

Environmental and Development Economics

Module 2 - How does development affect the environment?

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Lecture 2

How does development affect the environment?

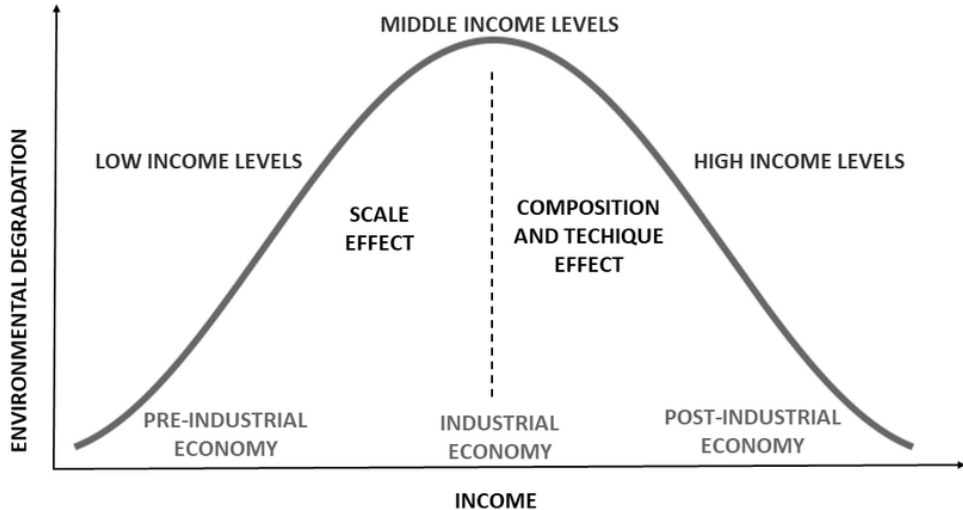
- ▶ State things that are due.

- ▶ **Guiding question:** how does economic development affect the environment?
- ▶ Descriptive overview
- ▶ Channel I: Income effects
 - ▶ Changes in consumption
 - ▶ Energy
 - ▶ Diet
- ▶ Channel II: Technology and Infrastructure
 - ▶ Agricultural productivity
 - ▶ Infrastructure
- ▶ Channel III: Institutions
 - ▶ Later in the course

How does economic development affect the environment?

- ▶ There is no one answer
- ▶ Choices that maximize economic growth often degrade environment
- ▶ But development expands our choice set (e.g. clean energy)
- ▶ Development also lowers $u'(c)$ (model from last week)
- ▶ Hence: the Environmental Kuznets Curve (EKC)

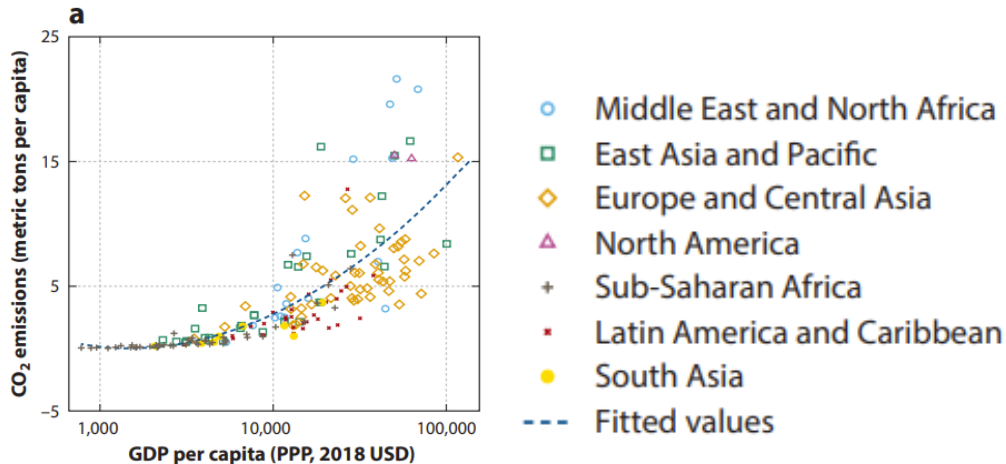
Environmental Kuznets Curve



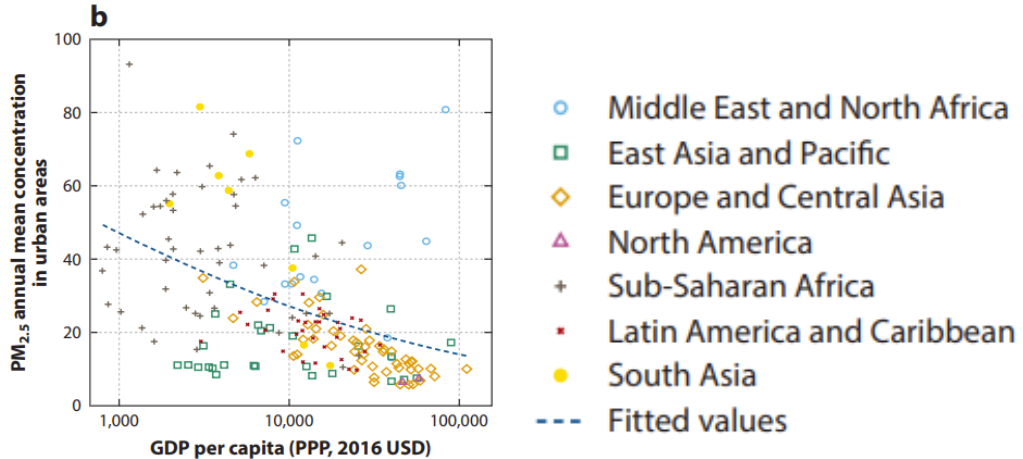
Discussion: Do you believe EKC?

- ▶ Mechanisms?
- ▶ Causality?
- ▶ Robustness?

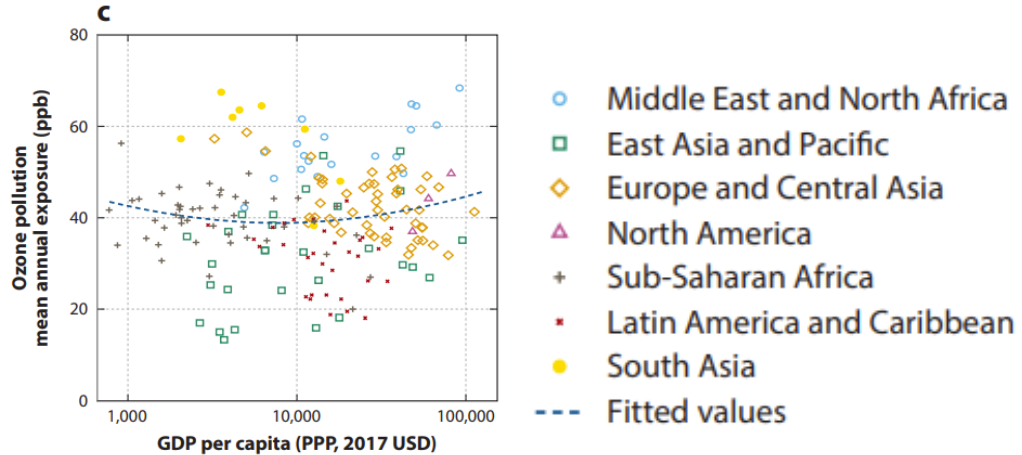
GDP and CO2



GDP and PM2.5



GDP and Ozone



- ▶ Macro correlations give limited answer
- ▶ Not the right question
 - ▶ What do we mean by development?
 - ▶ e.g. manufacturing ↑ pollution, but services may not
- ▶ Narrow, micro/empirical papers more helpful
 - ▶ even if they don't find consensus

Discussion: What is the ideal (quasi-) experiment?

- ▶ Can we randomize GDP?
- ▶ What variation in “development” can approximate the experiment?
 - ▶ what proxies development?
 - ▶ what does the proxy miss?
 - ▶ what (non-random) variation isolates causal relationship?
 - ▶ what does that variation miss?

Discussion: What about mechanisms?

- ▶ **Why and how** does economic development affect the environment?
- ▶ Theory is your friend: especially total derivatives. . .
 - ▶ Sometimes enough to rationalize mechanisms
 - ▶ Otherwise generate testable predictions
- ▶ Heterogeneity is your friend
 - ▶ But have a reason **why** you choose one channel over another

Question: What is the effect of income on environmental quality?

- ▶ Poverty alleviation may \uparrow demand for resource-intensive good
- ▶ Or, it may \uparrow demand for conservation
- ▶ Empirical challenge: household income is endogenous
- ▶ **This paper:** Exploit RDD eligibility to study impact of CCT on forests

Why can't we regress forests on income?

- ▶ Income correlated with other things that affect environment
- ▶ Like what?
- ▶ If we control for OVB, will panel fixed effects work?
 - ▶ No: small year-to-year income changes do not reflect “development”
 - ▶ Predictable short-term income fluctuations different than permanent change

Mexico's Oportunidades Program

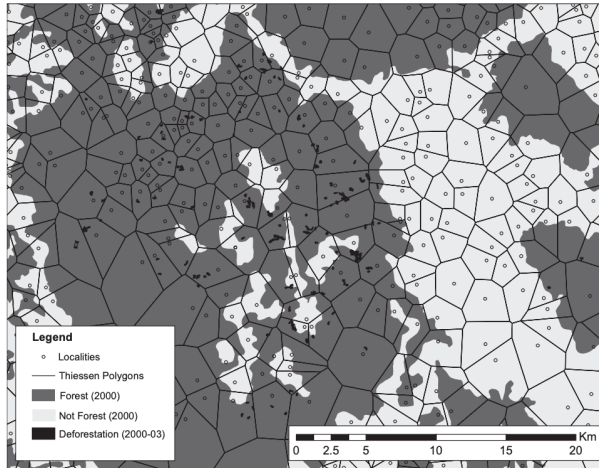
- ▶ **Goal:** Ambitious program to build human capital
- ▶ **Scope:** \$2.6 billion/year (half of anti-poverty budget)
 - ▶ Transfers are 1/3 of total income
 - ▶ Represents shift to new income path
- ▶ **Size:** 4 million households treated
- ▶ **Design:** Randomized rollout from 1998-2000
 - ▶ Afterwards, enrollment is non-random and based on **marginality index**
 - ▶ Two step: 1) village selected on index; 2) households selected in eligible villages

Results Preview: conditional cash transfer increases deforestation

- ▶ **Main variables:** village lat/lon, eligibility index, forest cover
- ▶ **Sample size:** 105,749 villages
- ▶ **Unit of analysis:** only village centroids provided?
 - ▶ How do we estimate village-level impacts? Thiessen polygons
 - ▶ Area around a point where every location closer to the point than to all others
 - ▶ Problems?

Thiessen/Voronoi Polygons

FIGURE 2.—ILLUSTRATION OF LOCALITY BOUNDARIES DEFINED USING THIESSEN POLYGONS



Eligibility cutoff: -1.2

FIGURE 3.—ENTIRE SAMPLE MINUS OBSERVATIONS WITH INDEX > 3

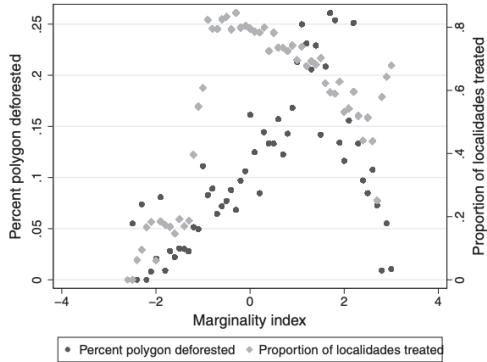
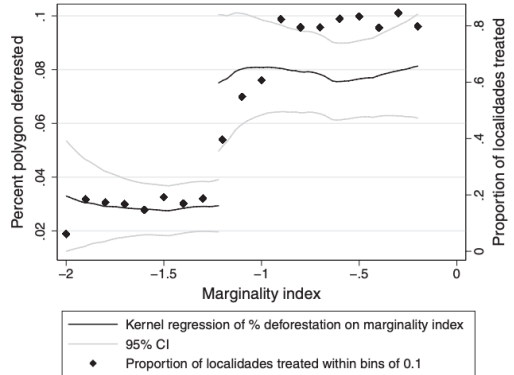


FIGURE 4.—KERNEL ESTIMATION OF DEFORESTATION ON MARGINALITY INDEX—RESTRICTED SAMPLE



- ▶ OLS:

$$\Delta f_i = \alpha + \delta E_i + \beta' X_i + \epsilon_i$$

where $E_i = 1$ if index $l_i > -1.2$

- ▶ RDD: Sample window $-2 < l < -0.2$

- ▶ Fuzzy RDD/IV:

$$\Delta f_i = \alpha + \delta T_i + \gamma l_i + \beta' X_i + \epsilon_i$$

$$\Delta T_i = \omega + \tau_1 E_i + \tau_2 E_i l_i + \tau_3 M_i + \tau_4 M_i l_i + \mu l_i + \Gamma' X_i + \epsilon_i$$

where T_i is enrollment dummy, E_i is eligibility cutoff dummy, l_i is index, M_i is dummy for region where enrollment increases rapidly

Identification Assumptions

Results: OLS and Sharp RDD

| | Tobit | | | OLS | |
|-------------------------------|----------------------|--------------------|-------------------|------------------------|------------------------|
| | % Polygon Deforested | | | Deforestation (0/1) | % Deforested (If 1) |
| | (1) | (2) | (3) | (4) | (5) |
| Eligible | .383 (.181)** | .549 (.295)* | .370 (.217)* | .013 (.008)* | .387 (.190)** |
| Marginality index | .523 (.041)*** | .753 (.077)*** | .219 (.189) | .031 (.003)*** | .069 (.075) |
| Index ² | | .069 (.072) | | .002 (.003) | .060 (.075) |
| Index ³ | | -.100 (.037)*** | | -.004 (.001)*** | -.022 (.025) |
| Index ⁴ | | -.002 (.015) | | -.0001 (.0005) | -.012 (.013) |
| Baseline area in forest, 2000 | -.0004 (.001) | -.0005 (.001) | .004 (.002)** | .0006 (.0001)*** | .005 (.001)*** |
| Ln(polygon area) | .947 (.042)*** | .954 (.042)*** | .728 (.068)*** | .046 (.002)*** | -.993 (.062)*** |
| Ln(total population in 1995) | .142 (.024)*** | .144 (.024)*** | .036 (.034) | .010 (.001)*** | -.040 (.025) |
| Ln(slope) | -.052 (.005)*** | -.053 (.005)*** | -.009 (.010) | -.003 (.0002)*** | -.029 (.006)*** |
| Ln(road density) | -.059 (.026)** | -.056 (.026)** | .025 (.053) | -.004 (.001)*** | -.010 (.027) |
| Observations | 58,587 | 58,587 | 15,758 | 58,587 | 5,545 |
| Ecoregion controls | Yes | Yes | Yes | Yes | Yes |

Results: Fuzzy RDD / Instrumental Variables

- Very strong first stage: eligibility \uparrow probability of enrollment ($\tau_1 = 0.8$)

| | IV Tobit | | | | IV OLS | |
|-------------------------------|------------------------|--------------------|--------------------|-------------------|---------------------|---------------------|
| | Full Estimation Sample | | | Restricted Sample | Deforestation (0/1) | % Deforested (If 1) |
| | (1) | (2) | (3) | (4) | (5) | (6) |
| Treated | .584 (.280)** | 1.293 (.715)* | | 1.038 (.609)* | .031 (.019)* | 1.264 (.680)* |
| Proportion treated | | | 3.453 (1.870)* | | | |
| Marginality index | .521 (.042)*** | .641 (.106)*** | .244 (.298) | -.072 (.339) | .028 (.003)*** | -.005 (.101) |
| Index ² | | .177 (.116) | .391 (.221)* | | .004 (.004) | .162 (.119) |
| Index ³ | | -.091 (.035)*** | -.053 (.031)* | | -.003 (.001)*** | -.036 (.030) |
| Index ⁴ | | -.010 (.015) | -.037 (.022)* | | -.0003 (.0005) | -.019 (.014) |
| Baseline area in forest, 2000 | -.0005 (.001) | -.0008 (.001) | -.001 (.001) | .003 (.002)** | .0006 (.0001)*** | .004 (.001)*** |
| Ln(polygon area) | .963 (.043)*** | .990 (.047)*** | 1.075 (.079)*** | .756 (.070)*** | .047 (.002)*** | -.948 (.065)*** |
| Ln(total population in 1995) | .055 (.051) | -.056 (.116) | -.305 (.245) | -.097 (.086) | .005 (.003) | -.262 (.120)** |
| Ln(slope) | -.054 (.005)*** | -.057 (.006)*** | -.064 (.008)*** | -.012 (.010) | -.003 (.0002)*** | -.033 (.007)*** |
| Ln(road density) | -.075 (.027)*** | -.092 (.033)*** | -.119 (.043)*** | .016 (.054) | -.005 (.001)*** | -.049 (.036) |
| Observations | 58,587 | 58,587 | 58,587 | 15,758 | 58,587 | 5,545 |
| Ecoregion controls | Yes | Yes | Yes | Yes | Yes | Yes |

- ▶ RD results show that CCT increases deforestation
- ▶ Why? What changes at the household level?
- ▶ Approach 1: Use experimental sample (Progresa) with household survey data
- ▶ Approach 2: Heterogeneity by road density to study role of market access

Mechanisms: Consumption channel

- ▶ Before (1997-1998) and after (2000) data on consumption
- ▶ 506 villages, 320 treated, 186 control
- ▶ Treatment at the village level

Difference in differences:

$$y_{it} = \gamma_0 + \gamma_1 Treat_i + \gamma_2 Post_t + \gamma_3 Treat_i \times Post_t + \epsilon_{it}$$

- ▶ y_{it} = consumption
- ▶ cluster at locality level

Hypothesis: $\gamma_3 > 0$

Mechanisms: Market channel

- ▶ Demand shock must be met by supply, which drives land use change
 - ▶ e.g. \uparrow consumption of milk met by \uparrow in grazing land
- ▶ But γ_3 captures partial equilibrium
 - ▶ Part of demand shock supplied locally. What about rest?
- ▶ Hypothesis: If consumption is driving deforestation
 - ▶ Then effect should be larger when infrastructure quality is low
 - ▶ Low market access: demand met by local supply
 - ▶ High market access: demand shock propagates across markets

Triple Differences:

Heterogeneity by infrastructure quality:

$$\begin{aligned} y_{it} = & \beta_0 + \beta_1 Treat_i + \beta_2 Post_t + \beta_3 Treat_i \times Post_t \\ & + \beta_4 Road_i + \beta_5 Road_i \times Treat_i + \beta_6 Road_i \times Post_t \\ & + \beta_7 Road_i \times Treat_i \times Post_t + \epsilon_{it} \end{aligned}$$

► $Road_i$ = inverse road density (km of road w/n 10km of village/polygon)

Hypothesis: $\beta_7 > 0$

Results: Increase in consumption of land intensive goods

| | Rooms in Home | | Days Ate Beef | | Days Drank Milk | |
|---|------------------|-----------------|--------------------|--------------------|--------------------|--------------------|
| | (1) | (2) | (3) | (4) | (5) | (6) |
| Treatment effect | .014 (.033) | .017 (.035) | .114 (.030)*** | .118 (.031)*** | .337 (.081)*** | .331 (.087)*** |
| Treatment \times inverse road density | | -.034 (.148) | | -.070 (.097) | | .183 (.669) |
| Village chosen to receive Progresa | .0001 (.037) | .002 (.038) | -.025 (.029) | -.031 (.030) | -.133 (.111) | -.143 (.118) |
| Posttreatment year | .053 (.028)* | .049 (.029)* | -.137 (.024)*** | -.138 (.025)*** | -.655 (.061)*** | -.664 (.065)*** |
| Inverse of road density | | .266 (.169) | | -.156 (.069)** | | .051 (.499) |
| Village \times inverse road density | | .043 (.236) | | .102 (.140) | | .232 (.682) |
| Posttreatment \times inverse road density | | .067 (.140) | | .016 (.068) | | .155 (.252) |
| Observations | 23,318 | 23,318 | 33,128 | 33,128 | 33,128 | 33,128 |
| Mean dependent Variable in baseline | 1.557 (0.930) | | 0.388 (0.661) | | 1.440 (2.367) | |

Higher beef and milk demand (land intensive products)

Demand-side impacts do not vary with market access (we didn't expect it to)

Results: No increase in local production

| | Number of Plots | | Log (1+ Total Hectares) | | Number of Cows | |
|---|--------------------|-------------------|-------------------------|-------------------|--------------------|--------------------|
| | (1) | (2) | (3) | (4) | (5) | (6) |
| Treatment effect | .030 (.039) | .031 (.040) | -.014 (.038) | -.015 (.039) | .092 (.057) | .036 (.057) |
| Treatment \times inverse road density | | -.107 (.210) | | .142 (.223) | | .936 (.522)* |
| Village chosen to receive Progresa | .014 (.056) | .037 (.057) | -.004 (.040) | .017 (.040) | -.004 (.087) | .058 (.085) |
| Posttreatment year | -.094 (.032)*** | -.077 (.033)** | .312 (.033)*** | .317 (.033)*** | -.239 (.046)*** | -.180 (.046)*** |
| Inverse of road density | | .833 (.161)*** | | .820 (.227)*** | | 2.122 (.799)*** |
| Village \times inverse road density | | -.263 (.317) | | -.217 (.258) | | -.760 (.872) |
| Posttreatment \times inverse road density | | -.275 (.149)* | | -.235 (.128)* | | -.982 (.402)** |
| Observations | 45,087 | 45,087 | 32,631 | 32,631 | 34,248 | 34,248 |
| Mean dependent | 0.824 | | 1.724 | | 0.604 | |
| Variable in baseline | (0.955) | | (3.535) | | (2.304) | |

Supply-side impacts do not vary with market access

Results: Deforestation higher in places with poor market access

| Dependent Variable | Low Density | | Medium Density | | High Density | |
|-----------------------|-------------|----------|----------------|--------|--------------|--------|
| | (%) | (0/1) | (%) | (0/1) | (%) | (0/1) |
| | (1) | (2) | (3) | (4) | (5) | (6) |
| Treated | 1.619 | .075 | .554 | .019 | 1.818 | .023 |
| | (.868)* | (.037)** | (.836) | (.030) | (1.472) | (.021) |
| Treated \times low | | | | | | |
| Low road density | | | | | | |
| Observations | 19,529 | 19,529 | 19,529 | 19,529 | 19,529 | 19,529 |
| Ecoregion controls | Yes | Yes | Yes | Yes | Yes | Yes |

Bigger impact where supply response is localized

Supports mechanism of increased demand for land-intensive goods

Thoughts?

- ▶ Form discussion group to think of alternate story
- ▶ Do we believe the story about increased demand for land-intensive goods?
 - ▶ Increased income relieves credit constraints
 - ▶ Expands capital, increases ag productivity on extensive margin b/c malfunctioning factor markets.
 - ▶ Market access may improve access to inputs, and reduce deforestation.
 - ▶ Consistent with papers results.
- ▶ Only focused on one mechanism
 - ▶ returns to off-farm labor
 - ▶ income may increase aggregate production and lead to deforestation?
 - ▶ market access may lead to migration, less population pressure and deforestation
 - ▶ also consistent story
- ▶ This all points to shortcomings of RCTs in general
 - ▶ we cannot answer why

Same results in Colombia! Malerba (2020)

- ▶ Question: What is the impacts of CCT on consumption, energy, and deforestation?
- ▶ Context: Familias en Accion CCT project (2001-2005)
- ▶ Design: Matched Difference in Differences
- ▶ Result: Increased beef and milk (land intensive) consumption
 - ▶ Mediated by markets
 - ▶ Negligible \uparrow in deforestation (counterintuitive)
 - ▶ No impact on CO₂

- ▶ CCT program launched in 2000
 - ▶ Non-random: Municipalities selected on amenities (banks, education, health)
 - ▶ 721 eligible: slow phase-in
- ▶ Study sample: 5,477 households
- ▶ Variation: compare outcome in treated hh before/after enrollment relative to control hh
- ▶ **Identification assumptions**

| Number of municipalities enrolled (annual) | |
|--|-----|
| 2000 | 2 |
| 2001 | 360 |
| 2002 | 244 |
| 2003 | 6 |
| 2004 | 0 |
| Total | 612 |

Results: Increased beef and milk consumption

Household DiD:

$$\Delta Y_i = \alpha + \delta T_i + X_i + \epsilon_i$$

| Variables | (1) Beef, days per week | (2) Beef, days per week | (3) Milk, days per week | (4) Milk, days per week |
|-----------------------------------|-------------------------------|-------------------------------|-------------------------------|-------------------------------|
| FA | 0.416** (0.166) | 0.879** (0.333) | 0.720** (0.282) | 1.099** (0.547) |
| FA *distance to closest market | | -0.010* (0.005) | | -0.008 (0.008) |
| Households | 2268 | 2268 | 2269 | 2269 |
| R-squared | 0.029 | 0.032 | 0.045 | 0.046 |

- ▶ Effect muted with low market access (very weak)
- ▶ Why is sample size declining from 5,477?

Design: Environmental Impacts

- ▶ Unit of analysis: Municipality
- ▶ Design: municipalities enrolled gradually over time
 - ▶ Many were not enrolled by 2004
- ▶ Define $Treat_i = 1$ if enrolled by 2004
- ▶ $Post_t = 1$ after treatment, zero for 2000

$$y_{it} = \beta_0 + \beta_1 Treat_i + \beta_2 Post_t + \beta_3 (Treat_i \times Post_t) + \beta_4 X_{it} + \epsilon_{it}$$

- Variation: Δ forest b/w 2000/05 for municipalities enrolled by 2004 rel. to non-enrolled

Results: No environmental impact

| Variables | (1) % forest | (2) % forest, incl. ref. | (3) % forest |
|--|----------------------|-----------------------------|----------------------|
| FA enrollment | 0.510* (0.271) | 0.369 (0.303) | 0.523* (0.293) |
| FA enrollment, number of years | | | |
| FA enrollment * distance to closest market | | | −0.000 (0.001) |
| 2005 | −2.166*** (0.266) | −1.479*** (0.309) | −2.167*** |
| Constant | 22.361*** (1.588) | 23.452*** (2.324) | 22.358*** (1.615) |
| R-squared | 0.370 | 0.196 | 0.370 |
| Observations | 1440 | 1440 | 1440 |

- ▶ Weakly positive deforestation (counterintuitive)
- ▶ **Why?**

Back to the question

How does development affect the environment?

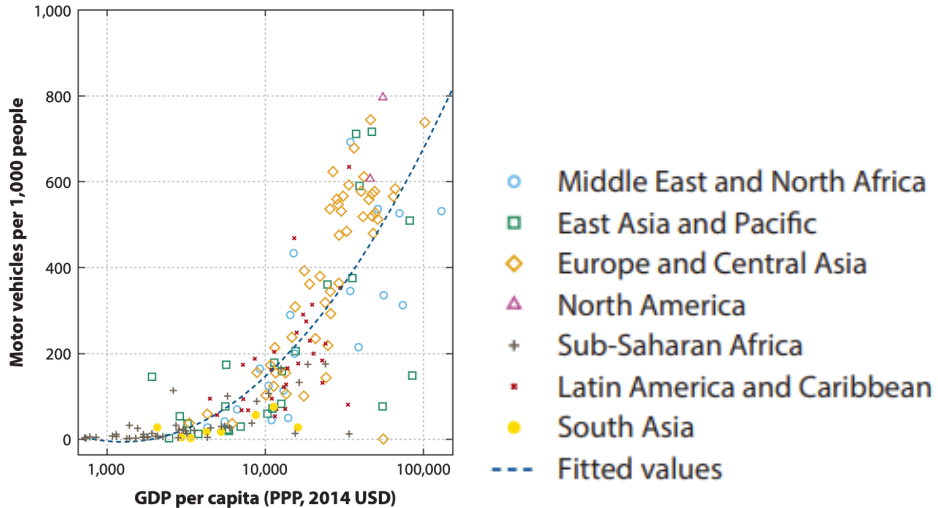
- ▶ Income
 - ▶ Land intensive consumption
 - ▶ **Energy**

- ▶ Next time: Technology and Infrastructure

Development, Energy, and the Environment

- ▶ Energy ladder: as income \uparrow , move from solid fuel \rightarrow gas \rightarrow electricity
 - ▶ Electricity may displace dirty energy (wood)
- ▶ But, electricity also increases total energy (fridge, AC)
 - ▶ energy footprint increases with income
- ▶ Unless generated from solar, wind, etc
- ▶ People also buy cars as they get richer. . .

GDP and Car Ownership

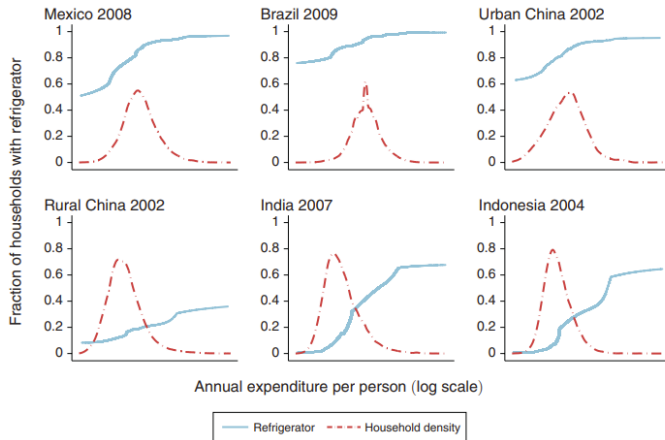


Gertler et al. (2016): Demand for energy-using assets among middle class

- ▶ **Question:** How does income affect energy demand?
- ▶ **Motivation:** 1.3 billion people live without electricity
- ▶ Should we expect linear climb up the energy ladder as incomes rise?
 - ▶ what about credit markets?

| | Electricity access (percent of population) | Refrigerators (share of households) |
|--------------------|--|---|
| Brazil | 98.7 | 0.93 |
| China | 99.7 | 0.69 |
| India | 75.0 | 0.13 |
| Indonesia | 73.0 | 0.17 |
| Mexico | 97.9 | 0.83 |
| Sub-Saharan Africa | 32.5 | 0.11 |
| Total | 70.8 | 0.38 |
| United States | 100.0 | 0.99 |

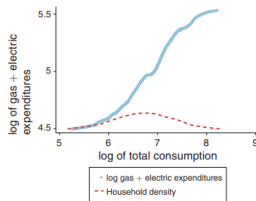
Non-linear relationship between income and fridges



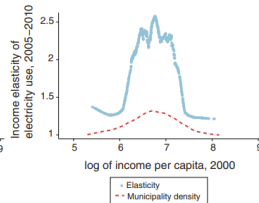
- Are fridges representative of energy-intensive assets?
- Can aggregating non-linearities → linearity?

S-shape appears robust

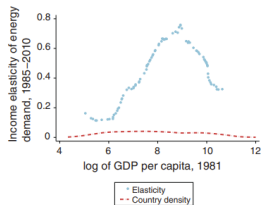
Panel A



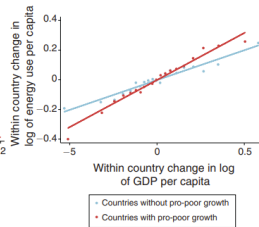
Panel B



Panel C



Panel D



How can we explain the S-shape?

Set up:

- ▶ Two periods with no discounting
- ▶ Agent i can consume two goods
 - ▶ non-durable gives per period utility: $u'(\cdot) > 0$, $u''(\cdot) < 0$
 - ▶ lumpy durable that gives static per period utility R , if owned
- ▶ Durable price = P ; non-durable is numeraire
- ▶ Let Y_1 , Y_2 be per period 1, 2 income
- ▶ Let $\bar{Y} = 1/2(Y_1 + Y_2)$ be average income

No credit constraints

Without credit constraints, if don't buy, total utility is:

$$u(\bar{Y}) + u(\bar{Y}) = 2u(\bar{Y})$$

If buy durable, spend P and spread cost equally across periods:

$$2u(\bar{Y} - \frac{P}{2}) + 2R$$

Purchase durable iff:

$$u(\bar{Y}) - u(\bar{Y} - \frac{P}{2}) \leq R$$

$u''(\cdot) < 0$ **implies acquisition increases in income**

1) With credit constraints: Buy in period 1

If don't purchase, total utility is:

$$u(Y_1) + u(Y_2)$$

If purchase in period 1:

$$u(Y_1 - P) + 2R + u(Y_2)$$

Purchase in period 1 iff:

$$\frac{u(Y_1) - u(Y_1 - P)}{2} \leq R$$

2) With credit constraints: Wait to buy in period 2

If don't purchase, total utility is:

$$u(Y_1) + u(Y_2)$$

If save in period 1, and buy in period 2:

$$2u(\bar{Y} - \frac{P}{2}) + R$$

Wait to buy in period 2 if:

$$u(Y_1) + u(Y_2) - 2u(\bar{Y} - \frac{P}{2}) \leq R$$

Intuition of Model

- ▶ $u''(\cdot) < 0$ means consumers gain from smoothing consumption
- ▶ Unconstrained household will buy in period 1
 - ▶ Use period 2 income (loan) to smooth consumption
- ▶ Credit constrained households cannot do this
 - ▶ If buy now, magnify consumption inequality across periods
 - ▶ Or, wait and buy in period 2, but delay utility gain
- ▶ First period buyers respond only to first period income (lemma 1)
- ▶ But increase in *cumulative* income increases buying through delay/saving (lemma 2)

Testable Predictions

- ▶ Prediction 0: Acquisition increasing in income
- ▶ Prediction 1: S-shaped curve in acquisition
- ▶ Prediction 2: Faster income growth leads to more period 2 adoption
- ▶ Prediction 3: Period 1 ownership depends on interaction of income and growth

Empirical Setting: Oportunidades (again)

- ▶ 320 randomly selected communities given early treatment (April 1998)
- ▶ 186 randomly selected communities given late treatment (October 1999)
- ▶ This paper: $N=506$ communities, 10,000 households surveyed (1997-2007)
- ▶ Compare early and late households
 - ▶ Similar b/c both groups selected on “vulnerability” characteristics
 - ▶ Strongly balanced on covariates

- **Prediction 0:** Probability of asset purchase \uparrow in income

$$h(a_{it}) = Pr(a_{it} = 1 | a_{it-1} = 0) = \gamma_0 + \alpha_1 Cl_{it} + \beta X_i + \beta_t F_i + R_{rt} + \epsilon_{it}$$

where $h(a)$ is the prob. that i buys a in time t conditional on not having it in $t - 1$.

- **Prediction 1:** S-shaped curve in acquisition

$$h(a_{it}) = Pr(a_{it} = 1 | a_{it-1} = 0) = \gamma_0 + \alpha_1 Cl_{it} + \alpha_2 Cl_{it}^2 + \beta X_i + R_{rt} + \epsilon_{it}$$

Hypothesis: $\alpha_2 > 0$

- **Prediction 2 and 3:** conditional on having same level of cumulative income, households which accumulated income slower are less likely to acquire asset; this effect is increasing in cumulative income

$$h(a_{it}) = Pr(a_{it} = 1 | a_{it-1} = 0) = \gamma_0 + \alpha_1 Cl_{it} + \alpha_3 Early_i + \alpha_4 (Early_i \times Cl_{it}) + \beta X_i + R_{it} + \epsilon_{it}$$

- where $Early_i = 1$ if i began receiving transfers 18 months before control households

Measurement issues

- ▶ Cumulative Income is sum of wage, farm, business income, CCT transfers
 - ▶ measurement error in wage data
 - ▶ cumulation requires interpolating between survey waves
- ▶ Double counting problem if households invest CCT money into business
- ▶ **Solution:** use cumulative CCT **transfers** to instrument cumulative **income**
 - ▶ Assumes transfers affect consumption only through income
 - ▶ Is this reasonable?

Identifying Variation

TABLE 2—OPORTUNIDADES BIMONTHLY SUPPORT LEVELS IN 2003 (*Pesos*)

| | | |
|--------------------------------|------|-------|
| Basic Support | 155 | |
| <i>Educational scholarship</i> | | |
| Grade | Boys | Girls |
| Third | 105 | 105 |
| Fourth | 120 | 120 |
| Fifth | 155 | 155 |
| Sixth | 205 | 205 |
| Seventh | 300 | 315 |
| Eighth | 315 | 350 |
| Ninth | 335 | 385 |
| Tenth | 505 | 580 |
| Eleventh | 545 | 620 |
| Twelfth | 575 | 655 |

Use maximum cumulative transfer as **instrument** for actual (cumulative) transfer

Plus variation from early/late enrollment

Results: Prediction 0

TABLE 4—BASIC RESULTS: REFRIGERATOR (*Income Effects*)

| | Discrete time hazard | | Household FE | Discrete time hazard | | Household FE |
|---|-------------------------|---------------------|---------------------|-------------------------|---------------------|---------------------|
| | OLS | IV | IV | OLS | IV | IV |
| | (1) | (2) | (3) | (4) | (5) | (6) |
| <i>Cumulative Transfers</i> | 0.018*** [0.005] | 0.020*** [0.007] | 0.047*** [0.008] | | | |
| <i>Cumulative Income</i> | | | | 0.003*** [0.001] | 0.016*** [0.005] | 0.034*** [0.007] |
| Observations | 30,414 | 30,414 | 30,258 | 30,414 | 30,414 | 30,258 |
| R^2 | 0.103 | | | 0.104 | | |
| Kleibergen-Paap Wald F -Stat on excluded variables | | 2,503 | 2,060 | | 92 | 108 |
| Number of households | | | 6,655 | | | 6,655 |

- For every 10,000 pesos, probability of buy fridge ↑ by 4.7 pp.

Results: Prediction 1

TABLE 5—BASIC RESULTS: REFRIGERATOR (*Nonlinear Income Effects*)

| | Discrete time hazard | | Household FE |
|---|-------------------------|----------------------|----------------------|
| | OLS (1) | IV (2) | IV (3) |
| <i>Cumulative Income</i> | −0.0007 [0.0011] | −0.0059 [0.0108] | 0.0132 [0.0132] |
| <i>Cumulative Income</i> ² | 0.0001*** [< 0.0001] | 0.0009** [0.0004] | 0.0008** [0.0004] |
| Observations | 30,414 | 30,414 | 30,258 |
| R^2 | 0.105 | | |
| Kleibergen-Paap Wald F -Stat on excluded variables | | 22 | 23 |
| Number of households | | | 6,655 |

- ▶ $\alpha_2 > 0$ implies convex relationship b/w cumulative income and asset purchase
- ▶ Since all households are poor, don't expect any to be past second inflexion point

Results: Prediction 2 and 3

- ▶ include only if time

Thoughts on paper?

Lessons from income/consumption channel?

- ▶ What do we learn about how development affects the environment?
- ▶ What are shortcomings of research designs? Measurement?
- ▶ External validity?

Research gaps

- ▶ Does earned income have different effect than transfers?
- ▶ Which externalities have highest income elasticities?
- ▶ Can the effect of income on $MWTP_e$ be separated from changes in consumption?

- ▶ How does development affect the environment?
 - ▶ **Technology and Infrastructure**

Lecture 3

How does development affect the environment?

Housekeeping

Recap

- ▶ Higher income leads to land-intensive **consumption**
 - ▶ Beef, milk, etc.
 - ▶ This requires clearing land for agriculture
 - ▶ Effect depends on market access
- ▶ Higher income leads to energy-intensive **consumption**
 - ▶ But the relationship is non-linear
 - ▶ Role of credit constraints

- ▶ Development increases access to **productive** capital
 - ▶ How do we think about environmental implications?
- ▶ Case I: Development increases agricultural productivity
 - ▶ Abman et al. (2023)
- ▶ Case II: Development spurs infrastructure development
 - ▶ Asher et al. (2020)

Technology and infrastructure

- ▶ As economies develop, technology and infrastructure improve
- ▶ Productivity boost means firms produce at lower cost
 - ▶ Outward shift in supply curve → bad for environment
 - ▶ But, technological progress reduces resource intensity → good for environment
- ▶ Most active areas of research:
 - ▶ agricultural productivity
 - ▶ infrastructure

Agricultural productivity and the environment

- ▶ **Boserup's hypothesis** : extensive margin
 - ▶ improvements in land productivity \uparrow ag land value
 - ▶ leads to pressure on forests to clear land for ag
- ▶ **Borluag's hypothesis**: intensive margin
 - ▶ *Under factor market constraints*, \rightarrow productivity spurs intensification
 - ▶ “spares” land for nature and puts less pressure on forests
- ▶ **Jevon's paradox**: general equilibrium
 - ▶ technology for increasing yield also lowers MC of producing given amount
 - ▶ Ag supply curve shifts out and may dominate gains in yield
 - ▶ Leads to net increase in land demand

Takeaway: this is an empirical question!

Conceptual framework

- ▶ Continuum of agents with heterogeneous outside options
 - ▶ Changes in returns to ag affect number of agents who select into farming
- ▶ Farmers allocate labor b/w farming on existing land, or cultivate new land
- ▶ Let γ denote household labor allocated to new land
- ▶ Production function for new and existing land are $f(\cdot)$ and $g(\cdot)$
- ▶ Exogenous productivity shock, Ω , improves productivity of existing land more

$$\frac{\partial g(l)}{\partial \Omega} \geq \frac{\partial f(l)}{\partial \Omega} \geq 0 \quad \forall l$$

Conceptual framework

- ▶ Farmers are factor market constrained (Conning and Udry, 2007)
 - ▶ Cannot hire labor from market
- ▶ Total deforestation given by new land cleared for agriculture:

$$D^* = n^* \cdot \gamma^*$$

- ▶ Differentiating with respect to Ω gives:

$$\frac{\partial D^*}{\partial \Omega} = \underbrace{n^* \frac{\partial \gamma^*}{\partial \Omega}}_{\leq 0} + \underbrace{\gamma^* \frac{\partial n^*}{\partial \Omega}}_{\geq 0}$$

- ▶ What does each term represent and what is the sign?

Abman et al. (2023): Agricultural Productivity and Deforestation

- ▶ **Goal:** To test $\frac{\partial D^*}{\partial \Omega}$, we need random variation in Ω
- ▶ **Question:** what is the impact of ag productivity on deforestation?
- ▶ **Context:** Large scale ag extension program in Uganda (2008-2013)
- ▶ **Design:** Leverage Spatial discontinuity in village eligibility
- ▶ **Results:** improvement in agricultural productivity; reduction in deforestation

Background: Why Uganda?

- ▶ Forest cover shrunk from 24% to 9% from 1990-2015
- ▶ Most forest land is **privately owned**
 - ▶ Land owners allowed to covert forest (Land Act, 2010)
 - ▶ Cannot use state powers to limit deforestation
- ▶ Need an indirect mechanism to incentivize conservation

Background: BRAC Extension Program

- ▶ Launched in 2008
- ▶ Aims to extend modern cultivation techniques to smallholders
- ▶ Two step treatment:
 - ▶ Training: “model farmers’ ’ trained in modern cropping, irrigation, weeding, pest control. Then, set up demo plot and train others
 - ▶ Extension: Promoters selected to sell subsidized HYV seeds in their villages
- ▶ **Key feature: Program limited to villages within 6km of each BRAC branch**

- ▶ Deforestation from GFC (Hansen et al., 2013)
 - ▶ Pro: high resolution (30m)
 - ▶ Con: 1) only captures loss not gain, 2) everything relative to 2000
- ▶ Lat/lon of villages within 12km of BRAC office (N=807 villages)
- ▶ Attribute forest pixel to village if w/n 400m of village centroid
 - ▶ Rationale: median household distance to village center = 400m
 - ▶ Problems?
- ▶ Agricultural survey on 7,781 households (451 villages) in 2011
 - ▶ Detailed cropping practices from last two seasons

Empirical Design: Spatial Regression Discontinuity

- ▶ Estimate IIT since we do not know true treatment designation
- ▶ Local linear regression to estimate left and right side of discontinuity

$$\beta = \lim_{z \uparrow 0} E[Y|z_i = z] - \lim_{z \downarrow 0} E[Y|z_i = z]$$

- ▶ where running variable, z_i , is distance of village from 6km cutoff
 - ▶ $z \leq 0$ means village had access to extension
- ▶ For estimation, use 2km bandwidth
 - ▶ Higher regression weights on points nearer to cutoff (triangle kernel)

Aside: RD with local linear regressions

- ▶ Left side of cutoff

$$Y_i = \alpha_L + \beta_L(z_i - c), \quad \text{for } z_i \leq c$$

- ▶ Right side of cutoff

$$Y_i = \alpha_R + \beta_R(z_i - c), \quad \text{for } z_i \geq c$$

- ▶ Treatment effect at threshold:

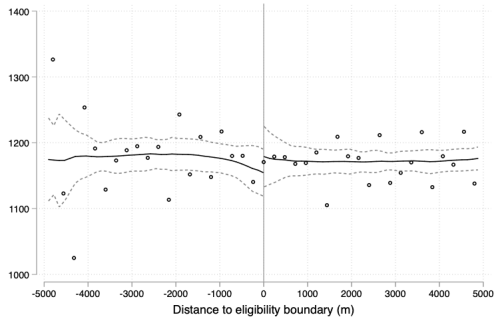
$$\beta = \alpha_R - \alpha_L$$

since α are the intercepts on each side of cutoff

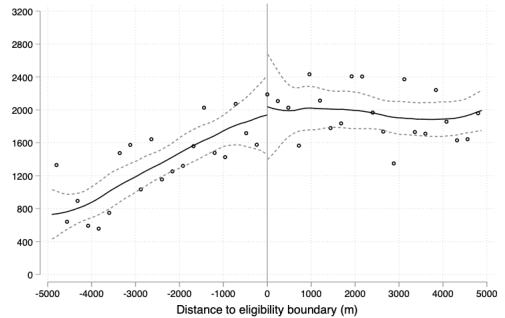
Identification Assumptions?

Identification Assumptions

(a) Elevation (meters)

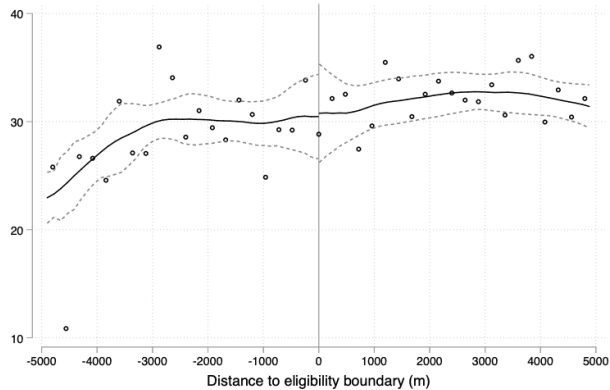


(b) Distance to nearest road (meters)



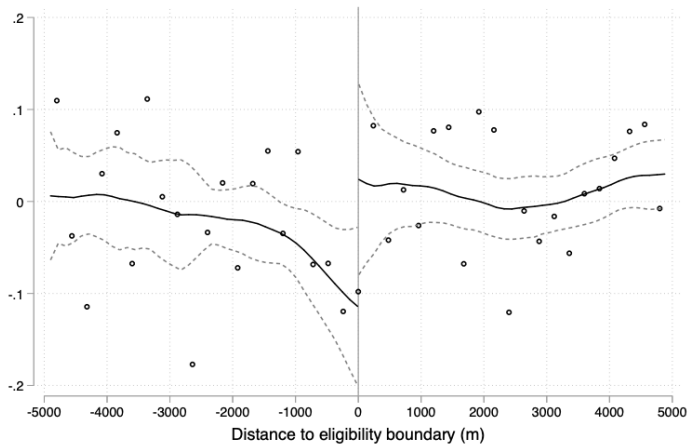
Identification Assumptions

(c) Baseline treecover



Results: 14% less deforestation in treated villages

(a) Residualized Forest Loss - Treatment period



Mechanisms: Intensification

| Outcome | (1) Manure Use | (2) Intercrop | (3) Crop Rotation | (4) Irrigation | (5) Weeding | (6) Fertilizer Use | (7) HYV Seeds | (8) Perennial crops |
|------------------|-----------------------|---------------------|-------------------------|------------------------|----------------------|--------------------------|---------------------|---------------------------|
| Program Eligible | 0.0977*** (0.0249) | 0.0590* (0.0308) | 0.0737*** (0.0251) | 0.0326*** (0.00829) | 0.0644** (0.0310) | -0.0171 (0.0161) | -0.0440 (0.0323) | -0.0264 (0.0351) |
| Obs | 2912 | 2912 | 2912 | 2912 | 2912 | 2912 | 2912 | 2912 |
| Control mean | 0.0731 | 0.796 | 0.797 | 0.0266 | 0.693 | 0.0725 | 0.356 | 0.351 |

- Techniques to address nutrient depletion **on existing land**
- No evidence of increased fertilizer, pesticides, or HYV seed adoption

Intensification vs. Extensification

| Outcome | (1) Revenue per acre | (2) IHS Cultiv ag area |
|------------------|----------------------------|------------------------------|
| Program Eligible | 0.325* (0.189) | 0.0262 (0.0514) |
| Obs | 2843 | 2907 |
| Control mean | 11.59 | 1.414 |

- ▶ Revenue/acre is a proxy for yields
- ▶ No extensification → binding constraint on land clearing

- ▶ Empirical issues
- ▶ Alternative explanations
- ▶ External validity (recall forest is private in Uganda)

Back to the Question

How does development affect the environment?

- ▶ Income
 - ▶ Land intensive consumption
 - ▶ Energy
- ▶ Access to capital
 - ▶ Technology
 - ▶ **Infrastructure**

Infrastructure and the Environment

- ▶ Roads can increase deforestation by:
 - ▶ ↓ MC (transport costs) of forest products, increasing supply
 - ▶ ↑ market access for agricultural products, causing extensification
 - ▶ ↑ land value, leading to agglomeration effects
- ▶ Roads can increase forest cover by:
 - ▶ ↑ access to forest product substitutes (i.e. clean fuel vs. firewood)
 - ▶ ↑ access to external labor/capital markets, lowering need for land-clearing
- ▶ Net effect is ambiguous

What about Infrastructure and Air Pollution?

- ▶ Lets build a conceptual framework
- ▶ Talk to your neighbor for 10 mins

Garg et al. (2023): Rural Roads, Labor Exits, and Crop Fires

- ▶ **Question:** what is the impact of roads on air quality?
 - ▶ what is the role of labor reallocation as a mechanism?
- ▶ **Context:** India builds last-mile roads using population cutoff
- ▶ **Design:** Regression discontinuity with crop fires as outcome
- ▶ **Results:** Roads \uparrow crop fires and PM2.5 levels in a village
- ▶ **Mechanism:** Roads \rightarrow labor exit \rightarrow higher wages \rightarrow labor-saving technology

Background

- ▶ Labor is misallocated in agriculture in developing countries
 - ▶ MP_{labor} lower in agriculture than other sectors
- ▶ Sources of misallocation is a big question in dev econ
- ▶ Many policies aimed at reducing misallocation
 - ▶ e.g. road-building (improve market access)
- ▶ Do these have externalities?
 - ▶ **innovation**: where do the externalities come from?

Context: India

- ▶ 14 of the 20 most polluted cities in the world are in India
- ▶ Important pollution source: crop burning
- ▶ Pros:
 - ▶ Clear rice straw residue & undergrowth b/w cropping seasons
 - ▶ Cheaper than labor/capital alternative
- ▶ Cons:
 - ▶ Carries civil/criminal penalty (weakly enforced)
 - ▶ Reduces soil nutrients and decreases ag productivity
 - ▶ Damages human health
- ▶ Roads may facilitate agricultural labor loss
 - ▶ Will farmers turn to crop burning, despite the costs?

- ▶ PMGSY aims to provide all-weather roads to all unconnected villages in India
- ▶ Prioritize larger villages first
 - ▶ Connect villages with population > 1000 by 2003
 - ▶ Connect villages with population > 500 by 2007
 - ▶ Connect villages with population > 250 afterwards
- ▶ State-by-state basis
 - ▶ If connected all large villages, can proceed to smaller ones immediately

- ▶ Roads: SHRUG database (N=11,151 villages w/o paved road at baseline)
- ▶ Fires: NASA EODIS -# fires w/n 10km of village polygon
- ▶ Pollution: Van donkelaar et al. (2016) measure PM2.5 at 10km resolution
- ▶ Infant mortality: NFHS-IV (2015-16)
 - ▶ child-birthyear panel from 2001-2013
 - ▶ Identify survey clusters (villages) within 50km of each PMGSY village

Empirical Strategy: Fuzzy RDD

First Stage:

$$\begin{aligned} Road_{vdst} = & \gamma_0 + \gamma_1 1(pop_{vds} \geq T) + \gamma_2 (pop_{vds} - T) \\ & + \gamma_3 (pop_{vds} - T) \times 1(pop_{vds} \geq T) + \theta X_{vds} + \mu_{d,h} + \rho_t + \epsilon_{vdst} \end{aligned}$$

Second Stage:

$$\begin{aligned} Y_{vdst} = & \beta_0 + \beta_1 Roads_{vdst} + \beta_2 (pop_{vds} - T) \\ & + \beta_3 (pop_{vds} - T) \times 1(pop_{vds} \geq T) + \delta X_{vds} + \eta_{d,h} + \omega_t + \epsilon_{vdst} \end{aligned}$$

- ▶ $Road_{vdst} = 1$ if village v received road by 2013
- ▶ Y_{vdst} is outcome (fires, pollution, mortality)
- ▶ pop_{vds} is village population in 2001
- ▶ T is treatment threshold (500 or 1000, depending on state)

Covariate Balance

Table 1: Main analysis sample: summary statistics, and balance and falsification tests

| Variable | Full sample | Below threshold | Over threshold | Difference of means | p-value on difference | RD estimate | p-value on RD estimate |
|--|-------------|-----------------|----------------|---------------------|-----------------------|-------------|------------------------|
| Primary school | 0.959 | 0.955 | 0.964 | 0.01 | 0.02 | -0.018 | 0.59 |
| Medical center | 0.166 | 0.155 | 0.177 | 0.02 | 0.00 | -0.097 | 0.14 |
| Electrified | 0.430 | 0.414 | 0.447 | 0.03 | 0.00 | -0.014 | 0.87 |
| Distance from nearest town (km) | 26.490 | 26.379 | 26.613 | 0.23 | 0.58 | -3.426 | 0.34 |
| Land irrigated (share) | 0.281 | 0.276 | 0.287 | 0.01 | 0.05 | -0.025 | 0.59 |
| Ln land area | 5.151 | 5.093 | 5.215 | 0.12 | 0.00 | -0.103 | 0.33 |
| Literate (share) | 0.457 | 0.454 | 0.461 | 0.01 | 0.01 | -0.012 | 0.62 |
| Scheduled caste (share) | 0.143 | 0.141 | 0.145 | 0.00 | 0.24 | -0.020 | 0.52 |
| Land ownership (share) | 0.733 | 0.733 | 0.732 | -0.00 | 0.75 | 0.013 | 0.72 |
| Subsistence ag (share) | 0.435 | 0.438 | 0.432 | -0.01 | 0.25 | 0.024 | 0.58 |
| HH income > INR 250 (share) | 0.754 | 0.752 | 0.757 | 0.00 | 0.37 | -0.023 | 0.63 |
| Outcomes at baseline: | | | | | | | |
| Annual fires (count) | 0.685 | 0.699 | 0.670 | -0.03 | 0.40 | 0.194 | 0.50 |
| Annual PM2.5 ($\mu\text{g}/\text{m}^3$) | 35.886 | 35.821 | 35.958 | 0.14 | 0.41 | -0.029 | 0.94 |
| Downwind infant mortality (0/1) | 0.067 | 0.068 | 0.067 | -0.00 | 0.28 | 0.002 | 0.63 |
| Other directions infant mortality (0/1) | 0.072 | 0.072 | 0.071 | -0.00 | 0.23 | -0.002 | 0.72 |
| Downwind PM 2.5 ($\mu\text{g}/\text{m}^3$) | 36.038 | 35.879 | 36.252 | 0.37 | 0.04 | 0.431 | 0.36 |
| Other directions PM 2.5 ($\mu\text{g}/\text{m}^3$) | 39.254 | 38.646 | 40.237 | 1.59 | 0.00 | 0.901 | 0.15 |
| N | 11151 | 5859 | 5292 | | | | |

Results

| | New road | Annual fire activity | | Annual average PM 2.5 | |
|----------------------|-----------------------------|----------------------|--------------------|-----------------------|--------------------|
| | (1) | (2) | (3) | (4) | (5) |
| | <i>1st stage</i> | RF | IV | RF | IV |
| Above threshold pop. | 0.230*** (0.017) | 0.366** (0.152) | | 0.106** (0.045) | |
| Road built | | | 1.567** (0.689) | | 0.470** (0.207) |
| N | 133,788 | 133,788 | 133,788 | 133,788 | 133,788 |
| Control group mean | 0.10 | 2.77 | 2.77 | 42.69 | 42.69 |

- ▶ Road access increases annual crop fires by 60%
- ▶ Road access increase PM2.5 level by $0.5 \mu\text{g}/\text{m}^3$ (1.1%)

Does pollution result also capture increased driving on new roads?

| Panel A: Winter harvest and post-harvest months | | | | | | | |
|---|---------------------|--------------------|-------------------|---------------------|--------------------|-------------------|-------------------|
| | Fires | All sources | | Biomass burning | | Other sources | |
| | (1) | (2) | (3) | (4) | (5) | (6) | (7) |
| | | BC | OC | BC | OC | BC | OC |
| Road built | 1.051*** (0.391) | 0.285** (0.130) | 3.146* (1.777) | 0.320*** (0.103) | 2.791** (1.180) | -0.010 (0.009) | -0.010 (0.009) |
| N | 133,788 | 133,788 | 133,788 | 133,788 | 133,788 | 133,788 | 133,788 |
| Control group mean | 1.45 | 30.47 | 120.81 | 1.17 | 13.46 | 29.30 | 107.35 |
| Panel B: Rest of the year | | | | | | | |
| | Fires | All sources | | Biomass burning | | Other sources | |
| | (1) | (2) | (3) | (4) | (5) | (6) | (7) |
| | | BC | OC | BC | OC | BC | OC |
| Road built | 0.276 (0.438) | -0.219 (0.351) | -0.115 (3.120) | 0.192 (0.246) | 0.886 (2.680) | -0.010 (0.009) | -0.010 (0.009) |
| N | 133,788 | 133,788 | 133,788 | 133,788 | 133,788 | 133,788 | 133,788 |
| Control group mean | 1.27 | 30.83 | 124.70 | 1.53 | 17.34 | 29.31 | 107.36 |

Mechanism: Roads increase ag wage rate

- ▶ Data: REDS Survey (1999 and 2006)
- ▶ Design: Diff-in-Diff (221 villages, 100 districts, 17 states)
 - ▶ Treat = 1 if v received road b/w 1999-2006; Post=1 in 2006

$$Y_{vt} = \alpha + \beta_1(Treat_v \times Post_t) + \gamma_v + \theta_t + \epsilon_{vt}$$

| | Wage rate | | | Log wage | | |
|--------------------|-------------------|---------------------|--------------------|-------------------|--------------------|--------------------|
| | (1) Male | (2) Female | (3) Average | (4) Male | (5) Female | (6) Average |
| Treat X Post | 0.722* (0.420) | 0.918*** (0.349) | 0.856** (0.352) | 0.012* (0.007) | 0.020** (0.008) | 0.016** (0.007) |
| N | 442 | 442 | 442 | 442 | 442 | 442 |
| Control group mean | 62.99 | 47.65 | 55.25 | 62.99 | 47.65 | 55.25 |
| R ² | 0.76 | 0.78 | 0.77 | 0.75 | 0.76 | 0.76 |

Mechanism: Roads induce labor exit

Back to RDD design:

| | Share of labor in agriculture | | Share of non-agricultural manual labor | |
|--------------------|----------------------------------|-----------------------------|---|-----------------------------|
| | (1) High rel. ag. wage | (2) Low rel. ag. wage | (3) High rel. ag. wage | (4) Low rel. ag. wage |
| Road built | -0.029 (0.046) | -0.245*** (0.093) | 0.030 (0.046) | 0.210** (0.092) |
| N | 5,402 | 5,483 | 5,402 | 5,483 |
| Control group mean | 0.49 | 0.46 | 0.45 | 0.46 |

- More labor exit in districts with lower baseline ag wages

Mechanism: places with more labor exit burn more biomass

| | High rel. ag. wage | | Low rel. ag. wage | |
|--------------------|--------------------|---------|-------------------|---------|
| | (1) | (2) | (3) | (4) |
| | Fires | PM 2.5 | Fires | PM 2.5 |
| Road built | 0.778 | 0.378* | 3.195** | 0.617* |
| | (0.721) | (0.220) | (1.439) | (0.367) |
| N | 62,880 | 62,880 | 67,740 | 67,740 |
| Control group mean | 2.68 | 45.38 | 2.88 | 40.09 |

- ▶ Fire/pollution impact greater in places with lower baseline ag wage
- ▶ Road generates more labor exit in these places

- ▶ Do you believe the story?
- ▶ Alternative explanations

Next week

- ▶ Presentations
- ▶ How does the environment affect development?
 - ▶ **Health**
 - ▶ Productivity