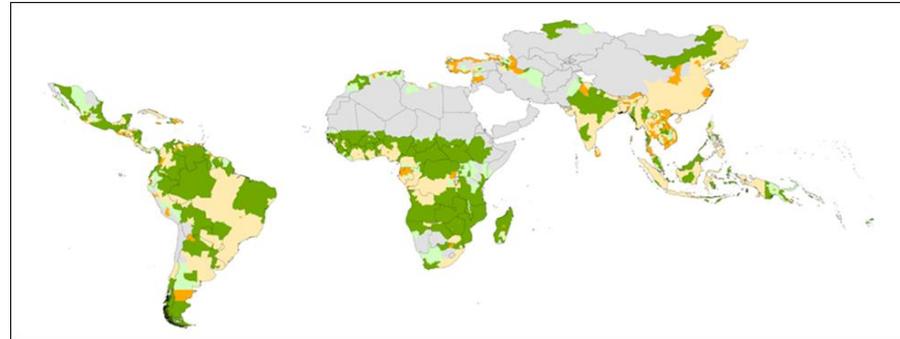


# Using Spatial Datasets to Improve Modeling of AR Costs, Carbon Accumulation, and Net Tree-Cover Gain in the Global South



**Jeffrey R. Vincent**

Nicholas School of the Environment, Duke University

*11<sup>th</sup> Forestry and Agriculture GHG Modeling Forum*

*North Carolina State University, Raleigh NC*

*March 6, 2024*

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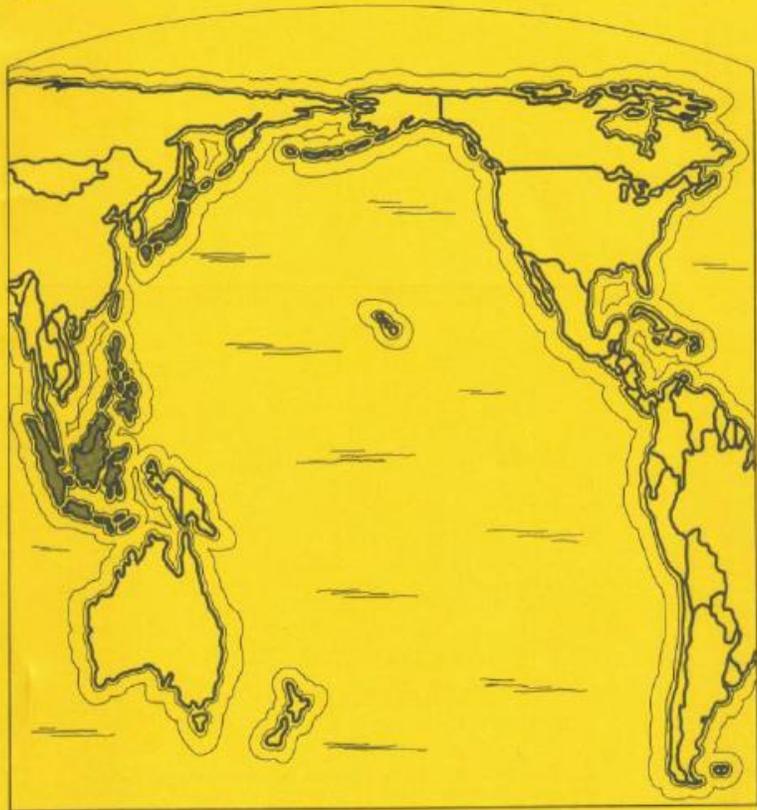


United States  
Department of  
Agriculture  
Forest Service  
Pacific Northwest  
Research Station  
General Technical  
Report  
PNW-GTR-210



# Modeling Japan – South Seas Trade in Forest Products

Jeffery R. Vincent



C I N T R A F O R

Working Paper

18

## An Economic Analysis of Short-Run Timber Supply Around the Globe

1988

P. A. Cardellichio, Y. C. Youn, C. S. Binkley,  
J. R. Vincent, D. M. Adams



CENTER FOR INTERNATIONAL TRADE IN FOREST PRODUCTS  
UNIVERSITY OF WASHINGTON  
COLLEGE OF FOREST RESOURCES AR-10  
SEATTLE, WASHINGTON 98195



## Economists are dismal scientists

- 2011–2023 search: *afforestation* OR *reforestation* OR *forest restoration*
  - Web of Science: 14,984 documents
  - EconLit: 186 documents
- 2011–2023 search: *deforestation*
  - EconLit: 906 documents

*Despite: > ½ of LMICs having  
net tree-cover gain during 2005 – 2020*



*Annual Review of Environment and Resources*  
**Forest Restoration in Low- and  
Middle-Income Countries**

Jeffrey R. Vincent,<sup>1</sup> Sara R. Curran,<sup>2</sup>  
and Mark S. Ashton<sup>3</sup>

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<sup>2</sup>Henry M. Jackson School of International Studies, University of Washington, Seattle, Washington 98195, USA; email: scurran@uw.edu

<sup>3</sup>The Forest School at the Yale School of the Environment, Yale University, New Haven, Connecticut 06511, USA; email: mark.ashton@yale.edu

# Research program 1 (micro): impact evaluations of AR programs

**PAKISTAN (2):** Billion Tree Afforestation Program (enclosures, plantations)

**NEPAL:** Plantation Registration Programme

**CHINA (2):** 3-North Shelter Forest Program; Yunnan Pilot Carbon Credit Program



**BANGLADESH:** Social Forestry Program

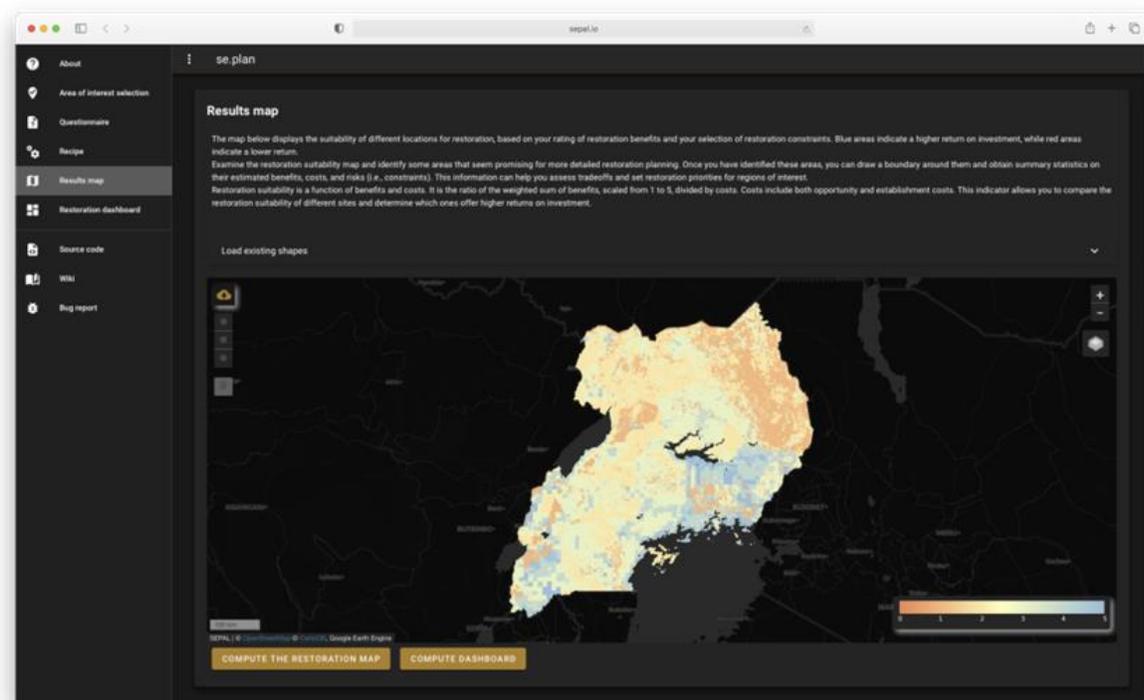
**UGANDA:** Sawlog Production Grant Scheme

**INDIA:** National Agroforestry Policy

**SRI LANKA:** National Agroforestry Programme

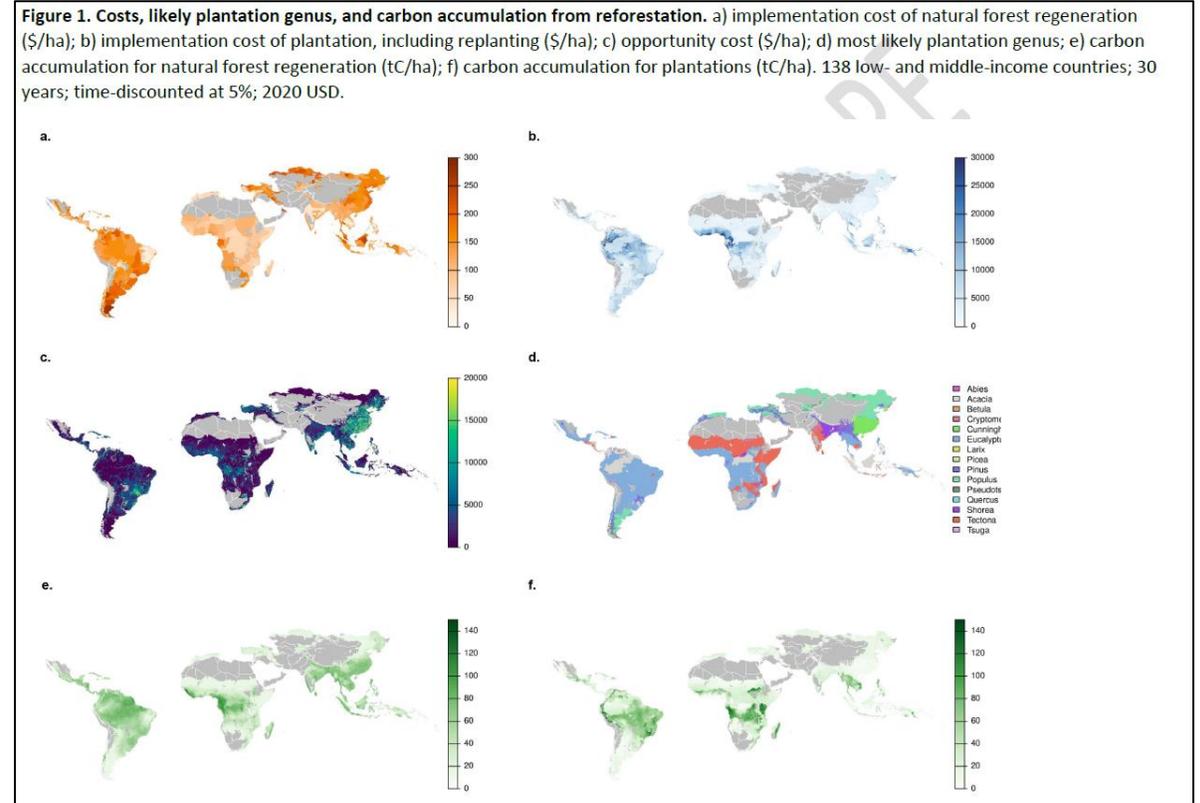
# Research program 2 (macro): LMIC-wide AR patterns and processes

- **se.plan**: spatially explicit forest restoration planning tool
  - FAO Open Foris SEPAL (<https://docs.sepal.io/en/latest/modules/dwn/seplan.html>)
  - Google Earth Engine-based
- Ongoing development since mid-2020



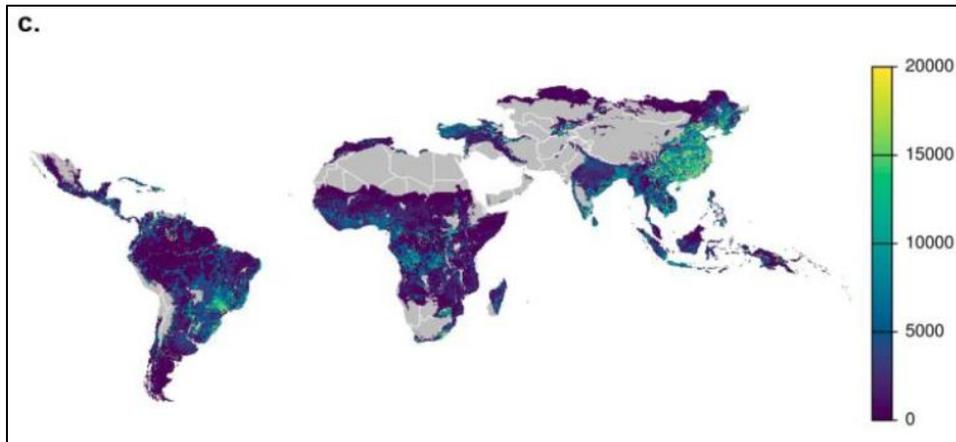
# Gridded datasets on AR costs, genera choice, and C accumulation rates

J. Busch, J.J. Bukoski, S.C. Cook-Patton, B. Griscom, D.J. Kaczan, M.D. Potts, Y.Y. Yi, and J.R. Vincent (in review),  
“Tree planting vs. natural forest regeneration: relative cost-effectiveness at mitigating climate change”

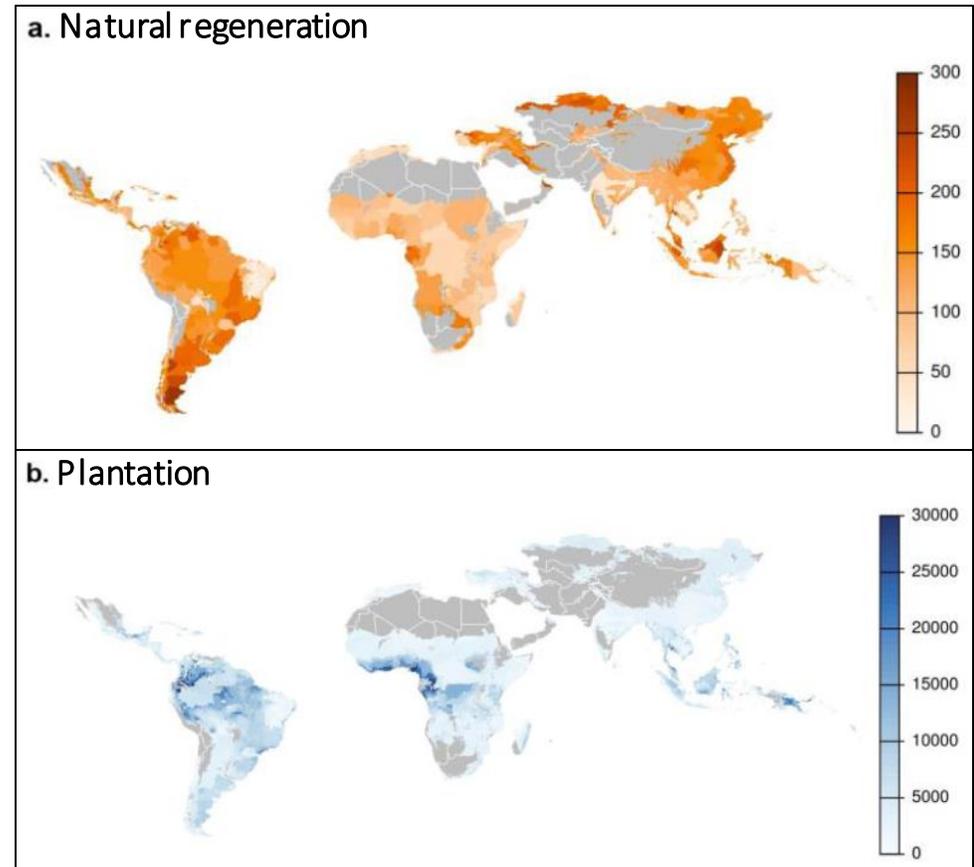


# AR costs

Opportunity cost: agricultural land rent (PV, 5%)  
[~10 km]



Implementation costs (= estab. + 3-5 yrs. maint.)  
[Level 1 subdivision]



# Workflow for opportunity cost: cropland

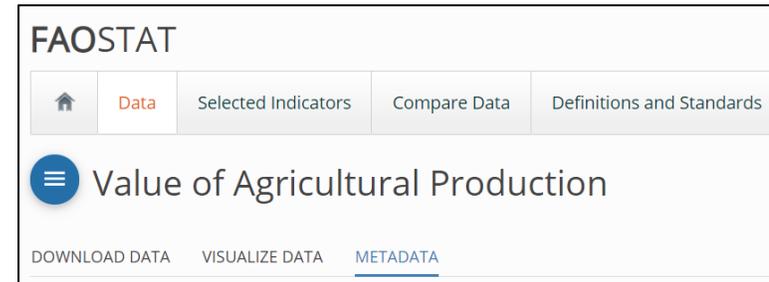
1. Obtain 2010\* gridded data on crop production value



<https://www.mapspam.info/>

\*2020 data now available

2. Update to 2019 using national data



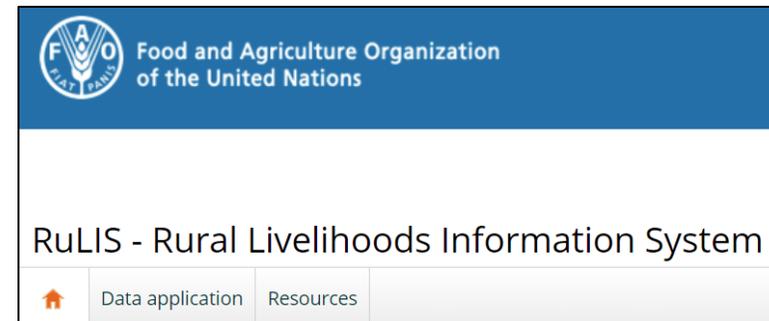
<http://www.fao.org/faostat/en/#data/QV/metadata>

4. Capitalize annual land rent using 5% real discount rate



<https://databank.worldbank.org/home>

3. Determine land-rent share using LSMS household survey data

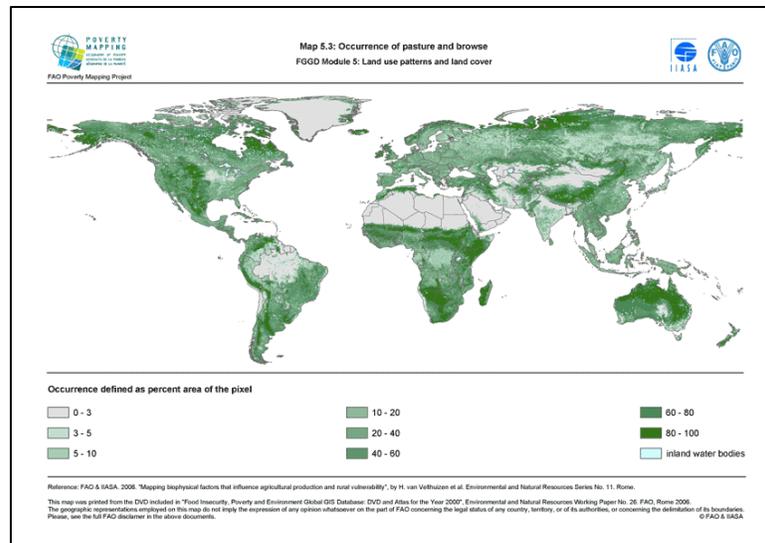


<http://www.fao.org/in-action/rural-livelihoods-dataset-rulis/en/>

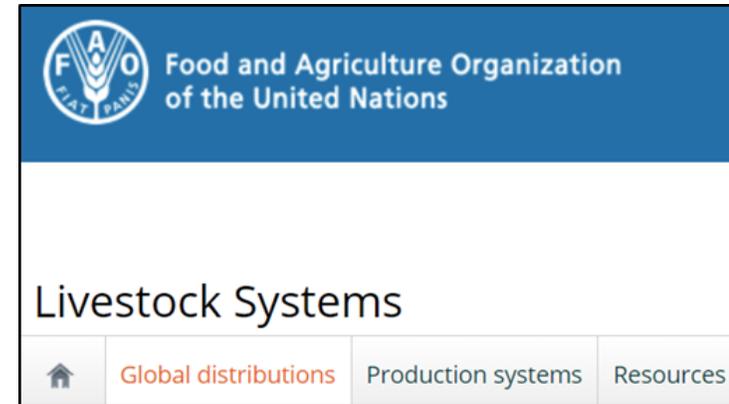
# Workflow for opportunity cost: pastureland

Similar to cropland value, but use different datasets, including:

1. 2000 FAO pasture map, updated to 2015 using spatial statistical methods
2. 2010 gridded livestock data, updated to 2019 using national data



<https://data.apps.fao.org/map/catalog/srv/eng/catalog.search#/metadata/913e79a0-7591-11db-b9b2-000d939bc5d8>

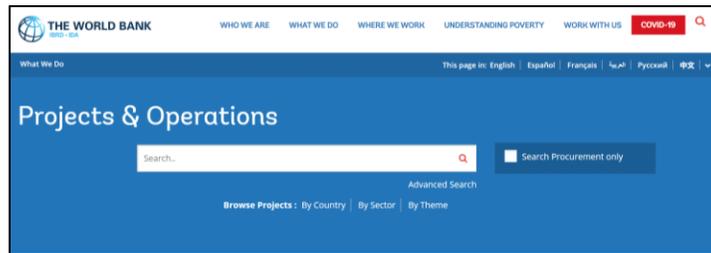


<http://www.fao.org/livestock-systems/global-distributions/en/>

*More confident in cropland OC estimates than pastureland OC estimates*

# Workflow for implementation costs

1. Extract 355 cost estimates from 99 documents (World Bank, other)



<https://projects.worldbank.org/en/projects-operations/projects-home>

2. Obtain spatial data on variables hypothesized to affect costs



<https://datadryad.org/stash/dataset/doi:10.5061/dryad.dk1j0>

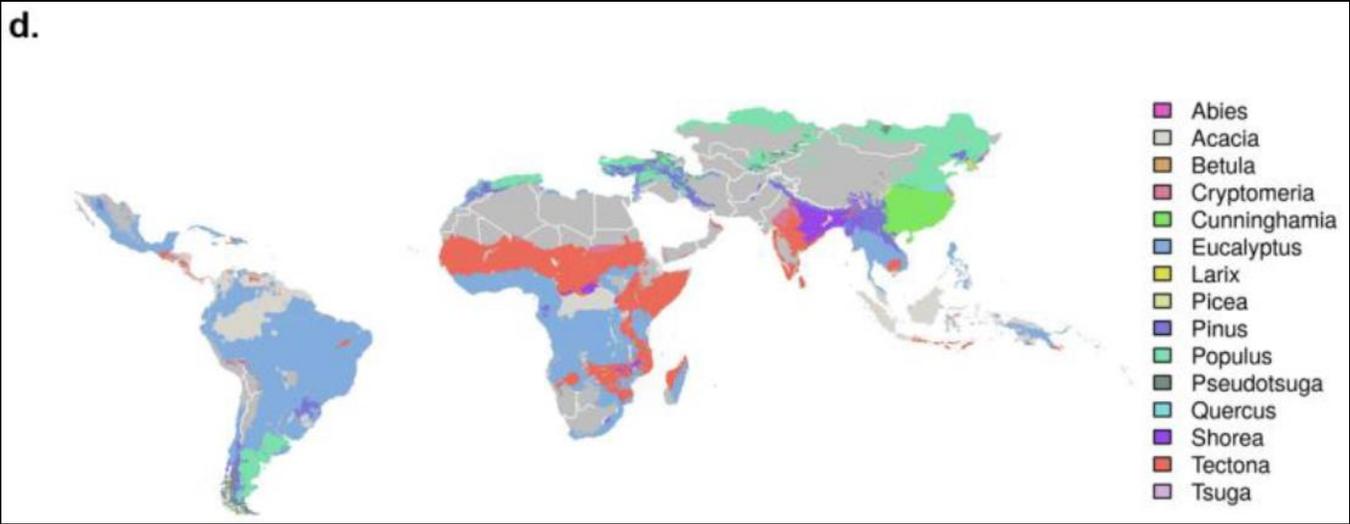
4. Predict costs for Level 1 subdivisions using statistical model



3. Use ML (lasso) to statistically relate cost estimates to spatial variables

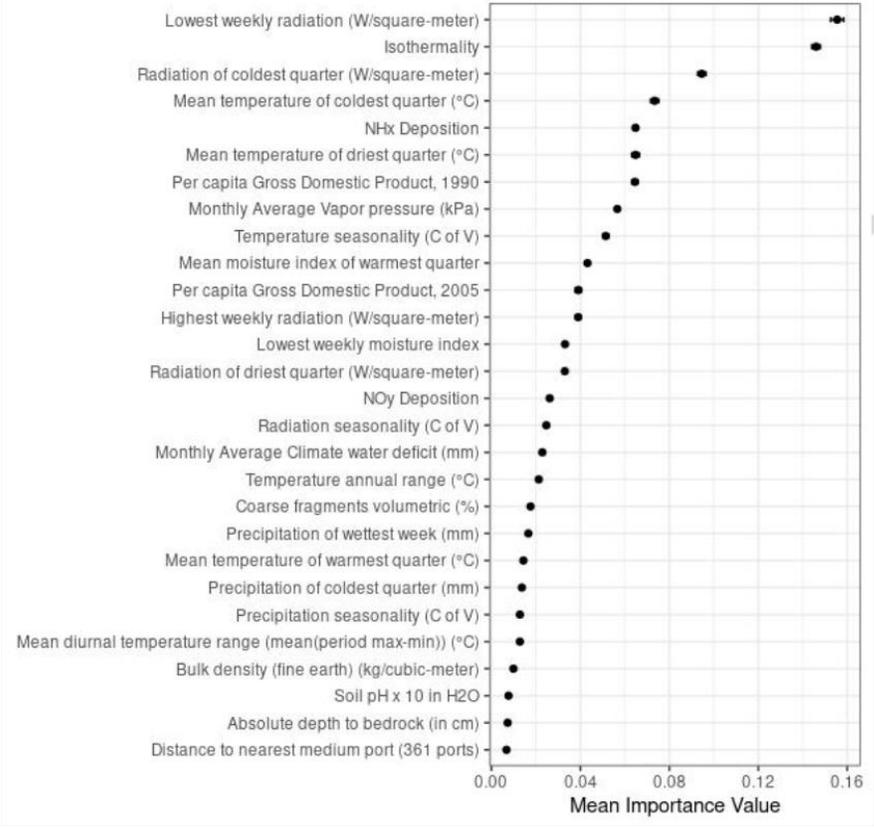
Variable	Coefficient
Dummy: regeneration method = planting	1.35
Dummy: regeneration method = passive natural	-0.916
Dummy: regeneration included native species	-0.356
ln(GDP per capita)	0.329
Dummy: cost estimate spans multiple years	0.206
Dummy: cost estimate not disaggregated by inputs or activities	-0.154
Area share: biome = tropical & subtropical dry broadleaved forest	-1.27
Area share: biome = Mediterranean forests and woodlands	0.725
Area share: biome = deserts and xeric shrublands	-1.04
Year of implementation	0.0123
Constant	-21.9

# Most likely plantation genus [1 km]

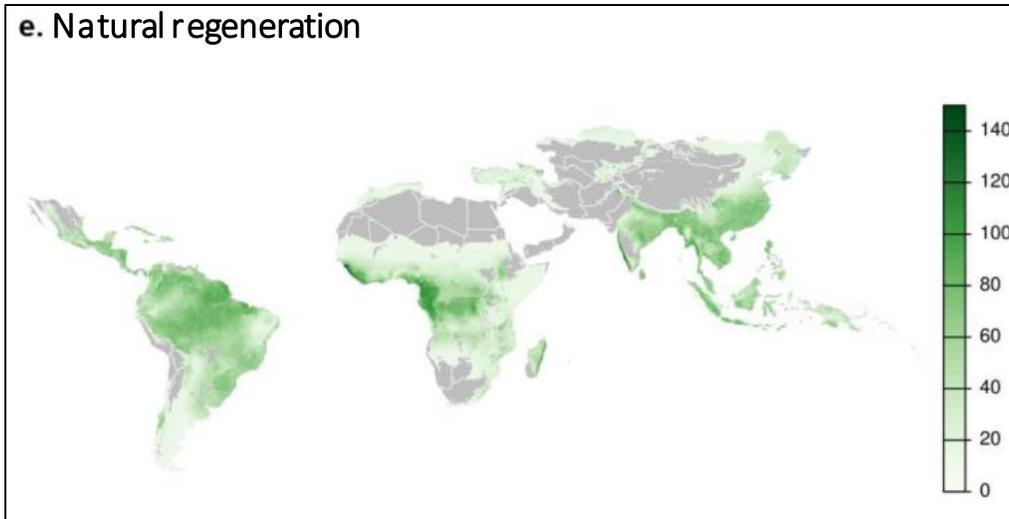


Primary data source: Harris et al., *Spatial Database of Planted Trees, Version 1* (WRI, 2019)

Figure S2. Parsimonious set of predictors for classification of most likely plantation type. Importance scores calculated by the Variable Selection Using Random Forests (VSURF) algorithm.



## Above-ground carbon accumulation functions [1 km]

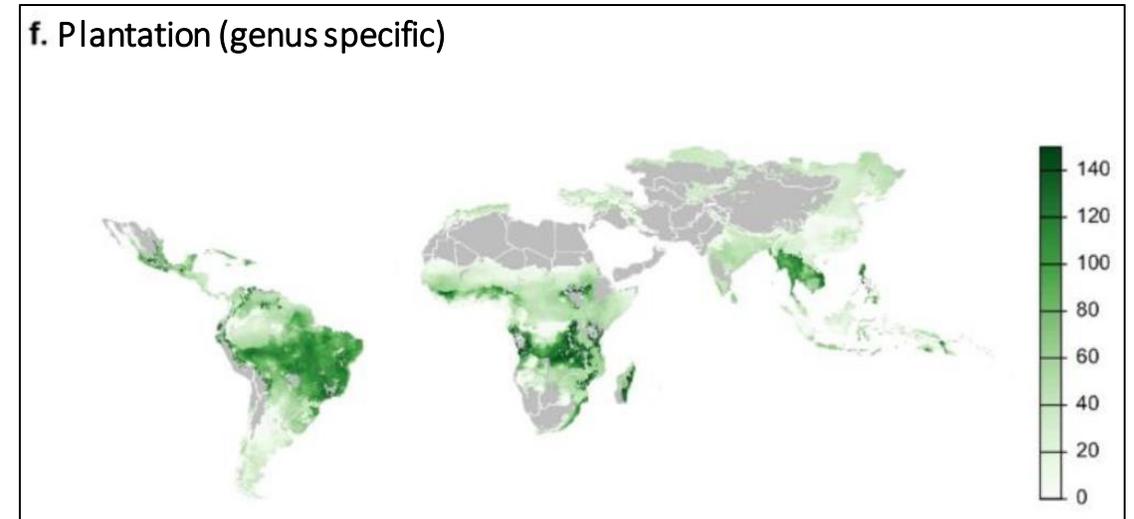


Primary data source: Cook-Patton et al., “Mapping carbon accumulation potential from global natural forest regrowth” (*Nature*, 2020)

- 2,309 observations from 410 LMIC sites

### Workflow:

1. Compile spatial data on 64 climatic and soil variables
2. Fit Chapman-Richards growth functions, with spatially varying parameters

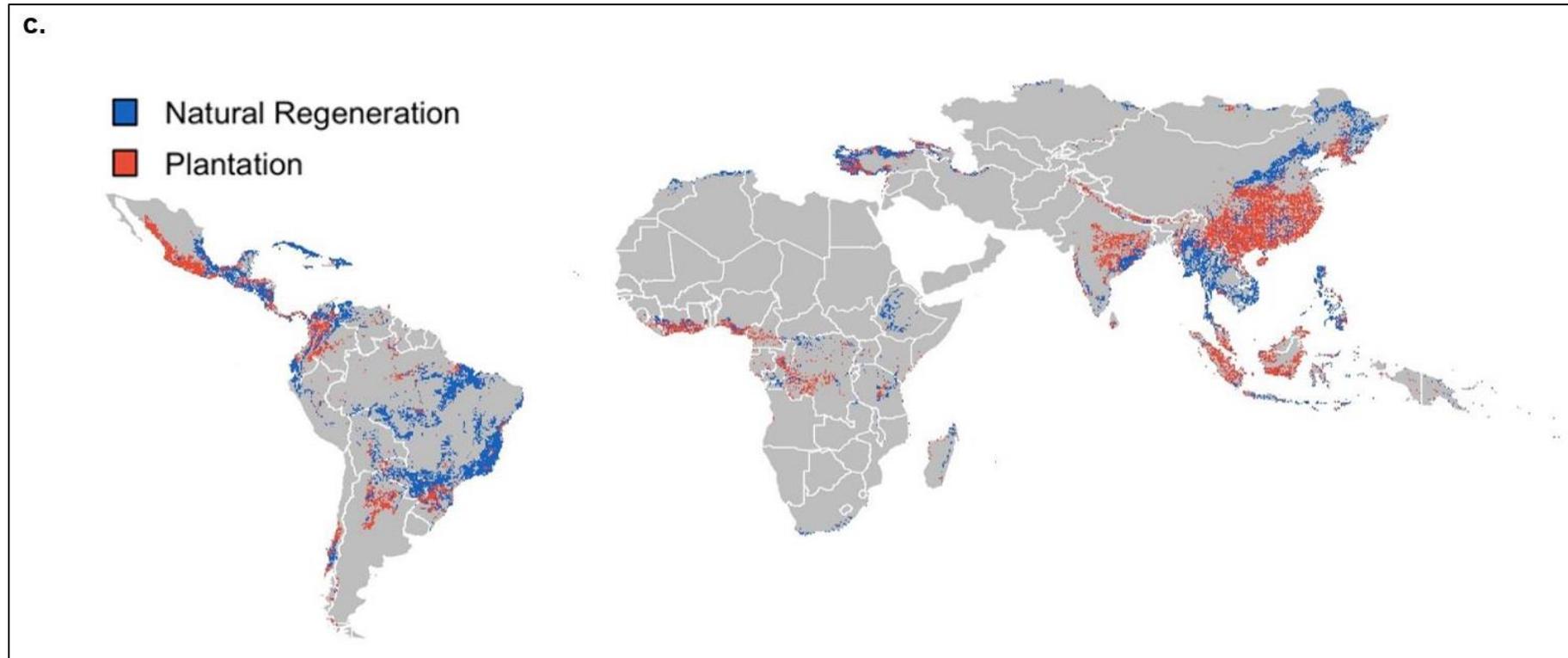


Primary data source: Bukoski et al., “Rates and drivers of aboveground carbon accumulation in global monoculture plantation forests” (*Nature Comm.*, 2022)

- 3,289 observations from 618 LMIC sites

# Which regeneration method offers lower cost sequestration, and where?

[Plantation: accounts for timber value and C storage in durable wood products]



*Natural regeneration (46%) and plantations (54%) have lower sequestration costs in roughly equal areas  
AR offers 3-10× more sequestration below \$20 – 50/tCO<sub>2</sub> than estimated by IPCC (2022)*

## **PROPOSAL FOR A SESYNC PURSUIT**

**Descriptive title:** Migration, Marginal Agricultural Land, and Tree-Cover Expansion in Low- and Middle-Income Countries

**Short title:** Migration and Tree Cover

**Name and contact information for PIs**

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University of Washington

# Demographic drivers of net 5-year tree-cover gain in LMICs

[Level 2 subdivisions; 3 years (2005, 2010, 2015); FE logit models w/ country trends]

Variables	Base model	Add population density	Add net migration rates	Add working-age share	Add youth & elderly shares
Tree cover gap	23.90*** (0)	23.92*** (0)	23.89*** (0)	23.93*** (0)	24.27*** (0)
GDP per capita	-1.95e-05 (0.480)	-1.97e-05 (0.475)	-1.89e-05 (0.491)	-1.89e-05 (0.487)	-1.92e-05 (0.471)

If indicated demographic variable increases by 1 standard deviation,  
then probability of net tree-cover gain increases by ...

- Population share: elderly 0.69
- Population share: youth 0.54
- Population density -0.22
- Net out-migration rate 0.16
- Net in-migration rate -0.001

A photograph of a field with several tall, thin trees and a fence in the foreground. The trees are green and appear to be young or saplings. The sky is overcast and grey. The text is overlaid on the image in a white, italicized font.

*Thank you,  
and please help me find a postdoc with GEE/GIS skills!  
([Jeff.Vincent@duke.edu](mailto:Jeff.Vincent@duke.edu))*