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international trade on
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Michael Hübler, Eduard Bukin and Yuting Xi

ABSTRACT

THE EFFECTS OF INTERNATIONAL TRADE ON STRUCTURAL CHANGE AND CO₂ EMISSIONS

Michael Hübler, Eduard Bukin and Yuting Xi

This article introduces a new econometric model that includes an innovative measure of intersectoral structural change. This model describes the structural convergence (or divergence) of sector share patterns across countries (from the North-South or global perspective) influenced by international trade. The econometric analysis applies panel data estimators with different types of fixed effects to the 2013 and 2016 releases of the World Input-Output Database (WIOD), covering the periods 1995–2009 and 2000–2014. The results show that international trade promotes structural convergence, which is enhanced by sectoral capital intensities. It seems, however, that in this millennium, structural divergence has been fostered by trade-induced specialization in CO₂-intensive production.

Keywords: structural change, international trade, CO₂ emissions, macro-econometrics, panel data, WIOD

JEL classification: C51, F14, F18, O11, O44

Michael Hübler
Justus Liebig University Giessen
Institute for Agricultural Policy and
Market Research, Center for
International Development and
Environmental Research (ZEU)
Senckenbergstr. 3
D-35390 Gießen, Germany
Leibniz University Hannover,
Germany
Email:
michael.huebler@agrار.uni-giessen.de
www.uni-giessen.de

Eduard Bukin
Justus Liebig University Giessen
Institute for Agricultural Policy and
Market Research
Senckenbergstr. 3
D-35390 Gießen, Germany
Email:
eduard.bukin@agrار.uni-giessen.de
www.uni-giessen.de

Yuting Xi
Leibniz University Hannover
Königsworther Platz 1
D-30167 Hannover, Germany
East China University of Science
and Technology, Shanghai, China
Email:
mf20080516@126.com
www.uni-hannover.de

1 Introduction

Economic growth, particularly that of large emerging economies such as China or India, increases not only output and income but also CO₂ emissions. Because of the resulting contribution to global warming, this effect shall be mitigated. Structural change, i.e., the shift in the economy's sectoral composition, can increase or decrease CO₂ emissions. Whereas the effects of international trade on productivity gains, economic growth and international technology diffusion (including energy- and CO₂-saving technologies) have been extensively researched,¹ research on intersectoral structural change and the resulting impact on CO₂ emissions remains scarce. Better knowledge of the latter effects would, however, be helpful for anticipating the international implications of national (climate, energy, trade and other) policies: can policy-induced sectoral shifts be expected to spill over to trading partners within a sufficiently large time horizon?

Empirically, intersectoral structural change has significantly contributed to changes in energy use and CO₂ emissions.² Econometric evidence of the effects of *international trade* on structural change, in general, and with respect to CO₂ emissions, in particular, however, seems to be widely missing. Some studies have examined the role of foreign direct investment (FDI) and European market integration in the convergence of European countries: in a working paper, Barrios et al. (2002) find convergence of per capita income and industry sector structure in the European Union supported by inward FDI; based on sectoral indices and descriptive statistics of exports, structural similarity and structural change, Crespo & Fontoura (2007) find similar results.

Against this background, our article tries to fill this research gap by exploring the *nexus between international trade and structural change* from a conceptual and econometric point of view. It contributes to the extant literature by conceptually describing the economic mechanisms of structural change driven by trade, by introducing a new econometric model with a new measure of structural change with (North-South) con-/divergence, by studying sectoral CO₂ emissions (in addition to sectoral outputs) and by exploiting the newest version of the large bilateral, bisectoral dataset of the World Input-Output Database (WIOD)³.

According to the Environmental Kuznets Curve (EKC) hypothesis, during the course of economic development, the sectoral structure of an economy shifts from agriculture

¹See, e.g., Coe et al. (1997); Saggi (2002); Keller (2004); Cole (2006); Perkins & Neumayer (2009); Havranek & Irsova (2011); Hübler & Glas (2014).

²See, e.g., Schäfer (2005); IEA (2007); Kahrl & Roland-Holst (2009); Li et al. (2014); Voigt et al (2014).

³<http://www.wiod.org/home>.

toward (heavy) industries, leading to higher emissions. It then shifts further toward advanced knowledge-based industries and services, leading to lower emissions. Consequently, Organisation for Economic Co-operation and Development (OECD) countries are expected to exhibit sectoral structures that create *ceteris paribus* lower economy-wide CO₂ emissions than those of emerging countries at medium stages of development. To examine this economic transition process, we first deploy a North-South setup with OECD and emerging countries and then extend it to all countries covered by the WIOD.

Our article particularly addresses the open question of whether exports from an (advanced) economy to another (emerging) economy make these two economies more similar or more different with regard to their sectoral structures and related CO₂ emissions. We denote these two alternatives as *structural convergence* and *structural divergence*. Assuming that both economies continue their economic growth process and that emerging economies catch up with industrialized countries, convergence decreases average CO₂ emissions across sectors, whereas divergence increases them.

Theoretically, the role of international trade in structural change is ambiguous. On the one hand, according to classic trade theories by Ricardo and Heckscher-Ohlin, different economies concentrate their production and exports on different sectors, resulting in *structural divergence*. Additionally, Krugman's New Economic Geography predicts the agglomeration of economic activities, which supports the emergence of specialization and clustering (Midelfart et al., 2003). Induced and directed technological change (Acemoglu, 2002, 2010) may reinforce sectoral heterogeneity across countries.

On the other hand, when knowledge or, more specifically, (energy- and CO₂-saving) technologies, spread across borders, supported by international trade, the use of similar technologies will result in similar productivities across sectors and similar sectoral structures within economies, resulting in *structural convergence*. Similarly, intensifying *interindustry* trade supports the emergence of similar sectoral structures across trading partners (Midelfart et al., 2003). Eventually, in the theoretical long-term equilibrium of a fully integrated world economy, the sectoral structures will be equalized.

The overview by Herrendorf et al. (2014) (Section 6.6.1 International Trade) reconciles the views of these two camps by arguing that in a country with high productivity growth, the development of the manufacturing sector share exhibits a hump shape, while in a country with low productivity growth, it exhibits a downward-sloping shape (Yi & Zhang, 2010). In accordance with the outcome of the Environmental Kuznets Curve theory, this theory implies that structural con- and divergence are theoretically possible

across economies depending on (sectoral) productivity growth and the phase of economic development.

Therefore, whether structural con- or divergence dominates, in general, across economies and sectors or, in particular, sectors at specific periods of time is an empirical question that needs to be answered. Compared with the existing literature, our analysis is, to the best of our knowledge, the first study showing econometrically that international trade, in general, promotes *structural convergence*. This result opposes classical Ricardian trade theory and supports the modern view of international trade as a driver of technology diffusion. We also find indications for structural divergence in terms of specialization in more or less CO₂-intensive production that began in this millennium. This insight is important for policy makers because it points to the outsourcing of CO₂-intensive production from industrialized to emerging economies, or so-called carbon leakage, which undermines climate policy efforts.

The article proceeds as follows. Section 2 derives the conceptual framework, Section 3 describes the data, and Section 4 presents the econometric results. Section 5 discusses the results, and Section 6 concludes the article. The supplementary online appendix provides further statistics and robustness check results.

2 Concept

In this section, we develop the econometric model of structural change driven by international trade and further determinants, first for one economy and then for two economies connected via trade. In the next step, we will discuss the theoretical effects of international trade on structural change and present two alternative testable hypotheses for the effect of international trade on structural change.

Sectoral one-economy model:

In economy e and sector c at time t , let sector size be denoted by Z_{ect} . Z_{ect} can be measured as the sectoral (gross) output value Y_{ect} , as physically measured CO₂ emissions E_{ect} or, alternatively, as physically measured (gross) energy use (or, in general, other suitable indicators). Given that the CO₂ intensity of production, measured as CO₂ emissions per output value, differs across sectors, the output and CO₂ shares will differ too. When the sectoral CO₂ intensities change due to efficiency gains, the corresponding output- and CO₂-based sector shares will change to different extents. Whereas CO₂ captures the emissions released within the sector, energy inputs lead to emissions in previous production

stages that are attributed to the corresponding sector in which they occur. Consequently, these two indicators differ in general.

Against this background, the sector share in the entire economy can be defined as follows:

$$\frac{Z_{ect}}{\sum_c Z_{ect}} = f \left(\frac{K_{ec(t-1)}}{L_{ec(t-1)}}, \frac{M_{ec(t-1)}}{Y_{ec(t-1)}}, \sigma_{ec}, \theta_t, \varepsilon_{ect} \right) \quad (1)$$

with $\sum_c Z_{ect}$ indicating the sum of the sizes of all sectors c in economy e at time t and $f(\dots)$ representing a function of the arguments explained below. Expecting a time lag between changes in the determinants on the right-hand side and their effect on the left-hand side, for the time being, let us assume one-period time lags denoted by $(t - 1)$.

$K_{ec(t-1)}$ is the value of the sectoral capital stock, and $L_{ec(t-1)}$ is the physically measured⁴ sectoral labor input. The capital-to-labor ratio indicates capital intensity and, indirectly, the technology intensity of production (given that capital embodies technologies). It is ex ante unclear whether a higher capital intensity is associated with a smaller or larger sector share because capital/technology and labor can be either substitutes or complements.

$M_{ec(t-1)}$ denotes the total value of intermediate goods imports from the rest of the world. Import intensity $\frac{M_{ec(t-1)}}{Y_{ec(t-1)}}$ indicates the strength of international (trade) connections. Again, it is ex ante unclear whether a higher import intensity is associated with a smaller or larger sector share. This relation is in the spotlight of this analysis and will be explained in detail in the following paragraphs.

σ_{ec} captures any other time-invariant economy- and sector-specific determinants within the cross-section, for example, sectoral productivity growth or the economy-wide infrastructure. θ_t captures any time-variant determinants that jointly affect all economies and sectors in the same way in each time period t , for example, an oil price shock or a pandemic shock. ε_{ect} captures any remaining unexplained random influences (noise).

Sectoral two-economy model:

Now, let us focus on the dyadic trade connections between specific sectors in specific countries and compare the two trading partners in terms of their sectoral structures and the determinants of these connections. For this purpose, let us label source countries of trade as s , source sectors as i , recipient countries of trade as r and recipient sectors as

⁴For example, it is measured as the number of persons working in a sector.

j . From this viewpoint and with this notation, our previous model is generalized to the following:

$$dz_{srjt} = f \left(dk_{srj(t-1)}, m_{srj(t-1)}, \sigma_{srj}, \theta_t, \varepsilon_{srjt} \right) \quad (2)$$

with $dz_{srjt} = \left| \left(\frac{Z_{rjt}}{\sum_j Z_{rjt}} - \frac{Z_{sjt}}{\sum_j Z_{sjt}} \right) / \frac{Z_{sjt}}{\sum_j Z_{sjt}} \right| = \left| \frac{Z_{rjt}}{\sum_j Z_{rjt}} / \frac{Z_{sjt}}{\sum_j Z_{sjt}} - 1 \right|$ representing the relative distance (the absolute normalized difference with a positive sign) between the (output, CO₂ or energy) shares of the same sector j in two countries s and r connected via trade, where $i = j$ is suppressed for simplicity. Although this expression implicitly captures the exporter and importer sides, it eventually reflects only their relative distance. The division by $\frac{Z_{sjt}}{\sum_j Z_{sjt}}$ renders dz_{srjt} independent of sector size; i.e., small and large sectors are weighted equally.

Similarly, $dk_{srj(t-1)} = \left| \left(\frac{K_{rj(t-1)}}{L_{rj(t-1)}} - \frac{K_{sj(t-1)}}{L_{sj(t-1)}} \right) / \frac{K_{sj(t-1)}}{L_{sj(t-1)}} \right| = \left| \frac{K_{rj(t-1)}}{L_{rj(t-1)}} / \frac{K_{sj(t-1)}}{L_{sj(t-1)}} - 1 \right|$ represents the relative distance between capital intensities in the sector j (with $i = j$) of the two trading partners. As a result, the expressions on the left and right sides match.

International trade is generalized as a bilateral, bisectoral $sirj$ relation. To keep the model tractable, we sum up the trade flows over source sectors i to obtain a bilateral trade relation with the intermediate goods imports of sector j in r from all sectors of s . Hence, $m_{srj(t-1)} = \frac{\sum_i M_{sirj(t-1)}}{Y_{rj(t-1)}}$ is the modified central trade measure under scrutiny.

Covering this extended dimensionality, the index of the time-invariant determinants is rewritten as σ_{srj} . Whereas the time-variant effects are denoted by θ_t as before, the error term now reads ε_{srjt} , which completes the model setup.

Testable hypotheses on trade and structural change:

There are basically two opposite possible approaches to the explanation of structural change and structural con-/divergence in the context of international trade.

The first approach refers to the classic trade theories of Ricardo and Heckscher-Ohlin. Countries specialize in the goods and, hence sectors for which they have a productivity-based comparative advantage or for which they have relatively abundant endowments with the required production factors. In this framework, sector shares and trade intensities may reflect sector-specific productivities (Eaton & Kortum, 2002). Induced factor- and sector-specific directed technological change (Acemoglu, 2002, 2010) may reinforce the heterogeneity of economic production depending on country-specific factor

endowments, policies affecting sectors in different ways and other economic conditions.⁵ New Economic Geography, popularized by Krugman, describes the agglomeration of economic activities. Local knowledge spillovers and linkages with customers and suppliers support the emergence of local specialization and clustering (Midelfart et al., 2003; Crespo & Fontoura, 2007). It follows from this theory that over time, countries shift their production toward different sectors. This implies, for structural change, that countries' sectoral structures diverge, i.e., become more different over time. Intermediate goods imports (M) and capital (K) accumulation are expected to enhance this effect. In this context, capital reflects technologies, knowledge and absorptive capacity (with respect to the adoption of technologies and knowledge). In terms of the previously defined model (Equation 2), the resulting hypothesis reads as follows:

$$H1: \text{International trade fosters structural divergence, i.e., } \frac{\partial(dz_{srjt})}{\partial(m_{srj(t-1)})} > 0.$$

H1 implies that sectoral distances become larger. In a Ricardian world with full specialization, in each economy, the shares of some or all but one sector will become zero, i.e., $\frac{Z_{ect}}{\sum_c Z_{ect}} = 0$, and thus, sectoral differences will diverge to the share of the sector of specialization, i.e., $dz_{srjt} = \max\{\frac{Z_{sjt}}{\sum_i Z_{sjt}}, \frac{Z_{srjt}}{\sum_j Z_{rjt}}\}$, where, in practice, both sector shares may become zero (no specialization in this sector among these two particular countries), and hence, $dz_{srjt} = 0$.

The second approach refers to international technology (knowledge) diffusion in the course of globalization with international trade and economic development. Accordingly, over time, countries' sectoral technologies and, hence, productivities converge, i.e., become more similar. Similarly, on the consumption side, the international spread of knowledge, culture, habits, tastes and preferences can be enhanced by international trade linkages, which will increase the similarity of consumers residing in different countries and, via changes in consumption patterns, increase the similarity of sectoral production structures. Additionally, interindustry trade supports the emergence of similar sectoral structures across trading partners because it allows for the exchange of different goods produced in different sectors via trade in terms of varieties of the same good within the same sector (Midelfart et al., 2003). This implies, for structural change, that countries' sectoral structures have a tendency to converge, i.e., become more similar over time.

⁵Whereas technological change normally increases sectoral productivity and, hence, sectoral output, it can increase or decrease sectoral (factor) inputs (of labor, capital, energy or CO₂ caused by fossil fuel inputs) depending on whether technological change is factor-augmenting or factor-saving.

Again, intermediate goods imports (M) and capital (K) accumulation are expected to enhance this effect. Accordingly, the resulting hypothesis reads as follows:

$$H2: \text{International trade fosters structural convergence, i.e., } \frac{\partial(dz_{srjt})}{\partial(m_{srj(t-1)})} < 0.$$

H2 implies that the sectoral distances become smaller. In the theoretical long-term equilibrium of a fully integrated world economy, $\frac{Z_{sjt}}{\sum_j Z_{sjt}} = \frac{Z_{rjt}}{\sum_j Z_{rjt}} \forall (s, r, j)$, and thus, $dz_{srjt} = 0$.

3 Data

In this section, we describe the data source and aggregation of the panel data in terms of countries and sectors.

Data source and setup:

In addition to using the newest 2016 release of the large dataset of the World Input-Output Database (WIOD)⁶, we deploy the 2013 release for comparison.⁷ We combine the World Input-Output Tables (WIOT) containing bilateral, bisectoral⁸ trade (in mill. 2010-US-\$, see below) data with socioeconomic accounts containing sectoral (gross) outputs (in mill. 2010-US-\$), labor (in 1,000 employment units) and capital (in mill. 2010-US-\$) data and with environmental accounts (the 2019 extension⁹ of the WIOD 2016) providing sectoral CO₂ emissions (in 1,000 tonnes) and sectoral energy use (in terrajoules) data. Following the model setup of the previous section, we sum up all intermediate good imports entering each sector across their sectors of origin while maintaining source-destination country dyads.

In the WIOD 2016, monetary values are expressed as 2010-US-\$, i.e., measured in constant prices of the base year 2010; similarly, in the WIOD 2013, monetary values are expressed as 1995-US-\$. They are created by applying the corresponding deflator¹⁰ and, in the case of output and capital, by converting the national currency values to US-\$ using the corresponding exchange rate contained in the WIOD. CO₂ emissions refer to direct

⁶<http://www.wiod.org/home>, Timmer et al. (2015, 2016).

⁷The WIOD 2013 and 2016 do not exactly match in terms of sectoral definitions; therefore, and to keep the number of observations computationally tractable, we deploy them separately.

⁸This means that international trade flows from any sector in any country to any sector in another country.

⁹Corsatea et al. (2019)

¹⁰We apply the WIOD deflator containing the price levels of intermediate inputs to discount trade values, price levels of (gross) output to deflate (gross) output values and price levels of (gross) value added to deflate capital values.

emissions caused by fossil-fuel-based energy use and process emissions released within the corresponding sector (excluding the indirect emissions embodied in intermediate inputs). This allows us to study the change in production technologies in each sector. An alternative measure, (gross) energy use, refers to the total direct energy input (consumption), including electricity consumption, in each sector.

We restrict the numerical setup to data sourced from the WIOD¹¹ to keep it as consistent as possible in terms of sector definitions and accounting methods and to keep the numerical requirements tractable. In the time dimension, where t denotes years, the 2013 release covers 1995 to 2009; the 2016 release covers 2000 to 2014. In the cross-section, our *North-South* setup includes 31 industrialized countries (OECD, North) in the WIOD 2013 and 34 industrialized countries in the WIOD 2016 versus 9 emerging countries (South) in both samples.¹² Depending on the available sectors in the original data source, we aggregate the sectors into 26 sectors f (equivalently, i or j) in the WIOD 2013 and 36 sectors in the WIOD 2016.¹³ Appendix C provides detailed sector lists and mappings, which results in over 140 thousand observations in the North-South sample and 870 thousand observations in the full sample of the WIOD 2016 and over 95 thousand observations in the North-South sample and 520 thousand observations in the full sample of the WIOD 2013. The *full sample* combines emerging and industrialized countries into 40 economies (countries) in the WIOD 2013 and 43 economies in the WIOD 2016.

In the full sample, each country exports to each other country; i.e., all countries are source s and recipient r at some point. However, in the North-South setup, industrialized (OECD) countries s export to emerging countries r .

¹¹This includes deflators and exchange rates.

¹²*Emerging countries (South) in the WIOD 2013 and 2016:* “BRA” Brazil, “BGR” Bulgaria, “CHN” Mainland China, “MEX” Mexico, “RUS” Russia, “TWN” Taiwan, “ROU” Romania, “IND” India, and “IDN” Indonesia.

Industrialized countries (North) in the WIOD 2013 and 2016: “AUS” Australia, “AUT” Austria, “BEL” Belgium, “CAN” Canada, “CYP” Cyprus, “CZE” Czechia, “DEU” Germany, “DNK” Denmark, “ESP” Spain, “EST” Estonia, “FIN” Finland, “FRA” France, “GBR” Great Britain, “GRC” Greece, “HUN” Hungary, “IRL” Ireland, “ITA” Italy, “JPN” Japan, “KOR” Republic of Korea, “LTU” Lithuania, “LUX” Luxembourg, “LVA” Latvia, “MLT” Malta, “NLD” The Netherlands, “POL” Poland, “PRT” Portugal, “SVK” Slovakia, “SVN” Slovenia, “SWE” Sweden, “TUR” Turkey, and “USA” The United States of America.

Additional industrialized countries (North) in WIOD 2016: “CHE” Switzerland, “HRV” Croatia, and “NOR” Norway.

¹³*Sectors in the WIOD 2013 and 2016:* A01 Agriculture, B Mining, C10-C12 Food, C13-C15 Textile, C16 Wood, C17 Paper, C19 Refined Petr., C20 Chemicals, C22 Rubber, C23 Minerals, C24 Metal, C26 Computers, C27 Electrical equip., C30 Transport equip., C33 Repair, D35 Energy, F Construction, G Trade, H49 Land transport, H50 Water transport, H51 Air transport, H52 Warehousing, H53 Post, I Accommodation, JKLMN Private Services, and OPQRS Public Services.

Additional sectors in the WIOD 2016: A02 Forestry, A03 Fisheries, C18 Printing, C21 Pharma., C25 Non machinery, C28 Machinery, C29 Vehicles, C31-C32 Furniture, E36 Water, and E37-E39 Waste.

Sectors T Household and U Household are discarded in both samples due to the absence of trade.

Descriptive statistics:

Figures 1 and 2 draw on the WIOD 2013 and 2016. They illustrate sectoral developments, computed as averages across countries, with each country group (emerging and industrialized countries or, in short, South and North). Sectoral developments refer to direct CO₂ emissions or (gross) output shares of each sector in total CO₂ emissions or (gross) output of the corresponding country. Each single dot represents one observation, the solid (for the WIOD 2013) or dashed (for the WIOD 2016) lines depict estimates obtained via locally estimated scatterplot smoothing (Cleveland et al., 1992), and the shaded areas indicate 95% confidence intervals. The investigation of descriptive statistics reveals sectoral developments, including South-North con-/divergence, which suggest a detailed econometric exploration of their drivers.

Figures 1 and 2 depict the selected sectors. In the energy sector (electricity, gas, water, steam and air conditioning supply; Figures 1a and 1b), emerging countries (in blue) exhibit larger shares than do industrialized countries (in red) in terms of both average output and CO₂ shares. While at the end of the time frame, output shares reveal a convergence tendency, and CO₂ shares show a divergence tendency. Whereas output shares move around 3%, CO₂ shares exceed 40% in emerging countries, which points to the high CO₂ intensity of the energy sector.

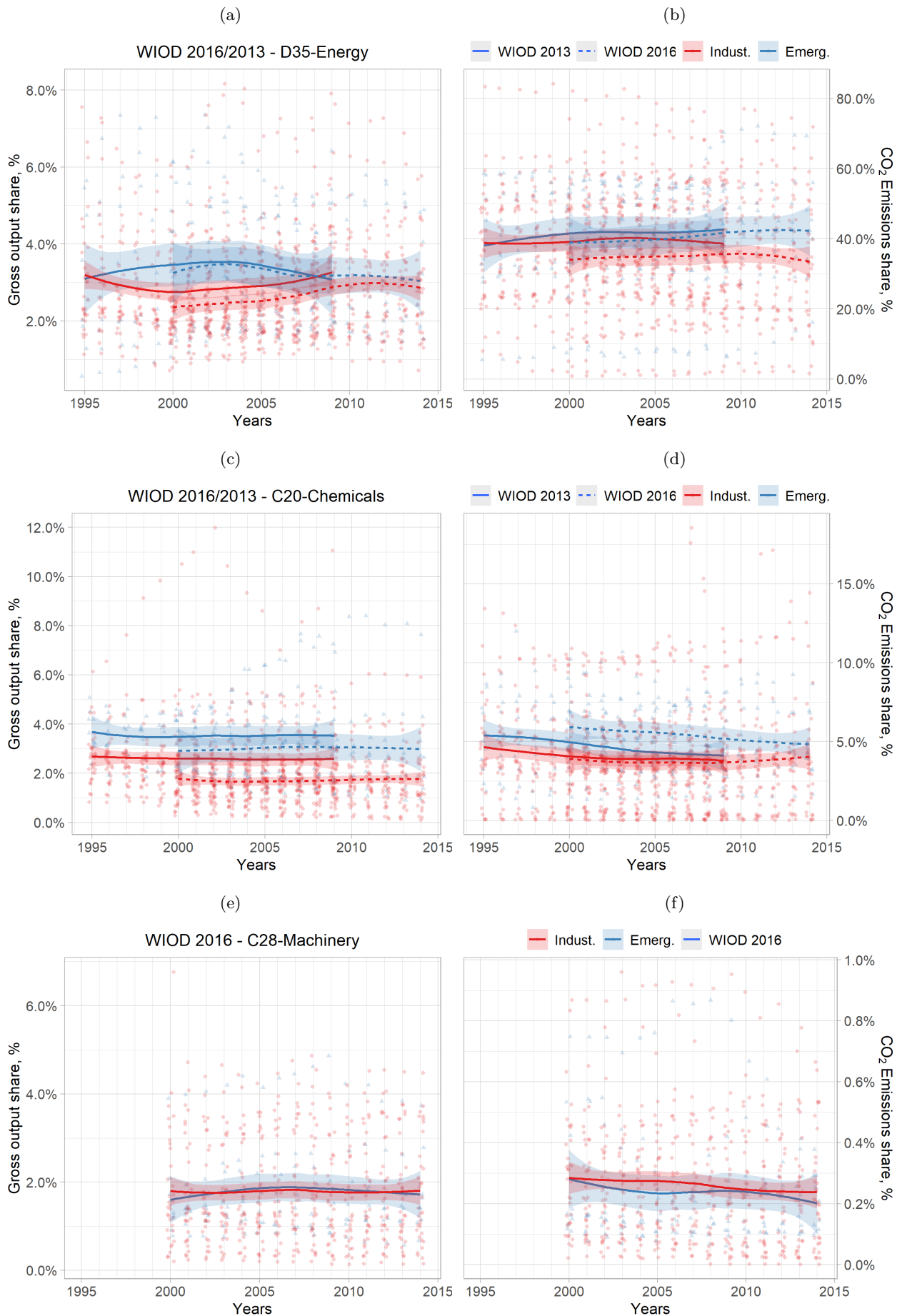
In the energy-intensive chemicals sector (Figures 1c and 1d), sector shares have similar sizes in terms of output, as in the energy sector, and are again larger in the South. In contrast, CO₂ shares are an order of magnitude lower than in the energy sector. The CO₂ shares in the South converge to those of the North. Compared with the energy-intensive chemical sector, the CO₂ share of machinery (Figures 1e and 1f) is another order of magnitude lower (about 0.2%). The output and CO₂ machinery sector shares of the South and North are nearly identical and change little over time.¹⁴

Regarding land transport (Figures 2a and 2b), the South exhibits larger output shares than the North but smaller CO₂ shares (with a slightly increasing trend in both regions), which indicates an advantage for the South with regard to CO₂ emission intensity.

In the construction sector (Figures 2c and 2d), emerging countries overtake industrialized countries in terms of output shares, while the opposite occurs in terms of CO₂ shares. This indicates significant CO₂ emissions reductions in the South, although the Southern CO₂ share exhibits a slightly increasing trend.

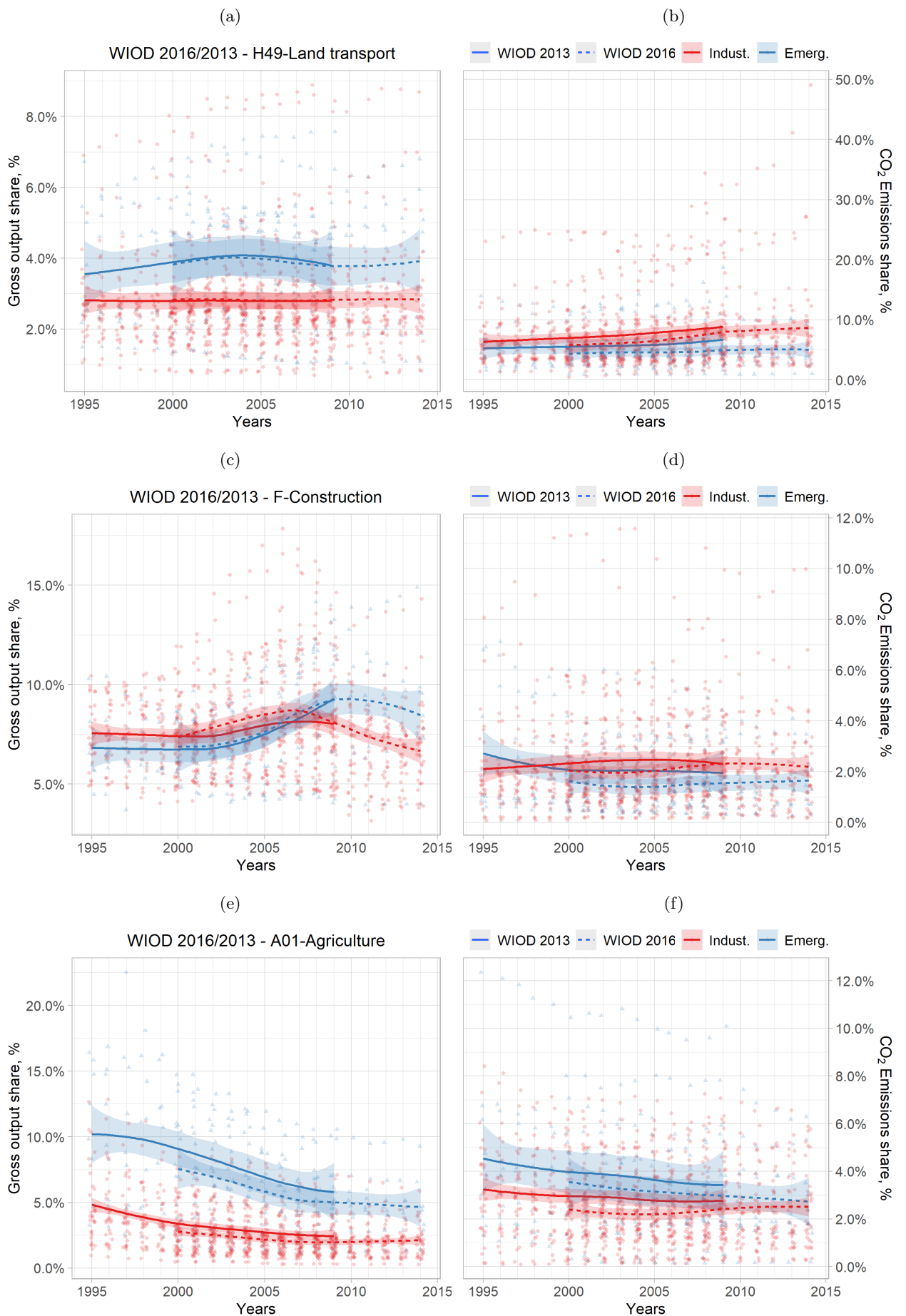
¹⁴The machinery sector is one of the sectors separated from the generic industry in the WIOD 2016 but not in the WIOD 2013.

Figure 1
Average sector shares in industrialized and emerging countries over time



Source: Own illustrations based on data taken from the WIOD 2013 and 2016 releases.

Figure 2
Average sector shares in industrialized and emerging countries over time



Source: Own illustrations based on data taken from the WIOD 2013 and 2016 releases.

In agriculture (Figures 2e and 2f), clearly, the southern output shares exceed those of the North more than do CO₂ shares, which indicates less CO₂ emission-intensive agricultural production in the South than in the North. Mostly southern output shares, but also CO₂ shares, converge to those in the North over time.

The next section will explore possible drivers of these sectoral developments (in the South relative to the North) in an econometric analysis. Appendix A.1 provides summary statistics of the economic indicators in the different data samples as they appear in the econometric analysis. Appendix A.2 uses the full sample to present all available data, showing the correlations among indicators appearing in one regression. Accordingly, all correlations are low, i.e., within ± 0.2 . The figure also illustrates the relation of the correlation partners by scatterplotting each indicator as a function of its partner. Red lines sketch the nonlinear relation between correlation partners by using a nonparametric smoothing algorithm. They indicate moderate relations between the regressors and dependent variables of the econometric model presented below. Histograms depict the distributions (the density of observations covering an area with a value of one) of the indicators. Accordingly, most of the logarithmic observations are located around zero.

4 Econometrics

This section first describes the econometric approach and test procedures for implementing it in an appropriate way and then presents the regression results.

Econometric approach:

To explicitly write out Equation 2, we assume a multiplicative model, take natural logarithms on both sides and rearrange the terms.

$$\ln(dz_{srjt}) = \alpha \cdot \ln(m_{srj(t-1)}) + \beta \cdot \ln(dk_{srj(t-1)}) + \delta \cdot \sigma_{srj} + \eta \cdot \theta_t + \varepsilon_{srjt} \quad (3)$$

with the relative distance between sector shares of the source and the recipient, $dz_{srjt} = \left| \frac{Z_{rjt}}{\sum_j Z_{rjt}} / \frac{Z_{sjt}}{\sum_j Z_{sjt}} - 1 \right|$; the sectoral import intensity, $m_{srj(t-1)} = \frac{\sum_i M_{sirj(t-1)}}{Y_{rj(t-1)}}$, which is in the spotlight in the analysis; and the relative distance between the sectoral capital-to-labor ratios of the source and of the recipient, $dk_{srj(t-1)} = \left| \frac{K_{rj(t-1)}}{L_{rj(t-1)}} / \frac{K_{sj(t-1)}}{L_{sj(t-1)}} - 1 \right|$.

These indicators are all constructed by using the WIOD. While we restrict the explicit inclusion of economic indicators to those with direct economic relevance and that are covered by the WIOD with the required high bilateral and bisectoral resolution, we deploy

a very large number of fixed effects, which exploit the technical (computational) limits of the estimation procedure.

To model triadic fixed effects, binary variables σ_{srj} and θ_t take a value of one for each bilateral sectoral trade relation and each year. Index srj combines source country s , recipient country r and recipient sector j characteristics, while t indicates the individual time dimension. The joint use of σ_{srj} and θ_t leads to a two-way fixed effects model, which in short will be denoted by srj & t . Alternatively, either σ_{srj} or θ_t can be used to implement single fixed effects models with cross-sectional fixed effects (in short, srj) or time fixed effects (in short, t). α , β , δ and η are the parameters to be estimated, while ε_{srjt} is the error term. Because of the log-log-specification, α and β represent elasticities, reflecting the effect of relative changes in import intensity or the capital-to-labor ratio on relative changes in the dependent variable.

If H1 holds, then it follows for Equation (3) that $\alpha > 0$ (and $\beta > 0$), and if H2 holds, then $\alpha < 0$ (and $\beta < 0$).

To examine whether the effect of international trade on structural change is enhanced or dampened by a higher relative capital-to-labor ratio (and the technologies embodied in capital), we add their multiplicative joint effect as follows:

$$\begin{aligned} \ln(dz_{srjt}) = & \alpha \cdot \ln(m_{srj(t-1)}) + \beta \cdot \ln(dk_{srj(t-1)}) + \gamma \cdot \ln(m_{srj(t-1)}) \cdot \ln(dk_{srj(t-1)}) \\ & + \delta \cdot \sigma_{srj} + \eta \cdot \theta_t + \varepsilon_{srjt} \end{aligned} \quad (4)$$

The interaction term $\gamma \cdot \ln(dk_{srj(t-1)}) \cdot \ln(m_{srj(t-1)})$, with parameter γ to be estimated, will be included in the main regressions but excluded from a robustness check.¹⁵

Test procedures:

We carry out the following standard test procedures. We check that the correlations among regressors are sufficiently low (i.e., within ± 0.2 ; see Appendix A.1 and the end of the previous section) to avoid multicollinearity. The standard F -tests for the null hypothesis of all estimated coefficients jointly being zero are reported for each regression (see the last rows in Tables 1 to 4). In the estimations yielding significant results, the F -statistics are, in most cases, (very) high. The regression results in the first column of Tables 1 and 2 exhibit insignificant F -statistics and, in most cases, insignificant t -

¹⁵To identify the overall effect of trade on structural change, we need to consider both the single effect and the joint effect (the marginal effect at a given capital-to-labor ratio) or refer to Equation 3.

statistics, indicated by missing asterisks, for single regressors as well. The R^2 values are low in all regressions, which hinges on the model specification with economic indicators being specified as shares, ratios or intensities, measured in relative and absolute terms.

Additionally, we carry out tests designed for panel data. We apply Fisher-type Augmented Dickey-Fuller unit root tests (Dickey & Fuller, 1979; Im et al., 2003) to ensure that the data are stationary. Consequentially, we test all dependent and independent variables in all datasets (the WIOD 2016 and 2013) and all subsamples (North-South and full sample) separately. We find that the unit root null hypothesis is always clearly rejected in favor of stationarity (excluding a time trend at the 0.00001 confidence level).

The standard Hausman test for fixed versus random effects clearly rejects the null hypothesis of consistent random effects in all specifications; therefore, we restrict our analysis to the use of fixed effects (dummy variables).

We apply F - and LM ¹⁶-tests evaluating different types of fixed effects against the null hypothesis of a pooled regression or reduced dimensionality (i.e., a reduced number) of fixed effects. Appendix A.3 presents the tests results. For all specifications, the F - and LM -tests clearly reject the null hypothesis of all fixed effects jointly being zero, i.e., the pooled regression. Similarly, the F - and LM -tests clearly reject the null hypothesis of fixed effects with reduced dimensionality; i.e., cross-sectional fixed effects plus time fixed effects (two-way fixed effects, srj & t) are preferable over cross-sectional fixed effects only (srj) or time fixed effects only (t).

For the choice between cross-sectional or time fixed effects versus two-way fixed effects, however, Kropko & Kubinec (2020) recommend the choice of a single type of fixed effects to enable a clear-cut interpretation of the estimation results with respect to variant and invariant effects in the time and cross-sectional dimension, instead of generating a mixture of both, which is difficult to interpret. Therefore, we use and compare the three fixed effects specifications (srj , t and srj & t). When using cross-sectional fixed effects, the variation remaining in the data is generated within the time dimension across years. When using time fixed effects, in contrast, the variation remaining in the data is generated in the cross-sectional dimension via differences between countries and sectors, which may be interpreted as a snapshot of the current situation or as long-term (equilibrium) effects. When using two-way fixed effects, both types of variation overlap, similar to a pooled regression (cf. Kropko & Kubinec, 2020).

¹⁶This means Lagrange Multiplier.

Table 1
Panel regression results with output shares using the WIOD 2016

	Dep. var.: relative distance between sectoral output shares $\ln(dz_{srjt})$					
	North-South			Full sample		
	<i>srj</i>	<i>t</i>	<i>srj & t</i>	<i>srj</i>	<i>t</i>	<i>srj & t</i>
Fixed effects						
Import intensity	-0.00919**	-0.08615*****	-0.02947*****	-0.00368*	-0.05888*****	-0.01815*****
$\ln(m_{srj(t-1)})$	(0.00462)	(0.00521)	(0.00484)	(0.00216)	(0.00222)	(0.00229)
Capital-to-labor rat.	-0.02336	-0.03883	-0.01067	-0.02236*****	0.01650	-0.02127*****
$\ln(dk_{srj(t-1)})$	(0.01739)	(0.04457)	(0.01733)	(0.00572)	(0.01103)	(0.00572)
Interaction term						
$\ln(m_{srj(t-1)}) \cdot$	-0.00183	-0.01759*****	-0.00095	-0.00370*****	-0.00800*****	-0.00378*****
$\ln(dk_{srj(t-1)})$	(0.00172)	(0.00478)	(0.00172)	(0.00071)	(0.00140)	(0.00071)
Num. of observat.	143435	143435	143435	871733	871733	871733
Degr. of freedom	133174	143418	133161	809344	871716	809331
R^2	0.00013	0.02835	0.00089	0.00020	0.01853	0.00052
F -stat.	1.901	117.103*****	12.436*****	13.333*****	487.013*****	35.522*****

Significance levels: * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$; **** $p < 0.005$; ***** $p < 0.001$. Robust standard errors clustered at the *srj*-level are reported in parentheses. *srj* indicates the combined dimensions of fixed effects for source country *s*, recipient country *r* and recipient sector *j* characteristics; *t* denotes the dimension of the individual time fixed effects; *srj & t* indicates the two-way fixed effects model.

Table 2
Panel regression results with output shares using the WIOD 2013

	Dep. var.: relative distance between sectoral output shares $\ln(dz_{srjt})$					
	North-South			Full sample		
	<i>srj</i>	<i>t</i>	<i>srj & t</i>	<i>srj</i>	<i>t</i>	<i>srj & t</i>
Fixed effects						
Import intensity	0.00679	-0.03892*****	-0.01090*	0.01331*****	-0.03945*****	-0.01309*****
$\ln(m_{srj(t-1)})$	(0.00575)	(0.00567)	(0.00588)	(0.00272)	(0.00252)	(0.00279)
Capital-to-labor ratio	0.00426	0.02679	-0.00667	-0.00785	0.03185**	-0.01428
$\ln(dk_{srj(t-1)})$	(0.03079)	(0.05910)	(0.03051)	(0.00926)	(0.01301)	(0.00919)
Interaction term						
$\ln(m_{srj(t-1)}) \cdot$	-0.00162	-0.01294**	-0.00245	-0.00086	-0.00455***	-0.00228**
$\ln(dk_{srj(t-1)})$	(0.00296)	(0.00622)	(0.00294)	(0.00113)	(0.00165)	(0.00113)
Num. of observat.	97774	97774	97774	541855	541855	541855
Degrees of freedom	90526	97757	90513	501375	541838	501362
R^2	0.00026	0.01256	0.00029	0.00016	0.01274	0.00019
F -stat.	2.042	34.316*****	2.164*	8.201*****	239.673*****	9.102*****

Significance levels: * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$; **** $p < 0.005$; ***** $p < 0.001$. See Table 1 for notes.

Table 3
Panel regression results with CO₂ shares using the WIOD 2016

	Dep. var.: relative distance between sectoral CO ₂ shares $\ln(dz_{srjt})$					
	North-South			Full sample		
	<i>srj</i>	<i>t</i>	<i>srj</i> & <i>t</i>	<i>srj</i>	<i>t</i>	<i>srj</i> & <i>t</i>
Fixed effects						
Import intensity	0.04035*****	-0.11613*****	0.00751	0.03647*****	-0.08952*****	0.00093
$\ln(m_{srj(t-1)})$	(0.00674)	(0.00590)	(0.00695)	(0.00301)	(0.00249)	(0.00320)
Capital-to-labor rat.	-0.05548**	-0.07578	-0.03318	0.00721	0.04137*****	0.01120
$\ln(dk_{srj(t-1)})$	(0.02329)	(0.04877)	(0.02309)	(0.00760)	(0.01231)	(0.00758)
Interaction term						
$\ln(m_{srj(t-1)}) \cdot$	-0.00600**	-0.01562***	-0.00446*	0.00107	-0.00080	0.00100
$\ln(dk_{srj(t-1)})$	(0.00242)	(0.00557)	(0.00239)	(0.00096)	(0.00157)	(0.00095)
Num. of observat.	143241	143241	143241	869736	869736	869736
Degrees of freedom	132980	143224	132967	807351	869719	807338
R^2	0.00139	0.03885	0.00016	0.00080	0.02058	0.00001
F -stat.	15.331*****	154.797*****	2.008	49.367*****	577.997*****	1.079

Significance levels: * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$; **** $p < 0.005$; ***** $p < 0.001$. Robust standard errors clustered at the *srj*-level are reported in parentheses. *srj* indicates the combined dimensions of fixed effects for source country *s*, recipient country *r* and recipient sector *j* characteristics; *t* denotes the dimension of the individual time fixed effects; *srj* & *t* indicates the two-way fixed effects model.

Table 4
Panel regression results with CO₂ shares using the WIOD 2013

	Dep. var.: relative distance between sectoral CO ₂ shares $\ln(dz_{srjt})$					
	North-South			Full sample		
	<i>srj</i>	<i>t</i>	<i>srj</i> & <i>t</i>	<i>srj</i>	<i>t</i>	<i>srj</i> & <i>t</i>
Fixed effects						
Import intensity	-0.01337**	-0.11022*****	-0.01433***	-0.00884*****	-0.11411*****	-0.01183*****
$\ln(m_{srj(t-1)})$	(0.00547)	(0.00722)	(0.00554)	(0.00260)	(0.00336)	(0.00267)
Capital-to-labor ratio	0.00233	-0.08977	0.00321	-0.00388	-0.02225	-0.00480
$\ln(dk_{srj(t-1)})$	(0.02695)	(0.07339)	(0.02698)	(0.00817)	(0.01835)	(0.00817)
Interaction term						
$\ln(m_{srj(t-1)}) \cdot$	-0.00108	-0.01630**	-0.00104	-0.00103	-0.00871*****	-0.00117
$\ln(dk_{srj(t-1)})$	(0.00251)	(0.00810)	(0.00251)	(0.00098)	(0.00245)	(0.00098)
Num. of observat.	95069	95069	95069	519131	519131	519131
Degrees of freedom	87920	95052	87907	479498	519114	479485
R^2	0.00031	0.03874	0.00034	0.00010	0.03558	0.00016
F -stat.	2.597*	88.327*****	2.872**	4.900****	465.869*****	7.691*****

Significance levels: * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$; **** $p < 0.005$; ***** $p < 0.001$. See Table 3 for notes.

Regression results:

Tables 1 to 4 present the main panel estimation results. We report heteroscedasticity- and serial-correlation-robust standard errors (Arellano, 1987) clustered at the *srj*-level throughout the regression analysis.

Based on the WIOD 2016, Table 1 uses output shares as the dependent variable. The statistically significant and negative coefficients of import intensity in all columns of Table 1 unequivocally confirm H2, stating structural convergence induced by international trade. The effect of the relative distance of the capital-to-labor ratio (in short, capital-to-labor ratio) on structural change is significant in the full sample estimations with fixed effects in the cross-section (*srj*) and in the cross-section plus time (*srj & t*) only. In these significant cases, the coefficients also confirm H2, stating structural convergence induced by more capital-intensive production.

In Table 1, the interaction term's coefficients are significant and negative, supporting H2 in the full sample and North-South sample estimations with time (*t*) fixed effects, but insignificant in the remaining two North-South sample results. Accordingly, simultaneously higher import and relative capital intensities jointly enhance convergence.

Table 2 shows the same estimations with output shares as the dependent variable based on the WIOD 2013, which provides a smaller number of observations than does the WIOD 2016, potentially reducing the statistical significance of the results. Therefore, the coefficient of import intensity becomes insignificant in the North-South sample with cross-sectional fixed effects (*srj*) and weakly significant and negative with two-way fixed effects (*srj & t*). The effect of import intensity becomes significant and positive in the full sample with cross-sectional fixed effects *srj*, supporting H1, stating structural divergence. Nonetheless, the majority of the estimates (*t* in the North-South sample and *t* and *srj & t* in the full sample) supports H2, stating structural convergence as before. The effect of the capital-to-labor ratio is significantly positive, in favor of H1, with time (*t*) fixed effects only.

Table 3 replaces the output shares used as the dependent variable by CO₂ shares (including emissions from fossil fuel use and process emissions), drawing on the WIOD 2016. Similar to the results with output shares and the WIOD 2013, the sign of the estimates depends on the choice of fixed effects. Cross-sectional fixed effects (*srj*) allow for variation in time and exhibit a positive effect of import intensity on structural change, i.e., divergence, as expressed by H1. This positive effect is, however, dampened by the

negative effect of the interaction of import intensity and the capital-to-labor ratio.¹⁷ Time fixed effects (t), in contrast, allow for variation in the cross-section and result in a positive effect, supporting structural convergence, as expressed by H2. The combination of both types of fixed effects (srj & t) and hence both opposing effects, not surprisingly, results in insignificant estimates. These findings refer to the North-South and full samples.

In contrast to these estimates for trade, the capital-to-labor ratio exhibits a significant and negative effect, supporting structural convergence with cross-sectional fixed effects (srj) in the North-South sample, but a significant and positive effect, supporting divergence with time fixed effects (t) in the full sample. It exhibits insignificant effects in the remaining cases. Nonetheless, the joint effect of import intensity and the capital-to-labor ratio is always (weakly) significant and negative, supporting convergence, in the North-South sample.

Table 4 deploys CO₂ shares as the dependent variable using the WIOD 2013. The results are similar to those deploying output shares using the WIOD 2016 presented in Table 1. The estimated coefficients of import intensity are significant and negative in all estimations, supporting H2, stating structural convergence. The effect of the capital-to-labor ratio is, however, always insignificant. The joint effect of import intensity and the capital-to-labor ratio expressed by the interaction term is always negative but statistically significant in specifications with time (t) fixed effects only.

All estimated coefficients represent elasticities, describing the impact of relative changes in a driver of structural change on relative changes in the (absolute) difference between the sector shares of the recipient and source country for international trade. The estimated (absolute) magnitudes of these elasticities vary between 0.02 and 0.05 among the statistically significant coefficients of the capital-to-labor ratio. The (absolute) magnitudes of the interaction terms are about an order of magnitude smaller. The variation in the (absolute) magnitudes of the coefficients of import intensity is substantial; the magnitudes vary between about 0.004 and 0.116.

5 Discussion

This section interprets and compares the regression results, particularly those of the main panel regressions presented in the previous section, those of alternative robustness checks and those of supplementary sectoral estimates.

¹⁷The sole effect of the import intensity without the interaction is positive as well; see Appendix Table B2.

Main regression results:

Basically, the results promote the view that international trade supports the international convergence of sectoral structures via the spread of knowledge, technologies, preferences and so forth, such that countries' sectoral structures become more similar. Similarly, more intensive utilization of capital and embodied technologies supports the international convergence of sectoral structures. Using the WIOD 2016 sample and output value shares as the dependent variable, this result holds unequivocally and significantly.

For the WIOD 2013 sample period from 1995 to 2009, however, the results provide an indication that over time, international trade has enhanced the international divergence of sectoral output structures. This means that in accordance with classical Ricardian trade theory, in the world economy, countries specialize in the production of different goods. Similarly, in the global (long-term) cross-section, relatively more intensive capital use seems to foster the sectoral divergence of output structures.

When considering CO₂ shares taken from the WIOD 2013 as the dependent variable, in contrast, the results clearly confirm the previous finding that trade fosters structural convergence.

For the later WIOD 2016 sample period from 2000 to 2014, however, the results point to a possible regime change. From global and North-South perspectives, the results indicate the international divergence of sectoral structures occurring via Ricardian specialization in more or less CO₂-intensive production. This result points to international outsourcing of CO₂-intensive production to emerging economies or so-called carbon leakage.

To obtain an impression of the time horizon required for the convergence dynamics to materialize and to become visible, we carry out crude estimates (Hübler & Glas, 2014). In the specifications with cross-sectional (srj) or two-way fixed effects (srj & t), the elasticities estimated for the effect of import intensity have an order of magnitude of about -0.01. This means that ceteris paribus, by solely focusing on the impact of trade (putting aside other confounders of sectoral changes), doubling the import intensity (increasing it by 100%) leads to an annual 1% decline in the relative distance between the sector shares in recipient and source countries. The resulting half time, i.e., the time to reduce the relative distance by 50%, is almost 70 years. The elasticities suggested by the specifications with time fixed effects (t) reach an order of magnitude of -0.1. The corresponding resulting half time amounts to seven years. In any case, despite the assumption

of this substantial increase in import intensity, the dynamics require decades or even centuries to approach the theoretical long-term equilibrium of internationally equalized sector shares. Consequently, we talk about long-term effects. This insight should be taken into account when considering the possible international impacts of national policies.

Robustness check results:

The following robustness checks exclude the interaction terms used so far, explore different time lags of the regressors and replace CO₂ emissions with energy use to construct the dependent variable. The robustness checks overall confirm the main panel regression results.

1. *Exclusion of the interaction term:* The results for the WIOD 2016 are presented in Appendix B.1. The estimated coefficients of import intensity are very robust to the exclusion of the interaction term between import intensity and the capital-to-labor ratio. The coefficients of the capital-to-labor ratio, in contrast, experience changes in significance levels and signs; particularly, most coefficients turn statistically significant and positive, supporting structural divergence (H1). This result nevertheless aligns with economic theory and intuition: more intensive capital use itself tends to result in increasing specialization in the activities that can be performed best with this capital and its embodied technologies; once, however, new goods, knowledge, technologies, etc., arrive from abroad, capital will incorporate technological improvements such that production will become more similar to that at the source of the goods, knowledge and technologies. Thus, international trade appears to be a prerequisite for international structural convergence (H1), while capital accumulation supports this trade-driven mechanism. This is reflected by the (if statistically significant) negative interaction terms.

2. *Different time lags of the regressors:* The results for the WIOD 2016 with the time (year) lags $t - 2$ and $t - 3$, instead of $t - 1$, for all regressors are presented in Appendix B.2. While some significance levels change, the results are barely qualitatively and quantitatively affected. As a notable exception, in the full sample with cross-sectional fixed effects (srj), $t - 3$ lags and output shares as the dependent variable, presented in Table B4, the effect of imports becomes significant and positive. Similarly, in the North-South and full samples with two-way fixed effects ($srj \& t$), $t - 2$ or $t - 3$ lags and CO₂ shares as the dependent variable, as presented in Tables B5 and B6, the positive effect of imports becomes significant.

3. *Energy shares as the dependent variable:* In the robustness check presented in

Appendix B.3, we replace the CO₂ emission shares by (gross) energy input shares. Compared with CO₂ emissions, (gross) energy use includes electricity and other non-CO₂-emitting energy inputs. In both samples, the WIOD 2013 and 2016, all statistically significant coefficients of import intensity and its interaction with the capital-to-labor ratio have a negative sign, supporting structural convergence (H2). The coefficients of the capital-to-labor ratio are always insignificant. Particularly, in the WIOD 2016, import intensity has a significant and negative effect on structural change with cross-sectional (srj) or time (t) fixed effects but not with two-way fixed effects (srj & t); in the WIOD 2013, import intensity has a significant and negative effect with time (t) fixed effects, and in the full sample, it has a weakly significant and negative effect with two-way fixed effects (srj & t). This means that regarding the impact of the imports estimated with the WIOD 2013, the energy share results are in line with the previous CO₂ share and output share results (except the single significantly positive effect of import intensity when using output shares). With the WIOD 2016, however, the energy share results confirm the previous finding of convergence (H2) being driven by imports with output shares but do not confirm the previous finding of divergence (H1) being driven by imports obtained with CO₂ shares and cross-section fixed effects (srj).

Sectoral regression results:

To understand how structural change actually occurs, we need to look at the sectoral level. To this end, we carry out the panel regressions separately for each sector j . As before, we use cross-sectional (sr), time (t) or two-way (sr & t) fixed effects based on the WIOD 2013 or 2016. Appendix B.4 presents the selected results of the sector-specific estimations. In each table, all available sectors are included and ordered by their CO₂ intensities, i.e., CO₂ emissions per output value, of trade recipient countries r . The left columns of the tables show the sector shares in trade recipient countries r at the beginning and end of the sample period in terms of emissions or output. This reveals whether the sectors were shrinking or expanding during the sample period.

Table B9, for example, indicates that the energy sector is the most CO₂-intensive sector, which slightly expanded between 2000 and 2014, and exhibits a significantly negative effect of import intensity (structural convergence) but a significantly positive effect (structural divergence) of the relative distance of the capital-to-labor ratio and its interaction with import intensity on the relative distance of output shares. The number of observations in this sector is 25,284, and the number of fixed effects is 1,806. Whereas

R^2 is low, the F -statistic for the null hypothesis of all estimated coefficients jointly being zero is high, clearly rejecting the null hypothesis.

The sectoral panel regression results overall confirm the cross-sectoral panel regression results. The estimates with the WIOD 2016 and output value shares as the dependent variable overall confirm the hypothesis of structural convergence (H2) being driven by trade. While in the estimations with cross-sectional fixed effects, about half of the coefficients of import intensity are statistically significant, most of which have a negative sign, in the specification with time fixed effects, most of them are significant, all with a negative sign.

In accordance with the previous cross-sectoral results, in the WIOD 2016 sample, the use of CO₂ shares as the dependent variable leads to mostly positive coefficients of import intensity among the relatively small number of statistically significant results when cross-sectional fixed effects (sr) are included. Based on the WIOD 2013 sample, in contrast, the significant estimates of the effect of imports are mostly negative, in favor of the convergence hypothesis (H2). This result supports the previous finding of structural divergence (H1), i.e., the specialization in more or less CO₂-intensive activities starting around the year 2000. When using time fixed effects (t), in contrast, more coefficients of imports become statistically significant, and all of them are negative, in favor of the convergence hypothesis (H2) as before.

There are less significant estimates of the effect of the capital-to-labor ratio than of import intensity. Although negative signs prevail for the capital-to-labor ratio, these estimated signs are mixed. The estimates of the interaction effect between import intensity and the capital-to-labor ratio show a similar picture as do the estimates of pure capital-to-labor ratio effects.

6 Conclusion

Classical Ricardian trade theory predicts differences between economic structures of countries engaged in international trade because each country specializes in specific sectors (goods) according to comparative advantages. Our results, however, oppose this view. Most results show that international trade leads to increasing similarity in terms of sectoral structure. This mechanism can be driven by the spread of knowledge, technologies, preferences, habits and so forth during the course of globalization. This finding holds when sector shares are measured using output value shares or energy input shares and, for the data sample covering the time period before the turn of the millennium, it holds when

using CO₂ emission shares. Running panel regressions for the available sectors separately confirms these cross-sectoral panel regression results.

Because advanced technologies need to be embodied in capital, this mechanism is enhanced when imports are accompanied by more intensive capital use, which is visible in the results. This means that the joint effect of import- and capital-intensive production enhances sectoral convergence, whereas the sole effect of a higher capital-to-labor ratio on structural convergence appears to be ambiguous because capital use without sufficient imports may embody old-fashioned technologies. Capital can thus be referred to as an enhancer of structural convergence.

The finding of structural convergence being driven by international trade basically means good news for climate and energy policy: over a sufficiently long time horizon, energy- and CO₂-emission-saving intersectoral structural change in industrialized countries will automatically spill over to emerging countries via international trade. The results, however, also provide an indication for increasing international specialization in more or less CO₂-intensive production starting at the turn of the millennium. This outcome is confirmed when running panel regressions for the available sectors separately. Among other possible reasons, this outcome might be driven by the international relocation of CO₂-intensive production to emerging economies, so-called carbon leakage, which might be fostered by climate policy measures in industrialized countries. However, this outcome does not hold when replacing CO₂ emission shares with energy input shares. Energy input shares refer to gross energy use including electricity and non-emission-relevant energy sources. They are affected by (total factor) productivity gains and energy-specific productivity gains. CO₂ emissions shares, in contrast, capture the emission intensity and decarbonization of energy supply and industrial production.

Nonetheless, this insight is somewhat alarming for climate policy makers because it implies that more stringent climate policy in industrialized countries may decrease direct CO₂ emissions in particular sectors of industrialized countries but increase them in the same sectors of emerging (or developing) countries, resulting in structural divergence. As long as no global climate policy solution is in place, this mechanism will weaken the effectiveness of unilateral climate policy in the globalized world economy. Therefore, in addition to the potentially productivity-enhancing and energy-saving effects of international trade (Cole, 2006; Perkins & Neumayer, 2009; Hübler & Glas, 2014), policies should directly strengthen the international transfer of environmentally friendly and, particularly, CO₂ emissions-saving technologies. The ultimate goal is a global climate policy solution

that avoids carbon leakage effects.

These empirical results should, however, be treated with caution, especially because some outcomes depend on the time frame (data sample) and choice of the exploited variation in the data (cross-section, time fixed effects or both). Furthermore, in addition to OECD countries, the dataset of the WIOD is limited to a number of heterogeneous emerging countries, where China and India are the dominant actors, and does not cover any developing countries. Although our results indicate structural convergence, such processes require a long time horizon, probably decades, to materialize. Therefore, the question remains as to whether international specialization in more or less CO₂-intensive production, as indicated by our analysis for this millennium, is a temporal statistical, negligible phenomenon or the beginning of a considerable long-term process.

Future research may address further drivers of structural change that may independently or in connection with international trade foster structural divergence or convergence. It may also include an extended set of countries once required data are available or selected countries or sectors with specific data as case studies.

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Supplementary online appendix

A Descriptive statistics and test procedures

A.1 Summary statistics

Table A1
Summary statistics for output shares as the dep. var. and $t - 1$ lags

Variable	Num. of obs.	Min.	Median	Mean	Std. dev.	Max.
WIOD 2016 North-South						
Gross Output relative distance between sector shares $\ln(dz_{srjt})$	143 435	-16.395	-0.542	-0.477	1.564	9.349
Capital-to-labor ratio $\ln(dk_{srj(t-1)})$	143 435	-10.945	-0.207	-0.397	0.822	6.468
Import intensity $\ln(m_{srj(t-1)})$	143 435	-23.115	-8.457	-8.900	2.938	-0.941
WIOD 2016 full sample						
Gross Output relative distance between sector shares $\ln(dz_{srjt})$	871 733	-16.395	-0.623	-0.664	1.450	9.817
Capital-to-labor ratio $\ln(dk_{srj(t-1)})$	871 733	-13.697	-0.262	-0.164	1.484	9.073
Import intensity $\ln(m_{srj(t-1)})$	871 733	-23.115	-7.569	-7.789	2.457	0.145
WIOD 2013 North-South						
Gross Output relative distance between sector shares $\ln(dz_{srjt})$	97 774	-11.286	-0.636	-0.641	1.420	10.749
Capital-to-labor ratio $\ln(dk_{srj(t-1)})$	97 774	-11.642	-0.147	-0.328	0.773	9.124
Import intensity $\ln(m_{srj(t-1)})$	97 774	-22.805	-7.953	-8.441	2.850	-1.027
WIOD 2013 full sample						
Gross Output relative distance between sector shares $\ln(dz_{srjt})$	541 855	-13.136	-0.716	-0.803	1.357	10.749
Capital-to-labor ratio $\ln(dk_{srj(t-1)})$	541 855	-11.642	-0.184	-0.020	1.534	10.863
Import intensity $\ln(m_{srj(t-1)})$	541 855	-22.805	-7.560	-7.782	2.397	0.873

Table A2
Summary statistics for CO₂ shares as the dep. var. and $t - 1$ lags

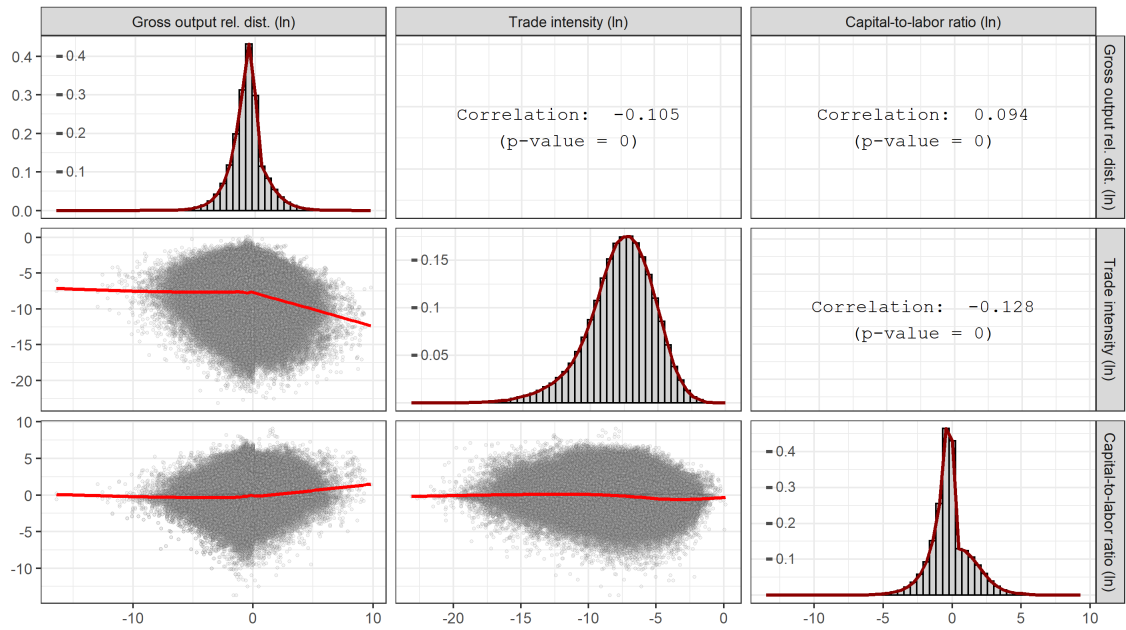
Variable	Num. of obs.	Min.	Median	Mean	Std. dev.	Max.
WIOD 2016 North-South						
CO ₂ emissions relative distance between sector shares $\ln(dz_{srjt})$	143 241	-10.802	-0.289	-0.122	1.690	14.854
Capital-to-labor ratio $\ln(dk_{srj(t-1)})$	143 241	-10.945	-0.207	-0.397	0.822	6.468
Import intensity $\ln(m_{srj(t-1)})$	143 241	-23.115	-8.456	-8.899	2.938	-0.941
WIOD 2016 full sample						
CO ₂ emissions relative distance between sector shares $\ln(dz_{srjt})$	869 736	-12.352	-0.267	-0.122	1.656	15.661
Capital-to-labor ratio $\ln(dk_{srj(t-1)})$	869 736	-13.697	-0.262	-0.163	1.484	9.073
Import intensity $\ln(m_{srj(t-1)})$	869 736	-23.115	-7.569	-7.789	2.457	0.145
WIOD 2013 North-South						
CO ₂ emissions relative distance between sector shares $\ln(dz_{srjt})$	95 069	-12.177	-0.293	-0.190	1.559	10.490
Capital-to-labor ratio $\ln(dk_{srj(t-1)})$	95 069	-11.642	-0.147	-0.329	0.776	9.124
Import intensity $\ln(m_{srj(t-1)})$	95 069	-22.805	-7.889	-8.386	2.843	-1.027
WIOD 2013 full sample						
CO ₂ emissions relative distance between sector shares $\ln(dz_{srjt})$	519 131	-12.177	-0.299	-0.197	1.564	10.998
Capital-to-labor ratio $\ln(dk_{srj(t-1)})$	519 131	-11.642	-0.183	-0.019	1.536	10.863
Import intensity $\ln(m_{srj(t-1)})$	519 131	-22.805	-7.530	-7.755	2.394	0.873

Table A3
Summary statistics for energy shares as the dep. var. and $t - 1$ lags

Variable	Num. of obs.	Min.	Median	Mean	Std. dev.	Max.
WIOD 2016 North-South						
Energy relative distance between sector shares $\ln(dz_{srjt})$	143 444	-12.190	-0.404	-0.312	1.628	13.421
Capital-to-labor ratio $\ln(dk_{srj(t-1)})$	143 444	-10.945	-0.207	-0.397	0.822	6.468
Import intensity $\ln(m_{srj(t-1)})$	143 444	-23.115	-8.457	-8.901	2.938	-0.941
WIOD 2016 full sample						
Energy relative distance between sector shares $\ln(dz_{srjt})$	871 719	-13.602	-0.411	-0.364	1.566	13.739
Capital-to-labor ratio $\ln(dk_{srj(t-1)})$	871 719	-13.697	-0.262	-0.164	1.484	9.073
Import intensity $\ln(m_{srj(t-1)})$	871 719	-23.115	-7.569	-7.789	2.457	0.145
WIOD 2013 North-South						
Energy relative distance between sector shares $\ln(dz_{srjt})$	97 657	-13.581	-0.534	-0.521	1.580	46.511
Capital-to-labor ratio $\ln(dk_{srj(t-1)})$	97 657	-11.642	-0.147	-0.328	0.774	9.124
Import intensity $\ln(m_{srj(t-1)})$	97 657	-22.805	-7.950	-8.439	2.850	-1.027
WIOD 2013 full sample						
Energy relative distance between sector shares $\ln(dz_{srjt})$	540 775	-13.583	-0.535	-0.546	1.524	47.244
Capital-to-labor ratio $\ln(dk_{srj(t-1)})$	540 775	-11.642	-0.184	-0.020	1.535	10.863
Import intensity $\ln(m_{srj(t-1)})$	540 775	-22.805	-7.558	-7.780	2.397	0.873

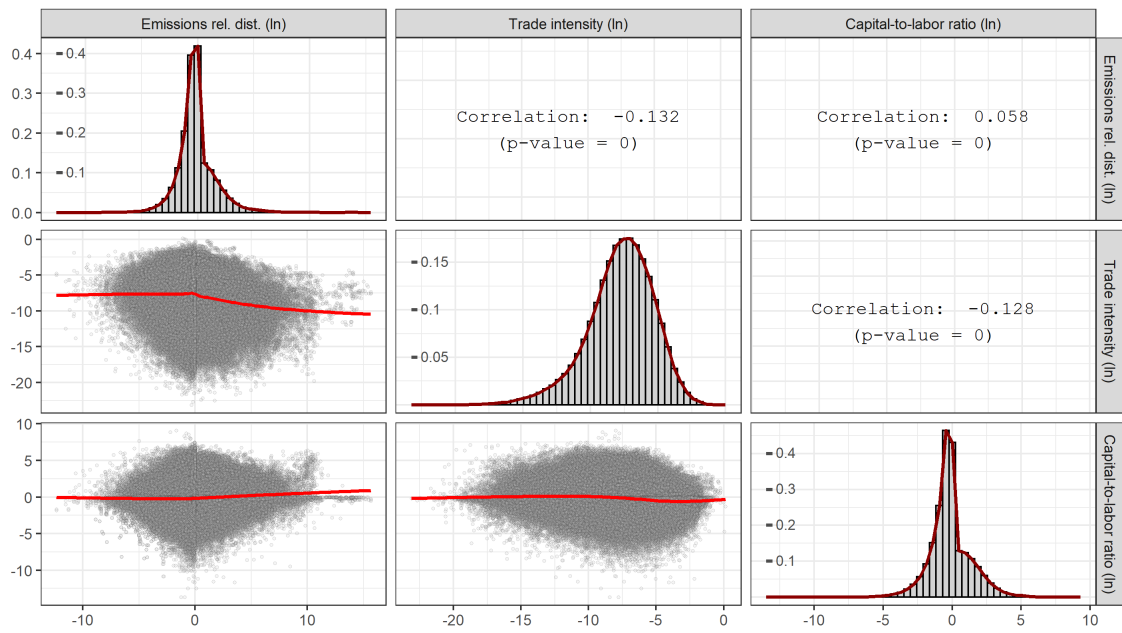
A.2 Correlations and distributions

Figure 3
Descriptive statistics with output shares using the WIOD 2016 full sample



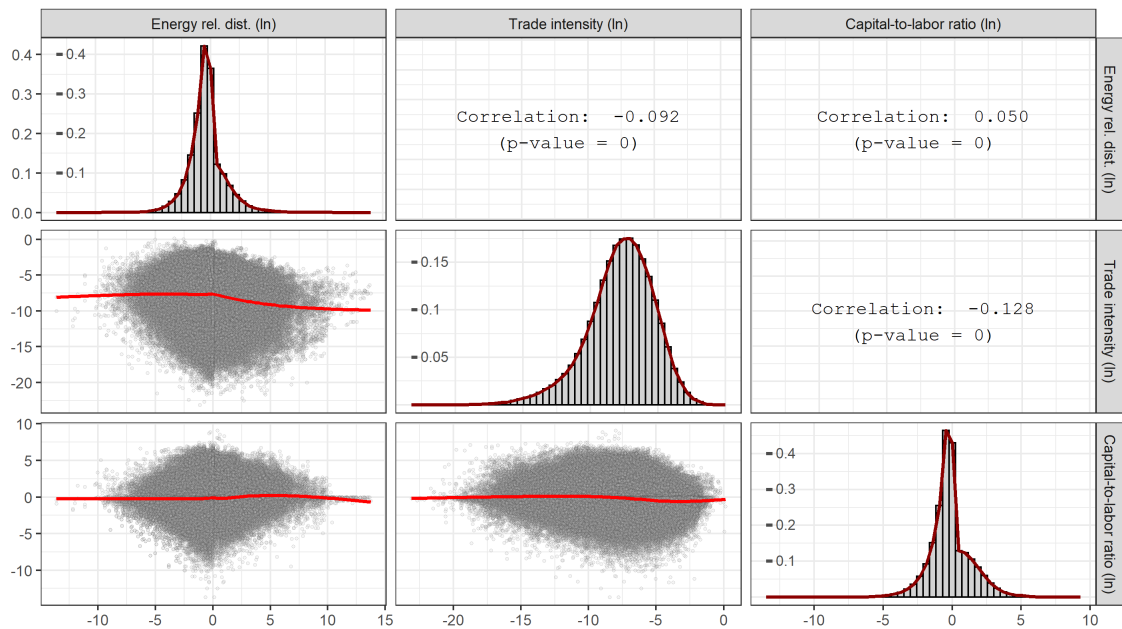
Source: Own illustrations based on data taken from the WIOD 2016 release. The numbers in the upper right area are correlations between the two indicators named on the upper horizontal and right vertical axis of the depiction matrix. The lower left area illustrates the relation of the correlation partners by scatter-plotting the indicator named on the right vertical axis as a function of the indicator named on the upper horizontal axis. The red lines in the scatter-plots in the lower left area have been created by using a non-parametric algorithm based on generalized additive models with integrated smoothness estimation procedures as suggested by Wood (2011, 2004). The histograms on the diagonal depict the distributions (density of observations covering an area with the size one) of each indicator separately.

Figure 4
Descriptive statistics with CO₂ shares using the WIOD 2016 full sample



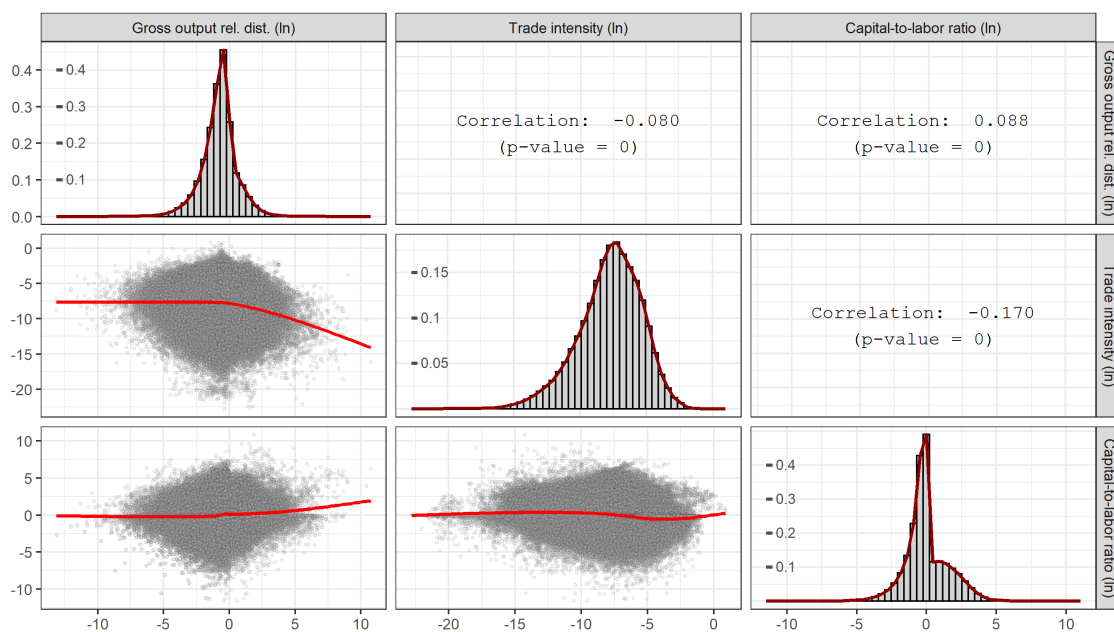
Source: Own illustrations based on data taken from the WIOD 2016 release. See before for notes.

Figure 5
Descriptive statistics with energy shares using the WIOD 2016 full sample



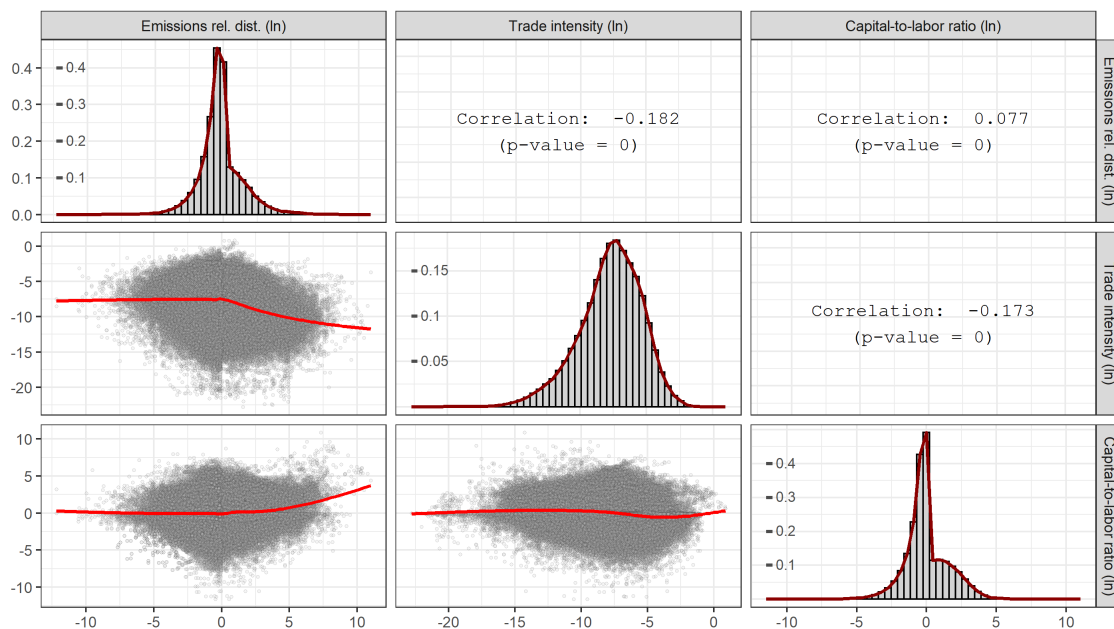
Source: Own illustrations based on data taken from the WIOD 2016 release. See before for notes.

Figure 6
Descriptive statistics with output shares using the WIOD 2013 full sample



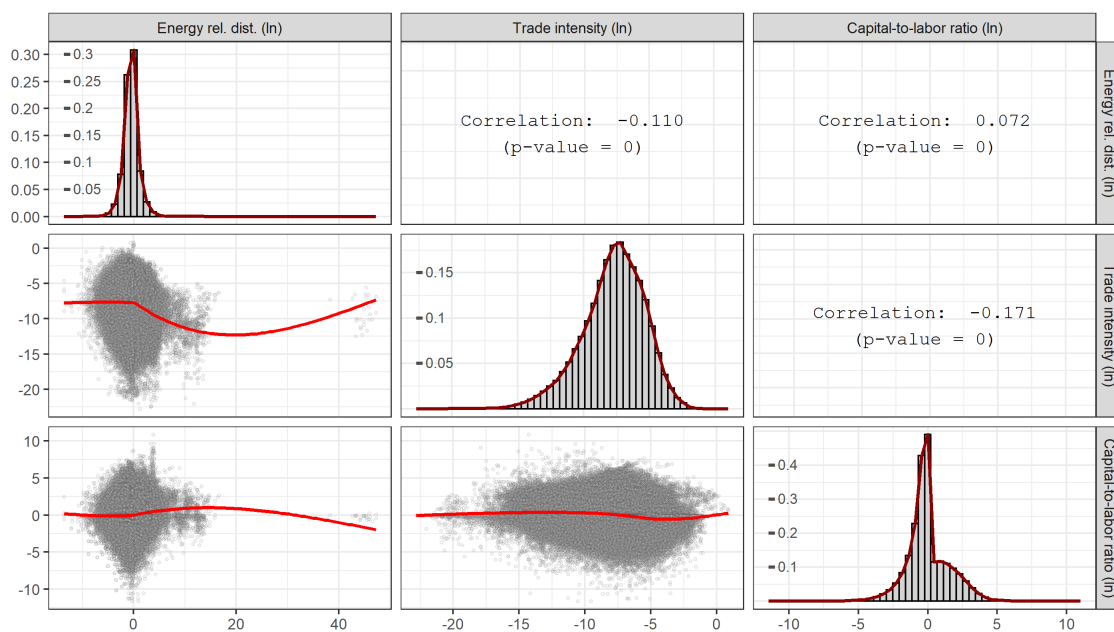
Source: Own illustrations based on data taken from the WIOD 2013 release. See before for notes.

Figure 7
Descriptive statistics with CO₂ shares using the WIOD 2013 full sample



Source: Own illustrations based on data taken from the WIOD 2013 release. See before for notes.

Figure 8
 Descriptive statistics with energy shares using the WIOD 2013 full sample



Source: Own illustrations based on data taken from the WIOD 2013 release. See before for notes.

A.3 Tests for panel specifications

Table A4
F-tests for fixed eff. (H1) vs. pooled or restricted fixed eff. (H0), output shares

Dataset	Sample	Interaction term	Method	H0 model	H1 model	Statistics
WIOD 2016	North-South	present	<i>F</i> -test for individual effects	pooled	<i>srj</i>	58.962***** (<i>df</i> = 10258; 133174)
WIOD 2016	North-South	present	<i>F</i> -test for two-way effects	pooled	<i>srj</i> & <i>t</i>	59.234***** (<i>df</i> = 10271; 133161)
WIOD 2016	North-South	present	<i>F</i> -test for time effects	pooled	<i>t</i>	502.930***** (<i>df</i> = 14; 143418)
WIOD 2016	North-South	present	<i>F</i> -test for two-way effects	<i>t</i>	<i>srj</i> & <i>t</i>	55.932***** (<i>df</i> = 10257; 133161)
WIOD 2016	North-South	present	<i>F</i> -test for two-way effects	<i>srj</i>	<i>srj</i> & <i>t</i>	50.332***** (<i>df</i> = 13; 133161)
WIOD 2016	Full sample	present	<i>F</i> -test for individual effects	pooled	<i>srj</i>	52.972***** (<i>df</i> = 62386; 809344)
WIOD 2016	Full sample	present	<i>F</i> -test for two-way effects	pooled	<i>srj</i> & <i>t</i>	53.086***** (<i>df</i> = 62399; 809331)
WIOD 2016	Full sample	present	<i>F</i> -test for time effects	pooled	<i>t</i>	3 184.077***** (<i>df</i> = 14; 871716)
WIOD 2016	Full sample	present	<i>F</i> -test for two-way effects	<i>t</i>	<i>srj</i> & <i>t</i>	49.883***** (<i>df</i> = 62385; 809331)
WIOD 2016	Full sample	present	<i>F</i> -test for two-way effects	<i>srj</i>	<i>srj</i> & <i>t</i>	119.087***** (<i>df</i> = 13; 809331)
WIOD 2013	North-South	present	<i>F</i> -test for individual effects	pooled	<i>srj</i>	47.240***** (<i>df</i> = 7245; 90526)
WIOD 2013	North-South	present	<i>F</i> -test for two-way effects	pooled	<i>srj</i> & <i>t</i>	47.635***** (<i>df</i> = 7258; 90513)
WIOD 2013	North-South	present	<i>F</i> -test for time effects	pooled	<i>t</i>	274.826***** (<i>df</i> = 14; 97757)
WIOD 2013	North-South	present	<i>F</i> -test for two-way effects	<i>t</i>	<i>srj</i> & <i>t</i>	45.447***** (<i>df</i> = 7244; 90513)
WIOD 2013	North-South	present	<i>F</i> -test for two-way effects	<i>srj</i>	<i>srj</i> & <i>t</i>	56.770***** (<i>df</i> = 13; 90513)
WIOD 2013	Full sample	present	<i>F</i> -test for individual effects	pooled	<i>srj</i>	41.858***** (<i>df</i> = 40477; 501375)
WIOD 2013	Full sample	present	<i>F</i> -test for two-way effects	pooled	<i>srj</i> & <i>t</i>	42.319***** (<i>df</i> = 40490; 501362)
WIOD 2013	Full sample	present	<i>F</i> -test for time effects	pooled	<i>t</i>	2 282.117***** (<i>df</i> = 14; 541838)
WIOD 2013	Full sample	present	<i>F</i> -test for two-way effects	<i>t</i>	<i>srj</i> & <i>t</i>	39.286***** (<i>df</i> = 40476; 501362)
WIOD 2013	Full sample	present	<i>F</i> -test for two-way effects	<i>srj</i>	<i>srj</i> & <i>t</i>	337.959***** (<i>df</i> = 13; 501362)
WIOD 2016	North-South	absent	<i>F</i> -test for individual effects	pooled	<i>srj</i>	59.216***** (<i>df</i> = 10258; 133175)
WIOD 2016	North-South	absent	<i>F</i> -test for two-way effects	pooled	<i>srj</i> & <i>t</i>	59.491***** (<i>df</i> = 10271; 133162)
WIOD 2016	North-South	absent	<i>F</i> -test for time effects	pooled	<i>t</i>	531.514***** (<i>df</i> = 14; 143419)
WIOD 2016	North-South	absent	<i>F</i> -test for two-way effects	<i>t</i>	<i>srj</i> & <i>t</i>	55.993***** (<i>df</i> = 10257; 133162)
WIOD 2016	North-South	absent	<i>F</i> -test for two-way effects	<i>srj</i>	<i>srj</i> & <i>t</i>	50.502***** (<i>df</i> = 13; 133162)
WIOD 2016	Full sample	absent	<i>F</i> -test for individual effects	pooled	<i>srj</i>	52.984***** (<i>df</i> = 62386; 809345)
WIOD 2016	Full sample	absent	<i>F</i> -test for two-way effects	pooled	<i>srj</i> & <i>t</i>	53.098***** (<i>df</i> = 62399; 809332)
WIOD 2016	Full sample	absent	<i>F</i> -test for time effects	pooled	<i>t</i>	3 178.550***** (<i>df</i> = 14; 871717)
WIOD 2016	Full sample	absent	<i>F</i> -test for two-way effects	<i>t</i>	<i>srj</i> & <i>t</i>	49.900***** (<i>df</i> = 62385; 809332)
WIOD 2016	Full sample	absent	<i>F</i> -test for two-way effects	<i>srj</i>	<i>srj</i> & <i>t</i>	118.767***** (<i>df</i> = 13; 809332)

Significance levels: * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$; **** $p < 0.005$; ***** $p < 0.001$. Independent variables are included with $t-1$ lags. *df* means degrees of freedom, where the first number is the difference of the *df* between $H0$ and $H1$, whereas the second number is the *df* in the model $H1$. *srj* indicates the combined dimensions of fixed effects for source country s , recipient country r and recipient sector j characteristics; t denotes the dimension of the individual time fixed effects; *srj* & t indicates the two-way fixed effects model. Alternative fixed effects specifications ($H1$) are tested against a pooled or a fixed effects specification with reduced dimensionality (a reduced number of fixed effects) ($H0$).

Table A5
LM-tests for fixed effects (H1) vs. pooled (H0), output shares

Dataset	Sample	Interaction term	H1 model	Honda (1985)	Breusch & Pagan (1980)	King & Wu (1997)	Gourieroux et al. (1982)
WIOD 2016	North-South	Present	<i>srj</i>	767.437*****	588 958.987*****	767.437*****	
WIOD 2016	North-South	Present	<i>t</i>	116.676*****	13 613.205*****	116.676*****	
WIOD 2016	North-South	Present	<i>srj & t</i>	625.162*****	602 572.192*****	143.915*****	602 572.192*****
WIOD 2016	Full sample	Present	<i>srj</i>	1 859.242*****	3 456 781.342*****	1 859.242*****	
WIOD 2016	Full sample	Present	<i>t</i>	710.500*****	504 809.627*****	710.500*****	
WIOD 2016	Full sample	Present	<i>srj & t</i>	1 817.082*****	3 961 590.969*****	737.275*****	3 961 590.969*****
WIOD 2013	North-South	Present	<i>srj</i>	599.445*****	359 334.853*****	599.445*****	
WIOD 2013	North-South	Present	<i>t</i>	86.353*****	7 456.829*****	86.353*****	
WIOD 2013	North-South	Present	<i>srj & t</i>	484.933*****	366 791.681*****	111.459*****	366 791.681*****
WIOD 2013	Full sample	Present	<i>srj</i>	1 370.589*****	1 878 513.841*****	1 370.589*****	
WIOD 2013	Full sample	Present	<i>t</i>	665.840*****	443 342.355*****	665.840*****	
WIOD 2013	Full sample	Present	<i>srj & t</i>	1 439.972*****	2 321 856.196*****	689.979*****	2 321 856.196*****
WIOD 2016	North-South	Absent	<i>srj</i>	769.744*****	592 505.474*****	769.744*****	
WIOD 2016	North-South	Absent	<i>t</i>	138.165*****	19 089.525*****	138.165*****	
WIOD 2016	North-South	Absent	<i>srj & t</i>	641.988*****	611 594.999*****	165.472*****	611 594.999*****
WIOD 2016	Full sample	Absent	<i>srj</i>	1 860.158*****	3 460 186.762*****	1 860.158*****	
WIOD 2016	Full sample	Absent	<i>t</i>	728.472*****	530 671.539*****	728.472*****	
WIOD 2016	Full sample	Absent	<i>srj & t</i>	1 830.438*****	3 990 858.301*****	755.258*****	3 990 858.301*****

Significance levels: * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$; **** $p < 0.005$; ***** $p < 0.001$. *LM* means Lagrange Multiplier. Independent variables are included with $t-1$ lags. *srj* indicates the combined dimensions of fixed effects for source country s , recipient country r and recipient sector j characteristics; t denotes the dimension of the individual time fixed effects; *srj & t* indicates the two-way fixed effects model. All alternative fixed effects specifications (H1) are tested against a pooled specification (H0).

Table A6
F-tests for fixed eff. (H1) vs. pooled or restricted fixed eff. (H0), *CO*₂ shares

Dataset	Sample	Interaction term	Method	H0 model	H1 model	Statistics
WIOD 2016	North-South	present	<i>F</i> -test for individual effects	pooled	<i>srj</i>	35.205***** (<i>df</i> = 10258; 132980)
WIOD 2016	North-South	present	<i>F</i> -test for two-way effects	pooled	<i>srj</i> & <i>t</i>	35.576***** (<i>df</i> = 10271; 132967)
WIOD 2016	North-South	present	<i>F</i> -test for time effects	pooled	<i>t</i>	413.170***** (<i>df</i> = 14; 143224)
WIOD 2016	North-South	present	<i>F</i> -test for two-way effects	<i>t</i>	<i>srj</i> & <i>t</i>	33.738***** (<i>df</i> = 10257; 132967)
WIOD 2016	North-South	present	<i>F</i> -test for two-way effects	<i>srj</i>	<i>srj</i> & <i>t</i>	89.175***** (<i>df</i> = 13; 132967)
WIOD 2016	Full sample	present	<i>F</i> -test for individual effects	pooled	<i>srj</i>	34.321***** (<i>df</i> = 62382; 807351)
WIOD 2016	Full sample	present	<i>F</i> -test for two-way effects	pooled	<i>srj</i> & <i>t</i>	34.593***** (<i>df</i> = 62395; 807338)
WIOD 2016	Full sample	present	<i>F</i> -test for time effects	pooled	<i>t</i>	1 406.484***** (<i>df</i> = 14; 869719)
WIOD 2016	Full sample	present	<i>F</i> -test for two-way effects	<i>t</i>	<i>srj</i> & <i>t</i>	33.548***** (<i>df</i> = 62381; 807338)
WIOD 2016	Full sample	present	<i>F</i> -test for two-way effects	<i>srj</i>	<i>srj</i> & <i>t</i>	366.870***** (<i>df</i> = 13; 807338)
WIOD 2013	North-South	present	<i>F</i> -test for individual effects	pooled	<i>srj</i>	60.958***** (<i>df</i> = 7146; 87920)
WIOD 2013	North-South	present	<i>F</i> -test for two-way effects	pooled	<i>srj</i> & <i>t</i>	60.927***** (<i>df</i> = 7159; 87907)
WIOD 2013	North-South	present	<i>F</i> -test for time effects	pooled	<i>t</i>	314.811***** (<i>df</i> = 14; 95052)
WIOD 2013	North-South	present	<i>F</i> -test for two-way effects	<i>t</i>	<i>srj</i> & <i>t</i>	57.796***** (<i>df</i> = 7145; 87907)
WIOD 2013	North-South	present	<i>F</i> -test for two-way effects	<i>srj</i>	<i>srj</i> & <i>t</i>	8.173***** (<i>df</i> = 13; 87907)
WIOD 2013	Full sample	present	<i>F</i> -test for individual effects	pooled	<i>srj</i>	63.633***** (<i>df</i> = 39630; 479498)
WIOD 2013	Full sample	present	<i>F</i> -test for two-way effects	pooled	<i>srj</i> & <i>t</i>	63.635***** (<i>df</i> = 39643; 479485)
WIOD 2013	Full sample	present	<i>F</i> -test for time effects	pooled	<i>t</i>	1 566.370***** (<i>df</i> = 14; 519114)
WIOD 2013	Full sample	present	<i>F</i> -test for two-way effects	<i>t</i>	<i>srj</i> & <i>t</i>	60.587***** (<i>df</i> = 39629; 479485)
WIOD 2013	Full sample	present	<i>F</i> -test for two-way effects	<i>srj</i>	<i>srj</i> & <i>t</i>	12.282***** (<i>df</i> = 13; 479485)
WIOD 2016	North-South	absent	<i>F</i> -test for individual effects	pooled	<i>srj</i>	35.333***** (<i>df</i> = 10258; 132981)
WIOD 2016	North-South	absent	<i>F</i> -test for two-way effects	pooled	<i>srj</i> & <i>t</i>	35.709***** (<i>df</i> = 10271; 132968)
WIOD 2016	North-South	absent	<i>F</i> -test for time effects	pooled	<i>t</i>	436.582***** (<i>df</i> = 14; 143225)
WIOD 2016	North-South	absent	<i>F</i> -test for two-way effects	<i>t</i>	<i>srj</i> & <i>t</i>	33.763***** (<i>df</i> = 10257; 132968)
WIOD 2016	North-South	absent	<i>F</i> -test for two-way effects	<i>srj</i>	<i>srj</i> & <i>t</i>	89.830***** (<i>df</i> = 13; 132968)
WIOD 2016	Full sample	absent	<i>F</i> -test for individual effects	pooled	<i>srj</i>	34.345***** (<i>df</i> = 62382; 807352)
WIOD 2016	Full sample	absent	<i>F</i> -test for two-way effects	pooled	<i>srj</i> & <i>t</i>	34.617***** (<i>df</i> = 62395; 807339)
WIOD 2016	Full sample	absent	<i>F</i> -test for time effects	pooled	<i>t</i>	1 438.606***** (<i>df</i> = 14; 869720)
WIOD 2016	Full sample	absent	<i>F</i> -test for two-way effects	<i>t</i>	<i>srj</i> & <i>t</i>	33.548***** (<i>df</i> = 62381; 807339)
WIOD 2016	Full sample	absent	<i>F</i> -test for two-way effects	<i>srj</i>	<i>srj</i> & <i>t</i>	366.912***** (<i>df</i> = 13; 807339)

Significance levels: * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$; **** $p < 0.005$; ***** $p < 0.001$. Independent variables are included with $t-1$ lags. *df* means degrees of freedom, where the first number is the difference of the *df* between *H0* and *H1*, whereas the second number is the *df* in the model *H1*. *srj* indicates the combined dimensions of fixed effects for source country *s*, recipient country *r* and recipient sector *j* characteristics; *t* denotes the dimension of the individual time fixed effects; *srj* & *t* indicates the two-way fixed effects model. Alternative fixed effects specifications (*H1*) are tested against a pooled or a fixed effects specification with reduced dimensionality (a reduced number of fixed effects) (*H0*).

Table A7
LM-tests for fixed effects (H1) vs. pooled (H0), CO_2 shares

Dataset	Sample	Interaction term	H1 model	Honda (1985)	Breusch & Pagan (1980)	King & Wu (1997)	Gourieroux et al. (1982)
WIOD 2016	North-South	Present	<i>srj</i>	670.558*****	449 647.635*****	670.558*****	
WIOD 2016	North-South	Present	<i>t</i>	150.747*****	22 724.547*****	150.747*****	
WIOD 2016	North-South	Present	<i>srj & t</i>	580.750*****	472 372.182*****	174.521*****	472 372.182*****
WIOD 2016	Full sample	Present	<i>srj</i>	1 652.904*****	2 732 092.600*****	1 652.904*****	
WIOD 2016	Full sample	Present	<i>t</i>	554.596*****	307 576.277*****	554.596*****	
WIOD 2016	Full sample	Present	<i>srj & t</i>	1 560.938*****	3 039 668.876*****	578.416*****	3 039 668.876*****
WIOD 2013	North-South	Present	<i>srj</i>	618.833*****	382 954.497*****	618.833*****	
WIOD 2013	North-South	Present	<i>t</i>	80.231*****	6 436.942*****	80.231*****	
WIOD 2013	North-South	Present	<i>srj & t</i>	494.313*****	389 391.439*****	106.438*****	389 391.439*****
WIOD 2013	Full sample	Present	<i>srj</i>	1 462.229*****	2 138 115.094*****	1 462.229*****	
WIOD 2013	Full sample	Present	<i>t</i>	365.754*****	133 776.123*****	365.754*****	
WIOD 2013	Full sample	Present	<i>srj & t</i>	1 292.580*****	2 271 891.217*****	392.000*****	2 271 891.217*****
WIOD 2016	North-South	Absent	<i>srj</i>	673.602*****	453 739.490*****	673.602*****	
WIOD 2016	North-South	Absent	<i>t</i>	163.710*****	26 801.069*****	163.710*****	
WIOD 2016	North-South	Absent	<i>srj & t</i>	592.069*****	480 540.558*****	187.585*****	480 540.558*****
WIOD 2016	Full sample	Absent	<i>srj</i>	1 653.949*****	2 735 546.616*****	1 653.949*****	
WIOD 2016	Full sample	Absent	<i>t</i>	571.463*****	326 570.123*****	571.463*****	
WIOD 2016	Full sample	Absent	<i>srj & t</i>	1 573.604*****	3 062 116.739*****	595.297*****	3 062 116.739*****

Significance levels: * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$; **** $p < 0.005$; ***** $p < 0.001$. *LM* means Lagrange Multiplier. Independent variables are included with $t-1$ lags. *srj* indicates the combined dimensions of fixed effects for source country s , recipient country r and recipient sector j characteristics; t denotes the dimension of the individual time fixed effects; *srj & t* indicates the two-way fixed effects model. All alternative fixed effects specifications (H1) are tested against a pooled specification (H0)

References of Appendix A

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B Robustness checks and sectoral regressions

B.1 Robustness: no interaction term

Table B1
Results excluding the interaction term with outp. shares using the WIOD 2016

Fixed effects	Dep. var.: relative distance between sectoral output shares $\ln(dz_{srjt})$					
	North-South			Full sample		
	srj	t	$srj \& t$	srj	t	$srj \& t$
Import intensity $\ln(m_{srj(t-1)})$	-0.00856* (0.00457)	-0.07970***** (0.00480)	-0.02916***** (0.00478)	-0.00422* (0.00216)	-0.05733***** (0.00218)	-0.01868***** (0.00229)
Capital-to-labor ratio $\ln(dk_{srj(t-1)})$	-0.00564 (0.00573)	0.13310***** (0.01516)	-0.00141 (0.00570)	0.00711***** (0.00193)	0.08094***** (0.00320)	0.00879***** (0.00192)
Num. of observat.	143435	143435	143435	871733	871733	871733
Degrees of freedom	133175	143419	133162	809345	871717	809332
R^2	0.00010	0.02748	0.00088	0.00008	0.01814	0.00039
F -stat.	2.279	171.688*****	18.603*****	8.627*****	723.199*****	43.524*****

Significance levels: * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$; **** $p < 0.005$; ***** $p < 0.001$. Robust standard errors clustered at the srj -level are reported in parentheses. srj indicates the combined dimensions of fixed effects for source country s , recipient country r and recipient sector j characteristics; t denotes the dimension of the individual time fixed effects; $srj \& t$ indicates the two-way fixed effects model.

Table B2
Results excluding the interaction term with CO₂ shares using the WIOD 2016

Fixed effects	Dep. var.: relative distance between sectoral CO ₂ shares $\ln(dz_{srjt})$					
	North-South			Full sample		
	srj	t	$srj \& t$	srj	t	$srj \& t$
Import intensity $\ln(m_{srj(t-1)})$	0.04244***** (0.00670)	-0.11040***** (0.00520)	0.00896 (0.00691)	0.03662***** (0.00302)	-0.08937***** (0.00243)	0.00108 (0.00321)
Capital-to-labor ratio $\ln(dk_{srj(t-1)})$	0.00273 (0.00836)	0.07693***** (0.01550)	0.01013 (0.00826)	-0.00130 (0.00260)	0.04783***** (0.00356)	0.00321 (0.00260)
Num. of observat.	143241	143241	143241	869736	869736	869736
Degrees of freedom	132981	143225	132968	807352	869720	807339
R^2	0.00125	0.03826	0.00008	0.00080	0.02058	0.00001
F -stat.	20.171*****	228.727*****	1.667	74.030*****	826.150*****	0.824

Significance levels: * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$; **** $p < 0.005$; ***** $p < 0.001$. See above for notes.

B.2 Robustness: different time lags

Table B3
Results with $t - 2$ time lags and output shares using the WIOD 2016

Fixed effects	Dep. var.: relative distance between sectoral output shares $\ln(dz_{srjt})$					
	North-South			Full sample		
	srj	t	$srj \& t$	srj	t	$srj \& t$
Import intensity	-0.00111	-0.08419****	-0.02144****	0.00263	-0.05774****	-0.01237****
$\ln(m_{srj(t-2)})$	(0.00464)	(0.00521)	(0.00482)	(0.00215)	(0.00222)	(0.00228)
Capital-to-labor ratio	-0.02926	-0.03768	-0.01727	-0.01809****	0.02093*	-0.01664****
$\ln(dk_{srj(t-2)})$	(0.01856)	(0.04511)	(0.01852)	(0.00574)	(0.01106)	(0.00573)
Interaction term						
$\ln(m_{srj(t-2)}) \cdot$	-0.00193	-0.01717****	-0.00114	-0.00307****	-0.00722****	-0.00312****
$\ln(dk_{srj(t-2)})$	(0.00185)	(0.00479)	(0.00185)	(0.00071)	(0.00139)	(0.00071)
Num. of observat.	133190	133190	133190	809491	809491	809491
Degrees of freedom	122929	133174	122917	747102	809475	747090
R^2	0.00009	0.02719	0.00050	0.00013	0.01791	0.00031
F -stat.	1.533	111.298****	6.917****	9.594****	467.532****	21.335****

Significance levels: * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$; **** $p < 0.005$; ***** $p < 0.001$. See before for notes.

Table B4
Results with $t - 3$ time lags and output shares using the WIOD 2016

Fixed effects	Dep. var.: relative distance between sectoral output shares $\ln(dz_{srjt})$					
	North-South			Full sample		
	srj	t	$srj \& t$	srj	t	$srj \& t$
Import intensity	0.00553	-0.08322****	-0.01429****	0.00764****	-0.05711****	-0.00772****
$\ln(m_{srj(t-3)})$	(0.00470)	(0.00527)	(0.00488)	(0.00214)	(0.00224)	(0.00228)
Capital-to-labor ratio	-0.02173	-0.04191	-0.01048	-0.01814****	0.02209**	-0.01622****
$\ln(dk_{srj(t-3)})$	(0.01856)	(0.04661)	(0.01854)	(0.00567)	(0.01123)	(0.00566)
Interaction term						
$\ln(m_{srj(t-3)}) \cdot$	-0.00081	-0.01762****	-0.00008	-0.00241****	-0.00679****	-0.00240****
$\ln(dk_{srj(t-3)})$	(0.00177)	(0.00494)	(0.00177)	(0.00069)	(0.00141)	(0.00069)
Num. of observat.	122945	122945	122945	747249	747249	747249
Degrees of freedom	112684	122930	112673	684860	747234	684849
R^2	0.00017	0.02675	0.00028	0.00011	0.01737	0.00012
F -stat.	2.407*	106.163****	3.811***	8.155****	444.959****	8.462****

Significance levels: * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$; **** $p < 0.005$; ***** $p < 0.001$. See before for notes.

Table B5
Results with $t - 2$ time lags and CO₂ shares using the WIOD 2016

	Dep. var.: relative distance between sectoral CO ₂ shares $\ln(dz_{srjt})$					
	North-South			Full sample		
	<i>srj</i>	<i>t</i>	<i>srj</i> & <i>t</i>	<i>srj</i>	<i>t</i>	<i>srj</i> & <i>t</i>
Fixed effects						
Import intensity	0.04982*****	-0.11596*****	0.01596**	0.04280*****	-0.08967*****	0.00735**
$\ln(m_{srj(t-2)})$	(0.00709)	(0.00590)	(0.00725)	(0.00316)	(0.00249)	(0.00333)
Capital-to-labor ratio	-0.06480***	-0.07718	-0.04314*	0.01248	0.04174*****	0.01739**
$\ln(dk_{srj(t-2)})$	(0.02444)	(0.05012)	(0.02425)	(0.00777)	(0.01241)	(0.00775)
Interaction term						
$\ln(m_{srj(t-2)}) \cdot$	-0.00658***	-0.01522***	-0.00510**	0.00122	-0.00073	0.00125
$\ln(dk_{srj(t-2)})$	(0.00255)	(0.00566)	(0.00253)	(0.00098)	(0.00157)	(0.00097)
Num. of observat.	132987	132987	132987	807418	807418	807418
Degrees of freedom	122726	132971	122714	745033	807402	745021
R^2	0.00208	0.03859	0.00031	0.00109	0.02072	0.00006
F -stat.	20.026*****	152.480*****	3.329**	61.609*****	576.981*****	5.248****

Significance levels: * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$; **** $p < 0.005$; ***** $p < 0.001$. See before for notes.

Table B6
Results with $t - 3$ time lags and CO₂ shares using the WIOD 2016

	Dep. var.: relative distance between sectoral CO ₂ shares $\ln(dz_{srjt})$					
	North-South			Full sample		
	<i>srj</i>	<i>t</i>	<i>srj</i> & <i>t</i>	<i>srj</i>	<i>t</i>	<i>srj</i> & <i>t</i>
Fixed effects						
Import intensity	0.06004*****	-0.11657*****	0.02328****	0.05258*****	-0.09003*****	0.01473*****
$\ln(m_{srj(t-3)})$	(0.00721)	(0.00599)	(0.00725)	(0.00326)	(0.00250)	(0.00339)
Capital-to-labor ratio	-0.07812****	-0.08330	-0.05871**	0.01794**	0.04097****	0.02269****
$\ln(dk_{srj(t-3)})$	(0.02533)	(0.05227)	(0.02507)	(0.00803)	(0.01273)	(0.00802)
Interaction term						
$\ln(m_{srj(t-3)}) \cdot$	-0.00723***	-0.01564***	-0.00599**	0.00147	-0.00093	0.00150
$\ln(dk_{srj(t-3)})$	(0.00264)	(0.00589)	(0.00261)	(0.00101)	(0.00161)	(0.00101)
Num. of observat.	122751	122751	122751	745260	745260	745260
Degrees of freedom	112490	122736	112479	682875	745245	682864
R^2	0.00303	0.03890	0.00054	0.00165	0.02106	0.00017
F -stat.	28.153*****	150.476*****	5.933*****	87.668*****	577.039*****	12.792*****

Significance levels: * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$; **** $p < 0.005$; ***** $p < 0.001$. See before for notes.

B.3 Robustness: energy shares

Table B7
Results with energy shares using the WIOD 2016

	Dep. var.: Relative distance between sectoral energy shares $\ln(dz_{srjt})$					
	North-South			Full sample		
	<i>srj</i>	<i>t</i>	<i>srj & t</i>	<i>srj</i>	<i>t</i>	<i>srj & t</i>
Fixed effects						
Import intensity	-0.00855*	-0.09061*****	0.00029	-0.01569*****	-0.05603*****	0.00390
$\ln(m_{srj(t-1)})$	(0.00471)	(0.00551)	(0.00499)	(0.00216)	(0.00243)	(0.00238)
Capital-to-labor ratio	-0.00133	0.02186	-0.00097	0.00555	0.00292	0.00441
$\ln(dk_{srj(t-1)})$	(0.01920)	(0.04341)	(0.01916)	(0.00588)	(0.01157)	(0.00588)
Interaction term						
$\ln(m_{srj(t-1)}) \cdot$	-0.00054	-0.00611	-0.00044	0.00159**	-0.00469****	0.00158**
$\ln(dk_{srj(t-1)})$	(0.00191)	(0.00480)	(0.00191)	(0.00072)	(0.00145)	(0.00072)
Num. of observat.	143444	143444	143444	871719	871719	871719
Degrees of freedom	133183	143427	133170	809334	871702	809321
R^2	0.00008	0.02724	0.00001	0.00024	0.00979	0.00009
<i>F</i> -stat.	1.225	111.384*****	0.111	22.283*****	249.368*****	7.909*****

Significance levels: * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$; **** $p < 0.005$; ***** $p < 0.001$. See before for notes.

Table B8
Results with energy shares using the WIOD 2013

	Dep. var. relative distance between sectoral energy shares $\ln(dz_{srjt})$					
	North-South			Full sample		
	<i>srj</i>	<i>t</i>	<i>srj & t</i>	<i>srj</i>	<i>t</i>	<i>srj & t</i>
Fixed effects						
Import intensity	0.00079	-0.08180*****	-0.01053	0.00204	-0.06467*****	-0.00689**
$\ln(m_{srj(t-1)})$	(0.00676)	(0.00656)	(0.00723)	(0.00309)	(0.00308)	(0.00329)
Capital-to-labor ratio	0.00773	-0.08222	0.00332	-0.00930	0.00702	-0.01176
$\ln(dk_{srj(t-1)})$	(0.03324)	(0.06247)	(0.03318)	(0.00948)	(0.01547)	(0.00947)
Interaction term						
$\ln(m_{srj(t-1)}) \cdot$	0.00123	-0.01684***	0.00089	-0.00003	-0.00590****	-0.00052
$\ln(dk_{srj(t-1)})$	(0.00310)	(0.00649)	(0.00310)	(0.00117)	(0.00201)	(0.00117)
Num. of observat.	97657	97657	97657	540775	540775	540775
Degrees of freedom	90418	97640	90405	500373	540758	500360
R^2	0.00001	0.02154	0.00011	0.00004	0.01529	0.00006
<i>F</i> -stat.	0.125	55.675*****	0.954	2.356*	244.154*****	3.356**

Significance levels: * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$; **** $p < 0.005$; ***** $p < 0.001$. See before for notes.

B.4 Sectoral regression results

Table B9
Sectoral results with output shares, sr fixed effects, WIOD 2016 full sample

Sector	CO ₂ share 2000 and 2014	Output share 2000 and 2014	CO ₂ intensity in 2014	Import intensity $\ln(m_{srj(t-1)})$	Capital-to-labor ratio $\ln(dk_{srj(t-1)})$	Interaction term $\ln(m_{srj(t-1)}) \cdot \ln(dk_{srj(t-1)})$	Num. of obs.	R ²	Num. of fix.-ef.	F-stat. p-value
D35-Energy	41.67% and 41.90%	2.36% and 2.38%	3.5285	-0.05641**** (0.00966)	0.10927**** (0.03356)	0.00874** (0.00357)	25284	0.0054	1806	15.051**** <0.001
C23-Minerals	7.63% and 11.04%	0.92% and 1.19%	1.8512	0.03798** (0.01647)	0.03845 (0.03799)	0.01032** (0.00511)	25284	0.0029	1806	6.120**** <0.001
H51-Air transport	3.62% and 2.42%	0.61% and 0.47%	1.0354	0.00770 (0.01037)	-0.04147 (0.02572)	-0.00769** (0.00311)	25284	0.0011	1806	3.646** 0.012
H50-Water transport	2.92% and 1.96%	0.42% and 0.41%	0.9648	0.00141 (0.00981)	-0.01228 (0.02854)	-0.00630* (0.00369)	25284	0.0027	1806	6.393**** <0.001
C24-Metal	8.31% and 10.33%	2.28% and 2.86%	0.7235	-0.02312*** (0.00856)	-0.00339 (0.01986)	-0.00462* (0.00245)	25284	0.0040	1806	9.618**** <0.001
C20-Chemicals	5.01% and 5.20%	2.34% and 2.71%	0.3844	0.02280 (0.01683)	-0.09182** (0.03658)	-0.02020**** (0.00506)	25284	0.0069	1806	11.533**** <0.001
E37-E39-Waste	0.46% and 0.55%	0.45% and 0.36%	0.3075	0.02415 (0.01776)	-0.11023** (0.05068)	-0.00780 (0.00529)	19684	0.0027	1406	2.898** 0.034
B-Mining	3.36% and 3.84%	2.22% and 2.50%	0.3066	-0.00162 (0.00877)	-0.01702 (0.02715)	-0.00507 (0.00335)	25284	0.0017	1806	3.224** 0.022
H49-Land transport	3.74% and 3.62%	2.49% and 2.37%	0.3057	-0.02917**** (0.00932)	0.10088*** (0.03638)	0.01052*** (0.00400)	25284	0.0027	1806	5.734**** <0.001
A02-Forestry	0.19% and 0.25%	0.24% and 0.19%	0.2612	-0.01257 (0.00968)	0.03307 (0.03099)	0.00314 (0.00343)	22960	0.0004	1640	0.901 0.440
C19-Refined Petr.	3.42% and 2.79%	2.42% and 2.41%	0.2318	-0.00758 (0.01122)	0.02842 (0.02469)	0.00265 (0.00309)	22728	0.0002	1722	0.653 0.581
C17-Paper	0.94% and 0.60%	0.83% and 0.69%	0.1749	0.00271 (0.01269)	-0.07327** (0.03027)	-0.01324**** (0.00407)	25284	0.0020	1806	4.726**** 0.003
C22-Rubber	0.25% and 0.88%	1.11% and 1.14%	0.1543	0.04954**** (0.01461)	-0.04295 (0.03222)	-0.00281 (0.00434)	25284	0.0022	1806	5.987**** <0.001
A01-Agriculture	2.22% and 1.87%	2.74% and 2.43%	0.1538	-0.03007**** (0.00884)	-0.01064 (0.02691)	-0.00197 (0.00318)	25284	0.0016	1806	4.284**** 0.005
E36-Water	0.20% and 0.13%	0.23% and 0.18%	0.1432	-0.00226 (0.01348)	-0.00909** (0.03895)	-0.00402 (0.00423)	22960	0.0030	1640	6.745**** <0.001
A03-Fisheries	0.15% and 0.13%	0.20% and 0.20%	0.1309	-0.00304 (0.00769)	-0.02037 (0.02393)	-0.00159 (0.00289)	22960	0.0001	1640	0.517 0.670
C31-C32-Furniture	0.20% and 0.33%	1.06% and 0.82%	0.0811	-0.00736 (0.01703)	-0.10274** (0.04226)	-0.01596**** (0.00546)	25284	0.0019	1806	3.206** 0.022
C16-Wood	0.25% and 0.22%	0.53% and 0.58%	0.0747	0.02448** (0.00961)	0.01142 (0.02436)	0.00350 (0.00313)	25284	0.0014	1806	4.216**** 0.006
H52-Warehousing	0.37% and 0.37%	0.98% and 1.06%	0.0701	-0.04308**** (0.01129)	-0.08582* (0.04455)	-0.00694 (0.00526)	25284	0.0033	1806	7.219**** <0.001
H53-Post	0.10% and 0.08%	0.36% and 0.25%	0.0605	-0.10666**** (0.01455)	0.05054 (0.04383)	0.00250 (0.00469)	19684	0.0145	1406	20.727**** <0.001
I-Accommodation	0.80% and 0.64%	2.75% and 2.40%	0.0539	0.00232 (0.01282)	-0.07985* (0.04446)	-0.01286** (0.00517)	25284	0.0017	1806	3.796**** 0.010
C13-C15-Textile	0.76% and 0.44%	1.45% and 1.63%	0.0535	0.05015**** (0.01286)	-0.00636 (0.03716)	-0.00049 (0.00493)	25284	0.0032	1806	5.710**** <0.001
C10-C12-Food	1.37% and 1.15%	4.05% and 4.35%	0.0528	0.02568** (0.01291)	-0.00759 (0.03587)	0.00178 (0.00460)	25284	0.0013	1806	2.824** 0.037
C25-Non machinery	0.37% and 0.43%	1.80% and 1.65%	0.0515	-0.03843** (0.01529)	-0.10442*** (0.03753)	-0.01042** (0.00487)	24108	0.0021	1722	5.113**** 0.002
OPQRS-Public Services	4.35% and 3.64%	16.12% and 14.56%	0.0500	0.16850**** (0.01608)	0.12116* (0.06449)	0.01375* (0.00704)	25284	0.0268	1806	37.267**** <0.001
F-Construction	1.30% and 1.17%	7.37% and 6.95%	0.0336	0.11834**** (0.01607)	-0.09621** (0.04601)	-0.00997* (0.00562)	25284	0.0082	1806	20.023**** <0.001
C18-Printing	0.17% and 0.05%	0.49% and 0.35%	0.0314	-0.03159** (0.01566)	-0.05187 (0.03312)	-0.00859** (0.00417)	24108	0.0010	1722	2.396* 0.066
C28-Machinery	0.44% and 0.35%	1.89% and 2.27%	0.0310	0.02388* (0.01434)	-0.01628 (0.03761)	-0.00604 (0.00469)	25284	0.0015	1806	2.973** 0.031
G-Trade	2.30% and 1.43%	9.86% and 9.24%	0.0309	-0.08608**** (0.01602)	-0.01120 (0.05227)	-0.00453 (0.00604)	25284	0.0061	1806	11.105**** <0.001
C30-Transport equip.	0.16% and 0.12%	0.67% and 0.97%	0.0239	-0.03916**** (0.00952)	0.01349 (0.02354)	0.00293 (0.00298)	24026	0.0019	1722	6.220**** <0.001
C27-Electrical equip.	0.18% and 0.17%	1.06% and 1.47%	0.0230	-0.05994**** (0.01698)	-0.10154*** (0.03915)	-0.01318** (0.00512)	24108	0.0039	1722	6.043**** <0.001
C33-Repair	0.04% and 0.02%	0.28% and 0.23%	0.0196	-0.09223**** (0.01777)	0.01825 (0.04428)	0.00195 (0.00539)	14656	0.0058	1056	9.530**** <0.001
JKLMN-Private Services	2.15% and 1.54%	22.51% and 22.07%	0.0140	0.01590* (0.00947)	-0.02957 (0.04190)	-0.00514 (0.00441)	25284	0.0012	1806	2.424* 0.064
C29-Vehicles	0.33% and 0.20%	2.49% and 3.02%	0.0136	0.00371 (0.01266)	-0.05204** (0.02393)	-0.00955**** (0.00327)	25284	0.0014	1806	4.049**** 0.007
C26-Computers	0.23% and 0.13%	1.53% and 2.66%	0.0096	-0.01368 (0.01903)	-0.06396* (0.03362)	-0.01071** (0.00468)	25284	0.0014	1806	2.333* 0.072
C21-Pharma.	0.04% and 0.03%	0.72% and 0.84%	0.0078	0.05409**** (0.01294)	0.01805 (0.02897)	0.00158 (0.00342)	22880	0.0031	1640	6.045**** <0.001

Significance levels: * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$; **** $p < 0.005$; ***** $p < 0.001$. Robust standard errors clustered at the sr -level are reported in parentheses. sr indicates the combined dimensions of fixed effects for source country s and recipient country r .

Table B10
Sectoral results with output shares, t fixed effects, WIOD 2016 full sample

Sector	CO ₂ share 2000 and 2014	Output share 2000 and 2014	CO ₂ intensity in 2014	Import intensity $\ln(m_{sr,t-1})$	Capital-to-labor ratio $\ln(dk_{sr,t-1})$	Interaction term $\ln(m_{sr,t-1}) \cdot \ln(dk_{sr,t-1})$	Num. of obs.	R ²	Num. of fix.-ef.	F-stat.	p-value
D35-Energy	41.67% and 41.90%	2.36% and 2.38%	3.5285	-0.04334**** (0.00763)	0.09905** (0.03936)	0.01101*** (0.00411)	25284	0.0098	14	11.141*****	<0.001
C23-Minerals	7.63% and 11.04%	0.92% and 1.19%	1.8512	0.01467 (0.00897)	0.02108 (0.04539)	0.00514 (0.00572)	25284	0.0011	14	1.578	0.193
H51-Air transport	3.62% and 2.42%	0.61% and 0.47%	1.0354	-0.04797***** (0.00880)	0.24485***** (0.04264)	0.01561**** (0.00517)	25284	0.0373	14	43.633*****	<0.001
H50-Water transport	2.92% and 1.96%	0.42% and 0.41%	0.9648	0.00223 (0.01231)	0.39265***** (0.06959)	0.03989***** (0.00769)	25284	0.0133	14	10.756*****	<0.001
C24-Metal	8.31% and 10.33%	2.28% and 2.86%	0.7235	-0.22146***** (0.01551)	-0.02754 (0.05655)	-0.01910** (0.00742)	25284	0.1488	14	89.863*****	<0.001
C20-Chemicals	5.01% and 5.20%	2.34% and 2.71%	0.3844	-0.17223***** (0.01096)	-0.00867 (0.04630)	-0.00116 (0.00645)	25284	0.1027	14	89.846*****	<0.001
E37-E39-Waste	0.46% and 0.55%	0.45% and 0.36%	0.3075	-0.09170***** (0.01172)	-0.03951 (0.06362)	-0.01292* (0.00766)	19684	0.0303	14	23.086*****	<0.001
B-Mining	3.36% and 3.84%	2.22% and 2.50%	0.3066	-0.10911***** (0.01405)	0.07089 (0.07077)	-0.00462 (0.00791)	25284	0.0470	14	34.683*****	<0.001
H49-Land transport	3.74% and 3.62%	2.49% and 2.37%	0.3057	-0.05739***** (0.01058)	0.13736** (0.05034)	0.01656**** (0.00541)	25284	0.0163	14	13.633*****	<0.001
A02-Forestry	0.19% and 0.25%	0.24% and 0.19%	0.2612	-0.02508* (0.01349)	-0.15543*** (0.05815)	-0.01353** (0.00658)	22960	0.0042	14	5.018****	0.002
C19-Refined Petr.	3.42% and 2.79%	2.42% and 2.41%	0.2318	-0.13351***** (0.01473)	-0.01596 (0.05778)	-0.02669***** (0.00675)	22728	0.1067	14	71.461*****	<0.001
C17-Paper	0.94% and 0.60%	0.83% and 0.69%	0.1749	-0.10016***** (0.01148)	0.04704 (0.06046)	-0.00279 (0.00771)	25284	0.0446	14	34.881*****	<0.001
C22-Rubber	0.25% and 0.88%	1.11% and 1.14%	0.1543	-0.06882***** (0.00958)	0.14147***** (0.04229)	0.02706***** (0.00556)	25284	0.0304	14	30.640*****	<0.001
A01-Agriculture	2.22% and 1.87%	2.74% and 2.43%	0.1538	-0.04839***** (0.01210)	0.16024***** (0.03862)	0.01868***** (0.00506)	25284	0.0120	14	10.238*****	<0.001
E36-Water	0.20% and 0.13%	0.23% and 0.18%	0.1432	0.00557 (0.01056)	-0.00243 (0.05006)	0.00513 (0.00528)	22960	0.0046	14	5.505*****	<0.001
A03-Fisheries	0.15% and 0.13%	0.20% and 0.20%	0.1309	-0.06044***** (0.01605)	-0.19152***** (0.06094)	-0.01260 (0.00795)	22960	0.0201	14	23.648*****	<0.001
C31-C32-Furniture	0.20% and 0.33%	1.06% and 0.82%	0.0811	-0.05137***** (0.01115)	0.12221** (0.05620)	0.01112 (0.00743)	25284	0.0126	14	12.856*****	<0.001
C16-Wood	0.25% and 0.22%	0.53% and 0.58%	0.0747	-0.05957***** (0.01152)	0.07876 (0.05260)	0.01623** (0.00717)	25284	0.0147	14	10.883*****	<0.001
H52-Warehousing	0.37% and 0.37%	0.98% and 1.06%	0.0701	-0.01767** (0.00881)	0.32972***** (0.05690)	0.01112* (0.00639)	25284	0.0881	14	84.879*****	<0.001
H53-Post	0.10% and 0.08%	0.36% and 0.25%	0.0605	-0.03249*** (0.00992)	0.06042 (0.04539)	0.00251 (0.00505)	19684	0.0062	14	6.103*****	<0.001
I-Accommodation	0.80% and 0.64%	2.75% and 2.40%	0.0539	-0.04907***** (0.01086)	-0.05217 (0.06993)	-0.00189 (0.00791)	25284	0.0091	14	8.399*****	<0.001
C13-C15-Textile	0.76% and 0.44%	1.45% and 1.63%	0.0535	-0.05297***** (0.01105)	0.03393 (0.04405)	0.01413** (0.00604)	25284	0.0157	14	13.558*****	<0.001
C10-C12-Food	1.37% and 1.15%	4.05% and 4.35%	0.0528	-0.05353***** (0.01125)	-0.09155* (0.05485)	-0.01312* (0.00676)	25284	0.0087	14	7.788*****	<0.001
C25-Non machinery	0.37% and 0.43%	1.80% and 1.65%	0.0515	-0.06579***** (0.00921)	-0.01658 (0.04567)	0.00218 (0.00587)	24108	0.0191	14	18.784*****	<0.001
OPQRS-Public Services	4.35% and 3.64%	16.12% and 14.56%	0.0500	-0.00005 (0.00979)	0.41014***** (0.06465)	0.02631***** (0.00708)	25284	0.0594	14	57.385*****	<0.001
F-Construction	1.30% and 1.17%	7.37% and 6.95%	0.0336	-0.01093 (0.00920)	-0.00033 (0.05052)	-0.00063 (0.00628)	25284	0.0004	14	0.529	0.663
C18-Printing	0.17% and 0.05%	0.49% and 0.35%	0.0314	-0.07195***** (0.00949)	-0.10609** (0.04776)	-0.01301** (0.00573)	24108	0.0157	14	19.364*****	<0.001
C28-Machinery	0.44% and 0.35%	1.89% and 2.27%	0.0310	-0.12850***** (0.01220)	0.03849 (0.05538)	-0.00031 (0.00751)	25284	0.0576	14	44.412*****	<0.001
G-Trade	2.30% and 1.43%	9.86% and 9.24%	0.0309	0.01568 (0.01012)	0.35655***** (0.06102)	0.03522***** (0.00694)	25284	0.0130	14	13.236*****	<0.001
C30-Transport equip.	0.16% and 0.12%	0.67% and 0.97%	0.0239	-0.16225***** (0.01423)	0.00903 (0.06325)	-0.00868 (0.00846)	24026	0.0798	14	58.545*****	<0.001
C27-Electrical equip.	0.18% and 0.17%	1.06% and 1.47%	0.0230	-0.11263***** (0.01183)	0.02180 (0.04401)	0.01123* (0.00618)	24108	0.0444	14	34.478*****	<0.001
C33-Repair	0.04% and 0.02%	0.28% and 0.23%	0.0196	-0.07785***** (0.01386)	0.03693 (0.06270)	0.01172 (0.00743)	14656	0.0180	14	11.219*****	<0.001
JKLMN-Private Services	2.15% and 1.54%	22.51% and 22.07%	0.0140	-0.00049 (0.01142)	0.33077***** (0.07599)	0.00969 (0.00852)	25284	0.0796	14	66.143*****	<0.001
C29-Vehicles	0.33% and 0.20%	2.49% and 3.02%	0.0136	-0.24186***** (0.01247)	-0.20807***** (0.05651)	-0.02784***** (0.00761)	25284	0.1307	14	131.440*****	<0.001
C26-Computers	0.23% and 0.13%	1.53% and 2.66%	0.0096	-0.16018***** (0.01368)	-0.04529 (0.05217)	-0.00240 (0.00739)	25284	0.0673	14	45.834*****	<0.001
C21-Pharma.	0.04% and 0.03%	0.72% and 0.84%	0.0078	-0.05188***** (0.01224)	-0.02398 (0.06773)	-0.01128 (0.00817)	22880	0.0143	14	9.936*****	<0.001

Significance levels: * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$; **** $p < 0.005$; ***** $p < 0.001$. Robust standard errors clustered at the sr -level are reported in parentheses. t denotes the dimension of the individual time fixed effects.

Table B13
Sectoral results with output shares, t fixed effects, WIOD 2013 full sample

Sector	CO ₂ share		Output share		CO ₂ intensity in 2014	Import intensity $\ln(m_{srjt(t-1)})$	Capital-to-labor ratio $\ln(dk_{srjt(t-1)})$	Interaction term $\ln(m_{srjt(t-1)}) \cdot \ln(dk_{srjt(t-1)})$	Num. of obs.	R^2	Num. of fix.-ef.	F-stat. p -value
	2000 and 2014	2000 and 2014	2000 and 2014	2000 and 2014								
D35-Energy	39.77% and 46.58%	2.43% and 2.21%	5.6674	-0.00286 (0.00850)	0.08579** (0.03818)	0.00311 (0.00409)	20918	0.0069	14	8.931***** <0.001		
C23-Minerals	7.05% and 7.09%	1.14% and 1.05%	1.8039	0.02941**** (0.00981)	0.07637 (0.04685)	0.01335** (0.00547)	20917	0.0059	14	7.579***** <0.001		
H51-Air transport	2.64% and 2.86%	0.53% and 0.45%	1.7024	-0.03630**** (0.00974)	0.27343**** (0.05380)	0.02018**** (0.00596)	20905	0.0264	14	22.885***** <0.001		
H50-Water transport	1.71% and 2.19%	0.33% and 0.44%	1.3342	-0.01418 (0.01378)	0.01129 (0.08292)	-0.00828 (0.01046)	20895	0.0069	14	4.233*** 0.005		
B-Mining	3.43% and 4.02%	1.14% and 0.94%	1.1501	-0.04110** (0.01464)	0.17399** (0.06742)	0.01612** (0.00809)	20914	0.0087	14	6.710***** <0.001		
C19-Refined Petr.	3.90% and 3.51%	1.21% and 1.39%	0.6808	-0.12802**** (0.01595)	-0.14875**** (0.04918)	-0.02852**** (0.00601)	18990	0.0841	14	50.718***** <0.001		
H49-Land transport	4.08% and 4.00%	2.40% and 2.11%	0.5103	-0.03521**** (0.01127)	0.13107** (0.05754)	0.00994 (0.00666)	20895	0.0109	14	9.253***** <0.001		
C24-Metal	8.43% and 7.56%	4.19% and 4.03%	0.5040	-0.12426**** (0.01176)	0.09708* (0.05500)	0.01843** (0.00733)	20916	0.0630	14	44.550***** <0.001		
C20-Chemicals	4.98% and 3.92%	3.01% and 3.08%	0.3421	-0.12816**** (0.01193)	-0.10473 (0.06424)	-0.00627 (0.00872)	20912	0.0570	14	41.305***** <0.001		
A01-Agriculture	3.02% and 2.24%	3.23% and 2.76%	0.2175	0.02377** (0.01207)	0.09459** (0.04299)	0.00884* (0.00514)	20917	0.0034	14	4.623*** 0.003		
C17-Paper	1.40% and 1.00%	2.29% and 1.74%	0.1546	-0.02981** (0.01175)	0.13391* (0.07204)	0.00559 (0.00874)	20918	0.0158	14	10.624***** <0.001		
H52-Warehousing	0.43% and 0.67%	1.14% and 1.16%	0.1543	0.00472 (0.01036)	0.33342**** (0.06345)	0.02916**** (0.00711)	20918	0.0187	14	12.465***** <0.001		
C16-Wood	0.29% and 0.28%	0.76% and 0.61%	0.1252	-0.07481**** (0.01300)	0.04254 (0.05649)	0.01407* (0.00738)	20896	0.0210	14	14.228***** <0.001		
C10-C12-Food	1.88% and 1.65%	4.56% and 4.03%	0.1100	-0.01908* (0.01105)	-0.01693 (0.06299)	-0.00232 (0.00795)	20920	0.0012	14	1.033 0.377		
OPQRS-Public Services	4.42% and 3.04%	9.74% and 8.78%	0.0929	-0.01898* (0.00987)	0.37286**** (0.06191)	0.02900** (0.00733)	20920	0.0925	14	73.815***** <0.001		
I-Accommodation	1.04% and 0.79%	2.78% and 2.37%	0.0892	-0.05756**** (0.01169)	-0.05339 (0.07535)	-0.01021 (0.00827)	20922	0.0126	14	9.275***** <0.001		
C13-C15-Textile	1.05% and 0.62%	2.01% and 1.87%	0.0887	0.02230** (0.01109)	0.08020* (0.04201)	0.01496** (0.00551)	20916	0.0045	14	6.821***** <0.001		
F-Construction	1.52% and 1.35%	7.24% and 5.78%	0.0627	-0.01627* (0.00925)	-0.12078** (0.05062)	-0.01222* (0.00642)	20918	0.0029	14	3.268** 0.021		
C33-Repair	0.29% and 0.16%	0.89% and 0.77%	0.0571	-0.02776** (0.01002)	-0.06014 (0.04199)	-0.00733 (0.00526)	20921	0.0030	14	3.819** 0.010		
C22-Rubber	0.42% and 0.28%	1.27% and 1.34%	0.0561	-0.06928**** (0.00919)	0.11078** (0.04400)	0.01616*** (0.00591)	20915	0.0235	14	22.793***** <0.001		
G-Trade	2.57% and 1.65%	10.94% and 10.00%	0.0443	0.01316 (0.00908)	0.33565**** (0.06490)	0.02908**** (0.00763)	20919	0.0211	14	16.520***** <0.001		
C26-Computers	0.65% and 0.42%	2.58% and 2.67%	0.0425	-0.13279**** (0.01208)	0.12740** (0.05817)	0.01751** (0.00779)	20918	0.0677	14	45.474***** <0.001		
H53-Post	0.33% and 0.37%	1.77% and 2.87%	0.0346	-0.00974 (0.00849)	0.50399**** (0.04943)	0.03359**** (0.00553)	20917	0.0820	14	105.330***** <0.001		
C30-Transport equip.	0.56% and 0.47%	3.84% and 3.93%	0.0325	-0.21758**** (0.01353)	-0.23680**** (0.05228)	-0.02339**** (0.00708)	20917	0.1293	14	90.727***** <0.001		
JKLMN-Private Services	3.64% and 2.96%	24.83% and 26.42%	0.0301	0.00559 (0.01055)	0.31611**** (0.06511)	-0.00099 (0.00705)	20920	0.1606	14	137.155***** <0.001		
C27-Electrical equip.	0.49% and 0.33%	3.75% and 7.21%	0.0121	-0.18146**** (0.01373)	-0.11920** (0.05390)	-0.00813 (0.00770)	20921	0.1004	14	58.781***** <0.001		

Significance levels: * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$; **** $p < 0.005$; ***** $p < 0.001$. See before for notes.

Table B14
Sectoral results with output shares, sr & t fixed eff., WIOD 2013 full sample

Sector	CO ₂ share		Output share		CO ₂ intensity in 2014	Import intensity $\ln(m_{srjt(t-1)})$	Capital-to-labor ratio $\ln(dk_{srjt(t-1)})$	Interaction term $\ln(m_{srjt(t-1)}) \cdot \ln(dk_{srjt(t-1)})$	Num. of obs.	R^2	Num. sr -fix.-ef.	Num. t -fix.-ef.	F-stat. p -value
	2000 and 2014	2000 and 2014	2000 and 2014	2000 and 2014									
D35-Energy	39.77% and 46.58%	2.43% and 2.21%	5.6674	-0.03640** (0.01461)	-0.01436 (0.05315)	0.00272 (0.00591)	20918	0.0016	1560	14	2.856** 0.036		
C23-Minerals	7.05% and 7.09%	1.14% and 1.05%	1.8039	0.02767** (0.01408)	0.03170 (0.06110)	0.00467 (0.00641)	20917	0.0006	1560	14	1.610 0.185		
H51-Air transport	2.64% and 2.86%	0.53% and 0.45%	1.7024	-0.05561**** (0.01225)	0.01645 (0.04455)	0.00776 (0.00555)	20905	0.0048	1560	14	9.403*** <0.001		
H50-Water transport	1.71% and 2.19%	0.33% and 0.44%	1.3342	-0.02468** (0.01115)	0.04845 (0.03688)	0.00026 (0.00473)	20895	0.0030	1560	14	5.663*** <0.001		
B-Mining	3.43% and 4.02%	1.14% and 0.94%	1.1501	-0.05679**** (0.01235)	0.00028 (0.04490)	0.00071 (0.00539)	20914	0.0031	1560	14	7.341*** <0.001		
C19-Refined Petr.	3.90% and 3.51%	1.21% and 1.39%	0.6808	-0.04001**** (0.01341)	0.04695 (0.04544)	0.00168 (0.00521)	18990	0.0021	1477	14	4.270*** 0.005		
H49-Land transport	4.08% and 4.00%	2.40% and 2.11%	0.5103	0.02486** (0.01158)	0.01513 (0.04924)	0.00261 (0.00538)	20895	0.0007	1560	14	1.869 0.133		
C24-Metal	8.43% and 7.56%	4.19% and 4.03%	0.5040	-0.01232 (0.01116)	-0.09452** (0.03726)	-0.01041** (0.00485)	20916	0.0009	1560	14	2.340* 0.072		
C20-Chemicals	4.98% and 3.92%	3.01% and 3.08%	0.3421	-0.02360* (0.01342)	-0.01171 (0.03586)	0.00417 (0.00495)	20912	0.0020	1560	14	4.623*** 0.003		
A01-Agriculture	3.02% and 2.24%	3.23% and 2.76%	0.2175	-0.01407 (0.01186)	0.04639 (0.03931)	0.00441 (0.00429)	20917	0.0004	1560	14	0.873 0.455		
C17-Paper	1.40% and 1.00%	2.29% and 1.74%	0.1546	-0.01086 (0.01463)	-0.05724 (0.04334)	-0.01100** (0.00528)	20918	0.0015	1560	14	3.024** 0.029		
H52-Warehousing	0.43% and 0.67%	1.14% and 1.16%	0.1543	0.00300 (0.01162)	0.04433 (0.03989)	0.00226 (0.00445)	20918	0.0006	1560	14	1.557 0.198		
C16-Wood	0.29% and 0.28%	0.76% and 0.61%	0.1252	0.02212* (0.01290)	0.03922 (0.03907)	0.00716* (0.00415)	20896	0.0012	1560	14	3.186** 0.023		
C10-C12-Food	1.88% and 1.65%	4.56% and 4.03%	0.1100	-0.01189 (0.01170)	-0.12345*** (0.03896)	-0.01545**** (0.00463)	20920	0.0011	1560	14	3.958*** 0.008		
OPQRS-Public Services	4.42% and 3.04%	9.74% and 8.78%	0.0929	0.00988 (0.01829)	-0.03505 (0.06274)	-0.00249 (0.00742)	20920	0.0002	1560	14	0.306 0.821		
I-Accommodation	1.04% and 0.79%	2.78% and 2.37%	0.0892	0.01855 (0.01329)	-0.00313 (0.03878)	-0.00424 (0.00425)	20922	0.0013	1560	14	2.956** 0.031		
C13-C15-Textile	1.05% and 0.62%	2.01% and 1.87%	0.0887	-0.02933** (0.01442)	-0.06231 (0.04312)	-0.00894 (0.00581)	20916	0.0012	1560	14	2.186* 0.088		
F-Construction	1.52% and 1.35%	7.24% and 5.78%	0.0627	-0.00433 (0.01721)	0.06697 (0.05685)	0.00010 (0.00673)	20918	0.0018	1560	14	3.505** 0.015		
C33-Repair	0.29% and 0.16%	0.89% and 0.77%	0.0571	0.02569 (0.01916)	-0.05807 (0.05622)	-0.01429** (0.00722)	20921	0.0030	1560	14	3.822*** 0.010		
C22-Rubber	0.42% and 0.28%	1.27% and 1.34%	0.0561	-0.05081** (0.02148)	-0.17336*** (0.05517)	-0.02497**** (0.00820)	20915	0.0034	1560	14	4.710**** 0.003		
G-Trade	2.57% and 1.65%	10.94% and 10.00%	0.0443	-0.01172 (0.01637)	0.08187 (0.05629)	0.00741 (0.00653)	20919	0.0005	1560	14	1.040 0.374		
C26-Computers	0.65% and 0.42%	2.58% and 2.67%	0.0425	0.01923 (0.01349)	-0.04543 (0.03678)	-0.00777 (0.00517)	20918	0.0009	1560	14	1.714 0.162		
H53-Post	0.33% and 0.37%	1.77% and 2.87%	0.0346	-0.00398 (0.01671)	-0.01001 (0.05821)	0.00472 (0.00647)	20917	0.0011	1560	14	2.008 0.111		
C30-Transport equip.	0.56% and 0.47%	3.84% and 3.93%	0.0325	-0.02139* (0.01280)	-0.13171**** (0.03922)	-0.01860**** (0.00474)	20917	0.0033	1560	14	6.384**** <0.001		
JKLMN-Private Services	3.64% and 2.96%	24.83% and 26.42%	0.0301	0.04672**** (0.01247)	-0.09209* (0.04903)	-0.00937 (0.00574)	20920	0.0031	1560	14	5.440**** 0.001		
C27-Electrical equip.	0.49% and 0.33%	3.75% and 7.21%	0.0121	-0.06203**** (0.01185)	-0.02043 (0.03288)	-0.00349 (0.00442)	20921	0.0047	1560	14	9.159***** <0.001		

Significance levels: * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$; **** $p < 0.005$; ***** $p < 0.001$. See before for notes.

Table B15
Sectoral results with CO₂ shares, *sr* fixed effects, WIOD 2016 full sample

Sector	CO ₂ share 2000 and 2014	Output share 2000 and 2014	CO ₂ intensity in 2014	Import intensity $\ln(m_{srj(t-1)})$	Capital-to-labor ratio $\ln(dk_{srj(t-1)})$	Interaction term $\ln(m_{srj(t-1)}) \cdot \ln(dk_{srj(t-1)})$	Num. of obs.	R ²	Num. of fix.-ef.	F-stat. p-value
D35-Energy	41.67% and 41.90%	2.36% and 2.38%	3.5285	0.02540** (0.01005)	0.05273* (0.03015)	0.00723** (0.00328)	25284	0.0017	1806	4.755**** 0.003
C23-Minerals	7.63% and 11.04%	0.92% and 1.19%	1.8512	0.02904** (0.01216)	0.01011 (0.03161)	0.00321 (0.00414)	25284	0.0011	1806	2.675** 0.046
H51-Air transport	3.62% and 2.42%	0.61% and 0.47%	1.0354	0.16105***** (0.01992)	0.13417**** (0.04088)	0.02282***** (0.00558)	25284	0.0219	1806	32.321***** <0.001
H50-Water transport	2.92% and 1.96%	0.42% and 0.41%	0.9648	0.08934***** (0.01844)	-0.01214 (0.03991)	0.00033 (0.00494)	25284	0.0067	1806	9.169***** <0.001
C24-Metal	8.31% and 10.33%	2.28% and 2.86%	0.7235	0.03513** (0.01432)	-0.03669 (0.02892)	-0.00500 (0.00382)	25284	0.0016	1806	2.490* 0.059
C20-Chemicals	5.01% and 5.20%	2.34% and 2.71%	0.3844	0.07891***** (0.01334)	-0.12236***** (0.03287)	-0.01853***** (0.00441)	25284	0.0066	1806	16.929***** <0.001
E37-E39-Waste	0.46% and 0.55%	0.45% and 0.36%	0.3075	0.17033***** (0.02484)	-0.03490 (0.05447)	-0.00592 (0.00677)	19684	0.0143	1406	16.389***** <0.001
B-Mining	3.36% and 3.84%	2.22% and 2.50%	0.3066	0.03819***** (0.01272)	0.06835** (0.03029)	0.00904** (0.00366)	25284	0.0030	1806	5.522***** <0.001
H49-Land transport	3.74% and 3.62%	2.49% and 2.37%	0.3057	0.01794 (0.01467)	-0.02463 (0.04384)	-0.00205 (0.00494)	25284	0.0004	1806	0.658 0.578
A02-Forestry	0.19% and 0.25%	0.24% and 0.19%	0.2612	-0.04636*** (0.01719)	0.06975 (0.05749)	0.00644 (0.00588)	22960	0.0012	1640	2.682** 0.045
C19-Refined Petr.	3.42% and 2.79%	2.42% and 2.41%	0.2318	-0.02504** (0.01082)	0.08483*** (0.03080)	0.00995*** (0.00362)	22572	0.0020	1722	4.191*** 0.006
C17-Paper	0.94% and 0.60%	0.83% and 0.69%	0.1749	0.03341** (0.01483)	-0.07164* (0.03969)	-0.01062** (0.00540)	25284	0.0013	1806	2.736** 0.042
C22-Rubber	0.25% and 0.88%	1.11% and 1.14%	0.1543	-0.00550 (0.02156)	-0.03367 (0.04485)	-0.00603 (0.00610)	25284	0.0002	1806	0.402 0.752
A01-Agriculture	2.22% and 1.87%	2.74% and 2.43%	0.1538	0.02509* (0.01474)	-0.00983 (0.04568)	0.00506 (0.00543)	25284	0.0020	1806	4.478*** 0.004
E36-Water	0.20% and 0.13%	0.23% and 0.18%	0.1432	0.09296***** (0.01782)	0.23379***** (0.06299)	0.02094**** (0.00700)	22960	0.0088	1640	13.269***** <0.001
A03-Fisheries	0.15% and 0.13%	0.20% and 0.20%	0.1309	-0.14218***** (0.02239)	-0.10159* (0.05676)	-0.01098* (0.00650)	22960	0.0100	1640	16.580***** <0.001
C31-C32-Furniture	0.20% and 0.33%	1.06% and 0.82%	0.0811	0.13782***** (0.01877)	-0.05385 (0.04732)	-0.01120* (0.00628)	25284	0.0066	1806	18.089***** <0.001
C16-Wood	0.25% and 0.22%	0.53% and 0.58%	0.0747	0.04383** (0.01621)	0.00943 (0.04768)	-0.00720 (0.00621)	25284	0.0030	1806	7.546***** <0.001
H52-Warehousing	0.37% and 0.37%	0.98% and 1.06%	0.0701	0.09663***** (0.02249)	0.00066 (0.06529)	0.00056 (0.00789)	25284	0.0041	1806	6.223***** <0.001
H53-Post	0.10% and 0.08%	0.36% and 0.25%	0.0605	-0.01155 (0.01555)	-0.02356 (0.04129)	-0.00117 (0.00457)	19388	0.0002	1406	0.601 0.614
I-Accommodation	0.80% and 0.64%	2.75% and 2.40%	0.0539	-0.00137 (0.01825)	-0.19628*** (0.07116)	-0.02290*** (0.00862)	25284	0.0016	1806	2.541* 0.055
C13-C15-Textile	0.76% and 0.44%	1.45% and 1.63%	0.0535	0.03046 (0.02249)	0.22294***** (0.05066)	0.03246***** (0.00741)	25284	0.0052	1806	7.570***** <0.001
C10-C12-Food	1.37% and 1.15%	4.05% and 4.35%	0.0528	-0.01288 (0.01208)	0.00420 (0.03126)	-0.00257 (0.00375)	25284	0.0007	1806	2.345* 0.071
C25-Non machinery	0.37% and 0.43%	1.80% and 1.65%	0.0515	0.05753***** (0.01566)	-0.01581 (0.03956)	-0.00029 (0.00488)	24108	0.0021	1722	5.182***** 0.001
OPQRS-Public Services	4.35% and 3.64%	16.12% and 14.56%	0.0500	0.00498 (0.01794)	-0.16771** (0.07367)	-0.01592* (0.00823)	25284	0.0011	1806	2.180* 0.088
F-Construction	1.30% and 1.17%	7.37% and 6.95%	0.0336	0.05428**** (0.01903)	-0.02648 (0.05524)	-0.00146 (0.00672)	25284	0.0017	1806	2.981** 0.030
C18-Printing	0.17% and 0.05%	0.49% and 0.35%	0.0314	0.03732* (0.01912)	0.01018 (0.04841)	0.00427 (0.00568)	24108	0.0009	1722	1.971 0.116
C28-Machinery	0.44% and 0.35%	1.89% and 2.27%	0.0310	0.03367* (0.01981)	0.07391* (0.04386)	0.00925 (0.00625)	25284	0.0008	1806	1.816 0.142
G-Trade	2.30% and 1.43%	9.86% and 9.24%	0.0309	0.03619** (0.01680)	-0.05278 (0.05207)	-0.00675 (0.00600)	24696	0.0011	1806	2.565* 0.053
C30-Transport equip.	0.16% and 0.12%	0.67% and 0.97%	0.0239	-0.01299 (0.01401)	-0.00034 (0.03600)	-0.00121 (0.00499)	23862	0.0002	1722	0.491 0.689
C27-Electrical equip.	0.18% and 0.17%	1.06% and 1.47%	0.0230	0.03927** (0.01646)	0.01661 (0.03606)	0.00173 (0.00490)	23534	0.0007	1722	2.093* 0.099
C33-Repair	0.04% and 0.02%	0.28% and 0.23%	0.0196	0.08656***** (0.02489)	0.15354** (0.07165)	0.01818** (0.00924)	14656	0.0037	1056	5.063**** 0.002
JKLMN-Private Services	2.15% and 1.54%	22.51% and 22.07%	0.0140	0.00468 (0.01530)	-0.04032 (0.05355)	-0.00664 (0.00616)	25284	0.0004	1806	0.820 0.483
C29-Vehicles	0.33% and 0.20%	2.49% and 3.02%	0.0136	0.02172 (0.02156)	-0.01531 (0.04605)	-0.00290 (0.00636)	25284	0.0003	1806	0.442 0.723
C26-Computers	0.23% and 0.13%	1.53% and 2.66%	0.0096	0.06488*** (0.02468)	-0.11658** (0.05021)	-0.02001*** (0.00741)	25200	0.0032	1806	4.124**** 0.006
C21-Pharma.	0.04% and 0.03%	0.72% and 0.84%	0.0078	-0.02848 (0.02513)	-0.13915** (0.05614)	-0.01166* (0.00692)	22800	0.0015	1640	3.847*** 0.009

Significance levels: * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$; **** $p < 0.005$; ***** $p < 0.001$. See before for notes.

Table B16
Sectoral results with CO₂ shares, *t* fixed effects, WIOD 2016 full sample

Sector	CO ₂ share 2000 and 2014	Output share 2000 and 2014	CO ₂ intensity in 2014	Import intensity $\ln(m_{xjt(t-1)})$	Capital-to-labor ratio $\ln(dk_{xjt(t-1)})$	Interaction term $\ln(m_{xjt(t-1)}) \cdot \ln(dk_{xjt(t-1)})$	Num. of obs.	R ²	Num. of fix.-ef.	F-stat. p-value
D35-Energy	41.67% and 41.90%	2.36% and 2.38%	3.5285	0.01912* (0.01085)	-0.13671** (0.05803)	-0.00888 (0.00586)	25284	0.0055	14	5.507***** <0.001
C23-Minerals	7.63% and 11.04%	0.92% and 1.19%	1.8512	-0.04952*** (0.01779)	0.05066 (0.06488)	0.02162** (0.00916)	25284	0.0188	14	14.097***** <0.001
H51-Air transport	3.62% and 2.42%	0.61% and 0.47%	1.0354	-0.09240**** (0.01410)	0.63058***** (0.05739)	0.05335***** (0.00649)	25284	0.0640	14	78.938***** <0.001
H50-Water transport	2.92% and 1.96%	0.42% and 0.41%	0.9648	-0.05583**** (0.01760)	0.46506**** (0.08023)	0.05239**** (0.00984)	25284	0.0146	14	13.433***** <0.001
C24-Metal	8.31% and 10.33%	2.28% and 2.86%	0.7235	-0.34499**** (0.01872)	-0.23350**** (0.06803)	-0.04744**** (0.00908)	25284	0.2114	14	139.411***** <0.001
C20-Chemicals	5.01% and 5.20%	2.34% and 2.71%	0.3844	-0.24028**** (0.01727)	0.04803 (0.06442)	0.01898** (0.00920)	25284	0.1138	14	66.483***** <0.001
E37-E39-Waste	0.46% and 0.55%	0.45% and 0.36%	0.3075	-0.09278**** (0.01264)	0.00463 (0.05971)	-0.00492 (0.00701)	19684	0.0223	14	20.824***** <0.001
B-Mining	3.36% and 3.84%	2.22% and 2.50%	0.3066	-0.20805**** (0.01357)	-0.05923 (0.05837)	-0.01006 (0.00680)	25284	0.1103	14	87.529***** <0.001
H49-Land transport	3.74% and 3.62%	2.49% and 2.37%	0.3057	-0.02410** (0.00899)	0.04315 (0.05673)	0.00547 (0.00615)	25284	0.0026	14	3.266** 0.021
A02-Forestry	0.19% and 0.25%	0.24% and 0.19%	0.2612	0.03600* (0.01543)	-0.18870** (0.07580)	-0.01437* (0.00826)	22960	0.0061	14	5.794***** <0.001
C19-Refined Petr.	3.42% and 2.79%	2.42% and 2.41%	0.2318	-0.20515**** (0.01849)	-0.00385 (0.07949)	-0.01619 (0.01043)	22572	0.0962	14	56.604***** <0.001
C17-Paper	0.94% and 0.60%	0.83% and 0.69%	0.1749	-0.16389**** (0.01488)	0.09357 (0.07918)	-0.00468 (0.01043)	25284	0.0835	14	63.606***** <0.001
C22-Rubber	0.25% and 0.88%	1.11% and 1.14%	0.1543	-0.10215**** (0.01047)	-0.05983 (0.04444)	-0.00397 (0.00634)	25284	0.0276	14	32.361***** <0.001
A01-Agriculture	2.22% and 1.87%	2.74% and 2.43%	0.1538	-0.04711**** (0.01136)	0.00664 (0.04665)	0.00906 (0.00585)	25284	0.0127	14	11.698***** <0.001
E36-Water	0.20% and 0.13%	0.23% and 0.18%	0.1432	-0.03464**** (0.01214)	0.14156** (0.06292)	0.01488** (0.00657)	22960	0.0039	14	4.333*** 0.005
A03-Fisheries	0.15% and 0.13%	0.20% and 0.20%	0.1309	-0.09388**** (0.02093)	-0.09552 (0.08849)	-0.00328 (0.01156)	22960	0.0120	14	12.445***** <0.001
C31-C32-Furniture	0.20% and 0.33%	1.06% and 0.82%	0.0811	-0.05278**** (0.01106)	-0.14210**** (0.05047)	-0.01446** (0.00645)	25284	0.0073	14	9.199***** <0.001
C16-Wood	0.25% and 0.22%	0.53% and 0.58%	0.0747	-0.06029**** (0.01148)	0.10262* (0.05194)	0.01626* (0.00704)	25284	0.0104	14	10.105***** <0.001
H52-Warehousing	0.37% and 0.37%	0.98% and 1.06%	0.0701	0.03613**** (0.01047)	0.19838**** (0.07064)	0.01533* (0.00782)	25284	0.0074	14	7.802***** <0.001
H53-Post	0.10% and 0.08%	0.36% and 0.25%	0.0605	-0.03886**** (0.01218)	0.05080 (0.04716)	0.01270** (0.00543)	19388	0.0084	14	7.230***** <0.001
I-Accommodation	0.80% and 0.64%	2.75% and 2.40%	0.0539	-0.11174**** (0.01195)	0.07752 (0.07292)	0.00899 (0.00857)	25284	0.0327	14	38.491***** <0.001
C13-C15-Textile	0.76% and 0.44%	1.45% and 1.63%	0.0535	-0.12280**** (0.01188)	0.22558***** (0.04662)	0.04411***** (0.00653)	25284	0.0557	14	47.761***** <0.001
C10-C12-Food	1.37% and 1.15%	4.05% and 4.35%	0.0528	-0.03708**** (0.01226)	0.13055* (0.06100)	0.01888** (0.00770)	25284	0.0092	14	8.014***** <0.001
C25-Non machinery	0.37% and 0.43%	1.80% and 1.65%	0.0515	-0.05627**** (0.01129)	0.05480 (0.05613)	0.01393** (0.00704)	24108	0.0138	14	12.105***** <0.001
OPQRS-Public Services	4.35% and 3.64%	16.12% and 14.56%	0.0500	-0.04212**** (0.01183)	0.13601** (0.06870)	0.00268 (0.00768)	25284	0.0201	14	18.393***** <0.001
F-Construction	1.30% and 1.17%	7.37% and 6.95%	0.0336	0.00902 (0.01063)	0.31232***** (0.06055)	0.02944***** (0.00725)	25284	0.0142	14	14.087***** <0.001
C18-Printing	0.17% and 0.05%	0.49% and 0.35%	0.0314	-0.07136**** (0.01095)	0.11838** (0.05601)	0.00396 (0.00718)	24108	0.0243	14	32.076***** <0.001
C28-Machinery	0.44% and 0.35%	1.89% and 2.27%	0.0310	-0.15145**** (0.01143)	0.09788* (0.05054)	0.00535 (0.00702)	25284	0.0686	14	72.338***** <0.001
G-Trade	2.30% and 1.43%	9.86% and 9.24%	0.0309	-0.00895 (0.01155)	0.42747***** (0.07334)	0.04134***** (0.00841)	24696	0.0199	14	18.858***** <0.001
C30-Transport equip.	0.16% and 0.12%	0.67% and 0.97%	0.0239	-0.18409**** (0.01429)	-0.06565 (0.06449)	-0.01049 (0.00870)	23862	0.0749	14	60.505***** <0.001
C27-Electrical equip.	0.18% and 0.17%	1.06% and 1.47%	0.0230	-0.18005**** (0.01232)	0.01111 (0.04937)	0.01326* (0.00695)	23534	0.0816	14	76.816***** <0.001
C33-Repair	0.04% and 0.02%	0.28% and 0.23%	0.0196	-0.05238**** (0.01316)	0.20647***** (0.06629)	0.02847***** (0.00754)	14656	0.0114	14	12.787***** <0.001
JKLMN-Private Services	2.15% and 1.54%	22.51% and 22.07%	0.0140	-0.01591 (0.01093)	0.18917*** (0.06772)	0.00722 (0.00772)	25284	0.0219	14	20.773***** <0.001
C29-Vehicles	0.33% and 0.20%	2.49% and 3.02%	0.0136	-0.18687**** (0.01151)	-0.12839** (0.05212)	-0.02613***** (0.00678)	25284	0.0899	14	97.291***** <0.001
C26-Computers	0.23% and 0.13%	1.53% and 2.66%	0.0096	-0.19911**** (0.01265)	-0.15341**** (0.05004)	-0.01088 (0.00711)	25200	0.0831	14	84.220***** <0.001
C21-Pharma.	0.04% and 0.03%	0.72% and 0.84%	0.0078	-0.02985** (0.01409)	0.21212** (0.08288)	0.00651 (0.00970)	22800	0.0182	14	15.923***** <0.001

Significance levels: * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$; **** $p < 0.005$; ***** $p < 0.001$. See before for notes.

Table C1
Sector mapping of the WIOD 2016 and 2013 datasets

Sector codes		WIOD 2016	Sector codes			WIOD 2013
Trade	Non-trade	Definition	Aggregate	Trade	Non-trade	Definition
r1	A01	Crop and animal production...	A01-Agriculture	c1	secA4B	Agriculture, Hunting, Forestry and Fishing
r2	A02	Forestry and logging	A02-Forestry	-	-	-
r3	A03	Fishing and aquaculture	A03-Fisheries	-	-	-
r4	B	Mining and quarrying	B-Mining	c2	secC	Mining and Quarrying
r5	C10-C12	Man. food, beverages and tobacco	C10-C12-Food	c3	sec15t16	Food, Beverages and Tobacco
r6	C13-C15	Man. textiles, wearing apparel and leather products	C13-C15-Textile	c4	sec17t18	Textiles and Textile Products
-	-	-	-	c5	sec19	Leather, Leather and Footwear
r7	C16	Man. wood and of products of wood and cork, except furniture	C16-Wood	c6	sec20	Wood and Products of Wood and Cork
r8	C17	Man. paper and paper products	C17-Paper	c7	sec21t22	Pulp, Paper, Paper, Printing and Publishing
r9	C18	Printing and reproduction of recorded media	C18-Printing	-	-	-
r10	C19	Man. coke and refined petroleum products	C19-Refined Petr.	c8	sec23	Coke, Refined Petroleum and Nuclear Fuel
r11	C20	Man. chemicals and chemical products	C20-Chemicals	c9	sec24	Chemicals and Chemical Products
r12	C21	Man. basic pharmaceutical products	C21-Pharma.	-	-	-
r13	C22	Man. rubber and plastic products	C22-Rubber	c10	sec25	Rubber and Plastics
r14	C23	Man. other non-metallic mineral products	C23-Minerals	c11	sec26	Other Non-Metallic Mineral
r15	C24	Man. basic metals	C24-Metal	c12	sec27t28	Basic Metals and Fabricated Metal
r16	C25	Man. fabricated metal products	C25-Non machinery	-	-	-
r17	C26	Man. computer, electronic and optical products	C26-Computers	c13	sec29	Machinery, Nec
r18	C27	Man. electrical equipment	C27-Electrical equip.	c14	sec30t33	Electrical and Optical Equipment
r19	C28	Man. machinery and equipment n.e.c.	C28-Machinery	-	-	-
r20	C29	Man. motor vehicles, trailers and semi-trailers	C29-Vehicles	-	-	-
r21	C30	Man. other transport equipment	C30-Transport equip.	c15	sec34t35	Transport Equipment
r22	C31-C32	Man. furniture; other manufacturing	C31-C32-Furniture	-	-	-
r23	C33	Repair and installation of machinery and equipment	C33-Repair	c16	sec36t37	Manufacturing, Nec; Recycling
r24	D35	Electricity, gas, steam and air conditioning supply	D35-Energy	c17	secE	Electricity, Gas and Water Supply
r25	E36	Water collection, treatment and supply	E36-Water	-	-	-
r26	E37-E39	Waste collection, treatment and disposal activities	E37-E39-Waste	-	-	-
r27	F	Construction	F-Construction	c18	secF	Construction
r28	G45	Wholesale and retail trade of motor vehicles	G-Trade	c19	sec50	Sale, mainten. repair of motor vehicles
r29	G46	Wholesale trade, except of motor vehicles and motorcycles	G-Trade	c20	sec51	Wholesale trade, except of motor vehicles
r30	G47	Retail trade, except of motor vehicles and motorcycles	G-Trade	c21	sec52	Retail Trade, Except of Motor Vehicles
r31	H49	Land transport and transport via pipelines	H49-Land transport	c23	sec60	Inland Transport
r32	H50	Water transport	H50-Water transport	c24	sec61	Water Transport
r33	H51	Air transport	H51-Air transport	c25	sec62	Air Transport
r34	H52	Warehousing and support activities for transportation	H52-Warehousing	c26	sec63	Other Supporting and Auxiliary Transport Activities
r35	H53	Postal and courier activities	H53-Post	c27	sec64	Post and Telecommunications
r36	I	Accommodation and food service activities	I-Accommodation	c22	secH	Hotels and Restaurants
r37	J58	Publishing activities	JKLMN-Private Services	c28	secJ	Financial Intermediation
r38	J59-J60	Motion picture, video and television programme ...	JKLMN-Private Services	c29	sec70	Real Estate Activities
r39	J61	Telecommunications	JKLMN-Private Services	c30	sec71t74	Renting of M&Eq and Other Business Activities
r40	J62-J63	Computer programming, consultancy and related activities	JKLMN-Private Services	-	-	-
r41	K64	Financial service activities	JKLMN-Private Services	-	-	-
r42	K65	Insurance, reinsurance and pension funding	JKLMN-Private Services	-	-	-
r43	K66	Activities auxiliary to financial services	JKLMN-Private Services	-	-	-
r44	L68	Real estate activities	JKLMN-Private Services	-	-	-
r45	M69-M70	Legal and accounting activities	JKLMN-Private Services	-	-	-
r46	M71	Architectural and engineering activities	JKLMN-Private Services	-	-	-
r47	M72	Scientific research and development	JKLMN-Private Services	-	-	-
r48	M73	Advertising and market research	JKLMN-Private Services	-	-	-
r49	M74-M75	Other professional, scientific and technical activities	JKLMN-Private Services	-	-	-
r50	N	Administrative and support service activities	JKLMN-Private Services	-	-	-
r51	O84	Public administration and defence; compulsory social security	OPQRS-Public Services	c31	secL	Public Admin and Defence, Social Security
r52	P85	Education	OPQRS-Public Services	c32	secM	Education
-	-	-	-	c35	secP	Private Households with Employed Persons
r53	Q	Human health and social work activities	OPQRS-Public Services	c33	secN	Health and Social Work
r54	R-S	Other service activities	OPQRS-Public Services	c34	secO	Other Community, Social and Personal Services
(r55)	T	Activities of households as employers	-	-	FC-HH	Final consumption expenditure by households
(r56)	U	Activities of extraterritorial organizations and bodies	-	-	secQ	Extra-territorial organizations and bodies