period, it was much lower than for households that continued to receive high quality supplies.

4 Empirical Approach and Results

Let Y denote the logarithm of household income and X be a vector of household-level controls. Define T to be the treatment variable, which represents the quality of power supply received by the household. Our specification in levels takes the following form:

$$Y_{idt} = \alpha + \delta_i + \delta_t + \delta_{dt} + \beta T_{idt} + \gamma X_{idt} + \varepsilon_{idt}$$
 (1)

where subscripts i, t and d denote household, time and district, respectively. Thus, δ_i , δ_t and δ_{dt} represent household, time and district-time fixed effects. ε_{idt} is the error term. The coefficient of interest is β . In order to eliminate the household fixed effect term and other time invariant factors, we difference over time to get

$$\Delta Y_{idt} = \Delta \delta_t + \Delta \delta_{dt} + \beta \Delta T_{idt} + \gamma \Delta X_{idt} + \Delta \varepsilon_{idt}$$
 (2)

Our panel is composed only of two waves, thus we can re-write equation (2) as

$$\Delta Y_{idt} = \alpha + \delta_d + \beta \Delta T_{idt} + \gamma \Delta X_{idt} + \Delta \varepsilon_{idt}$$
 (3)

The quality of power supply could be improving in districts that tend to grow faster than other districts. In this case $\Delta \varepsilon_{idt}$ is correlated with ΔT_{idt} and an OLS estimation of equation (3) does not identify the causal effect of an increase in the quality of power supply. We deal with this endogeneity issue by using an instrumental variable approach, following the seminal work of Angrist and Imbens.¹² The quality of power supply is instrumented with the increase of the density of transmission cables within the district relative to the national average, multiplied by the initial level of quality.

Proximity to the transmission network implies a higher probability of being connected to it and also be closer to generation points and therefore a higher probability of receiving better power supply. Given the inability to locate our households within the district for confidentiality reasons, multiplying the improvement of the transmission network by the initial quality of power supply of a household accounts for the heterogeneity in the distribution of households within a district. The construction of this instrument follows the methodology developed by Bartik (1991), and employed by Blanchard and Katz (1992), Bound and Holzer (2000) and Autor and Dugger (2003). The intuition behind the instrument is that, an upgrade of the transmission grid is more likely to affect a household already connected to the grid, yet with a low quality of power supply, than a household which was not connected at all. The instrument ΔZ is therefore constructed in the following way:

$$\Delta Z_{idt} = T_{id0} * \Delta G_{dt} \tag{4}$$

where G denotes the district level normalized density of transmission cables. The system

¹²See, for example, Angrist (1990), Angrist and Imbens (1994) and Angrist (1998).

of equations to be estimated is therefore:

$$\Delta Y_{idt} = \alpha + \delta_d + \beta \Delta T_{idt} + \Delta X_{idt} \gamma + \Delta \varepsilon_{idt} \tag{5}$$

$$\Delta T_{idt} = \varphi + \delta_d + \eta \Delta Z_{idt} + \Delta X_{idt} \gamma + \Delta u_{idt}$$
 (6)

where standard errors in all regressions are clustered at the district level in order to account for within-region heterogeneity. 13 The same methodology is then applied to the the probability of electrification, where T indicates the connection to the grid. 14

The expansion in transmission network is expected to increase the probability of electrification for households that are within a feasible radius of the infrastructure. However, the impact on quality of electricity is less straightforward as there are two potential effects. First, an expansion in the transmission and distribution network could increase the demand for electricity. If this is not complemented by an increase in generation capacity, the quality and reliability of electricity for households could decrease. On the other hand, if the electricity generation capacity is proportionally expanded, then we would observe an increase in the quality. While we are not able to obtain data on the location and capacity of power plants in India, the sign on the first stage would show which of these effects dominate.

The results for the first stage describing the relationship between household electrification and transmission cable density are presented in Table 5. The model described in Equation 6 is first estimated with no controls, then district fixed effects, household demographic controls and household asset controls are included in the model. The results suggest that cable density has a positive and robust impact on electrification. The results in column (1) imply that a one unit increase in the density of transmission lines increases the probability of electrification by approximately 5 percentage points. The impact increases to approximately 10 percentage points with the inclusion of district fixed effects and household controls.

On the right panel, the estimates for the impact on the quality of electricity are shown. They show that a one unit increase in the transmission cable density increases

 $^{^{13}}$ The results hold when we cluster the errors at the state or region level.

¹⁴The condition which needs to be satisfied in order for this identification to be consistent is that $E[T_{id0} * \Delta G_{dt} * \Delta \varepsilon_{idt}] = 0$. We can easily verify this by taking the limit over districts and individuals $\lim_{D,I \to \infty} \frac{1}{D*I} \sum_d \sum_i (T_{id0} * \Delta G_{dt} * \Delta \varepsilon_{idt}) = 0$. Since G_{dt} does not depend on the individual, we can extract it from the sum and re-write it as $\lim_{D,I \to \infty} \frac{1}{D*I} \sum_d \Delta G_{dt} \sum_i (T_{id0} * \Delta \varepsilon_{idt}) = 0$. In order for this to be verified we only need to argue that $\lim_{I \to \infty} \frac{1}{I} \sum_i (T_{id0} * \Delta \varepsilon_{idt}) = 0$, or equivalently that $E[T_{id0} * \Delta \varepsilon_{idt}] = 0$. This means that income shocks occurring between 1994 and 2005 need not to be correlated with the quality of power supply in 1994. While income shocks may be related to the improvement in the quality of power supply or the income trend may be related to the initial quality of power supply, it is difficult to argue why the initial quality should be correlated to shocks, especially over a period of 11 years. Therefore, we are confident in claiming that this condition is indeed satisfied.

the quality index by 15 percentage points.¹⁵ This implies that the increase in electricity demand due to additional transmission cables was not associated with a reduction in quality, and that the increase in demand induced by the expansion in the transmission cables was adequately compensated through increased production. The coefficients on the control variables show that an increase in the household size was associated with a larger probability of electricity connection and a higher quality. On the other hand, an increase in the number of children was associated with lower electricity outcomes while religion and asset ownership did not have a significant impact.

4.1 Electricity and Household Income

We now focus on the the impact of the quality of electricity supply on household income as specified in Equation 5. The results for OLS and IV estimations are reported in Table 6, the latter using the density of transmission cables multiplied by the initial state of electrification as instrument for the quality index. The OLS results with no controls, presented in column (1) indicate that higher electricity quality is associated with a 9 percent increase in household income. Inclusion of the district fixed effects and instrumenting the change in quality with the transmission cable instrument in column (4) increases the coefficient to 87 percent. From Table 2, we see that the quality of supply improved by 32 percentage points on average over the whole country between 1994 and 2005. These numbers imply that household incomes increased by roughly 28 percent over this 11-year period as a result of an increase in the quality of power supply. ¹⁶

We then include the demographic controls, asset controls, individually and together in columns (5) – (10). The results show that both the OLS and the IV coefficients are robust across specifications, with OLS estimates between 7.6% and 8.6%, while the IV estimates range between 87.3% and 89.4%. All the results consistently show that the impact of electricity quality on income is higher after instrumentation. A potential explanation relates to the type of infrastructure reflected in the coefficients. The IV estimates reflect the impact of the transmission network, while the OLS estimates incorporate the effect of both distribution and transmission networks. The smaller OLS estimates may thus be associated with the efforts of state governments to provide power to poorer areas. The transmission network, on the other hand, substantially improves the household's probability of receiving electricity, but it does not reflect the potential downward bias arising from state policies.¹⁷

¹⁵Note that the number of observations declines in a panel due to missing information on power quality. ¹⁶This number is obtained by multiplying the coefficient 0.887 by the improvement in the quality index of power supply, 0.32.

¹⁷In order to assess whether or not the results are robust to the choice of the 18 hour threshold, the main model is estimated with 16, 17, 19 and 20 hours as thresholds for high quality power supply. These results shown in Table A.1 suggest that our estimates are not very sensitive to this choice.

The coefficients on the other control variables have the expected signs. Per-adult equivalent household income is negatively associated with household size, with each additional individual reducing income by roughly 3 percent. The same is true for the proportion of children which also has a negative impact on income. The coefficient on livestock ownership is negative, indicating that the household is less likely to receive income from non-agricultural activities if it owns livestock assets. Home ownership has an insignificant effect as almost the entire sample are home owners, and the real estate rental market is likely to be very small or nonexistent in these small villages in a developing country.

These results imply that the impact of electrification on households cannot be considered independently of the quality of electricity. While the initial connection may be important in the sense that it can induce reallocation of labor and capital within the household, this impact will increase with quality. This highlights the importance of providing a high quality supply of power, as the potential benefits of electricity are not completely realized by just providing a grid connection and low quality power.

We now estimate the same model to investigate the impact of a connection to the grid on household income. The results are presented in Table 7. OLS results predict an impact of a connection to the grid of 6.7%, while IV results predict an impact of 55.4%. As before, instrumentation increases the impact several fold, implying that receiving electricity augments non-agricultural household income by more than 50%, and the estimate is largely robust to the inclusion of various household controls. Given that electrification rates increased by 16 percentage points between 1994 and 2005, these results imply that non-agricultural household incomes increased by 8.9% over this 11-year period as a result of increased access to electricity. The estimates are generally robust to the inclusion of household controls for demographic characteristics and asset ownership.

4.2 Further Robustness Checks

The robustness of the results are tested by using an alternative measure for the instrument, and by accounting for various factors that may violate the exclusion restriction. These results are presented in Tables 8 through 11 for the baseline specification with demographic controls, asset controls and district fixed effects. All standard errors are clustered at the district level. In each table, the first two columns represent the results of the baseline specification in columns (8) and (9) of Table 6 for the effect on electrification quality, and Table 7 for the effect of grid connection on household income.

The previous specification defines the transmission cable density as the total length of transmission lines in a district divided by the area of the district. It may be the case that larger districts have unpopulated areas with no cables, which can make our density measure smaller, but the populated areas may be well electrified. We thus alternatively define the transmission variable as the total length of cables per person instead of per square mile. The actual population of the district is used for the computation. These results are presented in columns (3) and (4) of Tables 8 and 9 for the effect of electrification quality and grid connection, respectively. The magnitude of the estimates is largely robust to this change, while the first stage F-statistics are estimated to be considerably larger. This indicates that cable length per person is a stronger instrument for the quality and grid connection, while the estimated impact on household income was not affected by this change.

Because we are focusing on a relatively long time period between 1994 and 2005, the results may be contaminated by other reforms that took place over these eleven years with differential impacts on some households. Specifically, the long run effects of trade liberalization that took place in 1991 would be fully reflected in the markets by 1994, and continue through the period of this study. The reduction in prices due to tariff cuts may have specifically affected agricultural households that live in coastal states. These states also tend to be more industrial and thus have a higher density of transmission cables, which can be seen from the Figure A.4. As a robustness check, we present results by excluding the coastal states of Gujarat, Maharashtra, Karnataka, Tamil Nadu, Andhra Pradesh, Orissa and West Bengal in order to make sure we are not picking up this effect. These results are presented in columns (5) and (6) of Tables 8 and 9. The magnitude of the coefficient for electricity quality increased to 0.985, while the magnitude for grid connection increased to 0.853 when compared to the baseline specification. The larger estimates imply that the effect is higher for inland states, and that the previous estimates may represent a lower bound for the impacts.

The political connections of the local government may bias our results if they actively lobby for higher investments in the transmission network. The regions with stronger political connections to the federal government may be more likely to receive electricity, and also grow faster due to better access to other resources as well. In order to check whether this is driving our results, we have excluded the states whose prime minister in 1994 belonged to the same party as the Indian prime minister at the federal level. The prime ministers of Andhra Pradesh, Assam, Haryana, Himachal Pradesh, Kerala, Maharashtra, Madhya Pradesh and Punjab belonged to the Indian National Congress party in 1994, that was in power in the federal government. We thus repeated the estimation by excluding these states. The results shown in columns (7) and (8) of Tables 8 and 9 suggest that the estimates are higher in the states that did not have the same party in power as the federal government. We can therefore conclude that political

¹⁸Ural Marchand, 2012.

alliance between the state and the federal government does not play a role in the decision to locate transmission lines.

In India, there are large disparities across states in terms of income levels and rates of economic growth. If high growth states experienced a relatively smaller improvement in electrification rates, then the estimates could be biased. We next excluded the three states where income per capita has increased the most between 1994 and 2005 according to the HDPI-IHDS surveys, i.e. Haryana, Himachal Pradesh and Kerala. The results are presented in columns (9) and (10). In addition, we excluded the five states with the highest growth rates, i.e. Haryana, Himachal Pradesh, Kerala, Punjab and Uttarakhand. These results are shown in columns (11) and (12). Both set of estimates are higher when high-growth states are excluded, implying that the marginal impact of grid connection and quality is higher in slower-growing areas.

The model is also estimated by excluding households that live in districts that are relatively urbanized. The incomes in these districts may be growing faster than others for reasons that are not related to electricity. The quality of power supply may be higher in these areas due to a higher number of power plants or higher investments in power transmission. We test for this by excluding the districts that include the 10 largest cities. In addition, we hand-select the districts that are neighbors of these districts and also exclude them in order to capture the full extent of their area of influence. The 10 biggest cities in India in decreasing order of size, are: Mumbai, Delhi, Bangalore, Hyderabad, Ahmedabad, Chennai, Kolkata, Surat, Pune, and Jaipur. His test captures parts of the transmission network that is likely not impacted by demand for urban and industrial centers. They are likely to represent the most cost effective route between a production site and a high demand pole or between two high demand poles. Thus the outcomes we observe should come from the transmission network only through electrification.

The results are presented in columns (3) and (4) of Table 11 for quality of electricity and in Table 11 for grid connection. Excluding the largest cities increases the impact of grid connection as well as electricity quality. The urbanized areas already have high grid connection rates and good quality electricity in the initial period. The increased coefficients after excluding urban areas imply that the marginal returns to additional improvements are smaller at high levels of electrification. The same exercise is repeated by excluding only the districts with the 10 largest cities, presented in Columns (5) and (6). Columns (7) and (8) exclude districts with the 15 largest cities, which also

¹⁹This results in elimination of the following districts from the sample: Maharashtra: the districts of Dhule, Pune, Ahmednagar, Solapur and Satara; Haryana: the districts of Sonipat, Gurgaon and Faridabad; Karnataka: the district of Bangalore rural, Tumkuru and Kolar; Tamil Nadu: the districts of Dharmapuri and Kancheepuram; Andhra Pradesh: the district of Medak; Gujarat: the districts of Ahmedabad, Kheda, Gandhinagar, Bharuch, Mahesana and Surendernagar; Rajasthan: the districts of Sikar, Alwar, and Sawai Madhopur; and finally West Bengal, the districts of South 24 Parganas and North 24 Parganas.

includes Lucknow, Kanpur, Nagpur, Indore and Thane. Similarly, columns (9) and (10) exclude districts with 20 largest, which also includes Bhopal, Visakhapatnam, Pimpri-Chinchwad, Patna and Vadodara. The magnitude of these coefficients are similar to the baseline estimation.

5 Concluding Remarks

The provision of reliable electricity is key to economic growth in the developing world. Previous studies of the effect of electrification have focused on new connections to the grid. In this paper we examine the effect of a grid connection as well as the quality of power supply on household incomes in rural India. These estimates can help determine spending priorities, for example between improving existing infrastructure or extending the grid to cover new areas.

Capturing the impact of electrification on income is difficult because of reverse causality and other endogeneity concerns. We use the variation in infrastructure, i.e. transmission lines, in order to isolate the causal relationship going from electricity provision to economic outcomes. We use a nationally representative dataset to show that the impact of electrification on households cannot be considered independently of the quality of the electricity supply. A grid connection increases non-agricultural incomes of rural households by about 9 percent during the study period (1994-2005). However, higher quality of electricity (in terms of fewer outages and more hours per day) increases non-agricultural incomes by about 29 percent during the same period. This highlights the importance of providing a high quality supply of power, as the potential benefits of electricity are not completely realized by only connecting households to the grid.

Our results suggest that policies that aim to provide reliable electricity to households may bring about significant economic benefits. These estimates can be used to perform cost-benefit analysis on infrastructure improvements, which can guide scarce capital assets into sector that yields the highest returns.

Although several recent studies have examined the impact of grid connections, this is the first paper providing an estimate of the effect of the quality of electricity supply on household incomes. However, significant work is still needed in order to fully understand the impact of electrification on welfare. Because of data limitations we were not able to precisely disentangle the various channels through which electrification may operate. For example, it will be useful to see how households change their labor allocation when they get reliable electricity and understand the spillover effects when the whole village gets connected to the grid.

Table 5: First stage regression on grid connection and quality index

		Gric	Grid connection	uc	Dependent variable:	riable:	Quali	Quality of Supply	ly	
	(1)	(2)	(3)	(4)	(5)	(9)	(7)	(8)	(6)	(10)
$\Delta Cable *Connection$	0.051***	0.051*** 0.100*** 0.010) (0.024)	(0.024)	* 0.100*** (0.024)	0.096*** 0.100*** 0.096*** (0.024) (0.024) (0.024)					
$\Delta Cable*Quality$						0.073** (0.031)	0.152*** (0.053)	$\begin{array}{cccccccccccccccccccccccccccccccccccc$	0.152*** (0.053)	0.149*** (0.052)
$\Delta Size$			0.010^{***} (0.002)	×	0.010*** (0.002)			0.005^{***} (0.002)		0.005*** (0.002)
$\Delta Children$			-0.104^{***} (0.020)		-0.104^{***} (0.020)			-0.074*** (0.015)		-0.073*** (0.015)
$\Delta Hindu$			0.024 (0.035)		0.024 (0.035)			-0.013 (0.026)		-0.012 (0.026)
$\Delta Home$				-0.009 (0.033)	-0.016 (0.032)				-0.000 (0.023)	-0.004 (0.023)
$\Delta Livestock$				0.014 (0.012)	0.005 (0.012)				0.017 (0.011)	0.012 (0.011)
$District\ F.E.$	ou	yes	yes	yes	yes	ou	yes	yes	yes	yes
Observations	9,785	9,785	9,785	9,785	9,785	7,181	7,181	7,181	7,181	7,181
F-stat first stage K-P F-stat (weak ident)	26.57 26.60	17.32 17.33	16.34 16.35	17.28 17.28	16.34 16.34	5.51	8.30	8.12	8.26 8.26	8.08

Notes: All estimations contain a constant. Standard errors in parentheses are clustered at the district level. *** p<0.01, *** p<0.05, * p<0.1. Instrument is the variation in the district deviation from the national density of transmission cables per squared kilometer multiplied by the initial state of the treatment. Hindu, Home and Livestock are dummy variables.

Table 6: Effect of quality of power supply on income

				Lo	Dependen	Dependent variable: Log of income per adult equivalent				
	OLS	IV_C	STO	IV_C	OLS	IV_C	STO	IV_C	STO	IV_C
	(1)	(2)	(3)	(4)	(5)	(9)	(7)	(8)	(6)	(10)
Δ Quality	0.092* (0.054)	1.133 (0.813)	0.081 (0.049)	0.867**	0.076 (0.047)	0.873** (0.391)	0.086*	0.898** (0.413)	0.079*	0.894** (0.395)
$\Delta Size$					-0.030** (0.005)	$-0.030^{***}-0.034^{***}$ (0.005) (0.006)			-0.028*** (0.005)	-0.033*** (0.006)
Δ Children					-0.361** (0.052)	$ \begin{array}{cccccccccccccccccccccccccccccccccccc$			-0.368*** (0.052)	-0.303*** (0.058)
$\Delta Hindu$					0.145 (0.096)	0.147 (0.094)			0.140 (0.096)	0.141 (0.094)
$\Delta Home$							$0.074 \\ (0.068)$	0.074 (0.073)	0.078 (0.067)	0.081 (0.072)
$\Delta Livestock$							-0.105** (0.026)	$\begin{array}{cccccccccccccccccccccccccccccccccccc$	-0.085*** (0.025)	
$District\ F.E.$	ou	no	yes	yes	yes	yes	yes	yes	yes	yes
Observations	7,181	7,181	7,181	7,181	7,181	7,181	7,181	7,181	7,181	7,181
F-stat first stage K-P F-stat (weak ident)		5.51 5.51		8.30		8.12 8.12		8.26 8.26		8.08 8.08

Notes: All estimations contain a constant. Standard errors in parentheses are clustered at the district level. **** p<0.01, ** p<0.05, * p<0.1. Income represents total household income excluding agriculture expressed in per adult equivalent terms. The instrument used is density of transmission cables multiplied by the initial quality of power supply. Harden and Liveschock are dummy variables. The Quality index takes a value of 0 for no electricity, 1 for high quality supply and 0.5 for low quality. The threshold for high quality power supply in 2005 is fixed at an average of 18 hours of power per day.

26

Table 7: Effect of grid connection on income

				Log	Dependent variable: of income per adult equiv	Dependent variable: Log of income per adult equivalent				
,	STO	IV_C	STO	IV_C	STO	IV_C	STO	IV_C	STO	IV_C
	(1)	(2)	(3)	(4)	(5)	(9)	(7)	(8)	(6)	(10)
$\Delta Grid$ connection	0.060*	0.823* (0.432)	0.064**	0.507** (0.253)	0.066**	0.534** (0.258)	0.067**	0.538**	0.067**	0.554** (0.263)
$\Delta Size$					-0.032^{***} (0.004)	$-0.032^{***}-0.037^{***}$ (0.004) (0.005)			-0.031^{***} .	-0.036*** (0.005)
$\Delta Children$					-0.372^{***} (0.045)	$-0.372^{***}-0.318^{***}$ (0.045) (0.052)			-0.377*** (0.045)	-0.322*** (0.052)
$\Delta Hindu$					0.104 (0.080)	0.090 (0.078)			0.101 (0.080)	0.086 (0.077)
$\Delta Home$							0.049 (0.052)	0.053 (0.056)	0.047 (0.051)	0.055 (0.056)
$\Delta Livestock$							-0.105** (0.023)	$\begin{array}{cccccccccccccccccccccccccccccccccccc$	-0.081^{***} (0.022)	
$District\ F.E.$	ou	ou	yes	yes	yes	yes	yes	yes	yes	yes
Observations	9,785	9,785	9,785	9,785	9,785	9,785	9,785	9,785	9,785	9,785
F-stat first stage K-P F-stat (weak ident)		26.57 26.60		17.32 17.33		16.34 16.35		17.28 17.28		16.34 16.34
		č								

Notes: All estimations contain a constant. Standard errors in parentheses are clustered at the district level. *** p<0.01, ** p<0.05, * p<0.01. Income represents total household income excluding agriculture expressed in per adult equivalent terms. Grid connection is instrumented using the district deviation from the national density of transmission cables per squared kilometer multiplied by the initial electrification condition. Hindu, Home and Livestock are dummy variables.

Table 8: Effect of quality of power supply on income – robustness tests

					Log	Dependent variable: Log of income per adult equivalent	variable: .dult equivalen	ę				
	$\begin{array}{c} \text{Baseline} \\ OLS & I \end{array}$	IV_C	Population OLS IV	ation IV_C	NoCoast OLS IV	$\frac{\mathrm{oast}}{IV_C}$	Political Conn. OLS IV_C	Conn. IV_C	$\begin{array}{cc} 3 \text{ high growth} \\ OLS & IV_C \end{array}$	$_{IV_C}^{\rm growth}$	$\begin{array}{ccc} 5 \text{ high growth} \\ OLS & IV_C \end{array}$	$_{IV_C}^{\rm growth}$
	(1)	(2)	(3)	(4)	(5)	(9)	(7)	(8)	(6)	(10)	(11)	(12)
Δ Quality	0.079*	0.894** (0.395)	0.079*	0.882** (0.346)	0.083	0.985**	0.163 (0.075)	1.643***	0.096*	1.039**	0.113**	1.075**
$\Delta Size$	_0.028***_ (0.005)	*_0.033*** (0.006)	_0.028***_ (0.005)	*_0.032*** (0.006)	_0.029***_ (0.005)	*_0.034*** (0.007)	-0.033*** (0.003)	*_0.039*** (0.004)	-0.031^{***}	*_0.036*** (0.006)	-0.030***_ (0.006)	*_0.035*** (0.006)
Δ Children	-0.368*** (0.052)	$-0.368^{***}-0.303^{***}$ (0.052) (0.058)	-0.368***_ (0.052)	*_0.304*** (0.057)	-0.324^{***} (0.067)	*_0.268*** (0.069)	-0.379*** (0.039)	*_0.256*** (0.046)	-0.389***_ (0.056)	*_0.308*** (0.062)	-0.396^{***} (0.061)	* -0.312 ** * (0.066)
$\Delta Hindu$	0.140 (0.096)	0.141 (0.094)	0.140 (0.096)	0.141 (0.094)	0.173 (0.144)	0.181 (0.138)	0.018 (0.097)	-0.032 (0.050)	0.126 (0.112)	$0.142 \\ (0.107)$	0.088 (0.145)	0.094 (0.136)
$\Delta Home$	0.078 (0.067)	0.081 (0.072)	0.078 (0.067)	0.081 (0.072)	0.058 (0.102)	$0.062 \\ (0.107)$	0.127 (0.076)	0.170** (0.073)	0.106 (0.070)	0.116 (0.079)	0.100 (0.074)	0.113 (0.083)
$\Delta Livestock$	-0.085^{**}	$-0.085^{***}-0.097^{***}$ (0.025) (0.028)	-0.085**	*_0.097*** (0.028)	-0.140^{***} (0.038)	* _0.165 * ** (0.043)	-0.061 (0.028)	-0.084^{***} (0.032)	-0.046^{*} (0.026)	-0.057* (0.031)	-0.043 (0.028)	-0.055^{*} (0.033)
$District\ F.E.$	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes
Observations	7,181	7,181	7,181	7,181	4,422	4,422	2,895	2,895	5,660	5,660	5,094	5,094
F-stat first stage K-P F-stat (weak ident)		8.08		43.52 43.54		7.40		15.42 4.63		6.90		7.08

Notes: All estimations contain a constant. Standard errors in parentheses are clustered at the district level. *** p<0.01, *** p<0.01, *** p<0.01. Income represents total household income excluding agriculture expressed in per adult equivalent terms. The instrument used is density of transmission cables multiplied by the initial quality of power supply. Hindu, Home and Livestock are dummy variables. The Quality index takes a value of 0 for no electricity, 1 for high quality supply and 0.5 for low quality. The threshold for high quality power supply in 2005 is fixed at an average of 18 hours of power per day.

28

Table 9: Effect of grid connection on income – robustness tests

					Log	Dependent variable: Log of income per adult equivalent	variable: .dult equivalen	÷				
	Baseline OLS	$_{IV_C}$	Population OLS IV_Q	IV_C	$\begin{array}{c} \text{NoCoast} \\ OLS & II \end{array}$	$_{IV_{C}}^{\mathrm{ast}}$	Political Conn. OLS IV_C	Conn. IV_C	$\begin{array}{cc} 3 \text{ high growth} \\ OLS & IV_C \end{array}$	$\stackrel{ m rowth}{IV_C}$	$\begin{array}{ccc} 5 \text{ high growth} \\ OLS & IV_C \end{array}$	$_{IV_C}^{ m rowth}$
	(1)	(2)	(3)	(4)	(5)	(9)	(7)	(8)	(6)	(10)	(11)	(12)
Δ Grid connection	0.067**	* 0.554** (0.263)	0.067**	0.566**	0.085**	0.853**	0.086 (0.065)	0.852***	0.067**	0.681** (0.287)	0.065**	0.752** (0.296)
$\Delta Size$	-0.031^{***}	$-0.031^{***}-0.036^{***}$ (0.004) (0.005)	-0.031^{***} (0.004)	$-0.031^{***}-0.036^{***}$ (0.004) (0.005)	-0.027***	$0.027^{***}-0.036^{***}$ (0.005) (0.007)	-0.038*** (0.005)	$0.038^{***}-0.044^{***}$ (0.005) (0.003)	-0.034^{***}	-0.034***_0.040*** (0.004) (0.006)	-0.033** (0.005)	-0.033***_0.040*** (0.005) (0.006)
Δ Children	-0.377^{**} ; (0.045)	$-0.377^{***}-0.322^{***}$ (0.045) (0.052)	-0.377^{***} (0.045)	$-0.377^{***}-0.321^{***}$ (0.045) (0.053)	_0.369***_ (0.059)	0.296*** (0.068)	-0.386^{**} (0.021)	$-0.386^{***}-0.296^{***}$ (0.021) (0.032)	-0.391^{***} (0.048)	0.314*** (0.059)	-0.396^{***}	*_0.314*** (0.063)
$\Delta Hindu$	0.101 (0.080)	0.086 (0.077)	0.101 (0.080)	0.085 (0.077)	0.153 (0.133)	0.153 (0.121)	0.041 (0.046)	-0.005 (0.057)	0.084 (0.089)	0.061 (0.086)	0.055 (0.106)	0.020 (0.102)
$\Delta Home$	0.047 (0.051)	0.055 (0.056)	0.047 (0.051)	0.055 (0.056)	0.036 (0.087)	0.036 (0.091)	$0.065 \\ (0.051)$	0.094** (0.044)	0.061 (0.053)	0.074 (0.061)	0.056 (0.055)	0.075 (0.065)
$\Delta Livestock$	-0.081^{**} ; (0.022)	$-0.081^{***}-0.085^{***}$ (0.022) (0.023)	-0.081^{***}	0.085*** (0.023)	-0.131^{***} (0.034)	$-0.131^{***}-0.141^{***}$ (0.034) (0.038)	-0.062* (0.020)	-0.059*** (0.021)	-0.050** (0.023)	-0.052** (0.025)	-0.050** (0.024)	
$District\ F.E.$	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes	
Observations	9,785	9,785	9,785	9,785	5,707	5,707	4,924	4,924	8,173	8,173	7,538	7,538
F-stat first stage		16.34		50.49		10.93		7.31		14.03		

K-P F-stat (weak ident) 16.34 50.51 10.93 10.93 14.04

Notes: All estimations contain a constant. Standard errors in parentheses are clustered at the district level. **** p<0.01, **** p<0.05, *** p<0.05, *** p<0.05, *** p<0.05, ** p<0.05, **

Table 10: Effect of quality of power supply on income (excluding urban centers and cities)

				Log	Dependen; of income per	Dependent variable: Log of income per adult equivalent				
	$\begin{array}{c} \text{Baseline} \\ OLS \end{array} \Gamma$	$_{IV_C}^{\rm eline}$	Urban areas OLS IV_C	$_{IV_C}^{\rm areas}$	$\begin{array}{c} 10 \text{ Cities} \\ OLS & IV \end{array}$	tties IV_C	$\begin{array}{c} 15 \text{ Cities} \\ OLS & IV_C \end{array}$	ities IV_C	$\begin{array}{c} 20 \text{ Cities} \\ OLS & II \end{array}$	ties IV_C
	(1)	(2)	(3)	(4)	(5)	(9)	(7)	(8)	(6)	(10)
Δ Quality	0.079*	0.894** (0.395)	0.062**	1.272**	0.079 (0.048)	0.901**	0.080*	0.892** (0.395)	0.077 (0.048)	0.895**
$\Delta Size$	-0.028*** (0.005)	**_0.033*** (0.006)	-0.030^{***} (0.005)	$-0.030^{***}-0.035^{***}$ (0.005) (0.007)	-0.028^{**}	0.028***_0.032*** (0.005) (0.006)	-0.028** (0.005)		-0.028*** (0.005)	-0.032^{***} (0.006)
Δ Children	-0.368^{***} (0.052)	**_0.303*** (0.058)	-0.346^{**} : (0.048)	$-0.346^{***}-0.259^{***}$ (0.048) (0.095)	-0.365*** (0.052) (* _0.301 * ** (0.058)	-0.366** (0.052)	$-0.366^{***}-0.302^{***}$ (0.052) (0.058)	-0.360*** (0.053)	-0.296*** (0.059)
$\Delta Hindu$	0.140 (0.096)	0.141 (0.094)	0.064 (0.065)	0.085* (0.051)	0.129 (0.096)	0.130 (0.094)	0.129 (0.096)	0.130 (0.093)	0.129 (0.096)	0.130 (0.094)
$\Delta Home$	0.078 (0.067)	0.081 (0.072)	0.090 (0.072)	0.086 (0.091)	0.071 (0.068)	0.070 (0.073)	0.071 (0.068)	0.070 (0.072)	0.074 (0.069)	0.075 (0.074)
$\Delta Livestock$	-0.085*** (0.025)	**_0.097*** (0.028)	-0.088* (0.035)	-0.116^{***} (0.034)	-0.088*** (0.026)	0.088***_0.099*** (0.026) (0.029)	-0.088** (0.026)	$-0.088^{***}-0.099^{***}$ (0.026)	-0.088^{***} (0.026)	-0.097*** (0.029)
$District\ F.E.$	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes
Observations	7,181	7,181	6,072	6,072	7,023	7,023	7,018	7,018	6,913	6,913
F-stat first stage		8.08		24.81		8.01		8.10		8.20

Table 11: Effect of grid connection on income (excluding urban centers and cities)

				Log	Dependen	Dependent variable: Log of income per adult equivalent	دب دب			
	$\begin{array}{c} \text{Baseline} \\ OLS \end{array} \Gamma$	$_{IV_C}$	$\begin{array}{ccc} \text{Urban areas} \\ OLS & IV_C \end{array}$		$\begin{array}{c} 10 \text{ Cities} \\ OLS & II \end{array}$	ties IV_C	$\frac{15~\mathrm{Cities}}{OLS}~II$	ities IV_C	$\begin{array}{cc} 20 \text{ Cities} \\ OLS & IV_C \end{array}$	ties IV_C
	(1)	(2)	(3)	(4)	(5)	(9)	(2)	(8)	(6)	(10)
$\Delta Grid$ connection	0.067**	0.554** (0.263)	0.062 (0.034)	0.640* (0.381)	0.065**	0.554**	0.063**	0.548** (0.262)	0.059**	0.552** (0.262)
$\Delta Size$	-0.031^{***}	*_0.036*** (0.005)	-0.032** (0.006)	-0.032***-0.038*** (0.006) (0.007)	-0.030**:	$-0.030^{***}-0.035^{***}$ (0.004) (0.005)	-0.030^{**} , (0.004)	$-0.030^{***}-0.035^{***}$ (0.004) (0.005)	-0.030*** (0.004)	-0.036** (0.005)
$\Delta \mathit{Children}$	-0.377*** (0.045)	*_0.322*** (0.052)	-0.373***- (0.036)	*-0.309*** (0.075)	-0.379***- (0.046)	*_0.326*** (0.053)	-0.382***- (0.046)	*_0.329*** (0.053)	-0.379*** (0.046)	-0.324*** (0.053)
$\Delta Hindu$	0.101 (0.080)	0.086 (0.077)	0.076* (0.028)	0.064^* (0.035)	0.093	0.079 (0.077)	0.093	0.079 (0.077)	0.093	0.078 (0.077)
$\Delta Home$	0.047 (0.051)	0.055 (0.056)	0.049 (0.057)	0.0547 (0.056)	0.039 (0.052)	0.046 (0.056)	0.041 (0.052)	0.049 (0.057)	0.037 (0.053)	0.045 (0.058)
$\Delta Livestock$	-0.081^{***}	*_0.085*** (0.023)	-0.077* (0.032)	-0.085*** (0.031)	-0.079^{**} : (0.023)	.0.079***_0.083*** (0.023) (0.024)	-0.079*** (0.023)	* _0.084*** (0.024)	-0.081^{***} (0.023)	-0.086** (0.024)
$District\ F.E.$	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes
Observations	9,785	9,785	8,383	8,383	9,585	9,585	9,561	9,561	9,444	9,444
F-stat first stage K-P F-stat (weak ident)		16.34 16.34		39.97 13.42		16.33 16.34		16.40 16.41		16.56 16.57

Notes: All estimations contain a constant. Standard errors in parentheses are clustered at the district level. **** p<0.01, *** p<0.05, ** p<0.1. Income represents total household income excluding agriculture expressed in per adult equivalent terms. The instrument used is density of transmission cables multiplied by the initial quality of power supply. Hindu, Home and Livestock are dummy variables.

References

- [1] The OECD List of Social Indicators. OECD, Paris, 1982.
- [2] World Energy Assessment: Energy and the Challenge of Sustainability. United Nations Development Program (UNDP), New York, 2000.
- [3] World Energy Outlook 2011. International Energy Agency, 2011.
- [4] J.D. Angrist. Lifetime Earnings and the Vietnam Era Draft Lottery: Evidence from Social Security Administrative Records. *American Economic Review*, 80(3):313–36, June 1990.
- [5] J.D. Angrist. Estimating the Labor Market Impact of Voluntary Military Service Using Social Security Data on Military Applicants. *Econometrica*, 66(2):249–288, March 1998.
- [6] D.A. Aschauer. Is Public Expenditure Productive? *Journal of Monetary Economics*, 23(2):177–200, March 1989.
- [7] D.H. Autor and M.G. Duggan. The Rise in the Disability Rolls and the Decline in Unemployment. *The Quarterly Journal of Economics*, 118(1):157–205, February 2003.
- [8] P. Balachandra. Dynamics of Rural Energy Access in India: An Assessment. *Energy*, 36(9):5556 5567, 2011.
- [9] T.J. Bartik. Who Benefits from State and Local Economic Development Policies? Klamazoo, MI, 1991.
- [10] G. Bensch, J. Kluve, and J. Peters. Rural Electrification in Rwanda An Impact Assessment Using Matching Techniques. Ruhr Ecnomic Papers 231, RUHR, December 2010.
- [11] A. Bhide and C. Rodriguez Monroy. Energy Poverty: A Special Focus on Energy Poverty in India and Renewable Energy Technologies. Renewable and Sustainable Energy Reviews, 15(2):1057 – 1066, 2011.
- [12] O.J. Blanchard and L.F. Katz. Regional Evolutions. *Brookings Papers on Economic Activity*, 23(1):1–76, 1992.
- [13] J. Bound and H.J. Holzer. Demand Shifts, Population Adjustments, and Labor Market Outcomes during the 1980s. *Journal of Labor Economics*, 18(1):20–54, January 2000.
- [14] M.H. Brown and R.P. Sedano. Electricity Transmission, A Primer. National Council on Electricity Policy, June 2004.
- [15] Taryn Dinkelman. The Effects of Rural Electrification on Employment: New Evidence from South Africa. *American Economic Review*, 101(7):3078–3108, December 2011.
- [16] E. Duflo and R. Pande. Dams. The Quarterly Journal of Economics, 122(2):601–646, 05 2007.

- [17] T. Garcia-Mila and T.J. McGuire. The Contribution of Publicly Provided Inputs to States' Economies. Regional Science and Urban Economics, 22(2):229–41, June 1992.
- [18] D. Holtz-Eakin. State-Specific Estimates of State and Local Government Capital. Regional Science and Urban Economics, 23(2):185–209, April 1993.
- [19] G.W. Imbens and J.D. Angrist. Identification and Estimation of Local Average Treatment Effects. *Econometrica*, 62(2):467–75, March 1994.
- [20] S.R. Khandker, D.F. Barnes, and H.A. Samad. Welfare Implications of Rural Electrification, A Case Study for Bangladesh. Policy Research Working Paper 4859, World Bank, March 2009a.
- [21] S.R. Khandker, D.F. Barnes, and H.A. Samad. Energy Poverty in Rural and Urban India. are the Energy Poor also Income Poor? Policy Research Working Paper 5463, World Bank, November 2010.
- [22] S.R. Khandker, D.F. Barnes, H.A. Samad, and N.H. Minh. Welfare Implications of Rural Electrification, Evidence from Vietnam. Policy Research Working Paper 5057, World Bank, September 2009b.
- [23] M. Lipscomb, A.M. Mobarak, and T. Barham. Development Effects of Electrification: Evidence from the Geologic Placement of Hydropower Plants in Brazil. American Economic Journal: Applied Economics, 5(2):200–231, April 2013.
- [24] V. Modi. Improving Electricity Services in Rural India. Working Paper 30, Center on Globalization and Sustainable Development, The Earth Institute at Columbia University, December 2005.
- [25] H. Oda and Y. Tsujita. The Determinants of Rural Electrification: The Case of Bihar, India. *Energy Policy*, 39(6):3086–3095, June 2011.
- [26] Directorate of Census Office of the Registrar General, editor. Census of India 2001. Government of India, New Delhi, India, 2001.
- [27] L.-H. Roller and L. Waverman. Telecommunications Infrastructure and Economic Development: A Simultaneous Approach. American Economic Review, 91(4):909–923, September 2001.
- [28] J.P. Rud. Electricity Provision and Industrial Development: Evidence from India. *Journal of Development Economics*, 97(2):352 367, 2012.
- [29] B. Ural Marchand. Tariff Pass-Through and the Distributional Effects of Trade Liberalization. *Journal of Development Economics*, 99(2):265–281, 2012.
- [30] USGS. Global Land Cover Facility (SRTM). In Shuttle Radar Topography Mission. Global Land Cover Facility, University of Maryland, 2004.
- [31] H. Yang. Overview of the Chinese Electricity Industry and its Current Uses. Working Paper 0617, CWPE, February 2006.

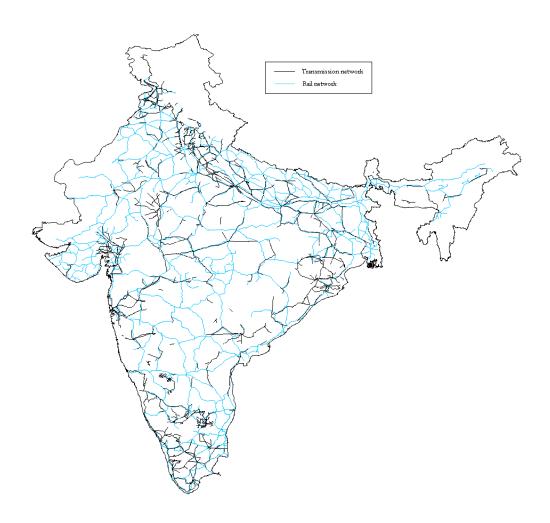
Table A.1: Effect of quality of power supply on income (alternate threshold values)

				Logo	Dependen	Dependent variable: Log of income per adult equivalent				
,	Baseline (18) OLS IV_C	e (18) IV _C	$\begin{array}{c} 16 \text{ hours} \\ OLS & I \end{array}$	$_{IV_{C}}^{\mathrm{urs}}$	$\frac{17 \text{ hours}}{OLS}$	IV_C	19 hours OLS I	IV_C	$\begin{array}{cc} 20 \text{ hours} \\ OLS & II \end{array}$	$_{IV_C}^{ m urs}$
	(1)	(2)	(3)	(4)	(5)	(9)	(7)	(8)	(6)	(10)
$\Delta Grid$ connection	0.079*	0.894** (0.395)	0.074 (0.048)	0.902** (0.391)	0.079 (0.048)	0.880** (0.379)	0.094*	0.980** (0.472)	0.098**	0.990** (0.481)
$\Delta Size$	-0.028^{**} ; (0.005)	$-0.028^{***}-0.033^{***}$ (0.005) (0.006)	-0.028*** (0.005)	-0.028^{***} -0.032^{***} (0.005) (0.006)	-0.028^{***}	$-0.028^{***}-0.032^{***}$ (0.005) (0.006)	-0.028** (0.005)	$0.028^{***} - 0.032^{***} (0.005) (0.006)$	-0.028*** (0.005)	-0.033*** (0.006)
Δ Children	-0.368***- (0.052)	*_0.303*** (0.058)	-0.368^{***} . (0.052)	*_0.299*** (0.058)	-0.368*** (0.052)	*_0.303*** (0.058)	-0.367^{***} (0.052)	*_0.304*** (0.059)	-0.367*** (0.052)	-0.305*** (0.058)
$\Delta Hindu$	0.140 (0.096)	0.141 (0.094)	0.139 (0.096)	0.124 (0.092)	0.139 (0.096)	0.132 (0.093)	0.138 (0.096)	0.122 (0.092)	0.138 (0.096)	0.123 (0.093)
$\Delta Home$	0.078 (0.067)	0.081 (0.072)	0.079 (0.067)	0.084 (0.071)	0.078 (0.067)	0.077 (0.072)	0.079 (0.067)	0.084 (0.074)	0.079 (0.067)	0.091 (0.075)
$\Delta Livestock$	-0.085^{***} . (0.025)	*_0.097*** (0.028)	-0.085^{***} . (0.025)	0.097*** (0.029)	-0.085*** (0.025) (*_0.097*** (0.028)	-0.086** (0.025)	$0.086^{***}-0.105^{***}$ 0.025) (0.029)	-0.086*** (0.025)	-0.107^{***} (0.030)
$District\ F.E.$	yes	yes	yes	yes	yes	yes	yes	yes	yes	yes
Observations	7,181	7,181	7,181	7,181	7,181	7,181	7,181	7,181	7,181	7,181
F-stat first stage K-P F-stat (weak ident)		8.08 8.09		7.97 7.97		8.25 8.25		7.06		6.95 6.95

Notes: All estimations contain a constant. Standard errors in parentheses are clustered at the district level. *** p<0.01, ** p<0.05, * p<0.1. Income represents total household income excluding agriculture expressed in per adult equivalent terms. The instrument used is density of transmission cables multiplied by the initial quality of power supply. Hindu, Home and Livestock are dummy variables. The Quality index takes a value of 0 for no electricity, 1 for high quality supply and 0.5 for low quality.

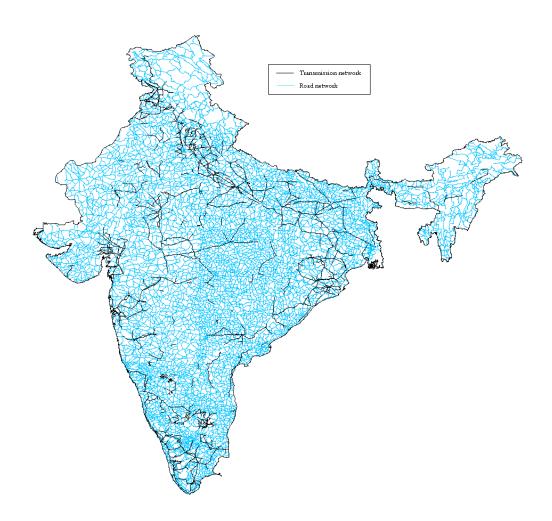
A Appendix Figures

Figure A.1: Transmission and rail networks



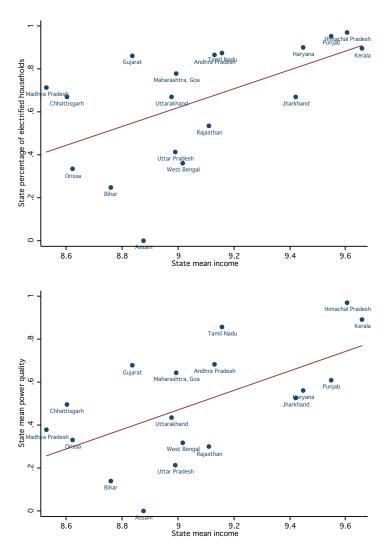
Source: ESRI ArcGIS World package and geocommons.

Figure A.2: Transmission and road networks



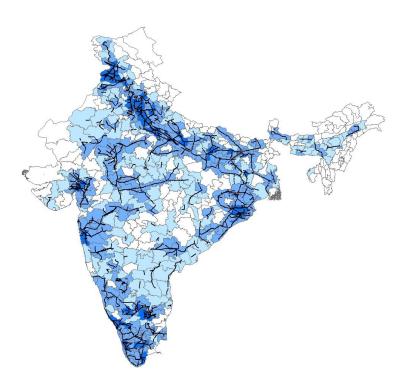
Source: ESRI ArcGIS World package and geocommons.

Figure A.3: Household Electrification Rate vs Mean Income by State, 2005



 $\underline{\underline{\text{Notes}}}$: We show the state-wise correlation between electricity outcomes and average household income in 2005. A linear fit for the scatter diagram is provided.

Figure A.4: Density of transmission cables per km^2 by district, 2005



Notes: Data on electrification rates is obtained from the India Human Development Survey (2005). The figure shows the change in the electrification rate in each district. Districts left white are not part of the survey sample. Source: ESRI ArcGIS World Package and Geocommons.