

Human and Nature: Economies of Density and Conservation in the Amazon Rainforest

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Tropical Forests and Indigenous Populations

- Home to much of the world's bio-diversity and natural resources
- Rapid deforestation and bio-diversity loss
- The Peruvian Amazon (and many other rainforest areas):
 - Without modern technology and external investments
 - Primary activities: agriculture (shifting cultivation), fishing, hunting



Human-Nature Interactions across Space in Rainforests

Policymaker's problems:

- Trade-off between rainforest conservation vs. local populations' welfare:

e.g. Cost of forest clearing $\uparrow \Rightarrow$ Agriculture income $\downarrow \Rightarrow$ Welfare \downarrow

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e.g. Sectoral reallocation: Cost of forest clearing $\uparrow \Rightarrow$ Fishing/hunting $\uparrow \Rightarrow$ Bio-diversity \downarrow

Spatial reallocation: Protected areas \Rightarrow Resource depletion in other locations \uparrow

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Research question

Are there policies that improve both local populations' welfare and ecological conservation?

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- ① Builds a spatial general equilibrium model with multiple rural sectors
- ② Estimates the model using data from river basins in the Peruvian Amazon
 - Agglomeration:** Population density $\uparrow \Rightarrow$ Productivity \uparrow
 - Congestion:** Population density $\uparrow \Rightarrow$ Productivity \downarrow

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 - Congestion:** Population density $\uparrow \Rightarrow$ Productivity \downarrow
- ③ Evaluates counterfactuals (protection policies & transport infrastructure)

This Paper: Preview of Findings

Estimating density externalities by exploiting the structure of river networks:

- Agglomeration in agriculture > Congestion in access to land
Concentration \Rightarrow Productivity \uparrow & Deforestation *per farmer* \downarrow
- Congestion with spatial spillovers in natural resource extraction

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Counterfactuals:

- Well-targeted place-based protection policies and transport infrastructure are complementary to improving human & ecological well-being:

Human welfare \uparrow Deforestation \downarrow Natural resource depletion \downarrow

e.g. Direction of deforestation impacts depends on where transport infrastructure is improved

Literature

1. Trade-offs between economic development and environmental goals

- **Agriculture and deforestation:** Abman et al. (2020); Angelsen (1999JDE; 2010PNAS); Carreira et al. (2022); Foster et al. (2002); Szerman et al. (2022)
- **Policy discussion of combating deforestation:** Alix-Garcia et al. (2015AEJ); Assunçao et al. (2022); Hsiao (2022); Jayachandran (2022); Sousa-Rodrigues (2019REStud)
- **Commons:** Dasgupta and Mäler (1995); Ostrom (1990); Ryan & Sudarshan (2022JPE)

Contributions

- Conservation in GE
- Agricultural productivity ↑ & Deforestation ↓
- Policies

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Contributions

- Conservation in GE
- Agricultural productivity ↑ & Deforestation ↓
- Policies

2. Economic geography and quantitative spatial/urban models

- **Environment in spatial models:** Balboni (2020); Costinot et al. (2016JPE); Cruz & Rossi-Hansberg (2021); Hollingsworth et al. (2022); Jedwab et al. (2022)
- **Agricultural trade:** Pellegrina (2022JDE); Sayre (2022); Sotelo (2020JPE)
- **Agglomeration economies:** Ahlfeldt et al. (2015ECTA); Duranton & Puga (2020JEP)

Contributions

- Agglomeration externality in rainforests
- Endogenize deforestation & wild resource depletion
- External validity

Setting & Data

2 Reasons why the Peruvian Amazon is an ideal setting

- ① Traditional ways of life in remote areas without modern technology and large-scale external investments (in contrast to the Brazilian Amazon)
 - ⇒ Attribute resource extractions to small-scale farmers and hunter-gatherers & focus on density externalities that they cause
 - * Small-scale deforestation recently increased in the Amazon (Kalamandeen et al. 2018)
- ② River networks almost solely constitute the transportation routes
 - ⇒ Identify key structural parameters by exploiting exogenous river shapes



Primary Livelihoods in the Peruvian Amazon

1. Agricultural sector:

- **Food crops** (manioc; farina; plantain)
- **Cash crops** (rice; maize; beans; sugar cane; vegetables; fruits)

2. Extractions of natural resource products:

- **Wild animal extractions:**

- Fishing
- Game meat (Hunting)

- **Forest products:**

- NTFP (Non-timber forest products)
- Timber

3. Non-food manufacturing items are typically from an urban center

Primary Data

1. Grid cell-level (1km × 1km) information:

- Geographical and remote sensing information
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 - Sectoral populations, prices, transport modes, and many others
- Community-level land footprint from satellite images

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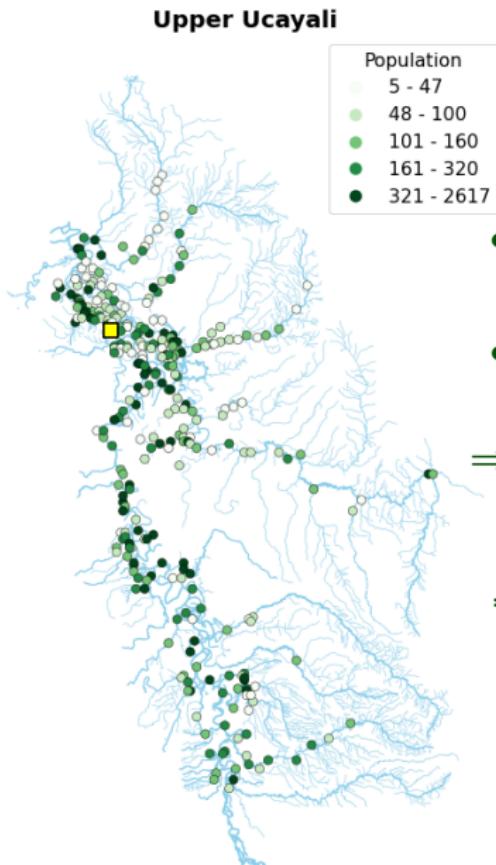
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3. National censuses by National Institute of Statistics and Informatics (INEI):

- Peru Population and Housing Census (2007, 2017)
 - Complement population information (esp. urban populations)
- Peruvian Agricultural Census (2012) → Technology use by all producers

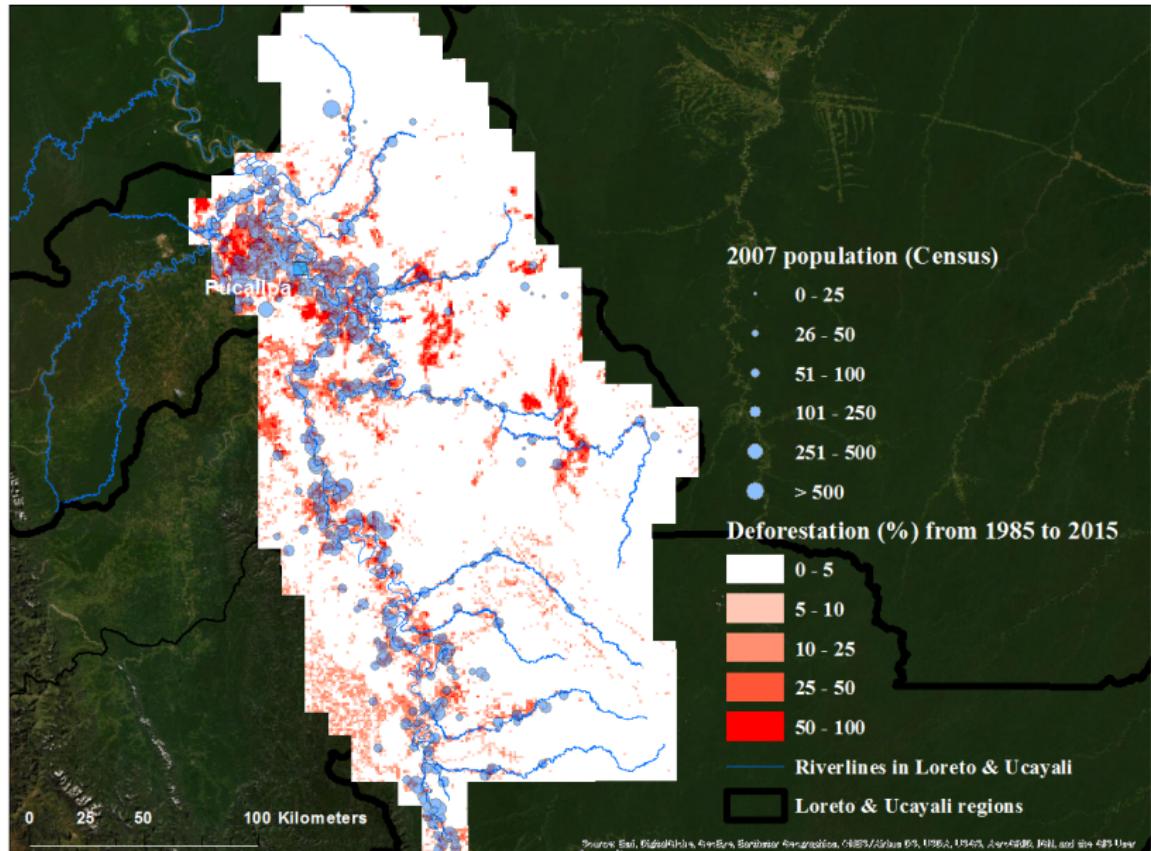
Facts

Spatial Concentration and Dispersion of Communities and Populations

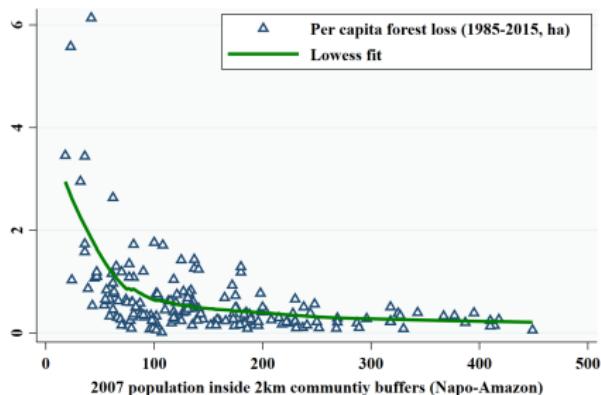
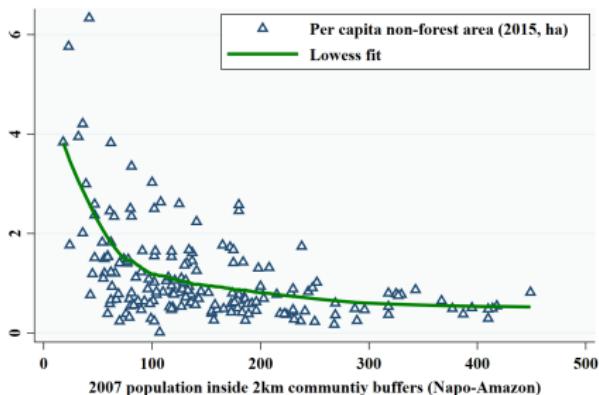


- The legend is based on quantiles (throughout the presentation)
 - $\approx 80\%$ of the rural communities have populations smaller than 320
 - ⇒ Suggestive of *agglomeration and congestion forces in economic activities*
-
- * Mostly indigenous and folk populations
Colonist settlements $\approx 1.4\%$ of communities

- (+) Human Settlements & Deforestation
- > 50% of total deforestation: within 5km from the census communities



Negative and Convex Relationship between Population & Per Capita Land Footprint



Suppose, for simplicity, take this as structural, then it implies:

- Congestion force in forest clearing (without land market)
- A mean-preserving reduction in the variance of the settlement size can reduce total deforestation

Stylized Facts: Summary

1. Spatial concentration and dispersion of populations and communities
⇒ Suggestive of *strong agglomeration and congestion forces of economic activities*

2. Negative & convex relationship b/w population & per capita land footprint
⇒ Suggestive of:
 - *Congestion force in forest clearing*
 - *Reducing total deforestation by spatial reallocation of populations*

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 - 3A. Agriculture is widely observed in both concentrated and dispersed areas
 - 3B. Natural resource extraction is observed more in communities with **lower population densities in the surrounding area** ▶ Detail
- ⇒ Suggests that *the spatial extent over which density externalities operate is distinct across sectors*

Model

Spatial Model of Rainforest Communities: Intuition

- The balance between concentration and dispersion forces determines the spatial distribution of economic activities

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 - Higher market access
 - Proximity to an urban center

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- The balance between concentration and dispersion forces determines the spatial distribution of economic activities
- Concentration forces:
 - Agglomeration in agricultural production
 - Higher market access
 - Proximity to an urban center
- Dispersion forces:
 - Congestion in forest clearing
 - Congestion in natural resource extraction

Setup

GE in a river basin, consisting of locations (1km^2 cells) $o, d \in \mathcal{I} = \mathcal{R} \cup u$:

- \mathcal{R} : the set of multiple **rural** locations & u : the urban center
- Locations differ in productivity fundamentals (e.g. soil quality; access to water) and connectivity to other locations in the river network

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3 Sectors:

- **Agriculture** (Ag): produced in **rural** locations
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Population:

- Each consumer inelastically supplies one unit of labor and gets wage (w_o)
- Perfectly mobile across locations while total population in the basin is fixed

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Prices & Trade:

- Perfect competition: Price = Marginal cost
- **Asymmetric trade costs** due to river orientations

Agriculture with Congestion & Agglomeration Externalities

Land access by forest clearing: $L_o(j) = \underbrace{A_{o,L} N_{o,Ag}^{-\mu_L}}_{\text{productivity}} \cdot N_{o,L}(j)$

Final output: $Q_{o,Ag}(j) = \underbrace{z_{o,Ag}(j) N_{o,Ag}^{\mu_{Ag}}}_{\text{productivity}} \cdot N_{o,C}(j)^{\gamma} L_o(j)^{(1-\gamma)}$

- $N_{o,L}$ & $N_{o,C}$: Employment for forest clearing & cropping
- $N_{o,Ag} = N_{o,L} + N_{o,C}$: Total employment for agriculture

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- $N_{o,Ag} = N_{o,L} + N_{o,C}$: Total employment for agriculture
- μ_L : Parameter governing **congestion forces in forest clearing** Why?
(consistent with the stylized fact)
- μ_{Ag} : Parameter governing **agglomeration forces in agricultural production and marketing**
- $A_{o,L}$: productivity fundamentals
- $z_{o,Ag}(j)$: Fréchet shock of variety j productivity
(θ : comparative advantage; $A_{o,Ag}$: absolute advantage)

Natural Resources with Congestion Externality across Space

$$Q_o^{Nr}(j) = z_{o,Nr}(j) \underbrace{\left[\sum_{d \in \mathcal{R}} D_{od}^{-\nu} N_{d,Nr} \right]^{-\mu_{Nr}}}_{\text{productivity}} \cdot N_{o,Nr}(j)$$

- People travel for natural resource extraction
- $N_{o,Nr}$: Total employment for natural resource extraction in o

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- People travel for natural resource extraction
- $N_{o,Nr}$: Total employment for natural resource extraction in o
- μ_{Nr} : Parameter governing **congestion forces with spatial spillovers from surrounding populations**
- D_{od} : River-equivalent distance between cells o & d in the shortest path
- ν : Parameter governing **spatial decay in accessing natural resources**
- $z_{o,Nr}(j)$: Fréchet shock of variety j productivity

Spatial Equilibrium Conditions

- ① Sectoral labor markets (Ag , Nr , M) clear
- ② The overall labor market clears (given the fixed total population in each basin)
- ③ Utility is equalized across populated locations (due to free labor mobility)
- ④ The total deforested area does not exceed the available land area

Estimation

Estimating the Model in a Sequential Procedure

Parameter	Description	Estimation strategy
δ_K	Elasticity of trade cost ($K = Ag, Nr, M$)	Commodity prices from the CC
$\lambda_{up}, \lambda_{land}$	Relative costs in terms of downstream river	Travel time and transport costs survey
σ	Within-sector elasticity of substitution	Expenditure information from ENAHO
$\bar{\sigma}$	Across-sector elasticity of substitution	Expenditure information from ENAHO
γ	Labor share in agricultural production	From the literature
θ	Trade elasticity	From the literature
μ_L	Congestion in forest clearing	Linear IV using the community-level data
μ_{Ag}	Agglomeration in agricultural production	Model inversion and linear IV
μ_{Nr}	Congestion in natural resource extraction	Model inversion and non-linear GMM
ν	Spatial decay in natural resource access	Model inversion and non-linear GMM
$\{A_{o,K}\}$	Absolute advantages ($K = Ag, Nr$)	Calibration
A_M	Absolute advantages in the urban sector	Calibration

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- ② Given the parameters obtained in the previous steps, invert the model to recover wages and productivities that rationalize the observable endogenous variables as a spatial equilibrium

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- ③ Employ GMM to estimate density externality parameters from inverted productivities obtained in the previous step

Step 2. Model Inversion

- Observable data: sectoral populations ($\{N_{o,Ag}\}$, $\{N_{o,Nr}\}$, $\{N_{u,M}\}$)
- Use the $2|\mathcal{R}| + 1 + |\mathcal{I}|$ equations from the spatial equilibrium conditions (sectoral labor market clearing + utility equalization across space) with the observables to solve for $2|\mathcal{R}| + 1 + |\mathcal{I}|$ unknowns (productivity composites + wages): $\{\tilde{A}_{o,Ag}\}$, $\{\tilde{A}_{o,Nr}\}$, $A_{u,M}$, $\{w_o\}$
→ Use the inverted productivity composites as data in the next step

Step 3. Density Externalities in Agriculture

- Inverted productivity composites of agriculture:

$$\tilde{A}_{o,Ag} \equiv \underbrace{A_{o,Ag} A_{o,L}^{(1-\gamma)\theta} K_1^\theta}_{\text{fundamentals}} \cdot \underbrace{N_{o,Ag}^{\tilde{\mu}_{Ag}\theta}}_{\text{externalities}} \quad \text{where } \tilde{\mu}_{Ag} \equiv \mu_{Ag} - (1-\gamma)\mu_L$$

- Procedure:

- ① Estimate $\tilde{\mu}_{Ag}$
- ② Estimate μ_L
- ③ Back out μ_{Ag}

Step 3. Agglomeration Externality in Agriculture via Linear IV

- Inverted productivity composites of agriculture:

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- Empirical specification:

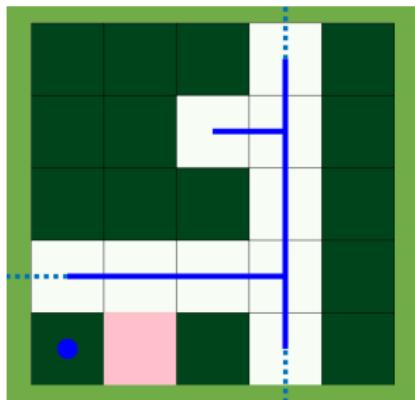
$$\ln \tilde{A}_{o,Ag} = \tilde{\mu}_{Ag}\theta \ln N_{o,Ag} + X'_o \beta + \phi_B + \epsilon_{o,Ag}$$

- “River Network Access” as an IV for $\ln N_{o,Ag}$:

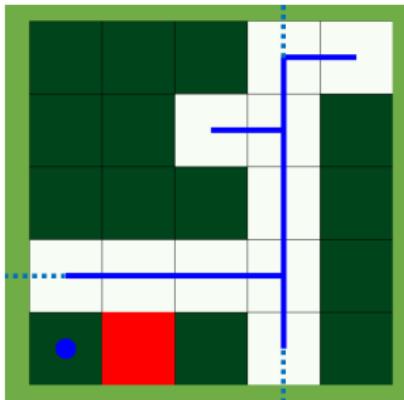
$$RNA_o \equiv \sum_{d \in RC} (\tilde{\tau}_{od})^{-\theta} \quad \text{where } RC: \text{grid cells with rivers}$$

Intuition of Identifying the Density Externalities

$RNA_o \equiv \sum_{d \in RC} (\tilde{\tau}_{od})^{-\theta}$: River Network Access (RC : white cells)



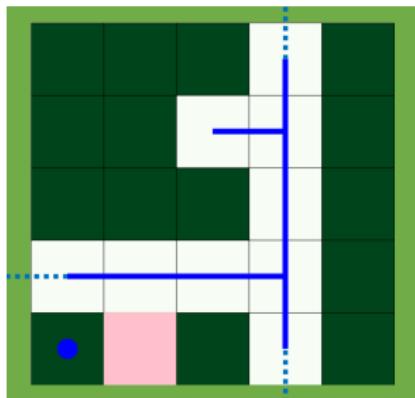
RNA at the pink cell <



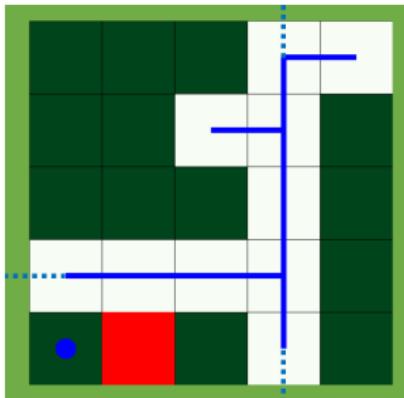
RNA at the red cell

Intuition of Identifying the Density Externalities

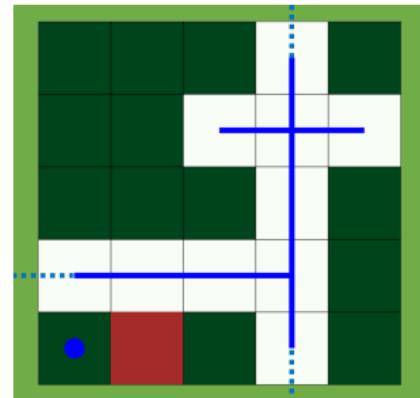
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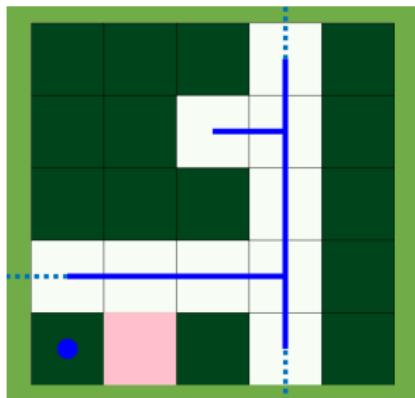
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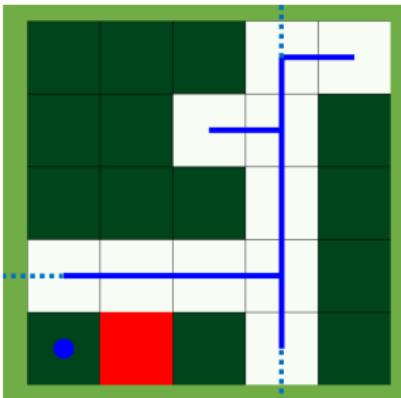
< RNA at the brown cell

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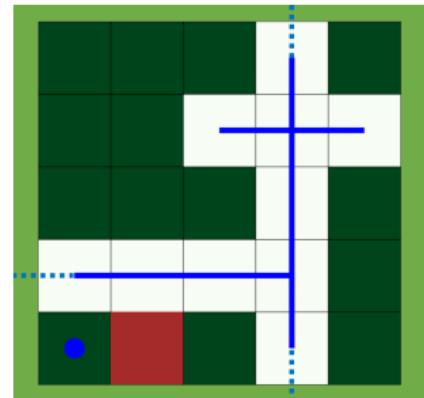
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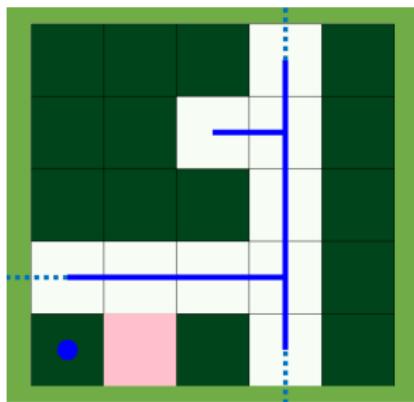


< RNA at the brown cell

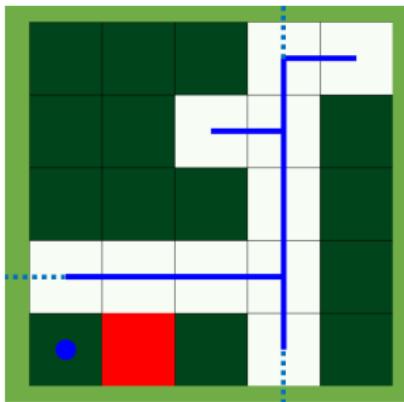
- **Independence:** Given the same observable agricultural conditions (river proximity, water areas, soil conditions, etc), unobservable productivity fundamentals are uncorrelated with the variation in RNA *that stems from exogenous river shapes in locations far away from the own location*

Intuition of Identifying the Density Externalities

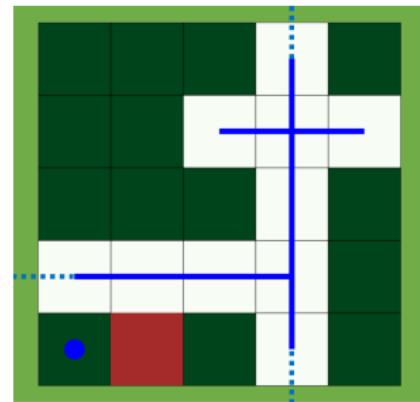
$RNA_o \equiv \sum_{d \in RC} (\tilde{\tau}_{od})^{-\theta}$: River Network Access (RC : white cells)



RNA at the pink cell



RNA at the red cell



< RNA at the brown cell

- **Independence:** Given the same observable agricultural conditions (river proximity, water areas, soil conditions, etc), unobservable productivity fundamentals are uncorrelated with the variation in RNA *that stems from exogenous river shapes in locations far away from the own location*
- **Exclusion:** RNA (as a market potential shifter) affects productivity only through its effect on employment and thus through externalities that arise

Step 3. Agglomeration Externality in Agriculture via Linear IV

- Inverted productivity composites of agriculture:

$$\tilde{A}_{o,Ag} \equiv \underbrace{A_{o,Ag} A_{o,L}^{(1-\gamma)\theta} \kappa_1^\theta}_{\text{fundamentals}} \cdot \underbrace{N_{o,Ag}^{\tilde{\mu}_{Ag}\theta}}_{\text{externalities}} \quad \text{where } \tilde{\mu}_{Ag} \equiv \mu_{Ag} - (1-\gamma)\mu_L$$

- Empirical specification:

$$\ln \tilde{A}_{o,Ag} = \tilde{\mu}_{Ag}\theta \ln N_{o,Ag} + X'_o \beta + \phi_B + \epsilon_{o,Ag}$$

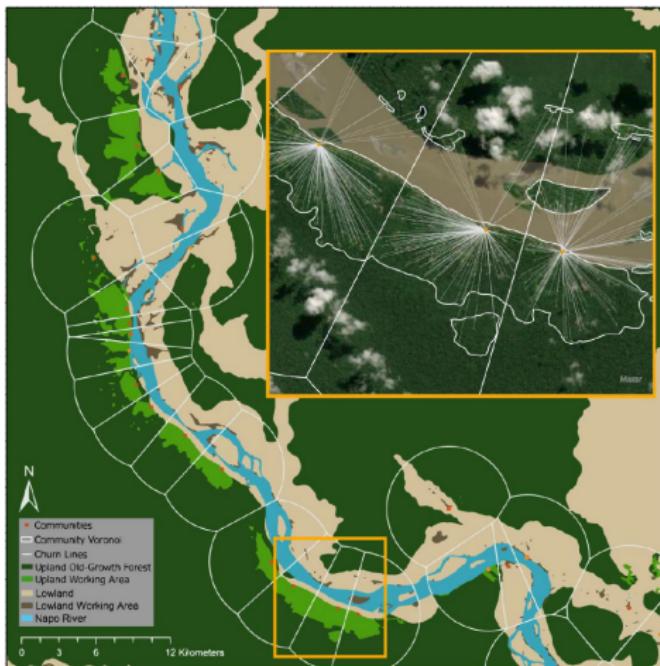
- “River Network Access” as an IV for $\ln N_{o,Ag}$:

$$RNA_o \equiv \sum_{d \in RC} (\tilde{\tau}_{od})^{-\theta} \quad \text{where } RC: \text{grid cells with rivers}$$

- Identifying assumption:

After controlling for geographic characteristics of the own location, productivity fundamentals are uncorrelated with accessibility to other locations

- A similar strategy for estimating μ_L due to common residual factors using community-level land footprint obtained from satellite images



Voronoi Polygons and Land Footprint around the Census Communities

Notes: To proxy community boundaries for agricultural land use, we partition land in the study area into voronoi polygons. Within each community voronoi polygon, we detect all patches of agricultural fields and secondary forests through satellite images. We then sum them up to calculate the land footprint of each community. See Coomes et al. (2021) for more details.

$$\text{Empirical specification: } \ln \frac{L_o}{N_{o,Ag}} = -\mu_L \ln N_{o,Ag} + X'_o \beta + \phi_B + \epsilon_{o,L}$$

3. Congestion in Natural Resource Extraction with Spatial Spillovers via Non-Linear GMM

- Inverted productivity composites of natural resource extraction:

$$\tilde{A}_{o,Nr} \equiv \underbrace{A_{o,Nr}}_{\text{fundamentals}} \cdot \underbrace{\left[\sum_d D_{od}^{-v} N_{d,Nr} \right]^{-\mu_{Nr}\theta}}_{\text{externalities}}$$

- Moment conditions:

$$\mathbb{E}[\epsilon_{o,Nr} \ln RCA_o] = 0 \quad \text{and} \quad \mathbb{E}[\epsilon_{o,Nr} \ln \left(\sum_{d|D_{o,d} \leq x} RCA_d \right)] = 0, \quad x \in X$$

- $\epsilon_{o,Nr}$: the residual variation in $\ln A_{o,Nr}$ (productivity fundamentals)
- $X = \{2, 5, 10, 25, 50, 75, 100 \text{ (km)}\}$
- Similar to the identification strategy by Ahlfeldt et al. (2015)
- Estimate v & μ_{Nr} by the two-step nonlinear GMM

▶ Across-sector externality?

3. Density Externalities in Agriculture and Natural Resource Extraction

Parameter	Point estimate	Standard error	Description
(A) Agriculture			
$\tilde{\mu}_{Ag}$	0.064	0.010	$= \mu_{Ag} - (1 - \gamma)\mu_L$
	<i>J test p-value = 0.648</i>		
μ_L	0.522	0.094	Congestion in forest clearing
μ_{Ag}	0.273		Agglomeration in agricultural production
(B) Natural resource extraction			
μ_{Nr}	0.335	0.042	Congestion in natural resource extraction
ν	0.593	0.075	Spatial decay of congestion externality
	<i>J test p-value = 0.821</i>		

Notes: Estimates of density externalities in agriculture (panel A) are based on the linear specification using $\ln RNA_o$ and the initial community existence in 1940 as instruments. Estimates of parameters governing congestion externality in natural resource extraction (panel B) are based on the non-linear GMM using $\ln RNA_o$ and $\{\ln \sum_{d|D_{o,d} \leq x} RNA_d\}$ for $x \in X = \{2, 5, 10, 25, 50, 75, 100\}$ as instruments.

- Robust to different IVs and samples (e.g. among small communities) ▶ Table
- Robust to different controls ▶ Table
- How important are these parameter values quantitatively? (next)

Quantitatively large effects of the agglomeration externality on

- improving human welfare & reducing deforestation
- at the expense of natural resource depletion

Counterfactual Outcomes without the Agglomeration Externality

Basin	Welfare (real wage)	Deforestation	Natural resource depletion
Napo	-11.2%	+55.5%	-0.6%
Pastaza	-8.9%	+24.1%	-0.7%
LowerUcayali	-12.7%	+21.4%	-2.8%
UpperUcayali	-7.1%	+17.7%	-1.2%

- $\mu_{Ag} \rightarrow 0$

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- $\mu_{Ag} \rightarrow 0$
- Spatial reallocation of agriculture: from concentration into dispersion because the congestion externality in forest clearing dominates
- Sectoral reallocation: more agricultural employment to feed the population given that consumption demands across Ag & Nr goods are complementary

Mechanisms behind the Agglomeration Externality in Agriculture

① Economies of scale in transport technology ▶ Detail

- Endogenous transport modes (different types of boats available)
- Endogenous transaction costs
- Trade costs decrease in origin populations
- Isomorphic to the original model

② Economies of scale in agricultural intensification ▶ Detail

Counterfactuals

Counterfactual Simulations: Overview

The aim is to design a “win-win” policy that simultaneously achieves:

- Local populations' welfare ↑
- Deforestation ↓
- Natural resource depletion ↓

Counterfactual policies:

- ① Resettlement and protection policies
- ② Improvement of river transport infrastructure
- ③ Combination of them

Resettlement Policies: Overview

Q. *Is it beneficial to concentrate the ecological footprint in fewer spots rather than having many small communities?*

Resettlement Policies: Overview

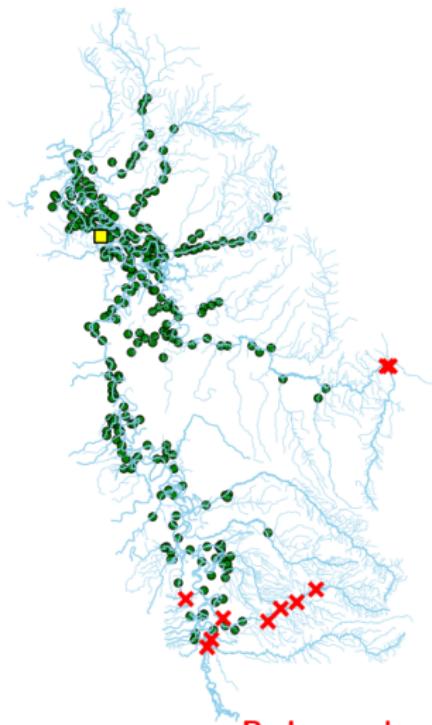
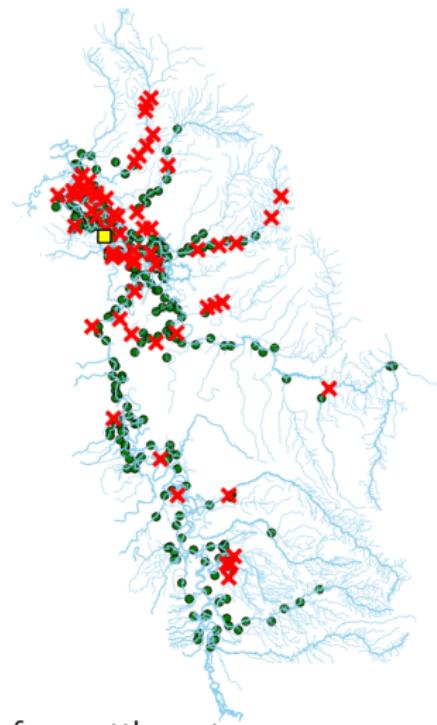
Q. *Is it beneficial to concentrate the ecological footprint in fewer spots rather than having many small communities?*

- (A)** Protecting the rural frontier (= maximum distance from the urban center to locations where human settlements are allowed)
- (B)** Not allowing for very small communities

Resettlement Policies: Overview

Q. Is it beneficial to **concentrate the ecological footprint in fewer spots rather than having many small communities?**

- (A)** Protecting the rural frontier (= maximum distance from the urban center to locations where human settlements are allowed)
- (B)** Not allowing for very small communities
- For meaningful comparison, **each policy directly treats an equal size of population**—2.5% of the total rural population in the benchmark equilibrium in each basin
- E.g.** (B) chooses rural locations to be treated in order, starting with the location that has the smallest population size, until the treated population reaches 2.5% of total population in each basin.

(A) Protecting the rural frontier**(B) Small communities not allowed**

Red x marks: Treated locations for resettlement

The comparison illustrates the planner's trade-off between mitigating different environmental costs

Basin	Welfare	Deforestation	Natural resource depletion
(A) Protecting the rural frontier			
Napo	-0.2%	-5.2%	-0.3%
UpperUcayali	-0.2%	-2.0%	-0.8%
(B) Not allowing for small communities			
Napo	-0.3%	-13.1%	+0.2%
UpperUcayali	-0.1%	-7.3%	+0.5%

(A)⇒ natural resource depletion ↓ but with a smaller deforestation impact

- Overall scope of natural resource extraction activities is narrowed
- Surrounding population density ↑ in most of populated areas
- Productivity ↓ due to the **congestion externality with spatial spillovers**

▶ Maps

(B)⇒ reduces deforestation the most but natural resource depletion ↑

Improving River Transport Infrastructure: Overview

- **High trade costs:**
 - Asymmetric transport costs due to river orientations
 - Seasonality of transport costs due to water level fluctuations
 - Slow speed of river boats
- **River transport infrastructure investments:**
 - Better quality boats
 - River dredging

Amazon Waterway Project: Government scheme with Chinese investment to deepen and widen the central parts of rivers to allow larger ships to travel

Improving River Transport Infrastructure: Overview

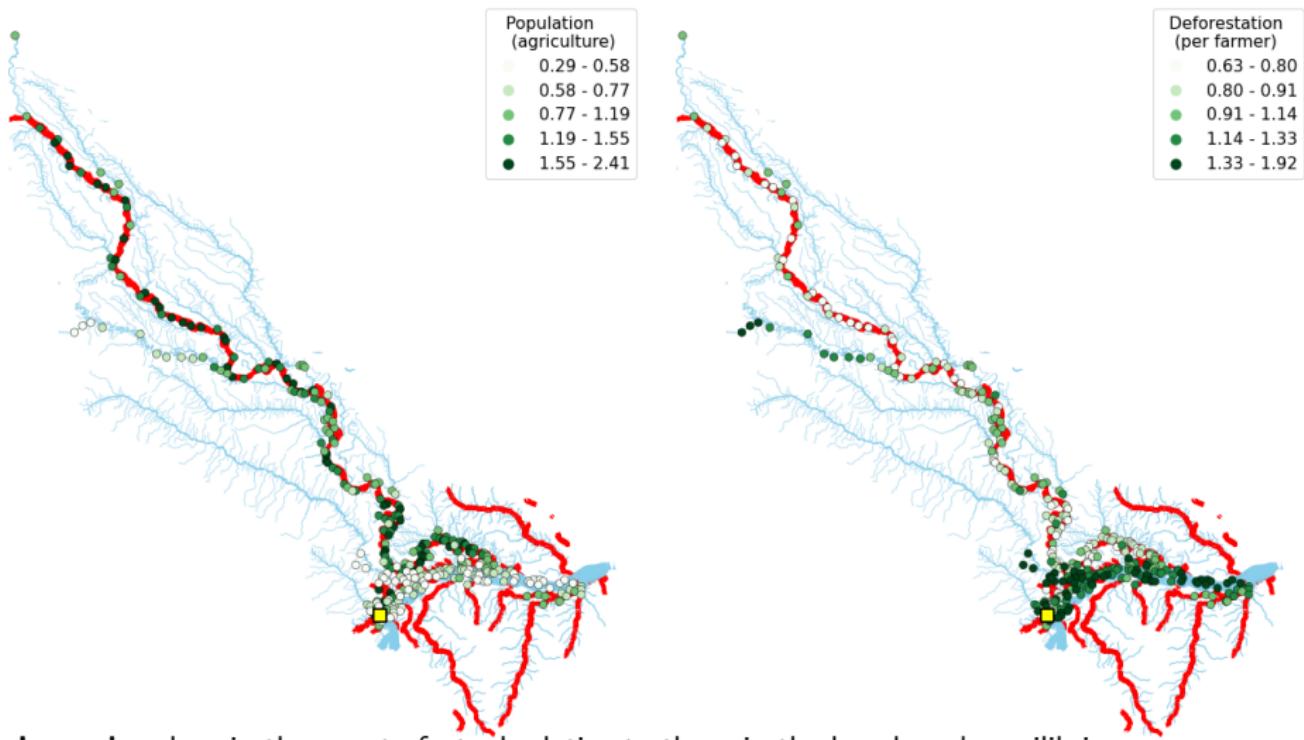
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 - Slow speed of river boats
- **River transport infrastructure investments:**
 - Better quality boats
 - River dredging
 - Amazon Waterway Project: Government scheme with Chinese investment to deepen and widen the central parts of rivers to allow larger ships to travel
- Replace the **downstream-river-equivalent distance** with:

$$D_{od} = D_{od,initial,down} + \lambda_{up} D_{od,initial,up} + \lambda_{land} D_{od,land} + \lambda_{upgraded} D_{od,upgraded}$$

where $\lambda_{upgraded} = 0.8$ & symmetric transport cost in the “upgraded” part

Connecting Hinterlands to the Central Area of a Basin ⇒

- Spatial reallocation of farmers toward remote areas
- Deforestation per farmer ↓ in remote areas



Legend: values in the counterfactual relative to those in the benchmark equilibrium

Red lines: transport infrastructure improvement

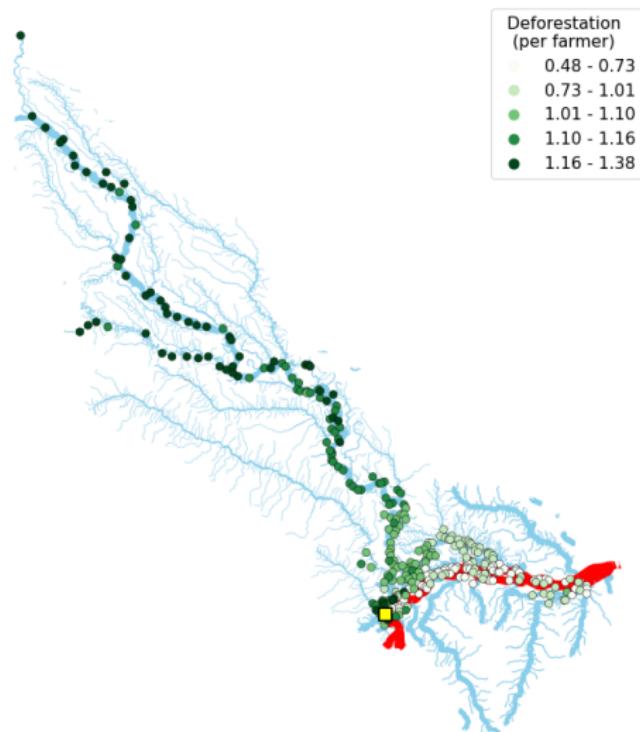
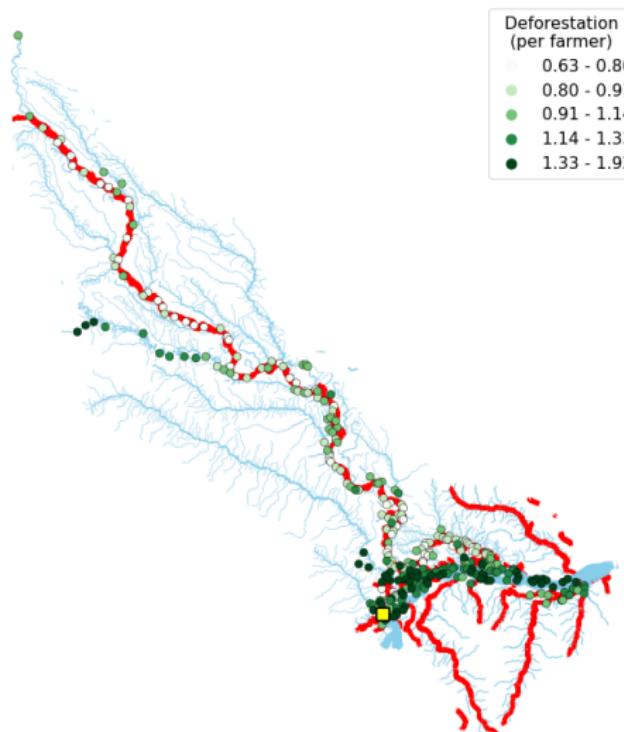
Connecting Hinterlands to the Central Area of a Basin ⇒

- Welfare ↑
- Deforestation ↓
- Natural resource depletion (↓)?

Basin	Welfare	Deforestation	Natural resource depletion
Napo	+1.8%	-1.1%	-0.3%
UpperUcayali	+1.2%	-1.1%	+0.3%

- The **agglomeration** benefits spread more evenly across the basin with more moderate-sized settlements
- Total deforestation decreases by reducing the variance of the settlement size in the river basin, given the structure of congestion forces in access to land (negative and convex between ag population and per-farmer deforestation)
- * The overall impact on natural resource depletion is unclear through GE effects

Connecting Hinterlands to the Center (left) vs. Concentrating the Infrastructure Investment in the Center (right)



Legend: values in the counterfactual relative to those in the benchmark equilibrium
Red lines: transport infrastructure improvement

Direction of deforestation impacts depends on where transport infrastructure is improved

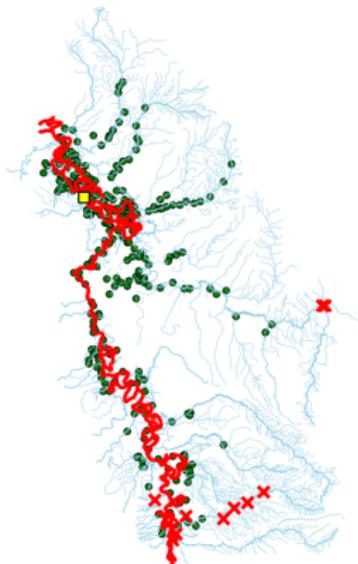
Basin	Welfare	Deforestation	Natural resource depletion
(A) Transport infrastructure improved by connecting hinterlands to the center			
Napo	+1.8%	-1.1%	-0.3%
(B) Transport infrastructure improved by concentrating investments in the center			
Napo	+1.0%	+6.0%	-0.4%

Improving the infrastructure only in densely populated areas \Rightarrow Deforestation \uparrow

- The population is more concentrated in the central area of the basin
- At the same time, **much smaller communities with much higher deforestation per farmer in hinterlands**

(A) Protecting the rural frontier &
(B) Transport infrastructure that connects hinterlands to the center

Basin	Welfare	Deforestation	Natural resource depletion
Napo	+1.6%	-6.7%	-0.6%
Pastaza	+1.0%	-4.7%	-0.3%
LowerUcayali	+2.1%	-1.0%	-2.4%
UpperUcayali	+1.0%	-3.1%	-0.5%

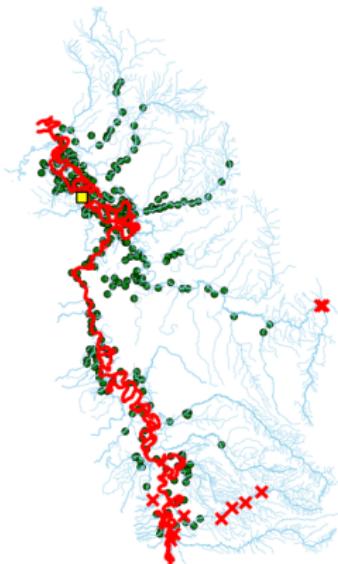


- The trade-offs are relaxed:
 - Welfare ↑
 - Deforestation ↓
 - Natural resource extraction ↓

(A) Protecting the rural frontier &

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- The trade-offs are relaxed:

- Welfare ↑
- Deforestation ↓
- Natural resource extraction ↓

(A) More compact basin for human settlements

⇒ congestion ↑ across populated areas

(B) Integrated between the center and hinterlands

⇒ Agglomeration benefits spread more evenly

Well-targeted protection policies and transport infrastructure are complementary to improving human and ecological well-being

Counterfactuals	Welfare	Deforestation	Natural resource depletion
(A) Protection policies			
i. Protecting the rural frontier	-	-	-
ii. Targeting the smallest communities	-	-	+
(B) Transport infrastructure			
i. Connecting hinterlands to the center	+	-	?
ii. Concentrating in the center	+	+	?
(A) i. + (B) i.	+	-	-

Conclusions

This paper:

- Applies a multi-sector spatial GE model to rainforest communities
- Estimates density externalities ([agglomeration](#) & [congestion](#))

RQ. *Are there policies that improve both local populations' welfare and ecological conservation?*

- * Combination of a protection policy and transport infrastructure investments
- * Spatial targeting matters for both
- * Even development is better for the environment than uneven development

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Future research:

- Dynamics: community formation, resource depletion, external investments
- Across-sector externalities
- Indigenous populations' values of their traditional ways of life

Appendix

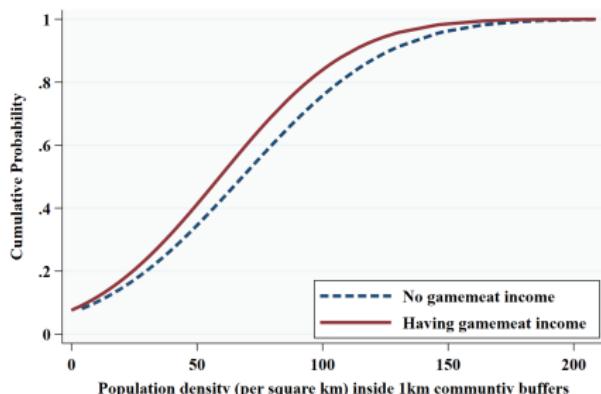
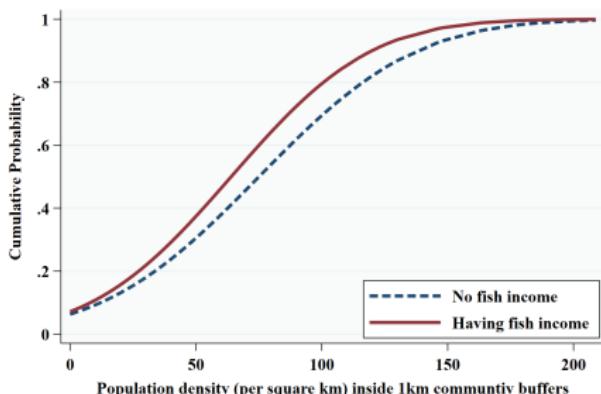
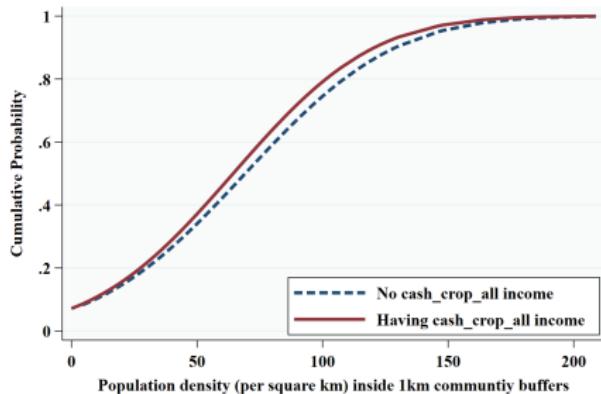
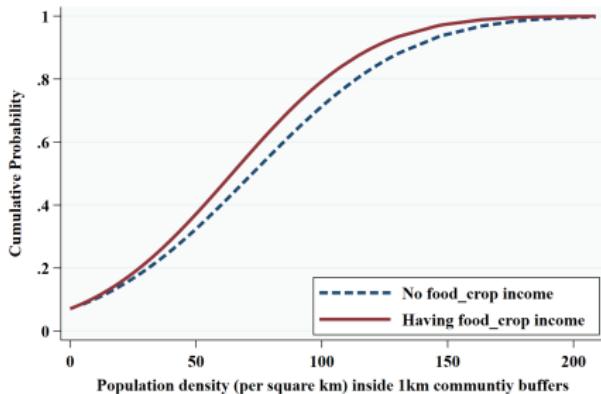
Static vs. Dynamic Variations in Forest Cover

- Shifting cultivation with the swidden-fallow cycle:
 1. Clear primary (old-growth) forests, burn the vegetation to obtain land plots, and plant crops
 2. When plots become no longer productive, plots are left in fallow and the secondary forest regrows
 3. After several years of fallow, such secondary forests become old-growth forest and farmers clear these forests again (→ 1.)

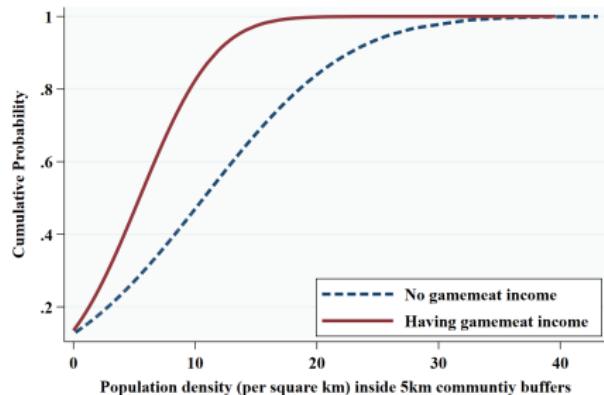
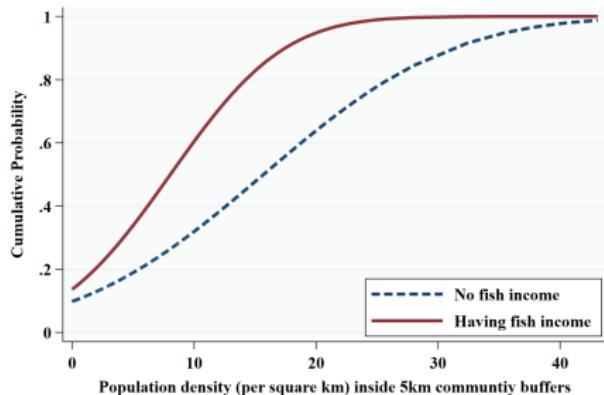
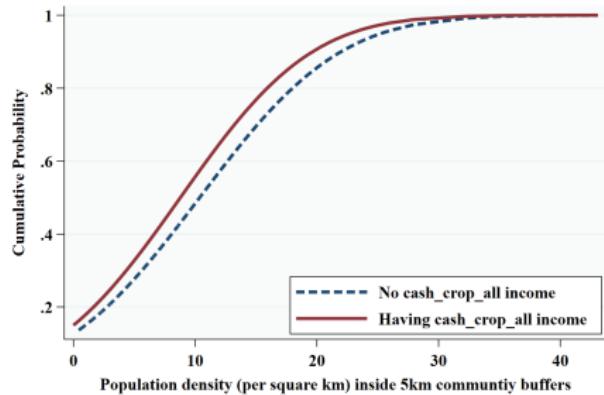
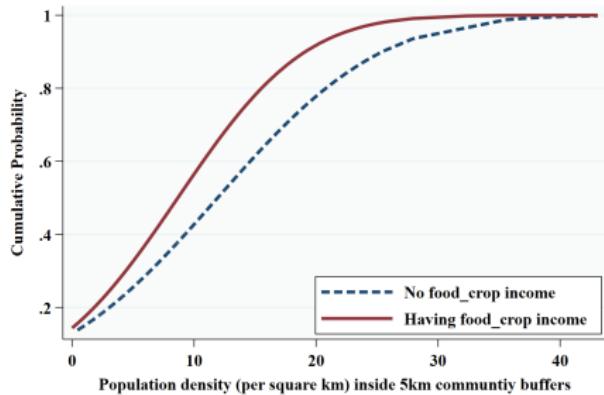
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 1. Clear primary (old-growth) forests, burn the vegetation to obtain land plots, and plant crops
 2. When plots become no longer productive, plots are left in fallow and the secondary forest regrows
 3. After several years of fallow, such secondary forests become old-growth forest and farmers clear these forests again (→ 1.)
- Although the deforested locations around the community are moving over time due to this cycle, at any given moment the stock of forest fallow and the total deforested area around the community remain relatively constant (Coomes et al. 2021)
 - ⇒ The cross-sectional relationship between the settlement size and deforestation has a more significant variation
 - ⇒ A static theoretical model

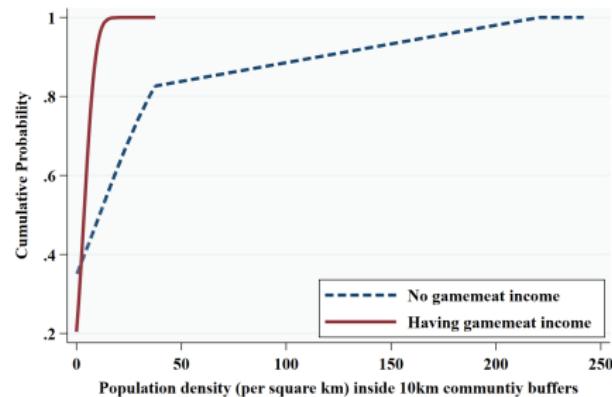
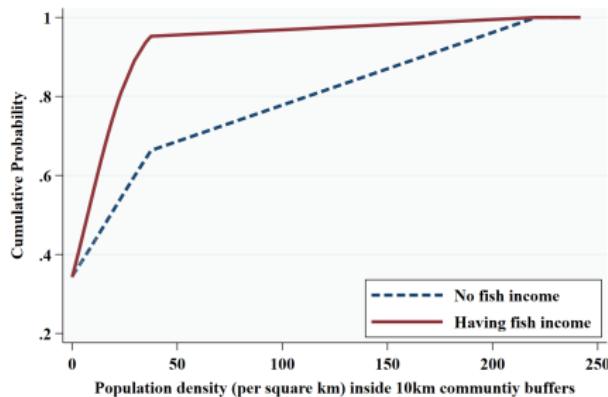
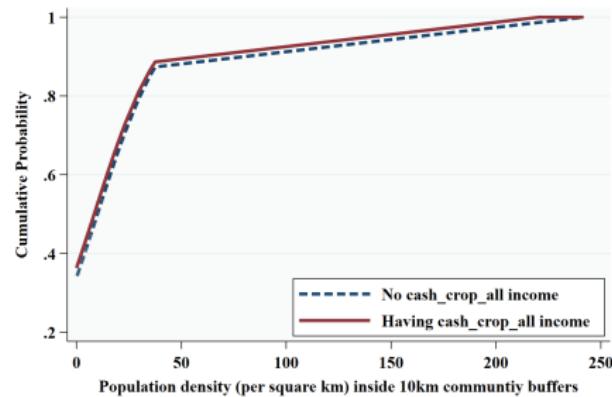
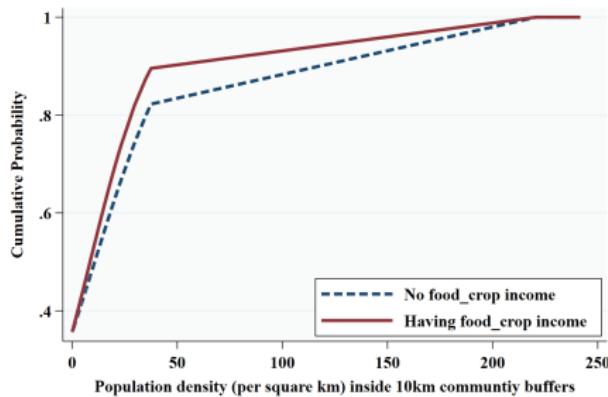
Population Density (1km buffer) & Activity/Occupation: Agricultural Sector vs. Wild Animal Extractions



Population Density (5km buffer) & Activity/Occupation: Agricultural Sector vs. Wild Animal Extractions



Population Density (10km buffer) & Activity/Occupation: Agricultural Sector vs. Wild Animal Extractions



Spatial Distribution of Agriculture and Natural Resource Extraction

- Population density in a smaller buffer of a community becomes close to the population density of that community itself
- A smaller population density in a large buffer of a community implies more sparseness of communities and less competition over natural resources
- **Spatial distributions of natural resource extractions become more contrastable from agricultural activities as the buffer size enlarges ($x = 5\text{km}, 10\text{km}$ compared to $x = 1\text{km}, 2\text{km}$)**

▶ Back

Congestion Forces in Land Access

▶ Back

- Farmers clear forests to obtain agricultural land only nearby their residential locations in the absence of the land market and property rights and with a high monitoring cost (mean/median of land footprint depths = 1 km/0.85km)
- The monitoring cost and the negotiation cost for allocating land areas to farmers increase as the agricultural population increases
- Therefore, deforested areas cannot increase proportionally to the increase in the population size in a community



- The cost may be very small among very small communities given the cooperative nature of forest clearing: consistent with the convexity

Step 1. Obtaining Parameters without Solving the Model

Parameter	Description	Estimation strategy	Value
δ_{Ag}	Elasticity of trade cost	Commodity prices from the CC	0.178
δ_{Nr}	Elasticity of trade cost	Commodity prices from the CC	0.137
δ_M	Elasticity of trade cost	Commodity prices from the CC	0.098
λ_{up}	Relative upstream-river travel cost	Travel time and transport costs survey	1.282
λ_{land}	Relative land travel cost	Travel time and transport costs survey	36.767
σ	Within-sector elasticity of substitution	Expenditure information from ENAHO	2.401
$\bar{\sigma}$	Across-sector elasticity of substitution	Expenditure information from ENAHO	0.752
γ	Labor share in agricultural production	From the literature	0.6
θ	Trade elasticity	From the literature	7.8

▶ Back

Geographic Controls

- River cell dummy
- Distance to the river point and its square
- Interaction between the above two
- River confluences
- Elevation
- Flood experience
- Geology measures
- Water (main and non-main) areas
- Distance to the urban center

▶ Back

Historical IV

IV: Community existence in its current location by 1940

- The primary reason for early settlement was the opportunity to extract natural resource products
 - The Amazon Rubber Boom:
 - Began the late 19th century, but collapsed around 1940
 - Significantly impacted initial settlements (Barham et al. 1996; Coomes 1995)
- ⇒ The locations of communities established before 1940 were likely to be determined primarily by natural resource endowments, not by advantages in agricultural productivity

▶ Back

Table: River Networks, Initial Communities, and Current Populations

	log($N_{o,Ag}$)			Community existence (1940)	
	(1)	(2)	(3)	(4)	(5)
log(RNA_o)	0.758*** (0.223)		0.711*** (0.218)	-0.0145 (0.0254)	0.0699 (0.0726)
Community existence (1940)		0.740*** (0.0983)	0.730*** (0.0980)		
Basin FE	Yes	Yes	Yes	Yes	Yes
Geographic controls	Yes	Yes	Yes	No	Yes
Mean (Dep. var.)	4.322	4.322	4.322	0.194	0.194
SD (Dep. var.)	1.192	1.192	1.192	0.395	0.395
R ²	0.154	0.195	0.206	0.094	0.117
Observations	893	893	893	904	899

Notes: Robust standard errors in parentheses. The sample includes 1 square km grid cells that have positive populations. Geographical controls include a dummy of high river orders (4 and 5), distance to the urban center, distance to the river, squared distance to the river, interaction terms of these two variables with a river cell dummy, elevation, river confluences, flood vulnerability, geology measures, and open water access measures.

* $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$.

(A) River Network Access

	(1) Water share: non-main channel	(2) River confluence: 1st×2nd or 2nd×3rd	(3) River confluence: 3rd×4th	(4) Flood vulnerability	(5) Pleistocene soil share	(6) Tertiary soil share
$\log(RNA_o)$	0.00751 (0.0161)	0.0470 (0.0470)	-0.0743 (0.0618)	-0.217 (0.307)	-0.0444 (0.0362)	-0.0498 (0.0474)
Mean (Dep. var.)	0.030	0.077	0.083	1.606	0.021	0.211
SD (Dep. var.)	0.087	0.266	0.277	1.606	0.115	0.344
R ²	0.068	0.095	0.137	0.130	0.057	0.735
Observations	899	899	899	899	899	899

(B) Early human settlements

	(1) Water share: main channel	(2) Water share: non-main channel	(3)	(4)	(5)	(6)
Community existence (1940)	0.0263 (0.0193)	-0.00352 (0.00698)	0.218 (0.142)	0.00191 (0.0268)	0.00777 (0.0122)	-0.0108 (0.0257)
Mean (Dep. var.)	0.109	0.030	1.606	0.584	0.021	0.211
SD (Dep. var.)	0.203	0.087	1.606	0.359	0.115	0.344
R ²	0.162	0.038	0.116	0.250	0.030	0.243
Observations	899	899	899	899	899	899

Basin FE

Yes

Yes

Yes

Yes

Geographic controls

Yes

Yes

Yes

Yes

Notes: Robust standard errors in parentheses. The sample includes 1 square km grid cells that have positive populations. In panel (A), geographical controls include a dummy of high river orders (4 and 5), distance to the urban center, distance to the river, squared distance to the river, interaction terms of these two variables with a river cell dummy, elevation, water share of main channel rivers, and floodplain soil share. In panel (B), geographical controls include a dummy of high river orders (4 and 5), distance to the urban center, distance to the river, squared distance to the river, interaction terms of these two variables with a river cell dummy, elevation, and river confluences.

* $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$.

[Back](#)

Across-Sector Externality?

- There may exist the across-sector externality—the effect of clearing forests for agriculture on the productivity of natural resource extraction
- We are not incorporating it. 3 comments:
 - ① The spatial extent of these sectors' activities is distinct:
 - Deforestation for agricultural land is distributed along the rivers: mean, median, max (land footprint depths) = 1 km, 0.85km, 5.5 km
 - Natural resources are also extracted in deep inland areas away from the river
 - ② Natural resource endowments are not significantly correlated with the community-level land footprint ▶ Table
 - ③ This model choice does not affect the inversion problem, but affects outcomes in counterfactual policy simulations:
 - We investigate policies that reduce total deforestation in a river basin
 - We can interpret these policies' welfare effects as lower bounds

Table: Natural Resource Endowments, Calibrated Productivity, and Community Land Footprint

(A)	Number of species found around a community				
	(1) Total	(2) Fish	(3) Timber	(4) NTFP	(5) Game
$\log(A_{o,Nr})$ (calibrated)	0.206*** (0.0306)	0.0220 (0.0334)	0.386*** (0.0407)	0.0488** (0.0204)	0.380*** (0.0437)
Mean (Dep. var.)	2.025	3.161	1.788	0.552	1.958
SD (Dep. var.)	1.145	1.163	1.676	0.893	1.636
R ²	0.059	0.150	0.126	0.349	0.213
Observations	909	909	909	909	909
(B)	Number of species found around a community				
	(1) Total	(2) Fish	(3) Timber	(4) NTFP	(5) Game
\log (land footprint)	0.0171 (0.0381)	-0.0653* (0.0383)	-0.0209 (0.0533)	0.0126 (0.0245)	0.0752 (0.0478)
Mean (Dep. var.)	2.021	3.147	1.796	0.555	1.956
SD (Dep. var.)	1.147	1.183	1.677	0.894	1.637
R ²	0.014	0.163	0.059	0.336	0.146
Observations	906	906	906	906	906
(C)	Number of species found around a community				
	(1) Total	(2) Fish	(3) Timber	(4) NTFP	(5) Game
\log (depth of land footprint)	0.0458 (0.0587)	-0.0358 (0.0541)	-0.135* (0.0771)	-0.0248 (0.0342)	0.0911 (0.0728)
Mean (Dep. var.)	2.070	3.168	1.873	0.550	1.964
SD (Dep. var.)	1.141	1.159	1.676	0.885	1.626
R ²	0.026	0.170	0.059	0.364	0.146
Observations	811	811	811	811	811
Basin FE	Yes	Yes	Yes	Yes	Yes
Geographic controls	No	No	No	No	No

Notes: Robust standard errors in parentheses. The unit of analysis is a community in the PARLAP Community Census (CC) in 2014. In panel (B), the land footprint represents the community-level land footprint within a voronoi polygon around the settlement, detected in satellite images. In panel (C), the land footprint depth represents the distance from the river to the furthest inland point in the community-level land footprint.

* $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$.

Table: Agglomeration Externality in Agriculture

	The calibrated value of $\log(A_{o,Ag})$				$N_o < 1000$			
	All locations							
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
$\log(N_{o,Ag})$	0.676*** (0.0207)	0.440** (0.171)	0.514*** (0.0809)	0.501*** (0.0790)	0.735*** (0.0196)	0.384** (0.169)	0.509*** (0.124)	0.464*** (0.109)
Basin FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Geographic controls	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
IV: RNA	No	Yes	No	Yes	No	Yes	No	Yes
IV: Historical	No	No	Yes	Yes	No	No	Yes	Yes
Mean (Dep. var.)	-0.096	-0.096	-0.096	-0.096	-0.172	-0.172	-0.172	-0.172
SD (Dep. var.)	4.578	4.578	4.578	4.578	4.614	4.614	4.614	4.614
First stage <i>F</i> -stat		11.502	56.653	31.005		15.298	35.632	22.822
Hansen's <i>J</i> test <i>p</i> -value				0.648				0.472
Observations	893	893	893	893	852	852	852	852

Notes: Robust standard errors in parentheses. The sample includes 1 square km grid cells that have positive populations. We use $\log(RNA_o)$ (IV: RNA) and the initial community existence in 1940 (IV: Historical) as instruments for $\log(N_{o,Ag})$. Geographical controls include a dummy of high river orders (4 and 5), distance to the urban center, distance to the river, squared distance to the river, interaction terms of these two variables with a river cell dummy, elevation, river confluences, flood vulnerability, geology measures, and open water access measures.

* $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$.

Table: Agglomeration Externality in Agriculture

	The calibrated value of $\log(A_{o,Ag})$						OLS
	IV						
	(1)	(2)	(3)	(4)	(5)	(6)	(7)
$\log(N_{o,Ag})$	0.434*** (0.0920)	0.519*** (0.0789)	0.519*** (0.0788)	0.521*** (0.0789)	0.509*** (0.0789)	0.501*** (0.0790)	0.676*** (0.0207)
log (Elevation)		2.341*** (0.171)	2.354*** (0.175)	2.324*** (0.179)	2.360*** (0.176)	2.397*** (0.177)	2.252*** (0.176)
River confluence (1st×2nd or 2nd×3rd)		0.0155 (0.0958)	0.0186 (0.0964)	0.0206 (0.0982)	0.0309 (0.0995)	0.0180 (0.0969)	
River confluence (3rd×4th)		-0.0356 (0.0724)	-0.0339 (0.0723)	-0.0246 (0.0730)	-0.0266 (0.0733)	0.0173 (0.0618)	
Flood vulnerability (1-4)		-0.0115 (0.0136)	-0.00947 (0.0137)	-0.0123 (0.0137)	-0.0123 (0.0130)	-0.0154 (0.0130)	
Water share: non-main channel				0.0806 (0.238)	0.123 (0.238)	-0.00146 (0.203)	
Water share: main channel				0.161 (0.122)	0.185 (0.121)	0.189 (0.120)	
Floodplain soil share					0.127** (0.0625)	0.126** (0.0575)	
Pleistocene soil share					0.175 (0.222)	0.333 (0.227)	
Basin FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Mean (Dep. var.)	-0.094	-0.096	-0.096	-0.096	-0.096	-0.096	-0.096
SD (Dep. var.)	4.576	4.578	4.578	4.578	4.578	4.578	4.578
First stage F-stat	28.030	29.419	29.974	29.634	30.770	31.005	
Observations	894	893	893	893	893	893	893

Notes: Robust standard errors in parentheses. The sample includes 1 square km grid cells that have positive populations. We use $\log(RNA_o)$ and the initial community existence in 1940 as instruments for $\log(N_{o,Ag})$. Other controls include distance to the urban center, distance to the river, squared distance to the river, and interaction terms of these two variables with a river cell dummy.

* $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$.

Table: Density Externality in Forest Clearing

	log (per capita land footprint)					
	All locations		$N_o < 1000$		$N_o < 500$	
	(1) OLS	(2) IV	(3) OLS	(4) IV	(5) OLS	(6) IV
$\log(N_{o,Ag})$	-0.650*** (0.0307)	-0.522*** (0.0940)	-0.654*** (0.0323)	-0.552*** (0.109)	-0.674*** (0.0346)	-0.545*** (0.123)
Basin FE	Yes	Yes	Yes	Yes	Yes	Yes
Geographic controls	Yes	Yes	Yes	Yes	Yes	Yes
Mean (Dep. var.)	0.929	0.929	0.956	0.956	0.981	0.981
SD (Dep. var.)	1.231	1.231	1.218	1.218	1.223	1.223
First stage <i>F</i> -stat		34.198		28.141		23.709
Hansen's <i>J</i> test <i>p</i> -value		0.987		0.896		0.969
Observations	895	895	878	878	847	847

Notes: Robust standard errors in parentheses. The unit of analysis is a community in the PARLAP Community Census (CC) in 2014. We use $\log(RNA_o)$ and the initial community existence in 1940 as instruments for $\log(N_{o,Ag})$. Geographical controls include a dummy of high river orders (4 and 5), distance to the urban center, distance to the river, squared distance to the river, interaction terms of these two variables with a river cell dummy, elevation, river confluence, flood vulnerability, geology measures, and open water access measures for a grid cell where each census community belongs.

* $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$.

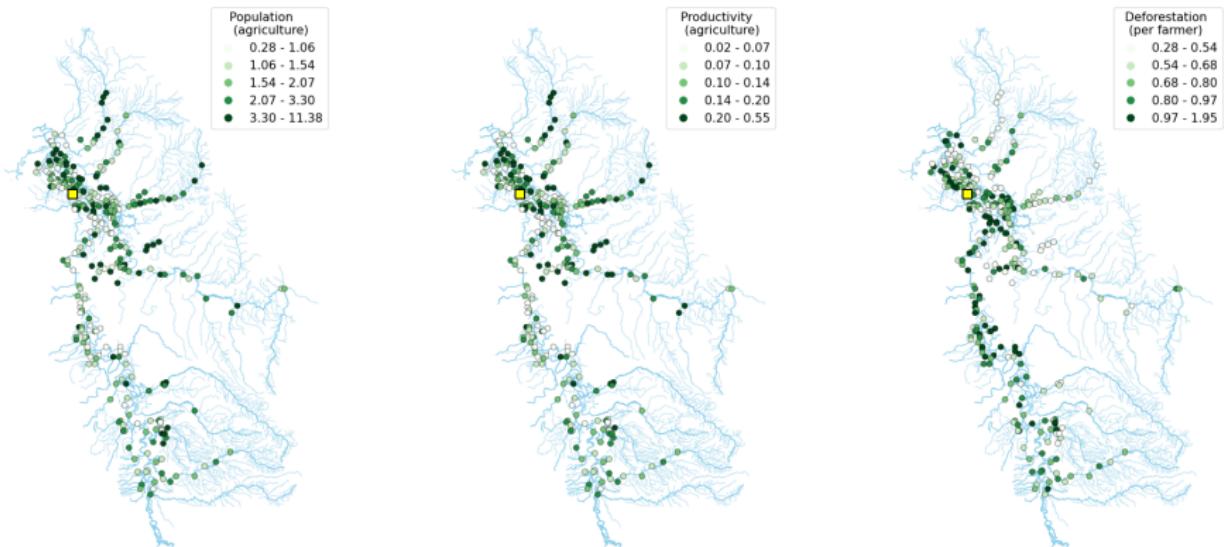
Table: Congestion Externality in Natural Resource Extraction with Spatial Spillovers

	The calibrated value of $\log(A_{o,Nr})$									OLS	
	IV										
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)		
$\log(N_{o,Nr})$	-2.127** (1.075)	-1.385 (0.939)	-0.879 (0.688)	-0.581 (0.558)	0.0960 (0.322)	0.278 (0.280)	0.573*** (0.208)	0.634*** (0.179)	0.606*** (0.184)	0.950*** (0.0516)	
$\log(\sum_{d D_{o,d} \leq 2km} N_{d,Nr})$	-0.573* (0.331)	-0.0343 (0.282)	0.0235 (0.236)	-0.0745 (0.143)	-0.0705 (0.122)	-0.0648 (0.0883)	-0.0745 (0.0822)	-0.0856 (0.0838)	-0.0856 (0.0611)	-0.0663	
$\log(\sum_{d D_{o,d} \leq 5km} N_{d,Nr})$	-0.596*** (0.189)	-0.286 (0.183)	-0.138 (0.106)	-0.130 (0.0888)	-0.120* (0.0637)	-0.111* (0.0597)	-0.112* (0.0613)	-0.132*** (0.0425)			
$\log(\sum_{d D_{o,d} \leq 10km} N_{d,Nr})$	-0.337** (0.141)	0.0345 (0.107)	0.0364 (0.0885)	0.0579 (0.0625)	0.0425 (0.0596)	0.0322 (0.0639)	0.0140 (0.0378)				
$\log(\sum_{d D_{o,d} \leq 25km} N_{d,Nr})$	-0.470*** (0.0918)	-0.357*** (0.0837)	-0.327*** (0.0584)	-0.294*** (0.0560)	-0.285*** (0.0571)	-0.165*** (0.0283)					
$\log(\sum_{d D_{o,d} \leq 50km} N_{d,Nr})$		-0.195*** (0.0610)	-0.0318 (0.0581)	-0.0548 (0.0526)	-0.0480 (0.0548)	-0.0619** (0.0242)					
$\log(\sum_{d D_{o,d} \leq 75km} N_{d,Nr})$			-0.280*** (0.0779)	-0.0758 (0.125)	-0.0407 (0.142)	-0.0989*** (0.0352)					
$\log(\sum_{d D_{o,d} \leq 100km} N_{d,Nr})$				-0.258* (0.141)	-0.439* (0.231)	-0.263*** (0.0498)					
$\log(\sum_{d D_{o,d} \leq 150km} N_{d,Nr})$					0.187 (0.171)	0.0970* (0.0567)					
Basin FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	
Geographic controls	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	
Mean (Dep. Var.)	0.337	0.337	0.337	0.337	0.337	0.337	0.337	0.337	0.337	0.337	
SD (Dep. Var.)	2.862	2.862	2.862	2.862	2.862	2.862	2.862	2.862	2.862	2.862	
Observations	894	894	894	894	894	894	894	894	894	894	

Notes: Robust standard errors in parentheses. The sample includes 1 square km grid cells that have positive populations. We use $\ln RNA_o$ and $\{\ln \sum_{d|D_{o,d} \leq x} RNA_d\}$ for $x \in \mathcal{X}$ as instruments when endogenous variables include $\log(N_{o,Nr})$ and $\{\ln \sum_{d|D_{o,d} \leq x} N_{d,Nr}\}$ for $x \in \mathcal{X}$. Geographical controls include a dummy of high river orders (4 and 5), distance to the river, squared distance to the river, interaction terms of these two variables with a river cell dummy, elevation, river confluence, flood vulnerability, geology measures, and open water access measures.

* $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$.

Spatial Distribution of Agriculture and Deforestation without the Agglomeration Externality



▶ Back

Economies of Scale in Transport Technology

- Consider an alternative model without the agglomeration externality in the production function but with **endogenous trade costs**:

$$\tilde{\tau}_{od,Ag} = N_{o,Ag}^{-\mu_{Ag}} \tau_{od,Ag}$$

- This model is **isomorphic to the original model**
- The trade cost can be decreasing in the origin population possibly because:
 - Large commercial boats ('lancha') are more likely to stop by
 - Collective investment in motor boats ('rapido')
 - The average transport cost charged is decreasing in the amount of products traded

Transport Modes in the Peruvian Amazon

Canue



Peque-peque (most widely available)



Lancha



Rapido (express motor boat)



Appendix
Table: Community Population and Availability of Transport Modes

	Availability of Transport Modes in a Community							
	Lancha		Colectivo		Rapido		Peque-peque	
	(1) OLS	(2) IV	(3) OLS	(4) IV	(5) OLS	(6) IV	(7) OLS	(8) IV
log($N_{o,Ag}$)	0.0469*** (0.0111)	0.144*** (0.0430)	0.0478*** (0.0115)	0.0280 (0.0383)	0.0522*** (0.0108)	0.0566* (0.0292)	-0.00528 (0.00576)	0.00418 (0.0156)
Basin FE	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Geographic controls	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Mean (Dep. var.)	0.492	0.492	0.386	0.386	0.110	0.110	0.972	0.972
SD (Dep. var.)	0.500	0.500	0.487	0.487	0.314	0.314	0.164	0.164
First stage F-stat	24.84462		24.84462		24.84462		24.84462	
Observations	906	906	906	906	906	906	906	906

Notes: Robust standard errors in parentheses. The unit of analysis is a community in the PARLAP Community Census (CC) in 2014. We use log(RNA_o) and the initial community existence in 1940 as instruments for log($N_{o,Ag}$). Geographical controls include a dummy of high river orders (4 and 5), distance to the urban center, distance to the river, squared distance to the river, interaction terms of these two variables with a river cell dummy, elevation, river confluence, flood vulnerability, geology measures, and open water access measures for a grid cell where each census community belongs.

* $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$.

- The data supports the fact that ‘peque-peque’ is most widely available
- Significant scale effects on the availability of ‘lancha’ and ‘rapido’
- Consistent results for the frequency of transport modes available and other proxies of transaction costs as well

Economies of Scale in Agricultural Intensification

- Test this using producer-level information from the Agricultural Census
- Modern technologies are limited: each of 24 listed modern technologies (except for boat) is used by <10% of agricultural producers
- Significant scale effects on:
 - direct inputs into land and crops (fertilizers, insecticides, herbicides, fungicides)
 - complementary equipment (sprayers)
 - crop processing technology to facilitate marketing (grain mill)

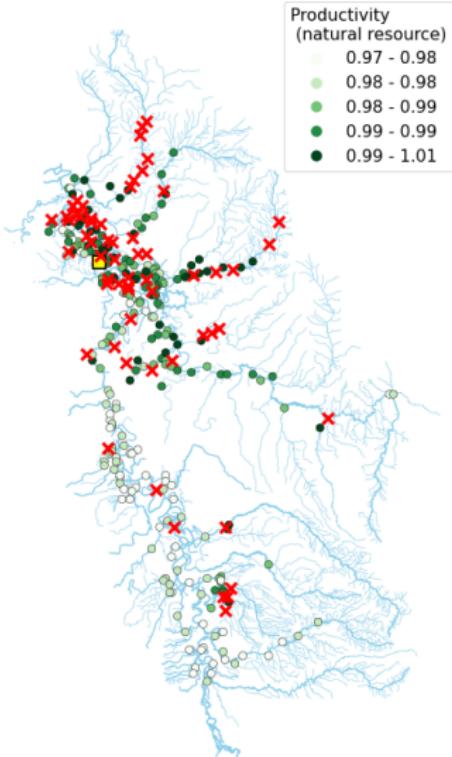
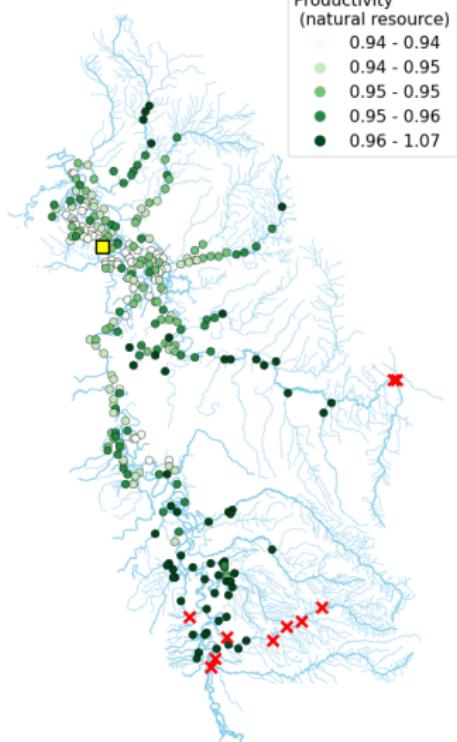
Table: Community Population and Modern Technology Use

(A) Basic infrastructure						
	(1)	(2)	(3)	(4)	(5)	(6)
	Irrigation	Certified seed	Crops have been certified organic	Electricity for agricultural work	Animals for agricultural work	Tractors for agricultural work
log($N_{o,Ag}$)	-0.00329* (0.00180)	-0.000857 (0.00430)	0.000692 (0.000584)	-0.000688 (0.000863)	0.00315 (0.00206)	0.000476 (0.000811)
Mean (Dep. var.)	0.013	0.064	0.001	0.003	0.010	0.002
SD (Dep. var.)	0.112	0.245	0.037	0.054	0.098	0.044
First stage F-stat	1649.082	1649.082	1649.082	1649.082	1649.082	1649.082
Observations	25827	25827	25827	25827	25827	25827
(B) Inputs into land and crops						
	(1)	(2)	(3)	(4)	(5)	(6)
	Guano/manure/ compost	Chemical fertilizers	Insecticides	Herbicides	Fungicides	Biologic control
log($N_{o,Ag}$)	0.000807 (0.00111)	0.00265** (0.00115)	0.0228*** (0.00353)	0.0314*** (0.00371)	0.0118*** (0.00219)	-0.00239 (0.00239)
Mean (Dep. var.)	0.005	0.004	0.040	0.051	0.012	0.020
SD (Dep. var.)	0.069	0.063	0.197	0.221	0.111	0.140
First stage F-stat	1649.082	1649.082	1649.082	1649.082	1649.082	1649.082
Observations	25827	25827	25827	25827	25827	25827
(C) Animal, electrical, or mechanical energy						
	(1)	(2)	(3)	(4)	(5)	(6)
	Iron plow of animal traction	Wooden plow of animal traction	Harvester	Foot plow	Motorized sprayer	Manual sprayer
log($N_{o,Ag}$)	-0.000796 (0.000523)	-0.000223 (0.000311)	-0.000229 (0.000282)	-0.000806 (0.000556)	0.00197** (0.000815)	0.0214*** (0.00401)
Mean (Dep. var.)	0.001	0.000	0.001	0.001	0.002	0.062
SD (Dep. var.)	0.035	0.022	0.025	0.035	0.043	0.241
First stage F-stat	1649.082	1649.082	1649.082	1649.082	1649.082	1649.082
Observations	25827	25827	25827	25827	25827	25827
(D) Electrical or mechanical energy						
	(1)	(2)	(3)	(4)	(5)	(6)
	Grain mill	Grass chopper	Thresher	Electric generator	Wheel tractor	Boat/canoe/ speedboat
log($N_{o,Ag}$)	0.00696*** (0.00194)	0.000462 (0.000448)	-0.00102 (0.000674)	-0.0103*** (0.00323)	0.000932 (0.000590)	-0.0187** (0.00746)
Mean (Dep. var.)	0.013	0.001	0.004	0.036	0.001	0.618
SD (Dep. var.)	0.111	0.025	0.061	0.186	0.030	0.486
First stage F-stat	1649.082	1649.082	1649.082	1649.082	1649.082	1649.082
Observations	25827	25827	25827	25827	25827	25827
Basin FE	Yes	Yes	Yes	Yes	Yes	Yes
Geographic controls	Yes	Yes	Yes	Yes	Yes	Yes

Notes: Robust standard errors in parentheses. The unit of analysis is a household in the 2012 Peruvian Agricultural Census. We use log(RNA_o) and the initial community existence in 1940 as instruments for log($N_{o,Ag}$). Geographical controls include a dummy of high river orders (4 and 5), distance to the urban center, distance to the river, squared distance to the river, interaction terms of these two variables with a river cell dummy, elevation, river confluence, flood vulnerability, geology measures, and open water access measures for a grid cell where each census community belongs.

* $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$.

(A) Protected areas by rural frontier (B) Small communities not allowed



Legend: values in the counterfactual scenario relative to those in the benchmark equilibrium