

# Green Development

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# Introduction

- Energy is a nontrivial factor in production and consumption
  - ▶ Energy share as an input in production is around 5%
  - ▶ Energy share in final consumption around 3.5%
- Fossil and renewable (green energy) yield different emissions, but also have distinct
  - ▶ **Geography of comparative advantage**
  - ▶ **Transport costs**: Grid versus pipelines and tankers
- Hence, the relative evolution of the technology to produce each type of energy can affect
  - ▶ The cost of energy
  - ▶ The locations where energy is produced and the resulting emissions
  - ▶ The geography of economic activity in the world
- **Can green energy lead to faster growth in parts of the developing world?**

# What We Do

- Develop a quantitative spatial model of the world economy with an explicit energy sector:
  - ▶ **Local investments in green and fossil technology**
  - ▶ **Costly energy trade**
  - ▶ **Capability of renewables and the evolution of local fossil fuel stocks**
- Quantify using data on
  - ▶ The geography of fossil fuel reserves
  - ▶ The geography of renewable capabilities ("amount of sun", "wind")
  - ▶ Realistic energy and good transportation network ("Grid", "Pipelines")
- Improvement in renewable technology, together with their geography, can be a local development opportunity
  - ▶ Depends on the grid, transport network, and energy policy
  - ▶ Isolated locations can generate their own energy without relying on trade or the grid
- Resulting quantitative model useful to study local energy markets and climate policy

# Literature Review

- Environmental Directed Technical Change

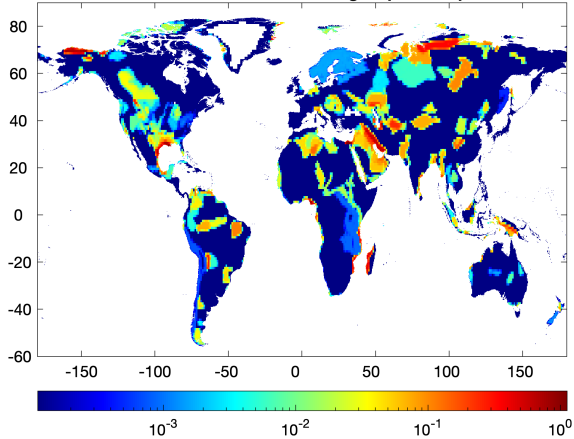
- ▶ Acemoglu et al. (2012, 2014, 2016, 2023), Aghion et al. (2016), Dechezleprêtre and Hémous (2022), Fried (2018), Hassler et al. (2021), Cicala et al. (2022)

- Spatial IAMs

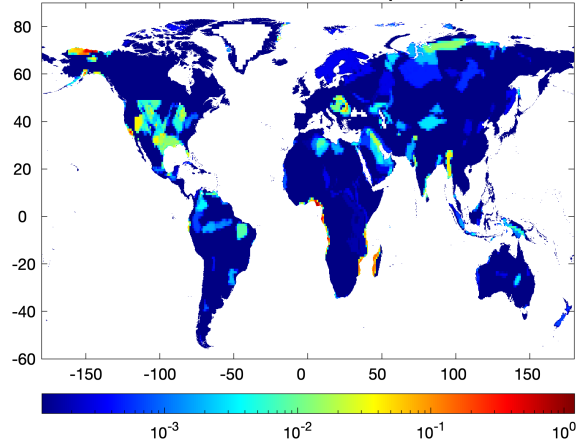
- ▶ **Arkolakis and Walsh (2023)**, Balboni (2025), Bilal and Rossi-Hansberg (2023), Conte et al. (2021, 2025), Cruz (2024), Cruz and Rossi-Hansberg (2025, 2024), Desmet and Rossi-Hansberg (2015), Desmet et al. (2021), Krusell and Smith Jr (2022), Nath (2025), Rudik et al. (2022)

# Comparative Advantage for Fossil Energy: Gas and Oil Reserves

Reserves conventional gas (PJ/km<sup>2</sup>)

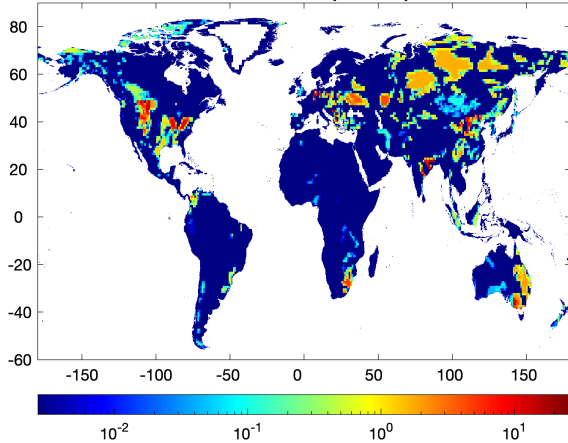


Reserves conventional oil (PJ/km<sup>2</sup>)

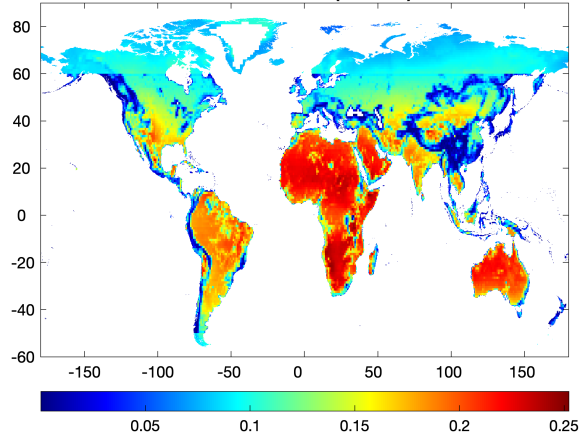


# Comparative Advantage for Fossil vs Green: Coal and Solar

Reserves coal (PJ/km<sup>2</sup>)

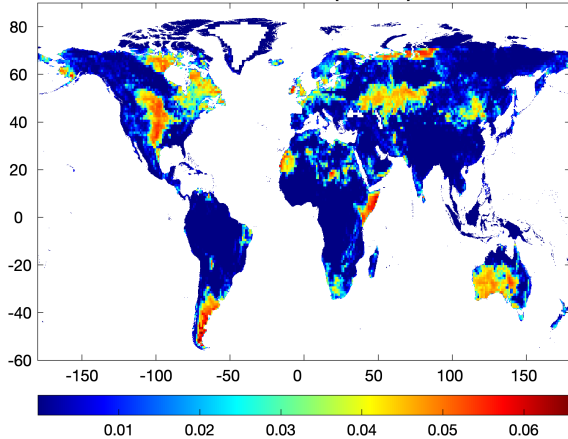


Reserves solar (PJ/km<sup>2</sup>)

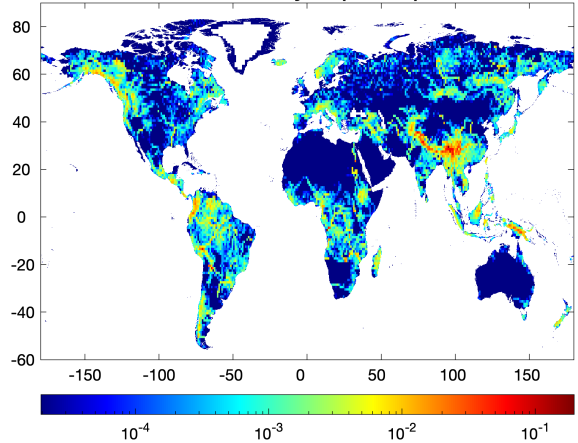


# Comparative Advantage for Green Energy: Wind and Hydro

Reserves wind (PJ/km<sup>2</sup>)

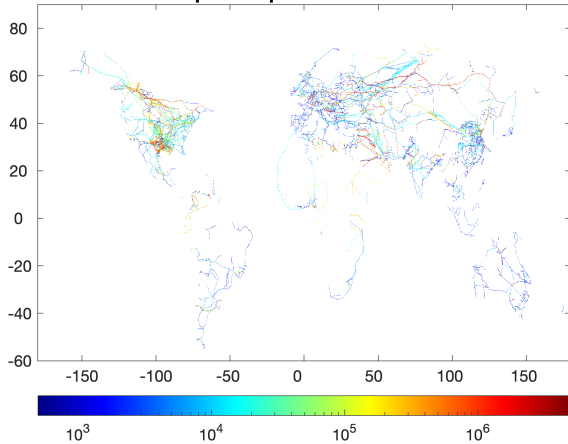


Reserves hydro (PJ/km<sup>2</sup>)

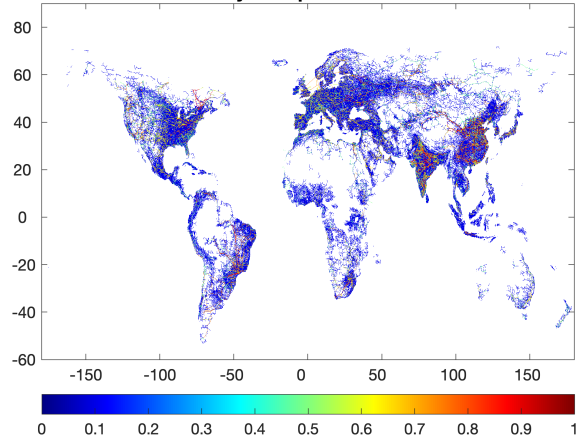


# Energy transportation network

Pipe transportation network



Electricity transportation network





# Preferences

- Household values amenities  $b_t(r)$ , consumption good  $c_t(r)$ , and **residential energy**  $e_t^h(r)$  as

$$u_t(r) = \bar{u} \bar{\varphi} \left( b_t(r) c_t(r)^\varphi e_t^h(r)^{1-\varphi} \right),$$

$$e_t^h(r) = \left( \varpi^{f,h}(r)^{\frac{1}{\epsilon^h}} e_t^{f,h}(r)^{\frac{\epsilon^h-1}{\epsilon^h}} + \varpi^{c,h}(r)^{\frac{1}{\epsilon^h}} e_t^{c,h}(r)^{\frac{\epsilon^h-1}{\epsilon^h}} \right)^{\frac{\epsilon^h}{\epsilon^h-1}},$$

$$b_t(r) = \bar{b}_t(r) L_t(r)^{-\lambda}, \quad \bar{b}_t(r) = \left( 1 + \Lambda^b(\Delta T_t(r), T_{t-1}(r)) \right) \bar{b}_{t-1}(r),$$

- where  $\bar{u} = (\iota + \gamma)/(\iota + \mu + \gamma)$  and  $\bar{\varphi} = \varphi^{-\varphi} (1 - \varphi)^{-(1-\varphi)}$  are constants
- Consumption and each type of energy are CES aggregators over a continuum of varieties (with elasticities of substitution  $\sigma^f, \sigma^c$ )
- Consider **taxes/subsidies** specific to each location  $r$ , period  $t$ , and energy source  $j$ ,  $\tau_t^{j,h}(r)$

$$\frac{P_t^{e,h}(v) e_t^h(v)}{1 - \varphi} = \frac{P_t^q(v) c_t(v)}{\varphi}, \quad \frac{e_t^{f,h}(r)}{e_t^{c,h}(r)} = \frac{\varpi^{f,h}(r)}{\varpi^{c,h}(r)} \left( \frac{(1 + \tau_t^{f,h}(r)) P_t^f(r)}{(1 + \tau_t^{c,h}(r)) P_t^c(r)} \right)^{-\epsilon^h}$$

# Mobility

- Households maximize intertemporal utility given by

$$V^w(r_0) = \sum_t \beta^t u_t(r_t) m(r_t)^{-1} \psi_{r_t}^\omega$$

- where  $m(r)$  denote migration costs of entering location  $r$  and  $\psi_{it}^\omega$  is an idiosyncratic iid Fréchet shock with shape governed by  $\Omega$
- The indirect utility in a period is then given by

$$u_t(r) = \bar{u} b_t(r) \left( \frac{w_t(r) + R_t(r)/L_t(r) + \Psi_t(r) + D_t(r)}{P_t^q(r)^\varphi P_t^{e,h}(r)^{1-\varphi}} \right),$$

where  $w_t(r)$  denotes the wage,  $R_t(r)$  land rents,  $\Psi_t(r)$  tax rebate,  $D_t(r)$  trade deficit (or surplus) and  $P_t^{e,h}(r)$  residential energy price index

- Fraction of households residing in each location

$$\frac{L_t(r)H(r)}{L_t} = \frac{[u_t(r)/m(r)]^{1/\Omega}}{\int [u_t(v)/m(v)]^{1/\Omega} dv}$$

# Consumption Good Technology

- Production function (of a variety) of consumption goods per unit of land

$$q_t(r) = z_t^q(r) a_t^q(r)^{\frac{1}{\theta^q}} L_t^{\text{prod},q}(r)^{\nu} L_t^{\text{innov},q}(r)^{\gamma} e_t^q(r)^{1-\nu-\mu-\gamma} h_t^q(r)^{\mu},$$

$$e_t^q(r) = \left( \varpi^{f,q}(r)^{\frac{1}{\epsilon^q}} e_t^{f,q}(r)^{\frac{\epsilon^q-1}{\epsilon^q}} + \varpi^{c,q}(r)^{\frac{1}{\epsilon^q}} e_t^{c,q}(r)^{\frac{\epsilon^q-1}{\epsilon^q}} \right)^{\frac{\epsilon^q}{\epsilon^q-1}},$$

- ▶  $h_t^q(r) = H_t^q(r)/H(r)$  is the share of land devoted to the production of goods
- ▶ Innovation costs given by  $\nu^q w_t(r)$ , expressed in terms of labor
- ▶ Each type of energy is a CES aggregator (with elasticities of substitution  $\sigma^f, \sigma^c$ )
- ▶  $z_t^q(r)$  is a productivity draw from an iid Fréchet with shape  $\theta^q$
- ▶  $a_t^q(r) = \bar{a}_t^q(r) L_t^q(r)^{\alpha^q}$  is good-specific productivity, evolving as

$$\bar{a}_t^q(r) = \left( 1 + \Lambda^a(\Delta T_t(r), T_{t-1}(r)) \right) \left( L_{t-1}^{\text{innov},q}(r) \right)^{\gamma} \left( \int_{v \in S} D^q(v, r) \bar{a}_{t-1}^q(v) dv \right)^{1-\xi^q} \left( \bar{a}_{t-1}^q(r) \right)^{\xi^q}$$

# Gravity in Goods Trade and Trade Balance

- Iceberg trade costs lead to gravity equation for bilateral trade flows in goods

$$\pi_t^q(s, r) = \frac{\bar{a}_t^q(r)(\ell_t^q(r)L_t(r))^{\alpha^q} mc_t^q(r)^{-\theta^q} \varsigma^q(r, s)^{-\theta^q}}{\int \bar{a}_t^q(v)(\ell_t^q(v)L_t(v))^{\alpha^q} mc_t^q(v)^{-\theta^q} \varsigma^q(v, s)^{-\theta^q} dv}$$

- ▶ where  $\varsigma^q(r, s)$  are goods transport costs,  $\ell_t^q(r)$  denotes the share of labor in the goods sector, and

$$mc_t^q(r) = w_t(r)^{\iota+\mu+\gamma} P_t^{e,q}(r)^{1-\iota-\mu-\gamma} L_t(r)^\mu$$

# Energy Production Technology

- Production function of energy type  $j \in \{f, c\}$  is a CES composite of different sources

$$\eta_t^j(r) = \left( \sum_{k \in K_j} \eta_t^{j,k}(r)^{\frac{\vartheta^j - 1}{\vartheta^j}} \right)^{\frac{\vartheta^j}{\vartheta^j - 1}},$$

$$\eta_t^{j,k}(r) = \frac{z_t^{j,k}(r) a_t^{j,k}(r)^{\frac{1}{\vartheta^j}}}{\kappa_t^{j,k}(r)} \left( L_t^{\text{prod},j,k}(r)^\iota L_t^{\text{innov},j,k}(r)^\gamma h_t^{j,k}(r)^\mu \right)^{\frac{1}{\iota + \mu + \gamma}}$$

- $z_t^{j,k}(r)$  is a productivity draw from an iid Fréchet with shape  $\theta^j$
- $\kappa_t^{j,k}(r)$  is the “cost” of the energy source
- $a_t^{j,k}(r) = \bar{a}_t^{j,k}(r) \left( \sum_{k \in K_j} L_t^{j,k}(r) \right)^{\alpha^j}$  is energy-specific productivity, evolving as

$$\bar{a}_t^{j,k}(r) = \left( 1 + \Lambda^a(\Delta T_t(r), T_{t-1}(r)) \right) \left( L_{t-1}^{\text{innov},j,k}(r) \right)^{\frac{\gamma}{\iota + \mu + \gamma}} \left( \int_{v \in S} D^j(v, r) \bar{a}_{t-1}^{j,k}(v) dv \right)^{1 - \xi^q} \left( \bar{a}_{t-1}^{j,k}(r) \right)^{\xi^q}$$

# Energy Production Technology

- $\eta_t^{j,k}(r)$  is the *flow* of resource  $j, k$  used in production
  - ▶  $\eta_t^{j,k}(r)$  cannot exceed the *stock* of resource,  $\Upsilon_t^{j,k}(r)$ , which evolves as

$$\Upsilon_{t+1}^{j,k}(r) = \Upsilon_t^{j,k}(r) - \mathbb{1}\{j = f\} \cdot \eta_t^{j,k}(r)$$

- ▶ Extraction of fossil fuels exhausts resources, but use of clean energy does not
- One unit of resource  $j, k$  costs

$$\kappa_t^{j,k}(r) = f^{j,k}(\Upsilon_t^{j,k}(r)H(r) - \mathbb{1}\{j = f\} \cdot \eta_t^{j,k}(r)H(r))$$

# Gravity in Energy Trade and Trade Balance

- Iceberg trade costs lead to gravity equation for bilateral trade flows in energy

$$\pi_t^j(s, r) = \frac{a_t^j(r) mc_t^j(r)^{-\theta^j} \varsigma^j(r, s)^{-\theta^j}}{\int a_t^j(v) mc_t^j(v)^{-\theta^j} \varsigma^j(v, s)^{-\theta^j} dv}$$

- where  $\varsigma^j(r, s)$  are energy transport costs, and

$$mc_t^j(r) := w_t(r) L_t(r)^{\frac{\mu}{\iota + \mu + \gamma}}$$

- We impose trade balance by location across sectors so

$$p_t^q(r, r) q_t(r) + \sum_j p_t^j(r, r) \eta_t^j(r) = P_t^q(r) c_t(r) L_t(r) + \sum_j P_t^j(r) \left( e_t^{j,h}(r) L_t(r) + e_t^{j,q}(r) \right)$$

# Government and Climate Evolution

- Tax collection is rebated at the cell level to households

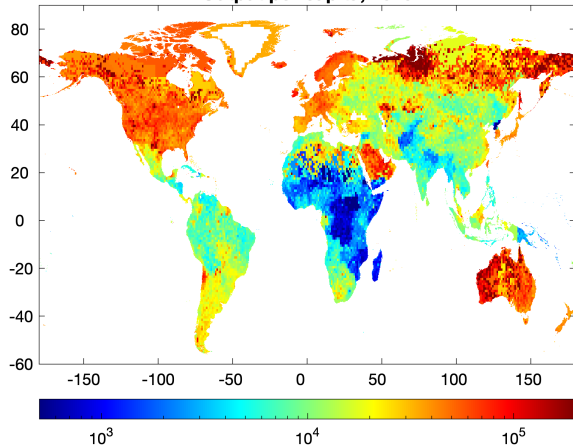
$$\Psi_t(r) = \frac{\left( \tau_t^{j,h}(r) P_t^j(r) e_t^{j,h}(r) L_t(r) + \tau_t^{j,q}(r) P_t^j(r) e_t^{j,q}(r) \right) H(r)}{L_t(r) H(r)}$$

- Climate module as in Estrada et al. (2022) and Meinshausen et al. (2020) model

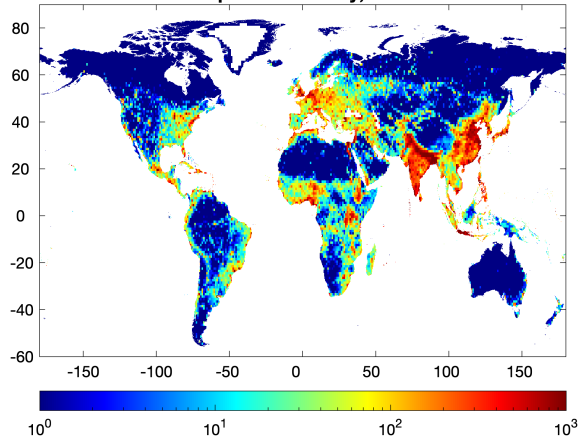


# GCP and population

Output per capita, 2020



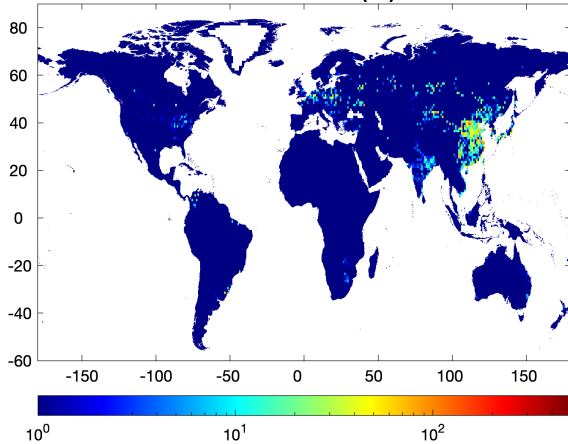
Population density, 2020



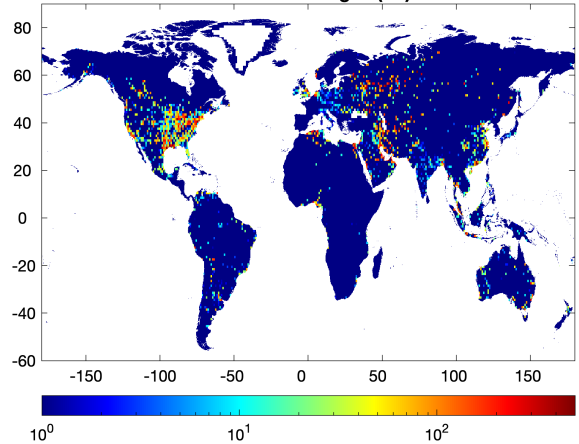
- GCP data from Rossi-Hansberg and Zhang (2025) in 2020 base year
- Input minimum GCP per capita and population by country in missing cells (20% of cells)

# Energy Production

Production coal (PJ)

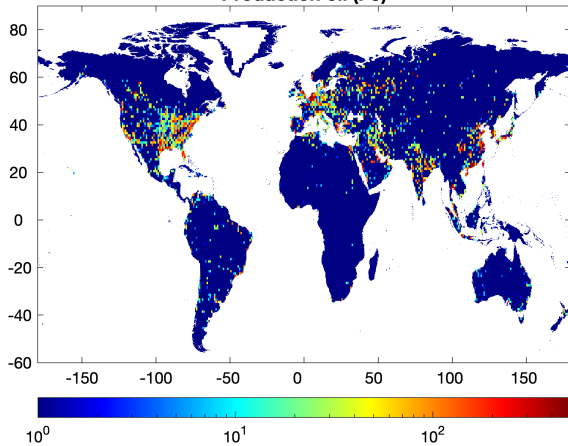


Production natgas (PJ)

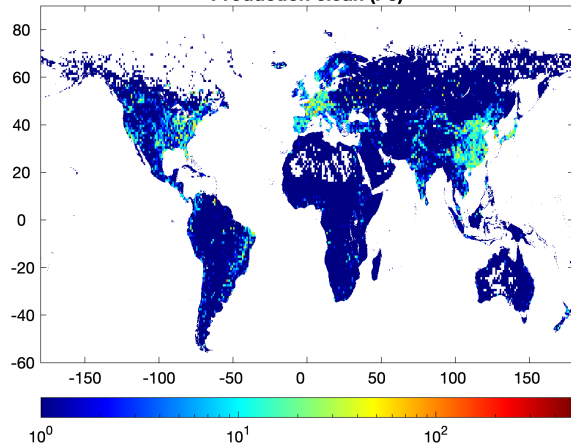


# Energy Production

Production oil (PJ)



Production clean (PJ)



# Extraction Cost Function for Fossil Fuels

- Parametrize the extraction cost,  $\kappa_t^{f,k}(\cdot)$ , as a function of reserves,  $\Upsilon_t^{f,k}(\cdot)H(\cdot)$

$$\kappa_t^{f,k}(r) = \exp \left( F_0^{f,k} + F_0^{f,k,r} \mathbb{1}\{\text{region}(r)\} + F_0^{f,k,s} \mathbb{1}\{\text{source}(r)\} \right) \left( \Upsilon_t^{f,k}(r)H(r) \right)^{F_1^{f,k}}$$

- ▶ Welsby et al. (2021) provides cumulative resource use and their associated extraction cost for coal, gas, and oil, categorized into 16 regions and source of fossil fuels
- Gas and oil highly sensitive to reserve scarcity, but not coal

# Extraction Cost Function: Coal, Natural Gas, and Oil

	Coal	Natural Gas	Oil
log(resource)	0.0786* (0.0463)	-0.1421*** (0.0181)	-0.0496*** (0.0103)
FE region	X	X	X
FE source		X	X
$N$	102	300	304
$R^2$	0.1489	0.5031	0.6483

Standard errors in parentheses

\* $p < 0.10$ , \*\* $p < 0.05$ , \*\*\* $p < 0.01$

# Clean Energy Cost Function

- Parametrize the clean energy costs,  $\kappa_t^{c,k}(\cdot)$ , as a function of suitability,  $\Upsilon_t^{c,k}(\cdot)H(\cdot)$ 
  - ▶ NREL (State and Local Planning for Energy) provides the Levelized Cost of Energy (usd/MWh) at the US county-level for 2020

$$\log \left( \kappa_t^{c,k}(r) \right) = F_0^{c,k} + F_1^{c,k} \log \left( \Upsilon_t^{c,k}(r)H(r) \right) + \epsilon_t^{c,k}(r)$$

- Solar is the most sensitive to suitability, and hydro the least

# Clean Energy Cost Function: Solar, Wind, and Hydro

	Commercial Solar	Residential Solar	Commercial Wind	Residential Wind	All Hydro
log(suitability)	-0.6628*** (0.0069)	-0.6795*** (0.0074)	-0.2672*** (0.0181)	-0.2181*** (0.0113)	-0.0090*** (0.0018)
weight pop	X	X	X	X	X
<i>N</i>	3,048	3,048	1,955	1,985	1,562
<i>R</i> <sup>2</sup>	0.7522	0.734	0.1006	0.1579	0.0155

Standard errors in parentheses

\* $p < 0.10$ , \*\* $p < 0.05$ , \*\*\* $p < 0.01$

# Trade Costs

- We calculate transport costs using the shortest path, and considering multimode options (Fuchs and Wong, 2022)
  - ▶ Important to consider multi-mode options for fossil
  - ▶ We use MatLab function `distances` to compute shortest path
- Parametrize transport adjacency,  $\mathcal{T}_m^j(\cdot)$ , and switching,  $\mathcal{S}_{m,n}^j(\cdot)$ , costs

$$\log \mathcal{T}_m^j(r, s) = \begin{cases} t_{0,m}^j + t_{1,m}^j \cdot \log \text{dist}(r, s) + t_{2,m}^j \cdot (\log T_m(r) + \log T_m(s)) > 0 & \text{if } r \text{ and } s \text{ are adjacent} \\ \infty & \text{otherwise} \end{cases}$$

$$\log \mathcal{S}_{m,n}^j(r, s) = \begin{cases} s_{0,m,n}^j + s_{1,m,n}^j \cdot (\log S_m(r) + \log S_n(r)) > 0 & \text{if } r = s \\ \infty & \text{otherwise} \end{cases}$$

- Consider six transport modes: road, road-rail-road, road-air-road, road-water-road, road-pipe-road, electricity lines
- Estimate to match predicted and observed transport shares using gravity equation in each sector



# Trade Mode for Goods and Fossil Energy

**Trade value by good and transportation mode**  
(billion usd 2017)

	Road	Rail	Water	Air	Pipes
Energy	1,030.03	625.47	294.20	0.03	388.05
Goods	12,584.61	2,782.41	397.21	645.31	0

# Electricity Trade Costs

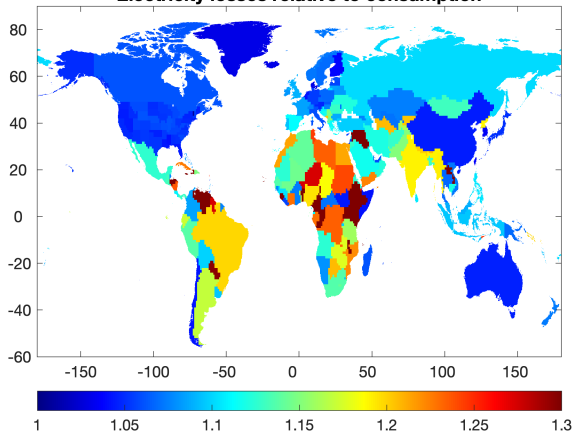
- Bohn et al. (1984) and Arkolakis and Walsh (2023) establish that electricity losses in line  $\ell$  are

$$\text{losses}_\ell \approx \left( \frac{\text{line length}_\ell}{\text{voltage}_\ell} \right) (\text{power flow}_\ell)^2$$

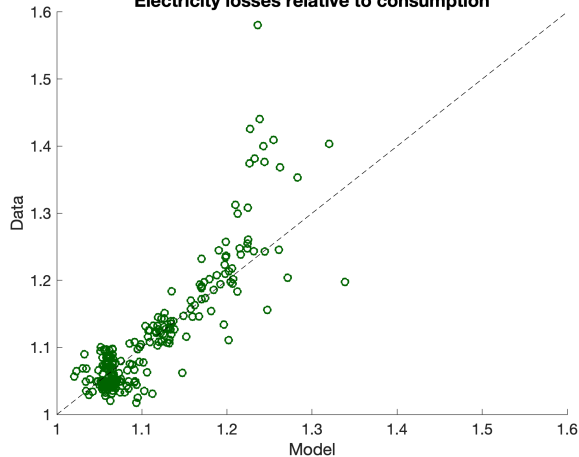
- US EIA provides yearly electricity generation, consumption, exports, imports, and distribution losses at the country- and US state-level (238 regions)
  - ▶ Allocate electricity consumption at the cell-level using the distribution of GDP, so that country-, USA state- and CHN province-level totals match EIA and NBSC data
- Flows depend on transport costs (losses) and transport costs depend in turn on power flows
  - ▶ Need an iterative procedure to find fixed point

# Electricity Trade Costs

Electricity losses relative to consumption

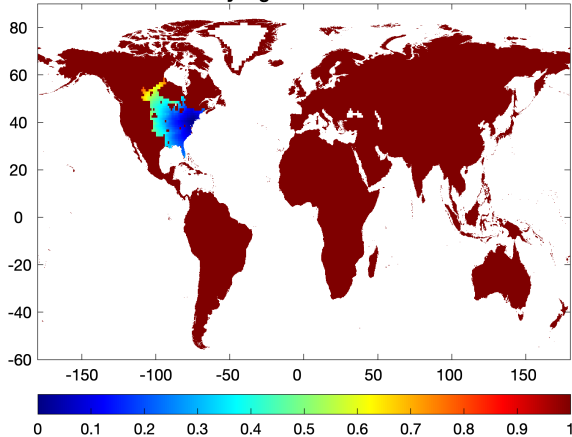


Electricity losses relative to consumption

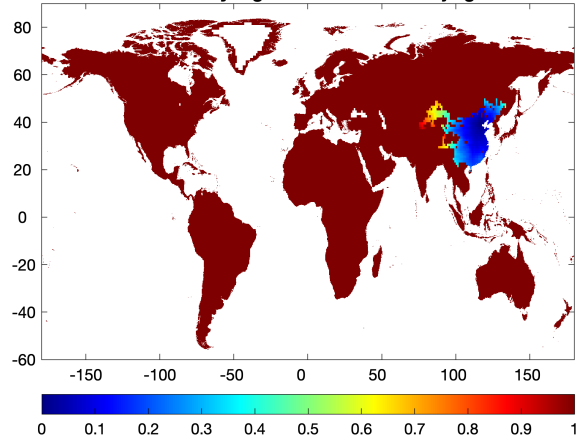


# Electricity Trade Costs

Electricity log trade cost from NYC



Electricity log trade cost from Beijing



# Growth Parameters

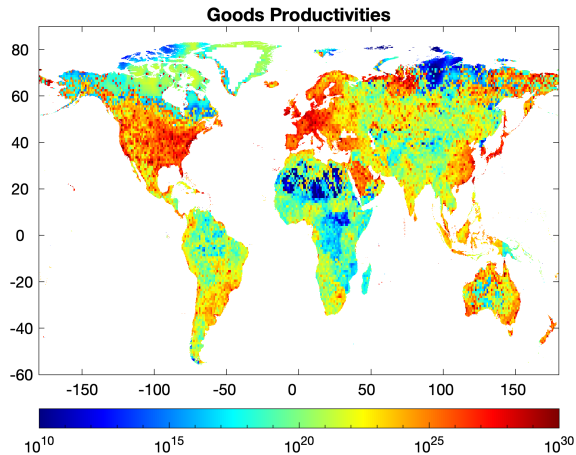
- Evolution of productivity in sectors  $x \in \{q, f, c\}$

$$\bar{a}_t^x(r) = \left(1 + \Lambda_t^a(r)\right) \underbrace{\left(\frac{\gamma/\nu^x}{\nu + \gamma} L_{t-1}^x(r)\right)}_{L_{t-1}^{\text{innov},x}(r)}^\gamma \left(\int_{v \in S} D^x(v, r) \bar{a}_{t-1}^x(v) dv\right)^{1-\xi^x} \bar{a}_{t-1}^x(r)^{\xi^x}$$

$$D^x(v, r) = \exp(-\aleph^x \text{dist}(v, r)) / \int_{s \in S} \exp(-\aleph^x \text{dist}(s, r)) ds$$

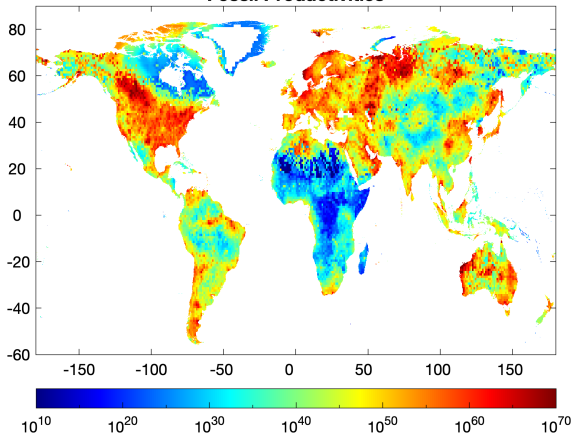
- Set the parameters  $\aleph^q, \aleph^f, \aleph^c, \nu^q, \nu^f, \nu^c$  to target:
  - ▶ Standard deviation of productivities in 2021 equal to that in 2020
  - ▶ Real GDP growth of 2% (Desmet et al., 2018)
  - ▶ CO<sub>2</sub> emissions growth of 3.7% (Arias et al., 2021) (2020-2030 in scenario SSP 5 RCP 8.5)
  - ▶ Clean energy growth of 6.0% (IEA, 2024) (2020-2050 in scenario Stated Policies, similar to growth rate during 2010-2022 of 6.9%)
  - ▶ Fix  $\gamma = 0.0024$  and  $\xi^q = \xi^f = \xi^c = 0.993$  as in Desmet et al. (2018)

# Good Sector Productivities

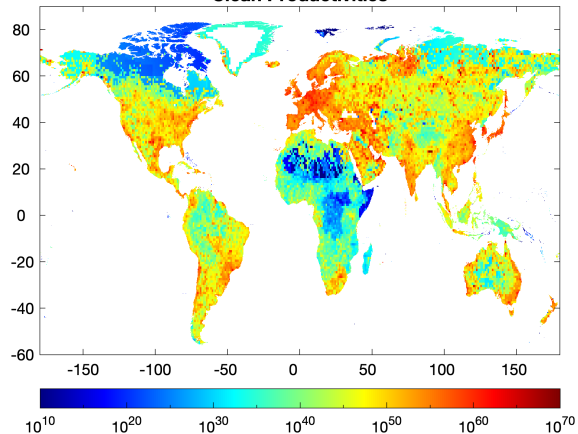


# Fossil and Clean Energy Productivities

**Fossil Productivities**

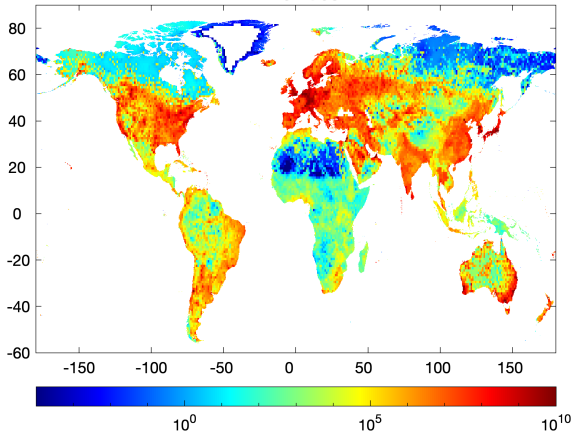


**Clean Productivities**

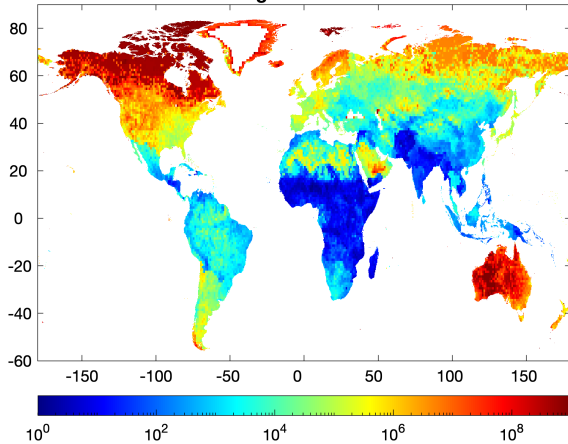


# Amenities and Migration Costs

Amenities



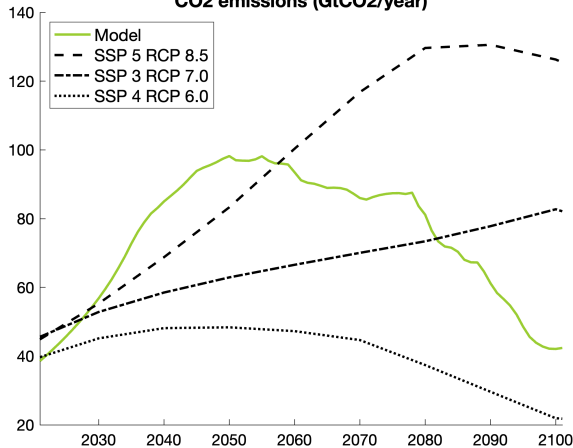
Migration costs



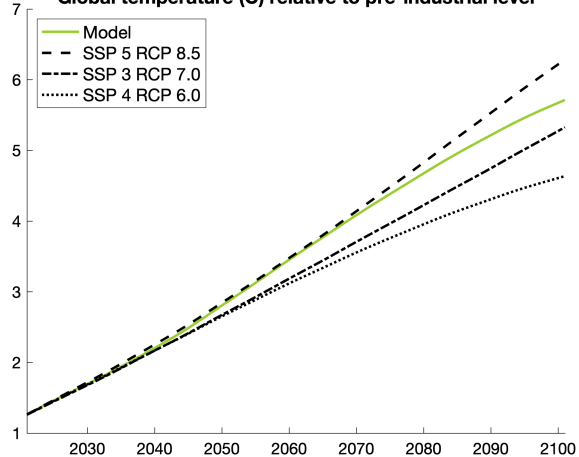


# Results: CO<sub>2</sub> Emissions and Temperature

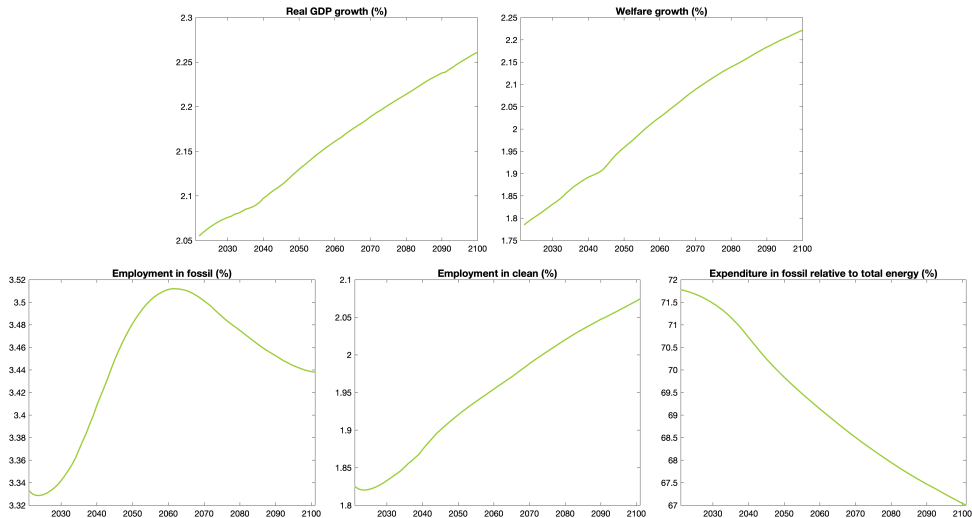
CO<sub>2</sub> emissions (GtCO<sub>2</sub>/year)



Global temperature (C) relative to pre-industrial level

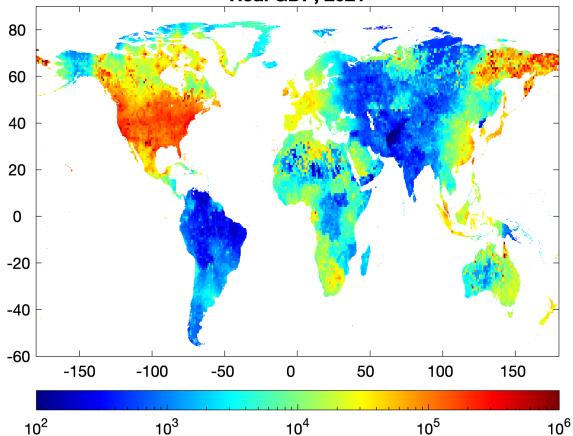


# Results: The Evolution of GDP, Welfare, Energy

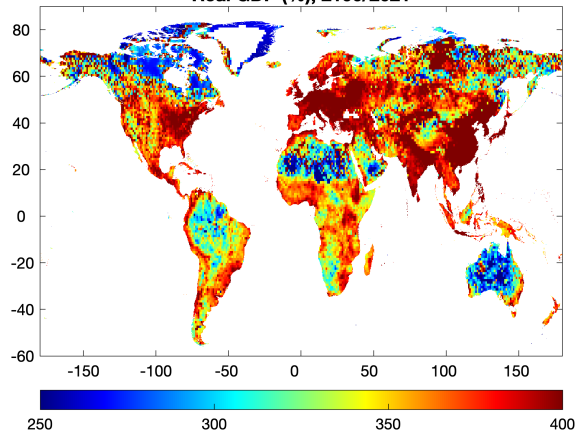


# Results: Real GDP per Capita

Real GDP, 2021

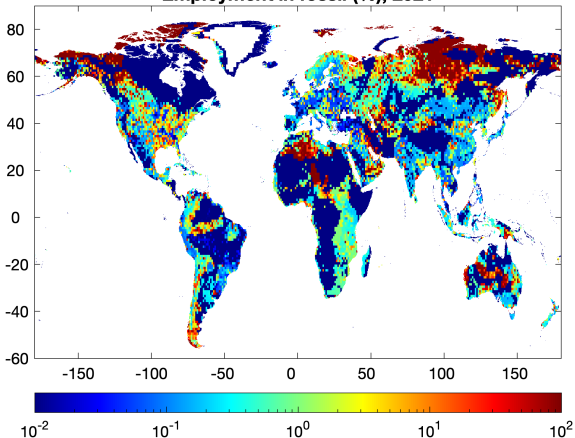


Real GDP (%), 2100/2021

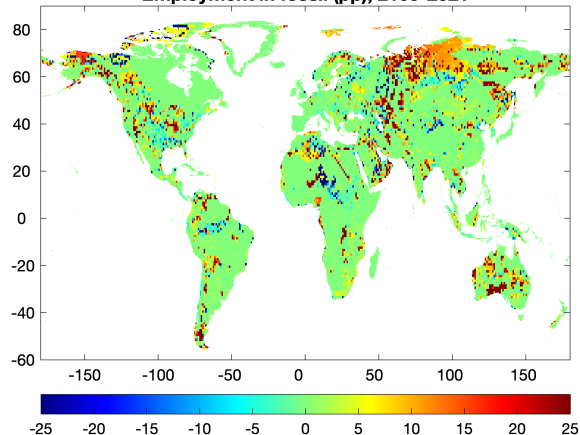


# Results: Employment in Fossil Fuel Industry

Employment in fossil (%), 2021

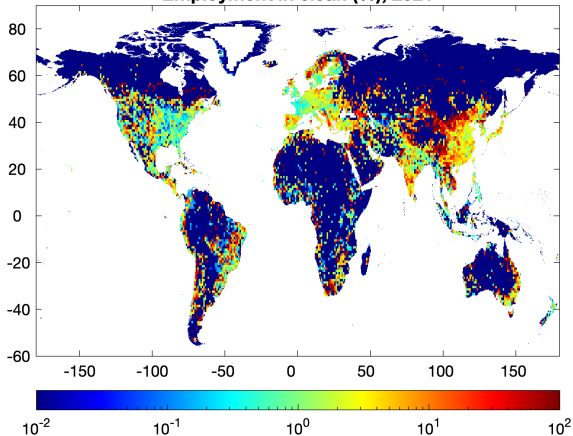


Employment in fossil (pp), 2100-2021

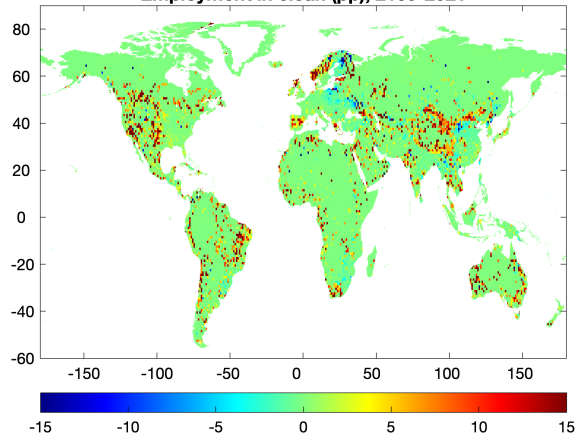


# Results: Employment in Clean Energy Industry

Employment in clean (%), 2021

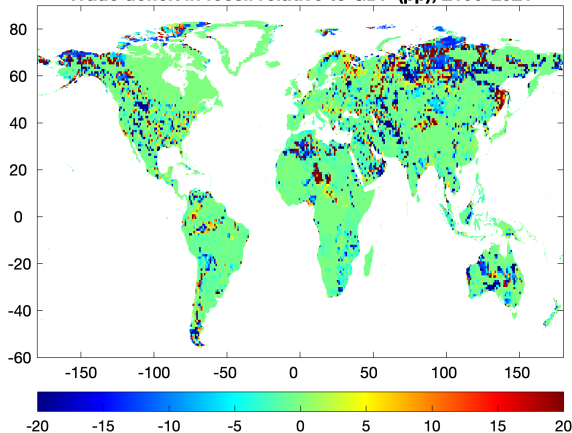


Employment in clean (pp), 2100-2021

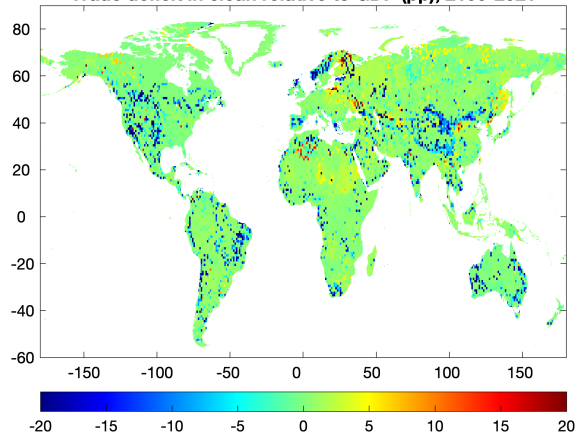


# Results: Trade Deficit in Fossil Fuel and Clean Energy Industry

Trade deficit in fossil relative to GDP (pp), 2100-2021

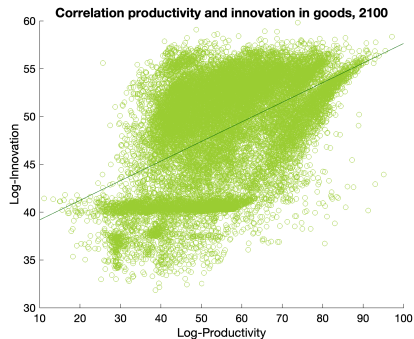
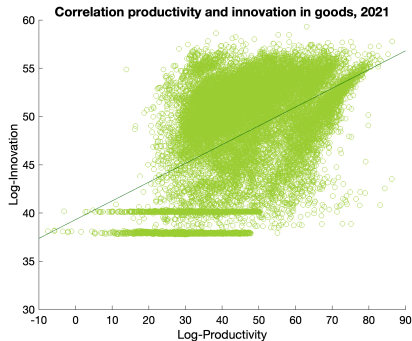


Trade deficit in clean relative to GDP (pp), 2100-2021



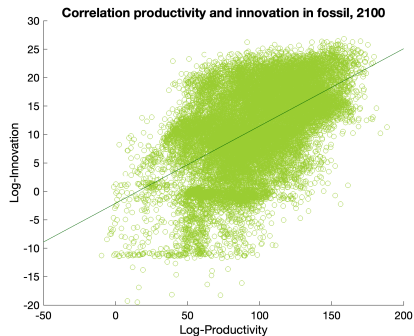
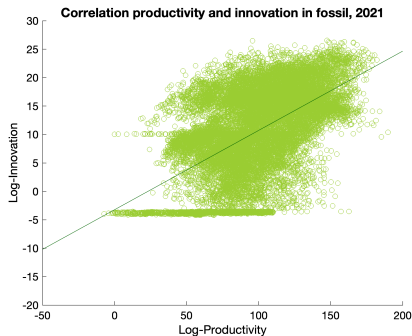
# Results: Productivity and Innovation in Goods

- Innovation in goods larger in productive locations with little change over time



# Results: Productivity and Innovation in Fossil Fuels

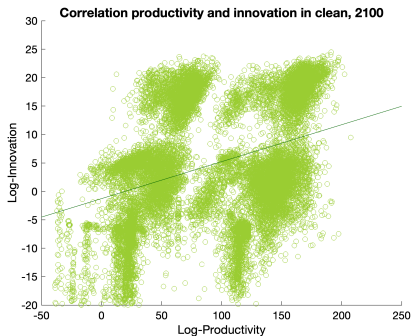
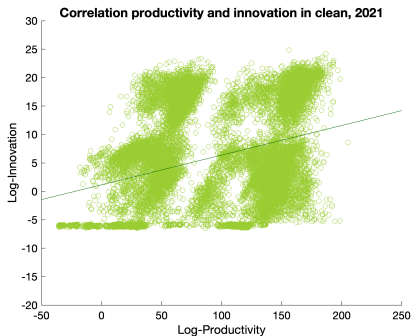
- Similar correlation between innovation and productivity. Similar over time.





# Results: Productivity and Innovation in Clean Energy

- Low and high productivity regions have a similar relationship with innovation
  - ▶ **Low productivity one due to comparative advantage**
  - ▶ **High productivity one due to initial productivity advantage and dynamic scale effects**
- Can explain the lack of **"Green Development"**



# Next Steps

- Resulting model can be useful to do many **potentially interesting counterfactuals**
  - ▶ Changes in clean energy policy in the developed world
    - ★ Could help developing countries gain comparative advantage in clean energy given their endowments
  - ▶ Technology diffusion incentives and policies
    - ★ Green energy technology in the developing world is not sufficient to generate green clusters
  - ▶ Changes in the grid and changes in energy trading costs
    - ★ Can high voltage grid connections between Africa and Europe help?
  - ▶ Implications of large energy innovations (fusion)

# Thank you!

# Appendix

# Resources and Energy Suitability

- James Oakleaf provided access to reserve data in yields/km<sup>2</sup> (Oakleaf et al., 2019, 2015)  
maps
  - ▶ Coal (tons/km<sup>2</sup>), conventional gas and oil (boe/km<sup>2</sup>), unconventional gas and oil (boe/km<sup>2</sup>), concentrated and photovoltaic solar power (MW/km<sup>2</sup>), wind (MW/km<sup>2</sup>), hydropower (MW/km<sup>2</sup>)
  - ▶ Transform projection from Mollweide (5km) to latitude-longitude (1°) coordinates (Snyder, 1987)
  - ▶ Express all sources in TJ/km<sup>2</sup>
  - ▶ Complete missing data for solar: interpolate solar potential by regressing it on elevation, latitude and longitude ( $R^2 = 0.61, 0.88$ )
- Adjust from fossil fuels reserves to resources with the region-level estimates of Welsby et al. (2021)
  - ▶ Differentiate between conventional and unconventional fuels
  - ▶ At the country- or US state-level, proportionally adjust reserves so that extraction/reserves lies between (global extraction/global reserves)/5 and (global extraction/global reserves)\*5

# Energy Extraction, Production and Consumption

- Energy balances at the country-level are taken from IEA (2023) and at the US-state level from the State Profiles and Energy Estimates of the U.S. Energy Information Administration energy
- Energy consumption
  - ▶ Industrial energy consumption in the IEA dataset is represented by the sum of the following flows: Agriculture, Fishing, Total Industry, Commercial and Public Services, Energy industry own use,
  - ▶ Residential energy consumption in the IEA dataset is represented by the flow of Residential
  - ▶ The flows of International marine bunkers, International aviation bunkers, Transport, Non-energy use and Final consumption not specified elsewhere are proportionally allocated between Industrial and Residential energy consumption

# Energy Extraction, Production and Consumption

- Accounting identity for energy balances (IEA, 2023)

$$\text{Production} + \text{Imports} - \text{Exports} - \text{International Transport} \pm \text{Stock changes} \\ = \mp \text{Transformation} + \text{Energy own use} + \text{Final uses} \mp \text{Stat diff}$$

- ▶ Production is that of primary energy, does not include secondary energy
- ▶ Transformation processes comprise the conversion of primary energy (e.g., crude oil) into secondary (e.g., gasoline) -inputs have negative values and outputs positive
- ▶ Energy industry own use covers the amount of fuels used by the energy producing industries (e.g., heating, lighting, operation of equipment)
- ▶ Final uses = Residential + Commercial + Domestic Transport + Non-energy
- ▶ Data at the country- or region-, product-, and year-level

# Energy Production and Consumption

- Disaggregate energy across states within the US with data from EIA
  - ▶ Differentiate between residential and commercial consumption
  - ▶ Allocate production of fossil fuels with production of raw sources
  - ▶ Allocate production of electricity with consumption



# Disaggregate energy across cells

- Construct the capacity and location of energy plants,  $K_t^j(\cdot)$  maps
  - ▶ Collect plant-level data from Global Energy Monitor (coal, oil and gas, solar, wind, hydro, geothermal, bioenergy and nuclear plants) (Global Energy Monitor, 2025a,b,c,d, 2024)
  - ▶ Consider all plants (including those canceled) and harmonize capacity units across energy sources
  - ▶ Supplement with World Resource Institute data (World Resource Institute, 2021) (remove duplicate observations: same energy source and same plant name, or same energy source and distance between plant locations is less than 1 km)
  - ▶ Eliminate fossil plants located in cells with no fuel yields
  - ▶ In cells with fuel yields but no plants, assume there is a plant with the lowest worldwide capacity
  - ▶ Normalize  $K_t^j(r) = \left( \frac{\int_{r \in \mathcal{R}} Y_t^j(r) dr}{\int_{r \in \mathcal{R}} K_t^j(r) dr} \right) K_t^j(r)$

# Energy Prices

- Supply costs and retail prices of coal, oil, natural gas and electricity at country-level from IMF Fossil Fuel Subsidies Data (Black et al., 2023) maps
- Prices of coal, oil, natural gas and electricity at US state-level from US Energy Information Agency
  - ▶ Aggregate across end-users (i.e., residential, industry) and fuel types (e.g., gasoline, kerosene)
  - ▶ Transform energy prices from MER to PPP with data from Rossi-Hansberg and Zhang (2025)
  - ▶ Carbon pricing is defined as Retail prices net of Value Added Taxes minus Supply costs  
-applicable VAT rate is the actual VAT times the share of final consumption

# Employment in Fossil Fuels and Clean Energy

- From the energy firm's optimality conditions

$$\ell_t^j(r) = \left( \frac{\iota + \gamma}{\iota + \gamma + \mu} \right) \left( \frac{p_t^j(r, r) \eta_t^j(r) H(r)}{w_t(r) L_t(r) H(r)} \right)$$

- Global average of fossil and clean employment: 2.75% and 3.85%
- IEA (2023) estimates global energy employment of 2%, equally distributed into fossil and clean

$$\begin{aligned} \min \quad & \int_{r \in \mathcal{R}} \sum_{j \in \mathcal{J}} W^j(r) \left( \hat{\ell}_t^j(r) - \ell_t^j(r) \right)^2 dr \\ \text{s. t.} \quad & \int_{r \in \mathcal{R}} w_t(r) L_t(r) H(r) \hat{\ell}_t^j(r) dr = \left( \frac{\iota + \gamma}{\iota + \gamma + \mu} \right) \int_{r \in \mathcal{R}} p_t^j(r, r) \eta_t^j(r) H(r) dr \quad \forall j \in \mathcal{J}, \\ & \hat{\ell}_t^f(r) + \hat{\ell}_t^c(r) \leq 0.9 \quad \forall r \in \mathcal{R}, \quad \hat{\ell}_t^j(r) \geq 10^{-6} \quad \forall j \in \mathcal{J} \quad \forall r \in \mathcal{R}, \\ & W^j(r) > 0 \quad \text{are weights} \end{aligned}$$

- Weights are such that in cells with no fossil yields,  $\hat{\ell}_t^f(\cdot)$  takes its minimum value

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