The Carbon Footprint of Multinational Production

Ezequiel Garcia-Lembergman Pontificia Universidad Católica de Chile Natalia Ramondo Boston University, CEPR & NBER

Andrés Rodríguez-Clare UC Berkeley, CEPR & NBER Joseph S. Shapiro UC Berkeley & NBER

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Emissions per Dollar Very Different Across Countries

$$\frac{\mathcal{E}_{l,s}}{Y_{l,s}} = \gamma_l + \delta_s + \varepsilon_{l,s}$$



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Example: The Steel Industry in Vietnam

Tenova (Italy): mini mill (electric arc furnace)





Kunming Iron & Steel (China): integrated mill (blast furnace)



MP China->Vietnam



Vietnam steel corporation (Vietnam)



• Comprehensive global data on MNEs and environment

• Affiliates from cleaner Home countries have lower emissions per dollar everywhere

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- Model of trade, MP and energy
 - Trade & MP (as Arkolakis et al., 18): closed-form & aggregation, GE
 - Multiple sectors & IO (as Caliendo & Parro, 14): energy, mining, other inputs
 - Energy-intensity technology choice (as Sun, 20): affects firm emissions worldwide

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Estimation

- Energy demand and supply elasticities: IV using US administrative micro data
- Technology choice elasticities: SMM to match new fact on emissions

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Carbon accounting

Consumption, Production, Extraction, Ownership

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Counterfactual exercises (not today)

e.g. MP autarky; MP liberalization; carbon taxes

What is New?

- Trade & environment Grossman & Krueger 1995; Copeland & Taylor 2004; Nordhaus 2015; Shapiro & Walker 2018; Kortum & Weisbach 2021; Farrokhi & Lashkaripour 2022
 - Analyze multinational production, not just trade
 - Incorporate in GE several mechanisms in the literature (e.g. technique, composition)
- Multinational production Helpman 1984; Markusen & Venables 2000; Helpman, Melitz, & Yeaple 2004; Ramondo & Rodriguez-Clare 2013; Arkolakis, Ramondo, Rodríguez-Clare, & Yeaple (ARRY) 2018; Sun 2020
 - Allow for energy-emissions link
- Carbon accounting Davis & Caldeira 2010; Peters et al. 2011; Zhang et al. 2020
 - Emissions accounting by ownership
- Second-best climate policy Goulder et al. 2012; Martin et al. 2014; Fowlie et al. 2016; Bohringer et al. 2016; Shapiro 2021
 - Policy for multinational production

Outline

• Data and new fact

Model

Estimation

Model-based results

Conclusions

Data on Emissions and MP: Sources

Aggregate data

• World Input Output Dataset (WIOD) and Exiobase/Eora

Emissions and energy consumption by industry-country-energy type

• Activity of Multinational Enterprises (AMNE)

Revenues by industry-origin country-host country

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• Firm and affiliate data

• Carbon Disclosure Project (CDP) and ORBIS

Emissions per dollar for each parent and country of production

• US Census of Manufactures and Manufacturing Energy Consumption Survey

Affiliates from Cleaner Countries Are Cleaner Everywhere

Firm f, home country i, host country I, industry s. E Emissions. Y Revenue

$$\log\left(\frac{\mathcal{E}_{fi,l,s}}{Y_{fi,l,s}}\right)^{CDP} = \beta_1 \log\left(\frac{\mathcal{E}_i}{Y_i}\right)^{WIOD} + X'_{f,l}\gamma + \delta_{l,s} + \epsilon_{fi,l,s}$$

Dependent variable:	Log firm CO ₂ rate					
Home log CO_2 rate	0.96*** (0.24)	1.07*** (0.22)	0.56* (0.30)	0.63** (0.25)	0.63** (0.23)	<mark>0.60**</mark> (0.29)
Host log CO_2 rate	0.89*** (0.09)	0.86*** (0.09)				
Firm log revenues	、					-0.48*** (0.08)
Observations	4,833	4,833	4,833	4,833	4,833	4,833 [́]
R-squared	0.05	0.24	0.28	0.48	0.63	0.70
# host countries	42	42	42	42	42	42
# home countries	32	32	32	32	32	32
Industry FE	no	yes	no	yes	-	-
Host country FE	no	no	yes	yes	-	-
Industry × host country FE	no	no	no	no	yes	yes

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Model: Notation and Preliminaries

• Many countries

• *i* home country of firms. *I* location of production. *n* destination of sales

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- Many sectors, input-output loop
 - Six energy sectors $s \in \mathcal{K}^E$

Mining sectors: coal, natural gas, crude oil (fossil fuels) $s \in \mathcal{K}^M$

Non-mining sectors: electricity, refined oil, gas distribution

• Non-energy sectors $s \notin \mathcal{K}^E$

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• Preferences: Cobb-Douglas across sectors $(\mu_{n,s})$; CES within each sector (σ_s)

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- Production in each mine has decreasing returns $\nu_s = 1 \eta_{l,\ell s} + \sum_{k \in \mathcal{K}} \eta_{l,ks} \in (0,1)$

$$q = B_{l,s} \ \ell^{\eta_{l,\ell s}} \prod_{k \in \mathcal{K}} q_k^{\eta_{l,k s}}$$

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- Trade: Armington (σ_s) . MP: exogenous $(M_{i,l,s})$. Perfect competition
- Emissions are generated exclusively by using fossil fuels

 $Q_{i,l,s}$ = production from mine (l, s) belonging to i

$$\mathcal{E} = \sum_{s \in \mathcal{K}^{\mathcal{M}}} \sum_{i,l} e_s Q_{i,l,s}$$

• Production (ε , γ): energy, non-energy inputs, labor

• Trade and MP: multivariate Pareto productivity (θ_s, ρ_s) , monopolistic competition (ν_s)

• Energy-intensity technology choice $(\tilde{\gamma}, \tilde{\varepsilon})$: firms have same technology everywhere

Model: Production Function for Non-mining Sectors

Energy vs non-energy inputs

$$q = \left(\left(q^E \right)^{\frac{\varepsilon - 1}{\varepsilon}} + \left(q^{NE} \right)^{\frac{\varepsilon - 1}{\varepsilon}} \right)^{\frac{\varepsilon}{\varepsilon - 1}}$$

$$\varepsilon \neq 1$$

Model: Production Function for Non-mining Sector

Energy inputs: coal, crude oil, natural gas, electricity, refined oil, gas distribution

$$q = \left(\left(\sum_{k \in \mathcal{K}^{\mathcal{E}}} \delta_{l,ks}^{\frac{1}{\gamma}} \left(q_{k} \right)^{\frac{\gamma-1}{\gamma}} \right)^{\frac{\gamma}{\gamma-1}\frac{\varepsilon-1}{\varepsilon}} + \left(q^{N\mathcal{E}} \right)^{\frac{\varepsilon-1}{\varepsilon}} \right)^{\frac{\varepsilon}{\varepsilon-1}}$$

$$\gamma \neq 1$$
 $\varepsilon \neq 1$

Model: Production Function for Non-mining Sector

Non-energy inputs: labor and other inputs

$$q = \left(\left(q^{\mathcal{E}} \right)^{\frac{\varepsilon - 1}{\varepsilon}} + \left(\ell^{\beta_{l,\ell s}} \prod_{k \notin \mathcal{K}^{\mathcal{E}}} q_k^{\beta_{l,k s}} \right)^{\frac{\varepsilon - 1}{\varepsilon}} \right)^{\frac{\varepsilon}{\varepsilon - 1}}$$

$$\beta_{I,\ell s} + \sum_{k \notin \mathcal{K}^E} \beta_{I,ks} = 1$$
 for all s

Model: Production Function for Non-mining Sector

A firm has productivity vector $\mathbf{z} \equiv (z_1, z_2, ..., z_N)$ and technology $\mathbf{a} \equiv (a_1, a_2, ..., a_{K^E}, a)$

$$q = \mathbf{z}_{l} \left(\left(\sum_{k \in \mathcal{K}^{\mathcal{E}}} \delta_{l,ks}^{\frac{1}{\gamma}} \left(\mathbf{a}_{k} q_{k} \right)^{\frac{\gamma-1}{\gamma}} \right)^{\frac{\gamma}{\gamma-1} \frac{\varepsilon-1}{\varepsilon}} + \left(\mathbf{a} \, \ell^{\beta_{l,\ell s}} \prod_{k \notin \mathcal{K}^{\mathcal{E}}} q_{k}^{\beta_{l,ks}} \right)^{\frac{\varepsilon-1}{\varepsilon}} \right)^{\frac{\varepsilon}{\varepsilon-1}}$$

$$\gamma \neq 1$$
 $\varepsilon \neq 1$ $\beta_{I,\ell s} + \sum_{k \notin \mathcal{K}^E} \beta_{I,ks} = 1$ for all s

- A firm from *i* draws productivity *z* from multivariate Pareto
 - $heta_{s}>\max\left\{1,1/\left(\sigma_{s}-1
 ight)
 ight\}$ and $ho_{s}\in\left[0,1
 ight)$

$$\Pr(Z_1 \le z_1, ..., Z_N \le z_n) = 1 - \left(\sum_{l=1}^N \left(T_{i,l,s} z_l^{-\theta_s}\right)^{\frac{1}{1-\rho_s}}\right)^{1-\rho_s}$$

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 - "Head-to-head" comparison (unit costs)

$$C_{i,n,s} = \min_{l} \tau_{i,ln,s} \frac{C_{i,l,s}}{z_l}$$

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$$C_{i,n,s} = \min_{l} \tau_{i,ln,s} \frac{c_{i,l,s}}{z_l}$$

Selection (marketing fixed costs)

$$Profits(C_{i,n,s}) - P_{n,s}F_{n,s} \geq 0$$

1

Model: Technology Choice for Non-Mining Firms

• A firm chooses its technology *a* from the set

 $\tilde{\varepsilon} \neq 1 \qquad \qquad \varepsilon + \tilde{\varepsilon} < 2$

$$\left(\boldsymbol{a}^{\mathcal{E}}
ight)^{1-\widetilde{arepsilon}}+\left(\boldsymbol{a}^{\mathcal{N}\mathcal{E}}
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$$\left(\sum_{k\in\mathcal{K}^{E}}a_{k}^{1-\tilde{\gamma}}\right)^{\frac{1-\tilde{\varepsilon}}{1-\tilde{\gamma}}}+a^{1-\tilde{\varepsilon}}\leq1$$

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• A firm chooses a before knowing z to maximize expected global profits

 \Rightarrow **a** is common across all (*i*, *s*) firms

Model: Optimal Technology Choice Across Energy Types

Slope of technology frontier (lhs) = Slope of iso-profit curve (rhs)

$$\left(\frac{a_{i,ks}}{a_{i,1s}}\right)^{1-\bar{\gamma}} = \frac{\sum_{l} \alpha_{i,l,ks} Y_{i,l,s}}{\sum_{l} \alpha_{i,l,1s} Y_{i,l,s}}, \quad \forall k \in \mathcal{K}^{\mathsf{E}}$$

- $\alpha_{i,l,ks} \equiv$ revenue share of k input for (i, l, s) firms equation
- $Y_{i,l,s} \equiv$ output of (i, l, s) firms
- $\sum_{l} \alpha_{i,l,ks} Y_{i,l,s} \equiv$ expected global costs of input k for (i, l, s) firms

Model: MP Autarky—Analytical Counterfactual

• No IO loop; exogenous technologies $(a_{i,s})$; no trade in energy; CRS in mining $(\nu_s = 0)$

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$$\frac{\mathcal{E}_{l,s}^{\prime}/L_{l,s}^{\prime}}{\mathcal{E}_{l,s}/L_{l,s}} = \frac{\mathcal{E}_{l,l,s}/L_{l,l,s}}{\mathcal{E}_{l,s}/L_{l,s}}$$

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• If $a_{i,s} = a_{j,s}$ for all i, j, turning off MP has no effect on sector-level emissions

- If clean country has MP in dirty country, turning off MP increases sector-level emissions
 - Clean country: emissions unchanged
 - Dirty country: resources move from clean foreign to dirty domestic firms
Model: Equilibrium

$$X_{l,s} = \mu_{l,s} X_l + \varsigma_s X_{l,s} + \sum_{i,k} \alpha_{i,l,sk} Y_{i,l,k}$$

Demand for s in l

$$Y_{i,l,s} = \sum_{n} \lambda_{i,ln,s} X_{n,s}$$

$$\sum_{i} Y_{i,l,s} = \sum_{i,n} \lambda_{i,ln,s} X_{n,s}$$

$$w_l L_l = \sum_{i,s} \alpha_{i,l,\ell s} Y_{i,l,s}$$

$$X_i = w_i L_i + \sum_{l,s} \prod_{i,l,s} + \Delta_i$$

Market clearing for i, I and $s \notin \mathcal{K}^M$

Market clearing for I and $s \in \mathcal{K}^M$

Labor market clearing in I

Final expenditure in *i*

- Parameters details
 - $\varepsilon = \gamma = 0.45$ IV using US Census micro-data for energy prices, quantities

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 - \Rightarrow Given $\varepsilon, \tilde{\varepsilon}$, MP and energy shares $\{\lambda_{i,l,s}, \alpha_{l,ks}\}$, and world-wide technology choice

Carbon Accounting with Multinational Production





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Carbon Accounting: Allocating Emissions

$$\mathcal{E}_{l}^{P} = \sum_{hi,jn,ks} \mathcal{E}_{hi,jln,ks} \quad \text{Production}$$

$$\mathcal{E}_{n}^{C} = \sum_{hi,jl,ks} \mathcal{E}_{hi,jln,ks} \quad \text{Consumption}$$

$$\mathcal{E}_{j}^{M} = \sum_{hi,ln,ks} \mathcal{E}_{hi,jln,ks} \quad \text{Mining}$$

$$\mathcal{E}_{i}^{O,P} = \sum_{h,jln,ks} \mathcal{E}_{hi,jln,ks} \quad \text{Ownership-Production}$$

$$\mathcal{E}_{h}^{O,M} = \sum_{i,jln,ks} \mathcal{E}_{hi,jln,ks} \quad \text{Ownership-Mining}$$

h, i = Country of ownership for inputs, outputs; j, l = Country of production for inputs, outputs n = Country of consumption; k, s = Industry for inputs, outputs

Carbon Accounting, By Type and Country



Carbon Accounting: The Role of Technology Choice

Ownership-based emissions in foreign countries



Final Remarks

• Multinational production and the environment

• Important, distinct issues from trade

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• What we offer so far

- Comprehensive global data on MNEs and the environment
- Flexible GE model of trade, MP, and energy to study climate change issues
- New estimates on key demand & supply energy elasticities

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What's next

- Optimal carbon taxes with MP
- Leakage through MP
- · Responsible sourcing and supply-chain externalities

Appendix

Emissions per \$Output Very Different Across Countries



Affiliates from Cleaner Countries Are Cleaner Everywhere

Firm f, home country i, host country I, industry s. \mathcal{E} Emissions. Y Revenue Back

$$\log\left(\frac{\mathcal{E}_{i,s}}{Y_{i,s}}\right)^{WOD} = \gamma_i + \delta_s + \varepsilon_{i,s} \qquad \text{vs.} \qquad \log\left(\frac{\mathcal{E}_{fi,l,s}}{Y_{fi,l,s}}\right)^{CDP} = \beta_i + \delta_{l,s} + \varepsilon_{fi,l,s}$$

$$\int_{0^{-}}^{5^{-}} + ND + CHN + CHN$$

Importance of Trade v. Multinational Production



Brucal, Javorcik, and Love (JIE 2019)



Parameters: Estimation Back

- 1. Energy-Type Substitution $\gamma \approx 0.45$: Energy quantities, prices, across states within firm
 - Data: US Mfg Energy Consumption Survey 2014; State Energy Database System

$$\ln\left(\frac{Q_{f,l,k}}{Q_{f,l,1}}\right) = -\gamma \ln\left(\frac{P_{l,k}}{P_{l,1}}\right) + \phi_{f,k} + \xi_{f,l,k}$$

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- 1. Energy-Type Substitution $\gamma \approx 0.45$: Energy quantities, prices, across states within firm
 - Data: US Mfg Energy Consumption Survey 2014; State Energy Database System

$$\ln\left(\frac{Q_{f,l,k}}{Q_{f,l,1}}\right) = -\gamma \ln\left(\frac{P_{l,k}}{P_{l,1}}\right) + \phi_{f,k} + \xi_{f,l,k}$$

- 2. Energy/Non-Energy Substitution $\varepsilon \approx 0.45$: Energy exp, prices across states within firm
 - Data: US State Energy Database System; US Census of Manufactures 2012

$$\ln\left(\frac{\alpha_{f,l}}{1-\alpha_{f,l}}\right) = (1-\varepsilon)\ln\left(\frac{P_{l,1}}{P_l^{NE}}\left(\frac{\alpha_{f,l,1}}{\alpha_{f,l}}\right)^{-\frac{1}{1-\gamma}}\right) + \phi_f + \xi_{f,l}$$

Parameters: Estimation Back

- 1. Energy-Type Substitution $\gamma \approx 0.45$: Energy quantities, prices, across states within firm
 - Data: US Mfg Energy Consumption Survey 2014; State Energy Database System

$$\ln\left(\frac{Q_{f,l,k}}{Q_{f,l,1}}\right) = -\gamma \ln\left(\frac{P_{l,k}}{P_{l,1}}\right) + \phi_{f,k} + \xi_{f,l,k}$$

- 2. Energy/Non-Energy Substitution $\varepsilon \approx 0.45$: Energy exp, prices across states within firm
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- 3. Decreasing returns in mining v = 0.25: Extraction costs, quantities by energy & region
 - Data: Welsby et al. (Nature 2021)

Vertex v, energy type k, region j:
$$v_k = \frac{\partial \ln p_k / \partial \ln E_k}{\partial \ln p_k / \partial \ln E_k + 1} \Rightarrow \ln p_{vj,k} = \zeta_k \ln E_{vj,k} + \mu_{j,k} + \zeta_{vj,k}$$

Parameters: Energy Type Substitution γ

• Extended to firms & states, model implies

$$\ln\left(\frac{Q_{f,l,k}}{Q_{f,l,1}}\right) = -\gamma \ln\left(\frac{P_{l,k}}{P_{l,1}}\right) + \phi_{f,k} + \xi_{f,l,k}$$

- Energy quantities $Q_{f,l,k}$: Manufacturing Energy Consumption Survey 2014
- Energy prices P_{1,k}: State Energy Database System
- Firm×energy type fixed effects $\phi_{f,k}$
- Electricity as reference energy type (k = 1)
- Notes
 - · Arbitrary autocorrelation (two-way cluster) within state and firm
 - · Excluded observations: administrative records, imputed values, zero electricity
 - Basic observation is firm×state (aggregate across establishments w/in state)
- Baseline estimate $\gamma \approx 0.45$

Parameters: Energy Type Substitution γ

$$\ln\left(\frac{Q_{f,l,k}}{Q_{f,l,1}}\right) = -\gamma \ln\left(\frac{P_{l,k}}{P_{l,1}}\right) + \phi_{f,k} + \xi_{f,l,k}$$

	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Price ratio term (γ)	0.409**	0.364**	0.400**	0.263**	0.401**	0.293**	0.415*	0.456**
	(0.159)	(0.177)	(0.155)	(0.105)	(0.162)	(0.124)	(0.245)	(0.184)
Plant level		х	Y					
Asinh			~	х		х		
Exclude coal				~	х	X		
Weighted							Х	
Instrument								х
N First stage F	4,600	7,000	4,600	9,000	4,400	6,000	4,600	4,600 651

Parameters: Energy Type Substitution γ

- Model-based analysis uses $\gamma = 0.45$
- Existing estimates?
 - Vermetten and Plantinga (1953) cross-section of US states: $\gamma pprox 2.1$ to 2.4
 - Serletis et al. (2010) translog with US time series: $\gamma = 0.25$ to 0.60
 - Cross-industry mean: 0.40
 - Standard value for CGE models (EPA, MIT EPPA model)
 - But time series confounding: inflation, growth, OPEC crisis, etc.

Parameters: Energy/Non-Energy Substitution ε

• Extended to firms, I = US state, our model implies

$$\ln\left(\frac{\alpha_{f,l}}{1-\alpha_{f,l}}\right) = (1-\varepsilon)\ln\left(\frac{P_{l,1}}{P_l^{NE}}\left(\frac{\alpha_{f,l,1}}{\alpha_{f,l}}\right)^{-\frac{1}{1-\gamma}}\right) + \phi_f + \xi_{f,l}$$

- Census of Manufactures 2012 administrative/confidential micro-data
- \$\alpha_{f,l}, \alpha_{f,l,1}\$ Energy-cost shares. Establishment-level spending on electricity, fuels, materials, value added
- *P*_{1,1} Price of energy type 1 (electricity). State Energy Data System (US Energy Information Agency)
- P_{l}^{NE} Price of non-energy. We use w_{l} for now
 - Microdata from 2012 Current Population Survey-ASEC
 - Mincer regression with state fixed effects
 - w_l^L are state fixed effects evaluated at reference category
- γ: from earlier estimates
- Baseline estimate $\varepsilon \approx 0.45$

Parameters: Energy/Non-Energy Substitution ε

$$\ln\left(\frac{\alpha_{f,l}}{1-\alpha_{f,l}}\right) = (1-\varepsilon)\ln\left(\frac{P_{l,1}}{P_l^{NE}}\left(\frac{\alpha_{f,l,1}}{\alpha_{f,l}}\right)^{-\frac{1}{1-\gamma}}\right) + \phi_f + \xi_{f,l}$$

	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Price ratio term	0.513*** (0.006)	0.404*** (0.007)	0.510*** (0.006)	0.791*** (0.047)	0.506*** (0.006)	0.421*** (0.007)	0.529*** (0.011)	0.526*** (0.007)
Bootstrap S.E.	(0.129)	(0.160)	(0.125)	(0.081)	(0.129)	(0.096)	(0.192)	(0.129)
Plant level Industry FE Asinh Exclude coal Weighted Instrument		Х	х	Х	х	X X	x	x
N First stage F	12,500	22,500	12,500	12,500	12,500	12,500	12,500	7,100 3121



Parameters: Energy/Non-Energy Substitution ε

$$\ln\left(\frac{\alpha_{f,l}}{1-\alpha_{f,l}}\right) = (1-\varepsilon)\ln\left(\frac{P_{l,1}}{P_l^{NE}}\left(\frac{\alpha_{f,l,1}}{\alpha_{f,l}}\right)^{-\frac{1}{1-\gamma}}\right) + \phi_f + \xi_{f,l}$$

• $\alpha_{f,l}$ on left and right-hand side: simultaneity bias if measurement error

- Solution: instrument $\alpha_{f,I}$ with lag from 2011 Annual Survey of Manufacturers
- γ is a generated regressor
 - Solution: bootstrap over 200 estimates of γ
- Other variations:
 - Firm v. establishment
 - · Zero values for energy share: inverse hyperbolic sine
 - · Coal often missing, some estimates exclude

Parameters: Decreasing Returns in Mining (v)

• Decreasing returns v_k in terms of inverse supply elasticity

 $\upsilon_k = \frac{\partial \ln p_k / \partial \ln E_k}{\partial \ln p_k / \partial \ln E_k + 1}$

• Regression version: Vertex v, energy type k, region j

 $\ln p_{vj,k} = \zeta_k \ln E_{vj,k} + \mu_{j,k} + \zeta_{vj,k}$

• Data source: Welsby et al. (Nature 2021)

• Decreasing returns (=resource cost share, rents): v = 0.25

• Source: pooled inverse elasticity $\zeta = 0.342 \ (0.025)$

Parameters: Decreasing Returns in Mining (v Raw Data)

 $\ln p_{vj,k}^{E} = \zeta_k \ln E_{vj,k} + \mu_{j,k} + \zeta_{vj,k}$



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Parameters: Decreasing Returns in Mining (v Raw Data)

Oil



Parameters: Decreasing Returns in Mining (v Raw Data)

Coal



Parameters: Decreasing Returns in Mining (v Raw Data)

Gas



Model: Revenue Energy Cost Shares Back

$$\alpha_{i,l,ks} = \frac{\left(\tilde{\rho}_{l,ks}/x_{i,ks}\right)^{1-\gamma}}{\sum_{k'\in\mathcal{K}^{\mathcal{E}}} \left(\tilde{\rho}_{l,k's}/x_{i,k's}\right)^{1-\gamma}} \frac{\left(\sum_{k'\in\mathcal{K}^{\mathcal{E}}} \left(\tilde{\rho}_{l,k's}/x_{i,k's}\right)^{1-\gamma}\right)^{\frac{1-\varepsilon}{1-\gamma}}}{\left(\sum_{k'\in\mathcal{K}^{\mathcal{E}}} \left(\tilde{\rho}_{l,k's}/a_{i,k's}\right)^{1-\gamma}\right)^{\frac{1-\varepsilon}{1-\gamma}} + 1}$$

where $x_{i,ks} \equiv a_{i,ks}/a_{i,s}$ and $\tilde{p}_{l,ks} \equiv \delta_{l,ks}^{\frac{1}{1-\gamma}} (p_{l,k}/w_l)$

Technology Choice: Illustration



Model: Optimal Technology Choice, Illustration



Model: Recovering Trilateral Expenditure Flows (Back)

$$X_{ln} = \sum_{i} X_{i,ln} \qquad Y_{i,l} = \sum_{n} X_{i,ln}$$

$$\boldsymbol{X}_{i,ln} = \frac{\phi_{i,l}\phi_{ln}}{\sum_{l'}\phi_{i,l'}\phi_{l'n}} \frac{\left(\sum_{l'}\phi_{i,l'}\phi_{l'n}\right)^{1-\rho}}{\sum_{i'}\left(\sum_{l'}\phi_{i',l'}\phi_{l'n}\right)^{1-\rho}} \boldsymbol{X}_{n}$$

$$\phi_{i,l} \equiv \left(M_i T_{i,l} (\tau_{i,l} c_{i,l})^{-\theta} \right)^{1-\rho} \qquad \phi_{ln} \equiv \left(\tau_{ln} \right)^{-\frac{\theta}{1-\rho}}$$

Model: Recovering Energy Cost Shares, Illustration

• Two inputs (energy, labor), one sector. Equilibrium:

$$x_{i}^{1-\tilde{\varepsilon}} = \frac{\sum_{l} \alpha_{i,l} \mathbf{Y}_{i,l}}{\sum_{l} (1-\alpha_{i,l}) \mathbf{Y}_{i,l}} \qquad \forall i$$

$$\alpha_{i,l} = \frac{1}{\tilde{\sigma}_s} \frac{\left(\tilde{p}_l/x_i\right)^{1-\varepsilon}}{\left(\tilde{p}_l/x_i\right)^{1-\varepsilon}+1} \qquad \forall i,l$$

$$\alpha_{l} = \sum_{i} \alpha_{i,l} \frac{\mathbf{Y}_{i,l}}{\sum_{i'} \mathbf{Y}_{i',l}} \qquad \forall l$$

where $x_i \equiv a_i^E / a_i^L$ and $\tilde{p}_l \equiv \delta_l^{\frac{1}{1-\varepsilon}} (p_l / w_l)$

• System of equations to solve for $\{x_i\}$, $\{\alpha_{i,l}\}$ and $\{\tilde{p}_l\}$ given data, $\tilde{\varepsilon}$, ε

Carbon Accounting with Multinational Production

$$\mathcal{E}_{hi,jln,ks} = \frac{e_{l,k}}{p_{l,k}} \chi_{hi,jl,ks} X^{f}_{i,ln,s}$$

• Emission rate (tons/\$):
$$\frac{e_{l,k}}{p_{l,k}} = \frac{e_k^{IEA}Q_{l,k}^{IEA}}{Y_{l,k}^{WIOD}}$$

- Leontief inverse: $\{\chi_{hi,jl,ks}\} = (I \{\alpha_{hi,jl,ks}\})^{-1}$ where $\alpha_{hi,jl,ks} \equiv \lambda_{h,jl,k}^{model} \alpha_{i,l,ks}^{model}$
- Final sales: $X_{i,ln,s}^f = \lambda_{i,ln,s}^{model} X_{n,s}^{f,WIOD}$