

Are Your Labor Shares Set in Beijing?

The View through the Lens of Global Value Chains*

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Abstract

We study the evolution of labor shares in 1995–2014, while taking into account international trade based on value added concepts. Declines in labor shares accelerate in 2001–2007, concurrently with global value chain (GVC) participation, after which there is no trend for both. We develop a gravity-based instrument for GVC participation and find that the acceleration in the decline in labor shares is caused by increased intensity of forward GVC participation. The insertion of China into GVCs has a disproportionately large effect through this mechanism. Declines in labor shares are shouldered mostly by less skilled workers in fabrication functions. Relatively capital abundant countries participate more in forward GVCs linkages, which is associated with greater upstreamness within GVCs and increases in capital intensity. Forward GVC participation is associated with international vertical integration of both upstream intermediate input production and of offshoring of downstream assembly.

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1 Introduction

The decline in labor shares in recent decades in many advanced economies has both caught the attention of academics and generated concern of policy makers. Apart from having important implications for economic modeling, the interest in declining labor shares fundamentally stems from its implications for income inequality.¹ Just like labor income, capital income accrues to people, but the ownership of capital is concentrated in the hands of relatively few; moreover, capital ownership among capital owners—and thus capital income—is more concentrated than human capital and labor income among workers.² A smaller share of value added paid to labor implies that income inequality among people rises. This is particularly acute given relatively weak productivity growth in recent times.³

Figure 1 illustrates that while declines in labor shares started in the 1980s, this evolution accelerated markedly in 2001–2007, when global value chain (GVC) participation accelerated as well. Both labor shares and forward GVC participation level-off after 2007.⁴ We argue that countries and industries that integrated forward differentially more saw greater declines in their labor shares. More offshoring of downstream tasks (including final assembly), as well as increases in exports of upstream intermediate inputs for use in foreign downstream production, both lower labor shares. This effect has large explanatory power in 2001–2007, and in particular for China’s trading partners.

Significant increases in international trade openness imply that studying the evolution of factor shares from a closed economy perspective—as does most of this literature (see Grossman and Oberfield (2021))—is bound to miss some of the underlying mechanisms. While there is a large literature that studies labor market effects of trade, recent work illustrates that standard gross trade sales statistics can be misleading, and that this has become particularly acute since China joined the World Trade Organization in 2001 and its subsequent increase of global production sharing.⁵

¹Changing shares contradict the first of the so-called “Kaldor facts” and lead to rejecting the Kaldor (1957) model of growth, along with other models that imply the same constancy of shares. Varying shares also have ramifications for computation of total factor productivity and long run macroeconomic projections.

²For example, see Piketty (2014), and up to date statistics from the World Inequality Database, <https://wid.world/>. This goes beyond the classic “functional inequality” between workers, “capitalists” and “rentiers”, due to Adam Smith and David Ricardo.

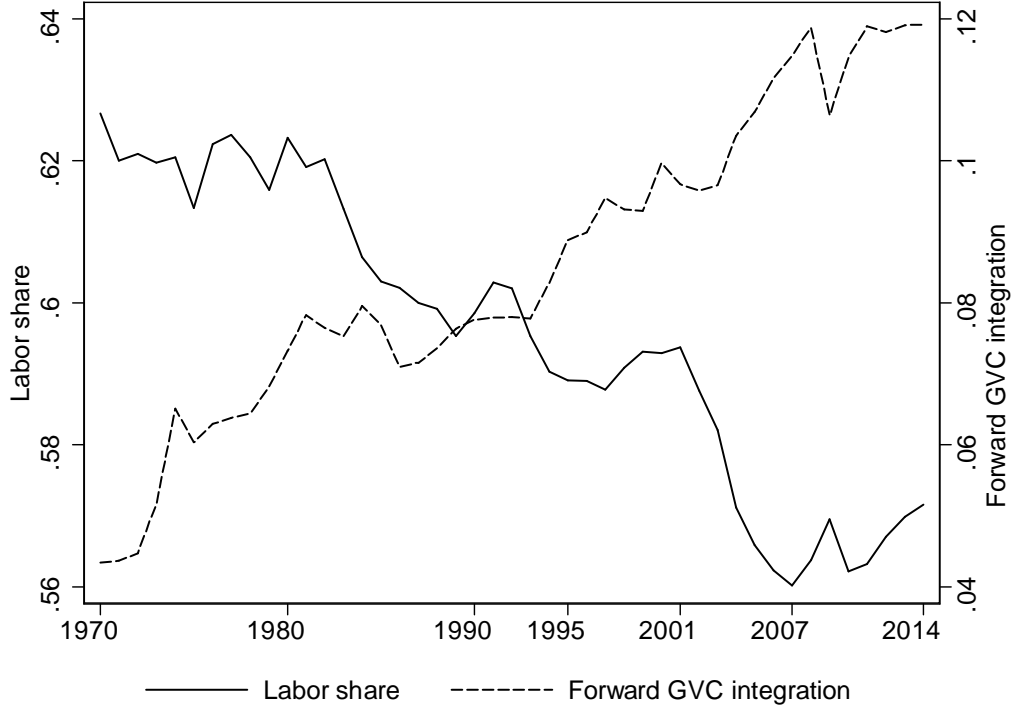
³An additional concern relates to how income inequality affects overall growth and political economy; see, e.g., Persson and Tabellini (1994), Alesina and Rodrik (1994), Alesina and Perotti (1996), and more recently Ostry and Berg (2011).

⁴In Appendix A we show that different concepts of the labor share in the Penn World Tables exhibit very similar trajectories, in particular the 2001–2007 acceleration in the decline in labor shares and the change in trajectory after 2007. Gutiérrez and Piton (2019) argue that after adjusting for (netting out) the housing sector, labor shares did not decline during 1970–2015 in a set of advanced economies plus China, India, and Russia—except in the US and Canada. However, we do find in their data that labor shares declined in 2001–2007—both using their “adjusted corporate sector” and “adjusted business sector” series—in similar magnitudes to our data.

⁵Trefler and Zhu (2010) show that taking into account intermediate inputs helps aligning factor content of trade predictions of the Heckscher-Ohlin model with the data. Ito, Rotunno, and Vézina (2017) show that predictions

While Freeman (1995) asked “Are your wages set in Beijing?”, the deepening of production sharing across international borders requires a different data approach in order to answer such questions.⁶

Figure 1. Labor Shares and Forward GVC Participation



Notes. The displayed series are year fixed effects from a regression of either labor shares or forward GVC participation on year fixed effects and country fixed effects, weighted by real GDP. Forward GVC Participation is exports of intermediate inputs in value added terms as a share of GDP. The sample for labor shares includes 39 countries that correspond to the WIOD 2013 release sample of countries, and the year fixed effects are adjusted by the weighted (by GDP) average in 1995. The labor share series encompasses compensation of employees and labor income of self-employed; the latter is based on part of mixed income. For China the labor share includes only compensation of employees. Source: Penn World Tables mark 9.1; see Feenstra, Inklaar, and Timmer (2015) for documentation on labor shares. The sample for forward GVC participation in 1995–2014 includes 39 countries that correspond to the WIOD 2013 release sample of countries. Data for 1995–2007 are from WIOD 2013 release. Data for 2007–2014 are from WIOD 2016 release, and the year fixed effects are adjusted by the weighted (by GDP) average in 2007 from the WIOD 2013 release. Data for 1970–1995 are from Johnson and Noguera (2012), and the year fixed effects are adjusted by the weighted (by

of Heckscher-Ohlin trade theory hold much better when using value added trade data versus gross trade values. Timmer, Miroudot, and de Vries (2019) show that revealed comparative advantage indices based on gross trade statistics deviate significantly from those based on and trade in value added, which are more sensible. Jakubik and Stolzenburg (2018) use data on trade in value added to revisit the estimates in Autor, Dorn, and Hanson (2013) of the effect of imports from China on local labor markets in the U.S.—and find significantly weaker effects. Using gross instead of value added export data is also one of the the conceptual flaws underlying the so-called Leontief (1953) paradox. See Johnson (2014) for a portrait of differences between gross trade and value added trade flows, as well as several implications. Koopman, Wang, and Wei (2012) and Kee and Tang (2016) demonstrate the consequences of the rise of China for mis-representation of gross trade data. Johnson (2018) provides a survey of these issues.

⁶As noted in the conclusion of Grossman and Rossi-Hansberg (2008): "...almost all current goods' trade data pertain to gross flows rather than to value added. The globalization of production processes mandates a new approach to trade data collection, one that records international transactions, much like domestic transactions have been recorded for many years."

GDP) average in 1995 from the WIOD 2013 release. Compared to the sample of 39 countries, the sample from Johnson and Noguera (2012) excludes Bulgaria, Cyprus, Czech Rep., Estonia, Lithuania, Luxembourg, Latvia, Malta, Russian Federation, Slovakia, Slovenia, Taiwan; these countries account for 2.8% of GDP of the 39 countries in the WIOD sample.

We study the timing and mechanisms that govern the relationship between labor shares and globalization using country and industry panel data from the World Input-Output Database (WIOD) in 1995–2014. This allows us to account for international trade in value added terms, the same concept in which labor shares are measured. While we find that increases in forward foreign GVC participation lowers labor shares, other dimensions of globalization (exports and imports of final goods and imports of inputs) do not affect labor shares significantly.

We estimate that the declines of labor shares are driven by forward GVC deepening only in 2001–2007, coinciding with China’s accession to the WTO in 2001. Moreover, China’s insertion into GVCs during this period has a larger effect on its trading partners than other countries with their corresponding trading partners. This is consistent with China attracting assembly tasks and importing intermediate inputs from its trading partners in 2001–2007. This relationship disappears after 2007, when China’s insertion into GVCs slows down, along with that of the rest of the world (Figure 1). Indeed, Chor, Manova, and Yu (2020) find that the upstreamness of imports by Chinese firms stops increasing after 2007.

In order to make causal inferences we develop an instrument based on a structural gravity relationship. The main threat to identification in our regressions is if supply shocks increase an industry’s exports and at the same time affect its labor share—for example, biased technological change, differential declines in the rental rate of capital, or changes in imported input costs. Structural gravity attributes variation in exports to variation in source country supply factors, destination country factors (which are determined by all source countries, in addition to the specific destination), and bilateral trade barriers. We estimate these factors and construct an instrument for GVC participation that is orthogonal to changes in supply factors. The internal validity of our regressions is bolstered by saturating them with fixed effects. Since the regressions are panels of long differences, adding fixed effects absorbs trends in economy and industry-wide variables. This strategy (including the instrument) addresses most of the concerns for identification that Grossman and Oberfield (2021) raise for studying the evolution of labor shares.

We then ask whether differences in factor endowments can help interpret the results. Sposi, Yi, and Zhang (2020) extend the model of Antràs and De Gortari (2020) to include Heckscher-Ohlin forces that operate within value chains. They assume that upstream stages of production are more capital intensive (which we document, especially in GVCs). When separating the geographical location of different stages of production is possible, declines in trade barriers cause relocation

of relatively capital-intensive upstream stages to relatively capital abundant countries.⁷ We find evidence that is consistent with this prediction: relative capital abundance predicts both the capital intensity and the volume of exports of intermediate inputs, and these associations strengthen over time. Related to this, we show that increases in the upstreamness of production are associated with reductions in labor shares, and that only the foreign-driven part of upstreamness accounts for this relationship.⁸

We then show that forward GVC-driven reductions in labor shares are shouldered mostly by labor in fabrication tasks, and to a lesser extent by management and marketing. In contrast, it is not driven by reductions in payments to labor engaged in research and development (R&D). This is consistent with the idea that selling final goods requires more interaction with consumers and more management control of production processes compared to producing intermediate inputs further up the value chain. It is also consistent with some degree of knowledge transfer that is involved in forward GVC participation, as argued by Baldwin (2016).

Finally, we examine briefly the roles of offshoring of assembly and multinational enterprises (MNEs). We find that offshoring of assembly of hitherto exported final goods, while contributing to the patterns we uncover, is not the main mechanism driving the results. Greater intensity of exports of intermediate inputs lowers labor shares, whether or not is it done for the purpose of this kind of offshoring. We find that MNEs are associated with forward GVC participation both in offshoring downward assembly, as well as in offshoring upstream input production.

Before turning to positioning our work within the literature, we wish to stress an under-emphasized fact: more than the entire drop in overall labor shares and cross-country variation therein is shouldered by unskilled labor. Labor shares decline on average 2.5 percent points in 1995–2007, with significant variation across countries summarized by a coefficient of variation of -1.37 (using GDP in 1995 as weights). In contrast, the share of *skilled* labor in GDP concurrently increased by 4.25 percent points with a much smaller coefficient of variation of 0.36 (data from EU KLEMS). Consistent with this, we find that most of the GVC-induced declines in labor shares are shouldered by labor in fabrication tasks, which is less skilled than labor engaged in management, marketing and R&D. Our finding of a negative association between GVC participation and labor shares while skilled labor shares increase is reminiscent of, *inter alia*, Richardson (1995) and especially Wood (1995). Consistent with this, Timmer, Los, Stehrer, and de Vries (2013) find that the rise of GVCs is associated with a shift towards skilled labor employment within total employment

⁷The main insight in Sposi, Yi, and Zhang (2020) can be deduced from Helpman (1984) if plants are thought of as downstream and headquarters upstream activities.

⁸Although we do not test this, this mechanism can raise labor shares everywhere if tasks or inputs that are relatively labor un-intensive within capital-abundant countries are offshored to labor-abundant countries, in which they are relatively labor intensive, as in Feenstra and Hanson (1997).

in 1995–2008.

The rest of the paper is organized as follows. We position our work in the relevant literature in Section 2. Section 3 describes the data and methodology. Section 4 describes changes in factor shares and in GVC participation. In Section 5 we investigate the role of GVC deepening in explaining declines in labor shares, as well as several mechanisms. We offer concluding remarks in Section 6.

2 Relationship to the literature

Our paper contributes to a large body of work that studies the determinants of labor shares. Since most countries have experienced declines in labor shares, it is plausible that the cause is common to all. One such common trend is globalization. While most previous research focuses on the United States and other developed economies, e.g., Blanchard (1997), Elsby, Hobijn, and Şahin (2013), Rognlie (2016), and pays little attention to international trade openness—Harrison (2005) studies the evolution of labor shares in a panel of countries, including less developed ones, and investigates, *inter alia*, the role of trade openness (aggregate gross imports + exports divided by GDP). Elsby, Hobijn, and Şahin (2013) suggest that offshoring may have contributed to the decline of the labor share in the United States, but do not study this beyond simple correlations. Böckerman and Maliranta (2012) and Panon (2020) study the role of exporters in accounting for the evolution of labor shares in Finnish and French manufacturing, respectively. Weinberger and Leblebicioglu (2018) study the effect of capital goods import liberalization in India, and find that this actually increased firm-level labor shares, probably by increasing quality of capital equipment while lowering its effective price. Compared to these papers, we stress the role of GVC participation and exporting of intermediate inputs, using value added trade data.

The closest paper to ours is chapter 3 in International Monetary Fund (2017). In contrast to our work, that paper does not distinguish the impact of forward from backward GVC participation (it aggregates them) on labor shares, and does not control for other forms of trade, e.g., in final goods in regression analysis. In addition, we identify causal effects by distinguishing the period in which GVC participation accelerated, and we use gravity-based instruments in order to overcome bias due to confounding factors. We also study the mechanisms underlying the relationship between GVCs and labor shares in greater detail.⁹ As discussed above, we find support for an Heckscher-Ohlin mechanism that operates within value chains, as suggested in the model of Sposi, Yi, and Zhang (2020). For example, Aichele and Heiland (2018) find in their calibrated model that reductions in trade costs *vis a vis* China that are associated with significant reductions in domestic value added

⁹Another difference is that International Monetary Fund (2017) mix several datasets that are based on different methodologies, whereas we rely on the WIOD for both labor shares and GVC measures.

shares for China’s trading partners. Our findings are consistent with this, and add that those declines in domestic value added are accounted for more by labor compensation than capital.

Another common trend, that has been proposed as an explanation for declines in labor shares by Karabarbounis and Neiman (2014), is the widespread decline in the price of investment. This may have caused a shift in expenditures towards capital if the elasticity of substitution between capital and labor were greater than unity, as Karabarbounis and Neiman (2014) estimate.¹⁰ Indeed, they document that in countries and industries where the decline in investment goods’ prices were deeper, labor shares dropped more. However, Glover and Short (2020) demonstrate how the estimator of the elasticity of substitution in Karabarbounis and Neiman (2014) is biased upwards; using a corrected estimator, they estimate an elasticity comfortably below unity. Oberfield and Raval (2014), using different econometric methodology, also estimate that the elasticity of substitution between labor and capital is less than unity.¹¹ Results in Bergholt, Furlanetto, and Faccioli-Maffei (forthcoming) also point toward strong capital-labor complementarity in US data. These findings weaken the argument on the importance of the decline in the price of investment in contributing to the decline in the labor share. We control for declines in the price of investment in our regressions, and find no evidence for this mechanism in our data.¹²

Variation in labor shares may also be related to endogenous directed technological change as proposed by Kennedy (1964) and Acemoglu (2003), where the decline in the relative price of capital leads to innovation that corrects initial changes in factors’ income shares. Acemoglu and Restrepo (2018) discuss the possible implications of technological change and robotization, and vom Lehn (2018) discusses how this manifests across occupations. Bergholt, Furlanetto, and Faccioli-Maffei (forthcoming) find an important role for automation in US data. Bentolila and Saint-Paul (2003) study variations in labor shares in OECD countries and show how they are linked to technological change, prices of imported materials and labor market frictions. Grossman, Helpman, Oberfield, and Sampson (2017) show that in a growth model with endogenous human capital accumulation, the decline in productivity growth can lead to declines in labor shares. More generally, biased technological progress can explain both variation in exporting and in labor shares, as we show below. In our analysis we develop a gravity model-based instrument that helps identifying the role

¹⁰Harrigan, Reshef, and Toubal (2021) find an elasticity between labor, capital and materials greater than 1 using French administrative firm-level data. When the elasticity of substitution between labor and capital is greater than unity and when factor markets are competitive, then a lower relative price of capital causes an increase in the share of expenditures on capital due to strong substitution towards capital usage. This explanation can also capture embodied technological change (computers, robots, etc.), as argued in Martinez (2018). See also Graetz and Michaels (2018).

¹¹Karabarbounis and Neiman (2014) consider the aggregate economy elasticity of substitution, while Oberfield and Raval (2014) consider only the elasticity in manufacturing. It is possible that the two differ markedly, as shown in Reshef (2013).

¹²In a previous version of this paper we showed that in the presence of capital-skill complementarity the relationship between the price of investment or the rental rate of capital and labor shares may not even be monotone.

of forward GVC participation separately from domestic technological change and changes in prices of inputs, *inter alia* imported intermediate inputs.¹³

Recent work studies the evolution of the labor share at the firm or establishment level, usually in manufacturing. Kyrrä and Maliranta (2008) and Autor, Dorn, Katz, Patterson, and Van Reenen (2020) consider the role of firm size distribution and increases in market share concentration in accounting for changes in labor shares, while Kehrig and Vincent (2021) assigns a much smaller role for changes in concentration in US manufacturing establishments, and Mertens (2019) attributes most of the decline in the labor share in German manufacturing to production process transformation, rather than firm concentration. Stansbury and Summers (2020) argue that declines in rent sharing with labor, due to declines in “labor power”, can help explain both the decline in labor shares and the decline in the NAIRU in the United States. Brooks, Kaboski, Li, and Qian (2019) argue that employers’ monopsony power in China and India lowers labor shares there, and that this effect has declined over time. Our sample covers mostly developed, mid-income and transition economies, but also important developing and emerging economies (e.g., India and China), and we perform our analysis both at the industry and country levels.

3 Data and methodology

The main source of data is the World Input-Output Database (WIOD). We outline the main features here and relegate other details to Appendix B. We use the WIOD 2013 release to compute statistics over the pre-2008 financial crisis period of 1995–2007. Along with detailed Input-Output tables for 40 countries and 35 industries (of which 14 are in manufacturing, ISIC rev. 3), the 2013 release also provides Socio-Economic Accounts with data on employment, labor compensation and capital stocks, all by country and industry. WIOD 2013 uses the 1993 System of National Accounts (SNA). We also use the more recent 2016 WIOD release, covering 43 countries and 56 industries (of which 14 are in manufacturing, ISIC rev. 4) to compute statistics for 2007–2014. WIOD 2016 uses the 2008 SNA.¹⁴ In both datasets the labor share includes compensation of employees and labor income of self-employed (part of “mixed income”).¹⁵ The labor shares from the WIOD data

¹³Declines in labor shares have been also related to structural change (Ngai and Pissarides (2007), Buera and Kaboski (2012), McAdam and Willman (2013)), the difference between capital returns and output growth (Piketty (2014)), deregulation of labor markets (Blanchard and Giavazzi (2003)), deregulation of bank branching in the U.S. (Weinberger and Leblebicioglu (forthcoming)), to dynamics in real estate values (Gutiérrez and Piton (2019)), and to fertility and immigration (d’Albis, Boubtane, and Coulibaly (2021)). Bengtsson, Rubolino, and Waldenström (2020) study the institutional determinants of the long run evolution of labor shares in a closed economy context. For business cycle properties of the labor share see McAdam and Willman (2013), Young (2004) and Mućk, McAdam, and Growiec (2018).

¹⁴For WIOD 2013 release documentation see Timmer, Dietzenbacher, Los, Stehrer, and de Vries (2015). For WIOD 2016 release documentation see Timmer, Los, Stehrer, and de Vries (2016). See <http://www.wiod.org/home> for further details on WIOD country coverage and data availability.

¹⁵WIOD 2013 release reports data until 2011, but the incidence of missing values for labor shares in the Socio-Economic Accounts increases significantly after 2009. This is not an issue in WIOD 2016. In both datasets and in

exhibit virtually identical trends at the country level as those from the Penn World Tables used to construct Figure 1. In Appendix A, we show that different concepts of the labor share exhibit very similar trajectories, in particular the 2001–2007 acceleration in the decline in labor shares and the change in trajectory after 2007.

Changes in the System of National Accounts preclude merging data from the two WIOD releases. However, for comparability of within-sample trends we reclassify WIOD 2016 release data to conform with the 2013 release in two dimensions. First, we allocate countries that appear in the 2016 release but not in the 2013 release to the “Rest of World” (ROW) category. Second, since the sectors in the 2016 release are more disaggregate, we aggregate them to the same level of the 2013 release.¹⁶ The correlation across the two WIOD releases in 2007 for labor shares and forward GVC participation is over 0.85, and the trends in the overlapping years (2000–2011) are remarkably similar.¹⁷

We drop from our dataset five non “strictly” private sector industries.¹⁸ We drop Poland from the analysis because it is an extreme outlier in 1995, and thus creates unreasonable variation over time for that country. Thus, our sample covers 39 countries and 30 private sector industries, of which 14 in manufacturing in the period 1995–2014.

3.1 Forward GVC participation

We measure forward GVC participation as payments to domestic factors that are generated by downstream foreign industries. This is driven by more than just direct exports of intermediate inputs and services to businesses, as it takes into account the entire network of GVCs, where value can “travel” across several borders (e.g., purchases from buyers of my output, etc.).¹⁹

We rely on the standard methodology of Leontief (1936), applied to an international setting. Gross output for any industry located in any country is the sum of intermediate input demand from all other industries located in all other countries, plus production of final goods. In matrix notation, this is $X = AX + Y$, where X is the vector of gross outputs, AX is intermediate demand and Y is final good production; A is the matrix of technical coefficients.²⁰ Rearranging leads to all years, the input-output matrices do not have missing values.

¹⁶The sectoral reclassification is important for comparability of changes over time across the two WIOD releases because the increase in the number of industries in WIOD 2016 is due to splitting industries into relatively upstream and relatively downstream industries.

¹⁷Table A1 and Table A2 in the appendix report country-level labor shares and skilled labor shares in 1995, 2007 and 2014.

¹⁸The private sector excludes the following industries: "Public Admin and Defence; Compulsory Social Security", "Education", "Health and Social Work", "Other Community, Social and Personal Services", "Private Households with Employed Persons".

¹⁹See Hummels, Ishii, and Yi (2001) and Yi (2003) on the increasing importance of vertical specialization and integration in world trade.

²⁰A typical entry of the A matrix a_{ij}^{od} indicates the value of input from industry i located in country o that is needed to produce one dollar worth of good j in country d .

$X = (I - A)^{-1}Y = BY$, where $B = (I - A)^{-1}$ is the Leontief (inverse) matrix, which takes into account the indirect production linkages across industries. The matrix B summarizes all value chains, be they domestic or global.²¹ It is useful to define final good production Y as a diagonal matrix, with the corresponding values on the diagonal and zeros elsewhere, implying that X is a matrix as well.

By pre-multiplying gross output terms (in US dollars) by a diagonal matrix of value added to gross output ratios (intensities) one obtains $VX = VBY$. The left hand side, VX , is industry value added produced and the right hand side, VBY , is demand for final goods in value added terms.²²

Each element of the VBY matrix contains all payments to factors that are employed in sector i in origin country o that contribute to the production of final goods in sector j in destination country d : $(vby)_{ij}^{od}$. Then $\sum_{d \neq o} \sum_j (vby)_{ij}^{od}$ summarizes payments to primary factors employed in country o in industry i due to exports (in value added terms) of intermediate inputs. We define forward foreign GVC participation for an industry i in country o as the share of exports of intermediate inputs in value added terms in value added: $forward_i^o = \sum_{d \neq o} \sum_j (vby)_{ij}^{od} / \sum_d \sum_j (vby)_{ij}^{od}$. The equivalent concept at the country level is $forward^o = \sum_{d \neq o} (vby)^{od} / GDP^o$, where $vby^{od} = \sum_i \sum_j (vby)_{ij}^{od}$ and $GDP^o = \sum_d (vby)^{od}$.²³ This follows the GVC income approach, which is theoretically consistent.²⁴

Figure 1 illustrates that the weighted average *forward* increases by 2.84 percent points in 1995–2007. All but one country (Latvia) in our sample exhibit increases in this period, to varying degrees. Figure 1 also illustrates that *forward* leveled off after 2007, which is a consequence of much more modest increases in some countries, and some reductions in others.²⁵

²¹ A typical entry of the B matrix b_{ij}^{od} indicates the value of production in industry i located in country o that is required in order to satisfy one unit of final demand for product j in country d , while taking into account direct and indirect intermediate demand from all other using industries.

²² By construction, summing all elements of VBY or of VX gives world GDP, i.e. the value of global expenditures on final goods accrues to primary production factors, which is also equal to their income. Summing all elements within the rows that pertain to a country's industries gives that country's GDP; summing all elements within the columns that pertain to a country's industries gives that country's production of final goods and services (in value added terms).

²³ Table A3 and Table A4 in the appendix report country-level forward GVC integration in 1995, 2007 and 2014. An alternative view of GVC deepening is backward linkages, which imply payments to foreign factors by domestic industries through supply of intermediate inputs and services. Since the world is a closed economy (and this is taken into account in the WIOD data), the global (and, therefore, average) forward and backward linkages are the same, although within each country there can be differences between the two. We discuss backward linkages in Appendix C.

²⁴ Johnson (2018) reviews different input-output approaches to study trade in value added. There are two complementary methodologies that describe how value added is traded, either from the consumption or the production perspective. The production perspective allocates the value added in a given location (and industry) of production to final goods production via global value chains of intermediate inputs. Since this traces the income generated by production of final goods to primary factors through intermediate input flows it is also named "GVC income" decomposition (Timmer, Los, Stehrer, and de Vries (2013)). Johnson (2018) acknowledges that by measuring the domestic and foreign content of domestic production, GVC income is conceptually closely tied to the literature on offshoring and trade in tasks. The third approach to trade in value added focuses on the decomposition of gross exports. Johnson (2018) argues that the theoretical motivations for a gross export decomposition are "unclear" and that some bilateral export decompositions may also generate inconsistencies, e.g., Koopman, Wang, and Wei (2014). For these reasons, the analysis presented in this paper builds on the GVC income approach.

²⁵ The slowdown in forward GVC deepening is consistent with Timmer, Los, Stehrer, and de Vries (2016), who rely

3.2 Upstreamness

Forward GVC participation manifests in a concurrent increase in “upstreamness”, defined as the average number of production stages between production in a particular industry in a given country and final good demand, across all possible value chains (domestic and global). Antràs and Chor (2018) use the following measure of upstreamness

$$U_i^r = 1 \times \frac{Y_i^r}{X_i^r} + 2 \times \frac{\sum_s \sum_j a_{ij}^{rs} Y_j^s}{X_i^r} + 3 \times \frac{\sum_s \sum_j \sum_t \sum_k a_{ij}^{rs} a_{jk}^{st} Y_k^t}{X_i^r} + \dots, \quad (1)$$

where, as above, Y is final demand, X denotes gross output, and a_{ij}^{rs} are technical coefficients, denoting the amount of output of industry i located in country r that is required to produce one unit of output of industry j located in country s . Miller and Temurshoev (2017) show that (1) is equal to the row-sum of the inverse Ghosh (1958) matrix G (which is related to the Leontief inverse B matrix), $U = G\iota$, where ι is a column vector of ones.²⁶ This is useful, because it permits to easily separate the part of upstreamness that is accounted for by domestic linkages and foreign ones:

$$U = U^D + U^F. \quad (2)$$

Here $U^D = G^D\iota$, where G^D is a block-diagonal matrix using only domestic demand. The foreign part is $U^F = [G - G^D]\iota = G^F\iota$.

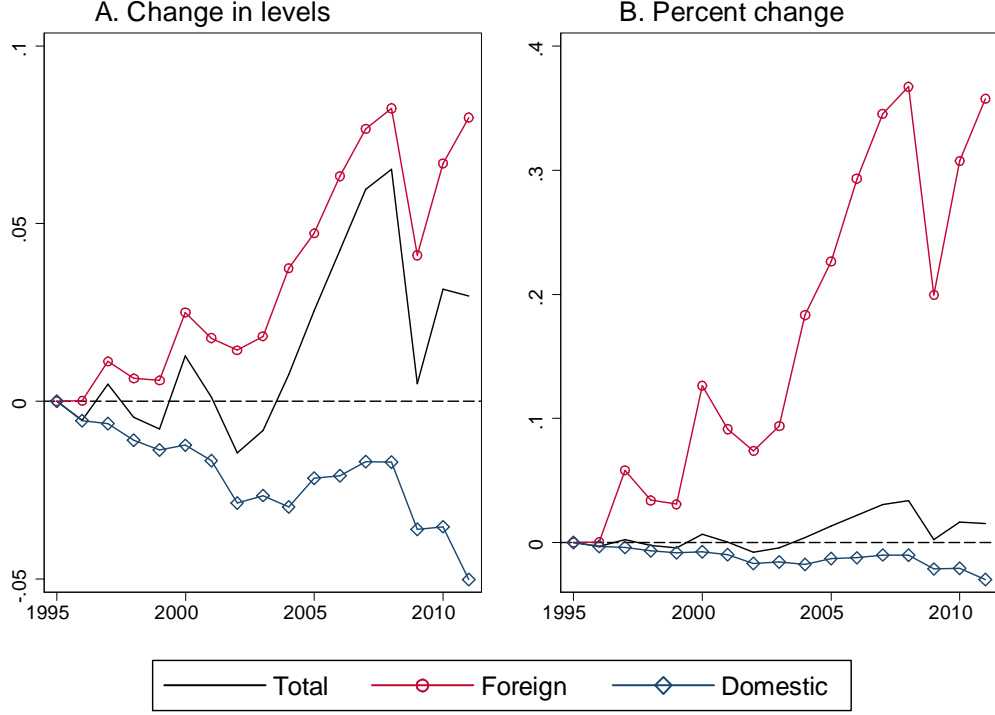
In Figure 2 we illustrate that global upstreamness increases, as in Antràs and Chor (2018), that the increase in U is entirely driven by U^F , as in Miller and Temurshoev (2017), and in addition, that most of this takes place in 2001–2007. Antràs, Chor, Fally, and Hillberry (2012) show that upstreamness is negatively correlated with labor shares in a cross section of U.S. manufacturing industries. We replicate this finding using the WIOD data in Section 5.6 below, which is an assumption in the model of Sposi, Yi, and Zhang (2020). We also show that this correlation holds within industries over time: industries that increase their upstreamness—and in particular foreign-driven upstreamness—exhibit declining labor shares. This is consistent with the mechanism in Sposi, Yi, and Zhang (2020), where in response to a reduction in the cost of geographical production fragmentation, the relatively capital abundant country specializes in more upstream and capital intensive steps in the value chain.²⁷

on the same data. On the manifestation of the "so-called" trade collapse from 2008 on value added trade see Bems, Johnson, and Yi (2011) and Nagengast and Stehrer (2016).

²⁶The "inverse Ghosh" matrix G relates changes in gross output across industries to changes in primary factor use in a particular industry, where primary factor use is value added: $X = G'V$. It is related to the Leontief inverse matrix B by the following formula $XGX^{-1} = B$. See chapter 12 in Miller and Blair (2009) for more details.

²⁷In Appendix D we study the evolution of U , U^D and U^F separately in manufacturing industries and in non-manufacturing industries. This indicates that the of upstreamness in non-manufacturing industries increases less than in manufacturing, but since it increases from a lower level, the growth rate is very similar to that in manufacturing. Importantly, the increase in both sectors is driven predominantly by foreign transactions (U^F). In Appendix D we also report analysis of splitting the foreign component of upstreamness by final using sector. Manufacturing is slightly

Figure 2. Upstreamness and its components



Notes. The displayed series are year fixed effects from a regression of either country-level upstreamness (U_{ct}), its domestic (U_{ct}^D) and foreign (U_{ct}^F) components, or their natural logarithms ($\ln U_{ct}$, $\ln U_{ct}^D$, $\ln U_{ct}^F$) on year fixed effects and country fixed effects, weighted by GDP in 1995. Country-level upstreamness indices are computed as the weighted average of industry values from (2) (using value added as weights), and $U_{ct} = U_{ct}^D + U_{ct}^F$. The sample includes 39 countries that correspond to the WIOD 2013 release, which is also the source of the data.

more important than non-manufacturing industries as a final user for manufacturing as a supplying industry —but the difference is not large. Overall, whether the final step of production is in manufacturing or in nonmanufacturing does not matter much.

4 Accounting for sources of change in labor shares

The goal of this section is to guide the regression analysis in Section 5. First, we wish to know what is the relative importance of within industry changes, changes in the GVC network, and overall composition of final good production in accounting for changes in labor shares. Second, we ask what is the relative importance of changes in global, compared to domestic, value chains. The interested reader is invited to delve into the details below. For the rest, the upshots are as follows.

Both within industry changes in factor intensities and changes in industry composition account for the evolution of labor shares. This leads us to study determinants of changes of labor shares both at the industry and at the national levels. Shifts towards more complex GVCs account for an important part of the decline in labor shares. The greater reliance on foreign sources of demand contributes to lower labor shares. This reinforces our emphasis on foreign sources of variation in explaining the evolution of labor shares.

4.1 Composition versus within-industry intensities

We split value added into labor and capital payments, as follows $V_L BY + V_K BY = V BY$, where V_f are diagonal matrices of factor shares in output.²⁸ Computing factor shares is straightforward. We split factor payments into the part that arises from payments by domestic final goods producers, and payments by foreign final goods producers. The domestic contribution is the block-diagonal part of $V_f BY$. The contribution of foreign industries is the off-block-diagonal part of $V_f BY$, and is akin to numerator of *forward* defined above, differentiated by factor f .

We decompose changes in factor shares into changes in within-industry intensities where factors are located V_f , and industry composition, determined by final good production across locations Y and the network of GVCs B . The change in $V_f BY$ (indeed, of any three conformable matrices) from period 1 to period 2 can be written as

$$\begin{aligned}
 V_{f2} B_2 Y_2 - V_{f1} B_1 Y_1 &= \Delta(V_f BY) \\
 &= (\Delta V_f) B_1 Y_1 + V_{f1} (\Delta B) Y_1 + V_{f1} B_1 (\Delta Y) \\
 &\quad + V_{f1} (\Delta B) (\Delta Y) + (\Delta V_f) B_1 (\Delta Y) + (\Delta V_f) (\Delta B) Y_1 \\
 &\quad + (\Delta V_f) (\Delta B) (\Delta Y) .
 \end{aligned} \tag{3}$$

where Δ denotes the element-by-element change operator.²⁹ Once we perform these decompositions for $V_f BY$, we compute the corresponding factor shares.

Before turning to the results of the decomposition we wish to flag an important caveat to

²⁸See, also Timmer, Erumban, Los, Stehrer, and de Vries (2014).

²⁹See Appendix E for proof.

analyzing changes in V_fBY : factor shares in any industry in any country are invariant to the using industries, or to whether the using industry is domestic or foreign. It is well-known that within industries exporters are more capital and skill intensive; this may bias our analysis of the role of GVC participation in the evolution of labor shares.³⁰

Table 1 reports the results of the decomposition of factor shares using (3) for 1995–2007. Panel A reports this for the entire economy, while Panel B focuses on manufacturing industries, which follow similar patterns but in greater magnitudes.³¹ Columns 1–4 report the shares of income accruing to capital and labor from domestic industries and from foreign industries. All other columns are derived from these. Column 5 reports the overall labor share in GDP. Column 6 reports the share in GDP arising from all international sources (this is the *forward* concept, displayed in Figure 1). Columns 7 and 8 report labor shares in payments by domestic and foreign industries, respectively. All numbers are weighted averages across countries using GDP in 1995 as weights.

Table 1 reveals several facts. First, on average, the increase of 2.45 percent points in capital shares is driven both by domestic industries (0.87 pp), and even more so by payments from foreign industries (1.57 pp). In manufacturing industries the increase of 4.04 percent points in capital shares is almost entirely driven by payments from foreign industries (3.81 pp). The overall decline in labor shares is driven by domestic industries (-3.72 pp), where the increase in payments from foreign industries (+1.27 pp) is far from enough to compensate for this decline (similarly, in manufacturing). Thus, the decline in labor shares is associated with a shift towards foreign sources of income (2.84 pp overall, and 7.13 pp in manufacturing).

Changes in industry composition account for a large part of the decline of labor shares. Of the overall average decline of 2.45 percent points in labor share, 1.06 percent points—or 43 percent of the actual change—are accounted for by within-industry changes in factor shares ($V_{2007}B_{1995}Y_{1995} - VBY_{1995}$).³² Changes in industry composition due to ΔB alone account for 0.47 percent points, and changes due to ΔY account for 0.44 percent points. Together, $\Delta(BY)$ accounts for 0.87 percent points decline in the labor share—which is 35 percent of the change. In manufacturing the split is

³⁰See, e.g., Bernard, Jensen, Redding, and Schott (2007) and Harrigan and Reshef (2015). Another caveat is that we do not make allowances for capital depreciation. International Monetary Fund (2017) demonstrate that although affecting levels, adjusting for this hardly alters trends in factor shares.

³¹Although all factors in Panel B are employed in manufacturing, services industries, both domestic or foreign, can also be a source of income for manufacturing.

³²Within-industry changes can be significantly driven by changes in firm composition, which are associated with globalization. This can be seen by juxtaposing the 4-digit SIC industry level analysis of U.S. manufacturing in Berman, Bound, and Griliches (1994) with the analysis of the plant level data that underlies the 4-digit SIC industries in Bernard and Jensen (1997). Virtually all firm level evidence indicates that exporting firms are significantly more capital and skill intensive (see, for example, Bernard, Jensen, Redding, and Schott (2007) for the U.S., Harrigan and Reshef (2015) in Chile). Therefore, variations in firm composition due to trade liberalization can also lower labor shares within-industries. See also Kehrig and Vincent (2021) for an establishment-level anatomy of the decline in the labor share with in U.S. manufacturing.

37 percent for within-industry changes and 40 percent for composition.³³

Both within industry changes in factor intensities and changes in industry composition account for the evolution of labor shares. This leads us to study determinants of changes of labor shares both at the industry and at the national levels.³⁴

4.2 Complex GVCs and foreign sources of demand

We study the relative importance of foreign value chains captured in B , and of foreign demand for final goods Y . We start by decomposing changes in B using Stone’s additive decomposition (see Appendix F for details). This allows us to write

$$V_f(\Delta B)Y = V_f(\Delta B^d)Y + V_f(\Delta B^x)Y + V_f(\Delta B^g)Y . \quad (4)$$

Here B^d captures all strictly *domestic* input-output linkages, B^x captures *bilateral exports* of intermediate inputs that cross borders only once, and B^g captures *complex global value chains* that cross borders more than once, including “return loops”.³⁵

Final goods production can be written as $Y = Y^d + Y^f$, where Y^d is *domestic* demand by the country producing the final goods and Y^f is *foreign* demand for final goods. Both Y^d and Y^f include demand for domestically-produced goods and for imports. Using this we can write

$$V_f B(\Delta Y) = V_f B(\Delta Y^d) + V_f B(\Delta Y^f) , \quad (5)$$

which allocates demands to production via B and then to factor payments via V_f .

Table 2 displays the results of the analysis for labor shares ($V_f = V_L$) in 1995–2007 for the entire economy level and separately for manufacturing. The four “Total” rows in columns 1–3 report labor shares in GDP that are paid by domestic industries, foreign industries, and overall in 1995; these are the same numbers in columns 2, 4 and 5 in Table 1. The “Total” rows in columns 4–6 report the changes in the same concepts. The rows above the “Total” rows indicate the contributions of

³³Karabarbounis and Neiman (2014) and International Monetary Fund (2017) find a smaller role for industry composition. There are at least two reasons for this. The first is that they use industry value added shares to aggregate industry-level value added labor shares. This can generate misleading results on the role of composition, because it does not take into account changes in composition due to sourcing decisions. The second reason for differences in results is variation in data sources, measurement, and level of aggregation. International Monetary Fund (2017) use only 10 industries, which mechanically causes more variation to occur within industries compared to our data, which include 35 industries. In the limit, if there is only one industry, all of the variation is within this single industry. The sample of countries is also different across studies. Karabarbounis and Neiman (2014) consider value added shares in corporate income, while we consider the entire economy.

³⁴We performed the same analysis for factor shares in 2007–2014 (Table A5 in the appendix). The main differences are that labor shares increase by 1 percent points, on average, in contrast to the decline in 1995–2007, and that within-industry changes account for more than all of the increase in labor shares in this period, while composition accounts for small reductions (in the same direction as in 1995–2007).

³⁵For example, consider a hypothetical German car door producer that ships doors to Czech Republic, where windows are manufactured and installed in the doors, which get shipped back to Germany and installed in cars that are either purchased domestically or are exported.

sub-components of either B or Y to their levels and changes in the corresponding columns.

We start with the Stone decomposition of B in levels. Overall, most payments to labor in 1995 are generated due to domestic linkages (91% for all industries and 80% in manufacturing). Almost all of labor payments originating in domestic industries occurs via to domestic linkages (B^d), while most of the demand from foreign industries occurs due to bilateral trade linkages (B^x) (roughly 84%). Complex GVCs (B^g) originate mostly from foreign industries; return loop value chains are much less important. These findings are consistent with Miroudot and Nordstrom (2015).

The novel findings are the contributions to changes, using (4). Column (6) of Table 1 implies that increasing complexity in how world demand for final goods is met by production (recall that Y is held constant) accounts for part of the decline in labor shares. The decline in labor shares in 1995–2007 that is due to ΔB is driven by a reduction in income from domestic industries (ΔB^d) that is not fully counterbalanced by both exports of intermediates (ΔB^x) and by more complex GVCs (ΔB^g). Complex GVCs account for more than 60% of this shift; in manufacturing the corresponding figure is more than 81%.

Turning to the breakdown of Y , we see that domestic demand (Y^d) accounts for the lion’s share of labor payments in 1995 (93% overall and 80% in manufacturing), although this declines by 2007 (91% overall and 74% in manufacturing). Changes in foreign demand (ΔY^f) in 1995–2007 increase labor payments both through domestic and foreign industries; the latter is due to how domestic labor participates in GVCs supplying foreign industries. Interestingly, changes in domestic demand (ΔY^d) reduce labor payments through domestic industries, while concurrently contributing to an increase in labor payments due to domestic demand for foreign final goods. The latter is due to domestic industries supplying labor intensive inputs to foreign industries. The changes in manufacturing are larger, and overall in the same direction as the entire economy.

Table 2 delivers two messages. The first is that shifts towards more complex GVCs account for an important part of the decline in labor shares. The second is that the greater reliance on foreign sources of income demand contributes to lower labor shares.³⁶

³⁶We performed the same analysis in 2007–2014 (Table A6 in the appendix). The main difference is that in manufacturing changes in both domestic (ΔB^d) and foreign linkages (ΔB^x and ΔB^g) contribute to declines in labor shares. In addition, changes in foreign demand in 2007–2014 account for small declines in labor shares, not increases as in 1995–2007.

5 Explaining changes in labor shares

In this section we study the role of forward GVC participation and juxtapose it with other explanations. We start with fitting the following industry-by-country regression in a panel of changes

$$\begin{aligned}\Delta LS_{ic,t} = & \beta_1 \Delta FWD_{ic,t} + \beta_2 \Delta BACK_{ic,t} + \beta_3 \Delta EXP_{ic,t} + \beta_4 \Delta IMP_{ic,t} \\ & + \beta_5 \Delta \ln q_{c,t} + \text{fixed effects} + \varepsilon_{ic,t},\end{aligned}\tag{6}$$

where ΔLS_{ic} is the change in labor share in industry i located in country c . The first four variables on the right hand side of (6) capture dimensions of globalization: $FWD_{ic} = forward_{ic}/VA_{ic}$ is the share of exports of intermediate inputs in value added, i.e., forward foreign GVC intensity (the industry-country equivalent of the country-level variable examined in Section 3.1); $BACK_{ic} = foreign_inputs_{ic}/inputs_{ic}$ is the share of imported inputs in total input purchases, i.e., backward foreign GVC intensity; $EXP_{ic} = y_{ic}^{-c}/VA_{ic}$ is export intensity of final goods in value added (y_{ic}^{-c} is final demand for industry i in country c from other countries $-c$); and $IMP_{ic} = y_{i,-c}^c/absorption_{i,c}$ is import intensity of final goods in total absorption ($y_{i,-c}^c$ is demand of country c for final goods i produced in other countries $-c$; and $absorption_{i,c}$ is given by local production plus imports minus exports of final good i in country c : $y_{i,c} + y_{i,-c}^c - y_{ic}^{-c}$). Finally, q_c is the relative price of capital equipment investment (only country-level variation).³⁷ ³⁸

We estimate versions of (6) for manufacturing industries and then for all private sector industries in 1995–2014. While we expect GVC deepening to affect manufacturing, GVCs should affect services and other non-traded sectors through input linkages. We split the sample into three periods: 1995–2001, 2001–2007 and 2007–2014. The split is determined by the timing of changes in the rate of decline of the labor share, which are evident in Figure 1. In the Appendix I we present a procedure that identifies 2001 as a natural split for the 1995–2007 sub-sample. The second split is mandated by the fact that cover the period before 2007 and the period after 2007 with different releases of the WIOD (see discussion in Section 3).

We report regressions with increasingly demanding combinations of fixed effects that absorb time, country and industry trends in labour shares. Our most demanding specifications of (6)

³⁷The data for q are from the Penn World Tables mark 9.0 (PWT, see Feenstra, Inklaar, and Timmer (2015)) and from the United States' Bureau of Economic Analysis (BEA), applying the same methodology as in Karabarbounis and Neiman (2014). See Appendix G for more details.

³⁸In order try to address the relationship between firm market share concentration and labor shares highlighted in Autor, Dorn, Katz, Patterson, and Van Reenen (2020) we experimented with adding to (6) the log of the number of firms at the country-by-industry level using data from UNIDO's INDSTAT 2019 edition database (we harmonize the 2-digit ISIC rev. 3 industry definition In INDSTAT to match the WIOD 2013 classification). This variable is very weakly correlated with labor shares and adding it does not alter the other coefficients. This is clearly an imperfect indicator of concentration, but we could not find other sources to compute industry concentration for so many industries and countries from 1995. For example, Orbis would allow us to do so only from 2004 and only for publicly available firms.

include country-industry fixed effects (with and without time fixed effects), which absorb any trends within these units, e.g., within-industry and country trends in use of housing (Gutiérrez and Piton (2019)). We report robust standard errors clustered by country and by industry (two-way clusters), to accommodate the fact that q_c varies only by country (Moulton (1990)), and to account for any cross-country correlations in the errors within industries (Cameron, Gelbach, and Miller (2011)).

We estimate (6) by weighted least squares (WLS) with VA_{ic} in 1995 as weights (choosing different years for weights changes little the results). This gives the regression coefficients an aggregate interpretation.³⁹ We also estimate versions of (6) at the country level, in order to address country-level labor shares directly. The difference between the two specifications is that the latter also allows for impacts on industry composition. We develop an instrument to address concerns for identification, as described below, and apply it both at the country-industry and at the country levels.

We find that the only robust explanatory variable for declines in labor shares is forward GVC participation. This happens through changes in sub-industry output mix.

Before turning to the results in detail, it is worth dwelling on why we do not identify an effect of backward GVC participation on labor shares, since every forward GVC flow for one country and industry is a backward flow for another.⁴⁰ The (vast) literature on the effects of offshoring focuses on different types of labor (skilled versus unskilled, etc.). We are not aware of a study that studies the effect of offshoring on the labor share.

We borrow from Sposi, Yi, and Zhang (2020), to which we add a notion of substitution between capital and labor, as in the production function used to derive equation (7) below. To keep things simple, imagine that there is only one industry and that within it there are only two stages of production: upstream component production and downstream assembly. Upstream production of components requires only capital and labor:

$$M = [\alpha K^\rho + (1 - \alpha) L^\rho]^{1/\rho}, \quad \rho \leq 1.$$

Assembly requires inputs in the form of components, and then again capital and labor, but at a lower capital intensity:

$$Y = F\left(M, [\beta K^\rho + (1 - \beta) L^\rho]^{1/\rho}\right)$$

Here we assume $\alpha > \beta$ and for simplicity we use the same elasticity of substitution for the K - L

³⁹See Solon, Haider, and Wooldridge (2015) on motivations for weighting.

⁴⁰This is true both conceptually and, as we verified in our data, at the aggregate global level: the aggregate value of backward GVC flows equals the aggregate value of forward GVC flows.

composite $\sigma = 1/(1 - \rho)$ (we cannot distinguish different elasticities in the data). Initially, both stages of production must occur in the same location, so that countries trade only final goods Y . Once it becomes feasible to separate the location of upstream component production and downstream assembly, countries specialize in their cost-driven comparative advantage. Which way each country goes will depend on factor endowments, and we address this part in the paper in Section 5.5. *Ceteris paribus*, countries that specialize in upstream production will both export more components and see a decrease in their labor share since $\alpha > \beta$; countries that specialize in downstream assembly will both export more final goods and see an increase in their labor share.

The forward GVC participation measure captures this change in the composition of activities within a country and industry. Our main interpretation of the negative regression coefficient is that increases in these measures reduce labor shares. But it can equally be interpreted as reductions in forward GVC participation intensity (read: more downstream assembly) increase the labor share.

Backward GVC participation is the share of payments to foreign primary factors due to the importation of intermediate inputs in total input expenditures. It is a share of expenditures on M (this is precisely correct in the framework above, because only primary factors can contribute to producing M). While an increase in this intensity is intuitively associated with more downstream assembly, this need not be the case and, in this instance, our “intuition” sets us off track.

The reasons for this are fourfold. First, backward GVC participation is an intensity within some level of M . An increase in downstream assembly should be associated with an increase in the overall use of M , but this is not captured in the backward GVC participation *intensity*. Second, there may be unmeasured quality differences between imported and domestic inputs, which make the expenditure share mis-measure the true productive services of inputs. Third, in a model with multiple production steps, the relationship between industry-level offshoring and changes in upstream/downstream intra-industry composition may be muted due to intra-value chain variation in input use intensity. Fourth, and perhaps most importantly, in our regressions we control for forward GVC participation, which “fixes”, so to speak, the sub-industry composition. If the main reason for which intra-industry labor shares change is compositional changes, then once this is fixed, the partial effect of backward participation is nil. Here we have an issue of differences between what model equilibrium predicts and what a regression can identify.

5.1 Baseline WLS estimates

Table 3 presents the baseline results. We start with specifications without any fixed effects, and then increase the saturation of the model. Including country-industry and period fixed effects absorbs specific panel dimension trends in GVC participation. Including country-period and industry fixed effects addresses, for example, general equilibrium constraints within a country. The spec-

ification with country and industry-period fixed effects identifies the regression coefficients from cross-country variation within industries and years, for example, due to country-specific policies or location.⁴¹

Overall, only the coefficient to ΔFWD has a similar magnitude and is statistically significant across almost all specifications, which considerably alleviates concerns for omitted variables bias. The only exceptions are in columns 8 and 9—but even there, once we instrument for ΔFWD we get large negative and statistically significant coefficient estimates (see the following subsection). The coefficient to $\Delta BACK$ is negative, but not precisely estimated and is smaller than that to ΔFWD , indicating a limited independent role for offshoring in reductions in labor shares.

The largest effect for manufacturing industries is found in column 7; it implies that a one percent faster increase in forward foreign GVC participation is associated with 0.43 percent point faster decline the labor share. Perhaps not surprisingly, the effect is smaller when considering all private sector industries, at 0.28. These are economically large effects. A manufacturing industry in a given country at the third quartile of ΔFWD sees its labor share drop by 2 percent points more than an industry at the first quartile, which is a fifth of the inter-quartile range for ΔLS . The equivalent computation for all private sector industries gives about half as much explanatory power.⁴²

5.2 Baseline W2SLS estimates

The results above imply a strong relationship between increases in forward GVC participation and declines in labor shares. Here we address concerns for causal interpretation of this relationship that may arise despite the high saturation of fixed effects. The main concern for endogeneity arises from unobserved biased technological change, or from differential reductions in the rental rate of capital. To make this concrete, suppose that the value added production function in some industry is

$$Y = [\alpha (\Omega_K K)^\rho + (1 - \alpha) (\Omega_L L)^\rho]^{1/\rho}, \quad \rho \leq 1,$$

where Ω_K and Ω_L capture capital and labor augmenting technological change, respectively. Then cost minimization in competitive factor markets implies that the labor share can be written as

$$LS = 1 - \alpha^\sigma (\Omega_K / r)^{\sigma-1}, \tag{7}$$

where r is the rental rate of capital and $\sigma = 1/(1 - \rho)$ is the elasticity of substitution between capital and labor. An increase in Ω_K or reduction in r will make the industry more competitive

⁴¹For example, the centrality-downstreamness nexus in Antràs and De Gortari (2020).

⁴²See descriptive statistics in Table A7. We make sure that our results are not driven by outliers. In unreported results we drop the top and bottom 1% observations of ΔFWD ; these yield virtually identical results.

and it will export more, and at the same time have a direct effect on LS unless $\sigma = 1$ (similar arguments can be made for changes in Ω_L and wages). We do not take a stand on whether σ is above or below unity, which implies a different direction of bias for the estimator of (6). Bias would similarly arise in the context of a gross output production function if intermediate inputs are not equally complementary to capital and labor, and their price declines due to, e.g., importing of intermediate inputs.

In order to deal with such concerns we construct an instrument based on structural gravity that is inspired by Aichele and Heiland (2018). We start with the technical coefficients matrix $A = [a_{ci}^{dj}]$, where c is a source country, d is a destination country, and i and j denote industries. The data underlying A are flows of intermediate inputs. We use tools from structural gravity to construct *changes* in A that are not influenced by *changes* in the source country and industry, thus eliminating technological change, but allowing for changes in destination and bilateral forces. We use this to construct changes in $B = (I - A)^{-1}$ and in value added flows VBV that are orthogonal to the direct effects of technological change or input prices.

Our objects of interest are the technical coefficients a_{ci}^{dj} , which are also the share of inputs that using industry j in destination country d sources from supplying industry i located in country c . We model these along the lines of Eaton and Kortum (2002),

$$\begin{aligned} a_{ci}^{dj} &= \frac{T_{ci}^j C_{ci}^{-\theta_i} (\tau_{ci}^{dj})^{-\theta_i}}{\sum_{c'i} (T_{c'i}^j / C_{c'i})^{\theta_i} (\tau_{c'i}^{dj})^{-\theta_i}} \\ &= \exp\left\{ \underbrace{\ln T_{ci}^j C_{ci}^{-\theta_i}}_{\alpha_{ci}^j} + \underbrace{[-\ln \sum_{c'i} (T_{c'i}^j / C_{c'i})^{\theta_i} (\tau_{c'i}^{dj})^{-\theta_i}]}_{\beta_i^{dj}} + \underbrace{(\tau_{ci}^{dj})^{-\theta_i}}_{\gamma_{ci}^{dj}} \right\}, \end{aligned} \quad (8)$$

where T denotes the level of technology, τ denotes bilateral trade barriers, the unit cost terms C may include inputs (both domestic and imported) and domestic primary factors, as in Caliendo and Parro (2015), and we allow different elasticities θ by industry. Note that in contrast to Caliendo and Parro (2015), we also allow the technology to differ by using industry. Note that the gravity methodology described in (8) cannot be applied directly to VBV , because bilateral value added flows cannot be decomposed into source, destination and bilateral components that are log-linearly separable. GVCs are influenced by all countries' productivities and costs and all bilateral trade barriers—potentially more than once, with different elasticities and in a nonlinear fashion.

We estimate (8) by PPML in five periods $t = 1995, 2001, 2007, 2007'$ and 2014, where 2007' denotes data from WIOD 2016:

$$a(t)_{ci}^{dj} = e^{\alpha(t)_{ci}^j + \beta(t)_i^{dj} + \gamma(t)_{ci}^{dj}} + \epsilon(t)_{ci}^{dj}, \quad (9)$$

where $\epsilon(t)_{ci}^{dj}$ is the deviation of the model from the data. We must make some restrictions in order to estimate bilateral trade barriers term $\gamma(t)_{ci}^{dj}$. We entertain three options. In the first, we set $\gamma(t)_{ci}^{dj} = \gamma_{ci}^{dj}$, which implies that we absorb all time-invariant trade barriers at the ci - dj level. In the second option we set $\gamma(t)_{ci}^{dj} = \gamma(t)_c^d$, which absorbs bilateral country-level trends. In these two options we estimate (9) as a three-way gravity model.⁴³ In the third option, we set $\gamma(t)_{ci}^{dj} = \gamma_1(t) \cdot \ln(\text{distance}_{cd}) + \gamma_2(t) \cdot \text{border}_{cd}$, where $\text{border}_{cd} = 1$ if c and d share a land border and zero otherwise. In this case we estimate (9) separately for each period t .⁴⁴ The 2SLS estimates using all three options are qualitatively similar. All three instruments remain strong across country-industry specifications and in most country-level specifications below, especially in manufacturing. We present results from the third option because it delivered first stage statistics that indicate a somewhat stronger instrument. All other results are available upon request.

Using the estimates we can construct

$$\begin{aligned} \text{Source} &: \hat{T}(t)_{ci}^j \hat{C}(t)_{ci}^{-\theta_i} = e^{\hat{\alpha}(t)_{ci}^j} \\ \text{Destination} &: \sum_{c'} \hat{T}(t)_{c'i}^j \hat{C}(t)_{c'i}^{-\theta_i} [\hat{\tau}(t)_{ci}^{dj}]^{-\theta_i} = \hat{\Phi}(t)_i^{dj} = e^{-\hat{\beta}(t)_i^{dj}} \\ \text{Bilateral} &: [\hat{\tau}(t)_{ci}^{dj}]^{-\theta_i} = e^{\hat{\gamma}_1(t) \cdot \ln(\text{distance}_{cd}) + \hat{\gamma}_2(t) \cdot \text{border}_{cd}}. \end{aligned}$$

For each period t (except for 1995) we build a counterfactual $\tilde{a}_{ci}^{dj}(t)$ by replacing the period t source elements by those estimated in period $t-1$. In doing so, we also replace for each destination country d and industry j the corresponding $\hat{T}(t)_{ci}^j \hat{C}(t)_{ci}^{-\theta_i}$ element inside $\hat{\Phi}(t)_i^{dj}$ with $\hat{T}(t-1)_{ci}^j \hat{C}(t-1)_{ci}^{-\theta_i}$ (which makes it vary also by source country c , $\tilde{\Phi}(t)_{ci}^{dj}$). Once this is done we build $\tilde{A}(t)$. We use this to build $\tilde{B}(t) = [I - \tilde{A}(t)]^{-1}$ and $V(t) \tilde{B}(t) Y(t)$ from which we compute $\widetilde{FWD}(t)$ using the same methodology as above. Our instrument is

$$\Delta Z_{ci,t} = \widetilde{FWD}(t)_{ci} - FWD(t-1)_{ci}. \quad (10)$$

The changes $\Delta Z_{ci,t}$ are driven by variation in demand and variation in bilateral trade barriers, and are orthogonal to variation in source country c and industry i supply shocks, which may be directly correlated with labor shares.

Table 4 reports the results of estimating (6) by weighted 2SLS (W2SLS), using (10) as an instrument. Overall, the instrument is strong, as indicated by the Kleibergen-Paap F statistic. We relegate the first stage regressions to Table A8 in the appendix, where we get the expected positive

⁴³Weidner and Zylkin (2021) show that when the time dimension is fixed, the point estimates produced by the three-way PPML gravity model have an asymptotic incidental parameter bias of order $1/N$, where N in our context is equal to $1170 = 39 \text{ countries} \times 30 \text{ industries}$. Thus, the bias is of order $1/1170 = 0.0008$, small enough to ignore.

⁴⁴In practice, the source and destination fixed effects are identified only up to a constant. We allocate this constant to $\hat{\tau}(t)_{ci}^{dj}$. Since a_{ci}^{dj} are shares and we estimate (9) by PPML, which preserves the sum of shares “within fixed effects”, this has no consequences for our procedure.

sign for the coefficient to ΔZ .

The W2SLS estimates of the coefficients to ΔFWD are negative and much larger than the WLS estimates in absolute value. This is consistent with upward bias in the WLS estimates in Table 3, which can arise from three reasons. First, classic measurement error would bias the estimator of the coefficient to ΔFWD towards zero. Second, measurement error in value added alone would bias the estimator of the coefficient to ΔFWD towards 1 (it is in the denominator of both ΔFWD and ΔLS). Third, unmeasured capital-augmenting technological change (an increase in Ω_K) or changes in the rental rate of capital r (despite controlling for $\Delta \ln q$ at the country level) would tend to increase the labor share if the elasticity of substitution between capital and labor is smaller than 1, while boosting exports.⁴⁵

We now discuss some threats to the validity of our instrument. If productivity growth is spatially correlated and factor-biased, then we may not be purging the counterfactual \widetilde{FWD} from correlation between export growth and factor bias. However, we find virtually no spatial correlation across source factors $\hat{\alpha}(t)_{ci}^j$ within a period, and especially not in changes, which is more important for the validity of the instrument. We explain in Appendix H our procedure for testing this.

Next, in order to assess the importance of potential violations of the exclusion restriction, we applied the Local-to-Zero (LTZ) approximation of Conley, Hansen, and Rossi (2012). We use this methodology to compute confidence intervals for the coefficient to ΔFWD that take into account uncertainty about the validity of the exclusion restriction for ΔZ . We first estimate by WLS a version of (6) that includes the instrument (10), and we store the standard error of the coefficient to the instrument, $\hat{\gamma}$, denoted $\hat{\sigma}_{\hat{\gamma}}$. We then use the LTZ approximation to construct confidence intervals for the W2SLS coefficient to ΔFWD using different multiples of $\hat{\sigma}_{\hat{\gamma}}$ —1, 10, 25 and 50—and two distribution functions for γ —normal and uniform. The confidence intervals do not include zero even at 50 times $\hat{\sigma}_{\hat{\gamma}}$. This indicates that even if the instrument is not strictly exogenous, it is strong enough to justify small violations of the exclusion restriction, and is thus useful for inference.⁴⁶

We checked whether outliers drive our results. We therefore repeatedly estimated (6), each time dropping a country, and industry, or a country-industry pair. The point estimates are generally not sensitive to these sample restrictions. In very few cases omitting a $c-i$ pair changes the point estimate by a significant magnitude. Even in these cases the point estimates remain negative and statistically significant. We report these in Appendix J.

⁴⁵Using (7), $\partial LS / \partial \Omega_K = -(\sigma - 1) \alpha^\sigma r^{\sigma-1} \Omega_K^{\sigma-2} > 0$ if $\sigma < 1$.

⁴⁶See Section III.C in Conley, Hansen, and Rossi (2012) for a discussion of this last point and of the LTZ approximation in general. Results for the WLS estimates of (6) that include the instrument (10), as well as the confidence intervals that use the LTZ approximation are reported in the appendix in Table A13 and Table A14, respectively.

Finally, we entertain specifications of (6) where we instrument for multiple variables. While this yields results that are in line with those reported in Table 4, they are much weaker, statistically: standard errors are generally larger and we never reject the null of weak instruments. We note that tests for weak instruments in the presence of multiple instruments are not as powerful as when there is only one endogenous variable (Stock and Yogo (2005)), or rest on a restrictive homoscedasticity assumption (Sanderson and Windmeijer (2016)). Therefore, we have less confidence in these results.

First, we add an instrument for $\Delta BACK$ by using the same $\tilde{A}(t)$ that was used to build the instrument for ΔFWD . The results are similar to those in Table 4 (see Table A9 and Table A10 in the appendix); the only difference is that in the sample that includes all private sector industries backward GVC integration weakly predicts higher labor shares (in the manufacturing sample it does not). Then we add also instruments for ΔEXP and ΔIMP . The instruments for the former are constructed by using a similar procedure to what we use to build $\tilde{A}(t)$.⁴⁷ The results are again similar to those in Table 4 (see Table A11 in the appendix), with negative and often statistically significant coefficients for ΔFWD and imprecise estimates of the coefficients to all other regressors.

5.3 The role of China

After establishing that the negative relationship between labor shares and forward GVC participation is causal in Table 4 we turn to the role of China in driving them. We show that the effect of forward GVC participation on reductions in labor shares is concentrated in 2001–2007, the period starting with China’s accession to the WTO, in which its participation into GVCs accelerated, and in which the decline in labor shares accelerates (Figure 1).

We report the results in Table 5, using the most saturated model; results from specifications with fewer fixed effects deliver the same message.⁴⁸ In columns 1 and 6 we repeat the baseline WLS results from Table 3, column 7. In columns 2 and 7 we interact ΔFWD with period fixed effects and find that the effect is present only in 2001–2007. In columns 3 and 8 we split ΔFWD into the part which is imported by China and another part that is imported by the rest of the world (RoW). The coefficient to ΔFWD to China is large and precisely estimated, while the coefficient to ΔFWD to RoW is smaller and not statistically significant at conventional levels of significance. In columns 4 and 9 we interact ΔFWD to China and to RoW with period fixed effects. In manufacturing we find that the effect of both ΔFWD to China and to RoW are precisely estimated only in 2001–2007. Although an F-test of equality of coefficients yields a p-value of 0.2 (the one-sided test yields a

⁴⁷We estimate by PPML gravity regressions for final good production $y_{ic}^d(t)$ with origin-country $c \times$ industry i fixed effects, destination-country $d \times$ industry i fixed effects and controlling for bilateral trade barrier proxies—for several cross-sections of t . We then predict $\tilde{y}_{ic}^d(t)$ by replacing source country $c \times$ industry i fixed effects in t by their lagged values. We build instruments for changes ΔEXP and ΔIMP by using $y_{ic}^d(t-1)$ and $\tilde{y}_{ic}^d(t)$, where $t = 1995, 2001, 2007, 2007'$ and 2014, where 2007' denotes data from WIOD 2016.

⁴⁸These results are available upon request.

p-value of 0.09), the coefficient for China is more than twice as large as the coefficient for RoW. The results are similar when we add all other private sector industries, with the only difference being that we also estimate an equally strong effect of China in 1995–2001. Here we reject the equality of the coefficients to China and RoW in 2001–2007 at the 5% level of significance.

Overall, the 2001–2007 period was one of rapid change, both in overall intensity of GVC participation (as seen in Figure 1), as well as in nature of GVC participation, following the accession of China to the WTO. The results reported in Table 5 imply that these changes have had strong effects on lowering labor shares for industries that integrated into GVCs, and in particular with China as a final downstream destination of exports of intermediate inputs.

5.4 Country-level regressions

The decompositions results in Table 1 imply that both within-industry changes and change in industry composition account for reductions in labor shares, and in equal measure. In order to take into account both margins, we estimate (6) at the country level.

We report the results in Table 6. Columns 1–4 report WLS estimates of (6) using value added in 1995 as weights, with different configurations of fixed effects. We report robust standard errors clustered by country. Overall, the results resemble those in Table 4, and the coefficients are larger for both manufacturing and for the entire economy. While the estimates are precise in all specifications in columns 1–4 in manufacturing, including both country and period fixed effects reduces precision for the entire economy, although the point estimate changes little.

In addition, we estimate (6) by W2SLS using the same instrument (10) aggregated up to manufacturing or the entire economy, reported in Table 6 in column 5, with first stage regressions reported in column 6. We report only the specification with country fixed effects. Different configurations of fixed effects yield similar results, and are reported in Table A12 in the appendix, but as with WLS, using both country and period fixed effects for the entire economy yields an imprecisely estimated coefficient to ΔFWD —but not in manufacturing. As with the W2SLS regressions at the country-industry panel, the instrument is not weak and the estimated coefficient to ΔFWD is larger compared to WLS. We discuss above in Section 5.2 potential explanations for this larger effect.⁴⁹

In unreported regressions we estimate that for manufacturing the effect of ΔFWD is concentrated in 2001–2007, with a greater effect for ΔFWD when China is the importer—similar to the results discussed in Section 5.3 and reported in Table 5 (and with larger coefficients). However, when estimating the same specifications for the entire economy, the coefficients to ΔFWD across

⁴⁹Different specifications of the instrument, discussed above in Section 5.2, yield similar results, although they prove to be significantly weaker instruments when applied to the entire economy, compared to manufacturing.

periods and destinations are much less precisely estimated.

Overall, the results at the country level corroborate those at the country-industry level.

5.5 The role of endowments

In this section we ask whether the patterns of forward GVC participation are associated with relative factor abundance (RKA). If relatively upstream production is more capital intensive (we corroborate this in Section 5.6), then Heckscher-Ohlin logic would imply that as trade barriers decline and GVCs rise, upstream activities should be attracted to relatively capital abundant countries. These countries should, therefore, export more intermediate inputs that are relatively capital intensive to less capital abundant countries, where labor-intensive downstream stages take place. This is the mechanism underlying the model in Sposi, Yi, and Zhang (2020).

In order to explore these ideas empirically we start by estimating the following regression

$$\frac{(V_K BY)_{od} - (V_L BY)_{od}}{GDP^o} = \beta \cdot RKA_{od} + \gamma' gravity_{od} + \alpha_o + \alpha_d + \varepsilon_{od} , \quad (11)$$

where $V_K BY_{od}$ is income accruing to capital installed in origin country o that originates from supplying intermediate inputs for final goods production in destination country d , and similarly for labor income in $V_L BY_{od}$. The dependent variable expresses the capital intensity (in value added terms) of intermediate input exports from o to d . We define relative capital abundance as

$$RKA_{od} = \ln \left(\frac{E_o^K}{E_o^K + E_d^K} \right) - \ln \left(\frac{E_o^L}{E_o^L + E_d^L} \right) ,$$

where E_i^f are expenditures on factor $f \in \{K, L\}$ in country $i \in \{o, d\}$. The coefficient of interest is β , and given the discussion above, we expect a positive relationship. We control for standard bilateral control variables in $gravity_{od}$: log distance, and indicators for a common border, colonial ties, common language, free trade agreements, and common currency.⁵⁰ We also include in (11) origin and destination fixed effects to control for overall attractiveness of o for capital intensive production, and overall tendency of d to produce final goods with capital intensive inputs structure (the source of income flowing from d to o). The results in Table 5 indicate that the effects of forward GVC participation are present only in 2001–2007, so we study these effects only in that period. We estimate (11) by OLS in two cross sections, in 2001 and in 2007. We report robust two-way clustered standard errors at the country o and country d level.

We report the results in columns 1 and 2 of Table 7. We find that greater RKA is associated with greater capital intensity of exports of intermediate inputs from o to d , and that this association has strengthened a bit from 2001 to 2007. Since we include both source and destination fixed effects,

⁵⁰Data from the CEPII gravity dataset.

this result is not driven by overall factor abundance *per se*.

We now turn to discuss the volume of these exports. We estimate gravity regressions

$$VBY_{od} = \exp\{\beta \cdot RKA_{od} + \gamma' gravity_{od} + \alpha_o + \alpha_d\} + \varepsilon_{od} , \quad (12)$$

where VBY_{od} is the value added of exports of intermediate inputs from origin country o used in destination country d for final good production. The other variables are as in (11). We estimate (12) by PPML, and report robust two-way clustered standard errors at the country o and country d level.

We see in columns 3 and 4 of Table 7 that greater RKA is strongly associated with greater exports of intermediate goods, and that this association, too, has strengthened from 2001 to 2007. We interpret this strengthening of the effect of RKA on both the volume and the capital intensity of forward GVC participation as a consequence of reductions in barriers to the geographical spread of GVCs, in line with the predictions in Sposi, Yi, and Zhang (2020).

Finally, we investigate whether the effect of RKA is different for direct exports, VB^xY_{od} , compared to indirect exports of intermediate inputs, VB^gY_{od} , where $VBY_{od} = VB^xY_{od} + VB^gY_{od}$ for $o \neq d$ (see (4) above). We see in columns 5 and 6 of Table 7 a stronger association of RKA with direct exports of intermediate goods, VB^xY_{od} , which is reasonable given the bilateral nature of RKA_{od} (and, similarly, stronger effects of distance and common language), versus the more complex and indirect GVCs involved in VB^gY_{od} .⁵¹

Overall, the results in Table 7 imply that the pattern of expansion of forward GVC participation is responsive to factor endowments. Relatively capital abundant countries tend to integrate forward, increasing the capital intensity of their exports of intermediate inputs, which contributes to lower labor shares.

5.6 Changes in positioning within GVCs: upstreamness

Industries and countries that increased the intensity of supplying intermediate inputs to foreign industries have seen their labor share decline. This is consistent with the idea that production that is more upstream, i.e. farther away from final goods production or final assembly is less labor intensive. It is also the key assumption in Sposi, Yi, and Zhang (2020). Here we investigate this assumption.

We fit two sets of regressions using data in 2001–2007, the period in which the relationship between forward GVC participation and labor shares is strongest. First we fit country by industry

⁵¹We also see that membership in the same free trade agreement and sharing a common currency increases VB^gY_{od} but not VB^xY_{od} . This may be associated with the fact that these integration tools are multilateral in nature, fostering a greater "pool" of countries through which intermediate inputs can flow at low costs. The negative association of common border with VB^gY_{od} is mechanical, since VB^gY_{od} necessarily passes through third countries.

by year panel regressions. Then we fit cross section long difference regressions of changes from 2001 to 2007. The dependent variable is labor shares and the explanatory variables are either upstreamness (1) or its domestic and foreign components (2) (in levels or in changes). We also control for downstreamness, which is computed in an analogous way as upstreamness, except that it pertains to distance from the first stage of production. This control is important since, as Antràs and Chor (2018) demonstrate, upstreamness and downstreamness are positively correlated, and together they indicate the overall position of an industry within its average value chain. We use different sets of fixed effects (FEs) in the panel regressions in order to illustrate different sources of variation. We report robust standard errors clustered both by country and by industry.

We report the results in Table 8. We start with manufacturing industries. Column 1, with only year fixed effects, demonstrates that upstreamness (U) is associated with lower labor shares in the cross section—but this is not precisely estimated. This is consistent with Antràs, Chor, Fally, and Hillberry (2012), who document a similar negative correlation in a cross section of U.S. manufacturing industries. Since we control for downstreamness, this implies that length (or fragmentation) of value chains *per se* is not strongly associated with labor shares. In columns 2 and 3 we find that this association is stronger over time within countries and industries. In column 4 we find that the domestic component of upstreamness (U^D) is associated with lower labor shares in the cross section. In contrast, in columns 5 and 6 we see that increases in the foreign component of upstreamness (U^F) are associated with decreases in labor shares over time, not changes in the domestic component. These results are corroborated in columns 7 and 8 using long differences, where we see that the association between upstreamness and declines in labor shares is driven by the foreign component. When we turn to the results for all private sector industries we find similar patterns. The only substantive difference is that we find a significant positive association between increases domestic upstreamness and changes in labor shares in column 8.

Overall, the results show that increases in upstreamness that are driven by exports of intermediate inputs to foreign industries are negatively associated with labor shares. As industries re-position farther away from final good production and assembly, they require less labor and more capital.

5.7 Changes in functional specialization

In this section we ask what types of labor tasks account for the declines in labor shares documented above. We draw on data from Timmer, Miroudot, and de Vries (2019), that allows splitting labor income into four categories of functional specialization: management (MGT), research and

development ($R\&D$), fabrication (FAB), and marketing (MKT). This allows us to write

$$\Delta LS_{ic} = \Delta MGT_{ic} + \Delta R\&D_{ic} + \Delta FAB_{ic} + \Delta MKT_{ic} .$$

Inspection of the occupational classifications that underlie the four functions reveals that they are particularly meaningful in manufacturing, but less so in non-manufacturing industries. For example, outside of manufacturing the marketing category is very heterogeneous, resembling a residual of the other, more consistently defined functions.

As above, we focus on the period in which the association between labor shares and forward GVC participation is strongest. Thus, we fit the following long difference regressions in 2001–2007, both in manufacturing and in all private sector industries:

$$\Delta FUNCTION_{ic} = \beta^F \Delta U_{ic}^F + \beta^D \Delta U_{ic}^D + \delta \Delta D_{ic} + \text{fixed effects} + \varepsilon_{ic} \quad (13)$$

and

$$\begin{aligned} \Delta FUNCTION_{ic} = & \gamma_1 \Delta FWD_{ic} + \gamma_2 \Delta OFF_{ic} + \gamma_3 \Delta EXP_{ic} + \gamma_4 \Delta IMP_{ic} \\ & + \kappa_1 \Delta \ln q_c + \text{fixed effects} + \varepsilon_{ic} , \end{aligned} \quad (14)$$

where $FUNCTION \in \{MGT, R\&D, FAB, MKT\}$, D_{ic} is the measure of downstreamness in industry i and country c . We weigh regressions by value added in 2001 and cluster standard errors by country and by industry. In order to ease the exposition of (14), we report only estimates of γ_1 ; other coefficients are not precisely or robustly estimated, and imply much weaker effects.⁵² We estimate (13) and (14) with no fixed effects, country effects and industry effects; results using both country and industry fixed effects are not precisely estimated, which is not surprising given the very demanding specification (only one cross section of changes). Since all variables are in changes, the coefficients are comparable across all specifications.

We report the results in Table 9. Across all specifications, we find that ΔU^F and ΔFWD are negatively associated with ΔMGT , ΔMKT and—to a much greater extent—with ΔFAB . One reason for the larger effect on fabrication tasks is that increases in upstreamness and GVC participation are driven by offshoring of assembly. Expenditure on management, design, marketing and post-sales services also decline as we move up the value chain, away from final good production, but less so than for fabrication tasks. These differences are also mostly statistically significant, especially in manufacturing, as can be seen in the p-values reported in Table 9.⁵³

⁵²Full results are available in Table A15 in the appendix.

⁵³These tests are based on regressions that pool all functions and allow for a full set of interaction terms for all variables, including fixed effects, for each function. We also perform such tests for a less demanding specification, in which we restrict the industry and country fixed effects to be the same across functions, and add only function fixed effects. In this case we strongly reject the null hypothesis of equality of coefficients to ΔU^F and ΔFWD for FAB

Our interpretation is that selling final goods requires more interaction with consumers and more management control of production processes compared to producing intermediate inputs further up the value chain. More generally, as Baldwin (2016) argues, moving production stages within a value chain across borders requires moving some management and marketing together with fabrication tasks, because GVC participation involves technology transfer to where production activities are relocated.⁵⁴ In contrast, we do not find any effect of forward GVC participation on the share of labor expenditures in R&D. This is what we would expect to find if forward GVC participation is associated with multinationals offshoring assembly, where parents perform almost all of the R&D for the multinational, a pattern that is well documented by several sources.⁵⁵

A corollary of these findings is that R&D becomes more intensive *within* labor in industries that integrate forward into foreign GVCs. Interestingly, when estimating (13) in the sample with all private sector industries (but not in manufacturing) we find a positive association between changes in downstreamness ΔD with $\Delta R\&D$. Together with the results on upstreamness, this implies that R&D becomes more intensive in industries that increase the overall length of the value chains they participate in. However, the association of downstreamness with R&D is not strong enough to increase labor shares significantly, as seen in Table 8.⁵⁶

Overall, forward GVC participation and greater upstreamness are associated with declines in labor shares because they require less expenditure on fabrication, management and marketing labor activities.

5.8 Offshoring of assembly of exports of final goods

We start by asking whether the effect of an increase in the export intensity of intermediate inputs is due to offshoring of assembly of final goods that would have otherwise been exported. Offshoring of assembly of final goods that would have been consumed domestically is controlled for by ΔIMP . The first stage regressions in Table 4 do not reveal any correlation between ΔIMP and ΔFWD .

We compute a modified version of ΔFWD , denoted ΔFWD^* , in which we subtract exports of final goods from value added in the denominator of FWD . Offshoring of assembly may increase the numerator of FWD if domestic inputs are shipped to foreign assembly sites. At the same time,

compared to other functions in manufacturing industries, while the differences are statistically weaker in the sample that includes all private sector industries. We discuss these specifications in detail in Appendix K and report the latter regressions and p-values in Table A16.

⁵⁴Indeed, this technology transfer has been hailed by many economists as one of the largest benefits of GVC integration for less developed countries. The point here is that this technology transfer cannot be achieved without some transfer of management and marketing tasks.

⁵⁵For example, about 80% of R&D expenditures by U.S. multinationals in 2014 was performed by parents located in the U.S.; see National Sciences Board (2018), Chapter 4.

⁵⁶We also estimated (14) by W2SLS using the instrument (10). Within manufacturing industries we find similar results to those in Table 8, except that the coefficients are generally larger; as in Table 4, the instrument is strong. However, the results are much less precise in the sample of all private sector industries. These results are available upon request.

it may reduce the denominator if it causes fewer exports of final (assembled) goods. If offshoring of labor-intensive assembly of hitherto exported final goods is an important mechanism, then ΔFWD^* should have a smaller effect on the reduction in labor shares. The results in columns 5 and 10 of Table 5 imply that, compared to the baseline results for ΔFWD in columns 1 and 6, there is a slightly weaker association ΔFWD^* with labor shares in manufacturing, but a slightly stronger association in all private sector industries.⁵⁷ The coefficient to ΔEXP (changes in exports of final goods intensity) increases dramatically in columns 5 and 10. This is because the numerator is precisely the part that is deducted from the denominator of ΔFWD^* .

In unreported regressions we tried to detect a different effect of ΔFWD^* through two channels: value added bilateral exports of intermediate inputs that cross borders only once and are absorbed in the direct destination country—versus when it is embodied in exports of the destination country, i.e. complex GVCs that cross borders more than once. Due to the high collinearity, we could not separately identify these effects. In our data 75 percent of ΔFWD^* accrues to value that crosses borders only once, on average.⁵⁸

We conclude that offshoring of assembly of final goods exports is one mechanism through which forward GVC participation lowers labor shares, although probably not the most important one.

5.9 Multinationals and cross-border capital ownership

Given their importance in the world economy, we now study the role of multinational enterprises (MNEs) in forward GVC participation.⁵⁹ For example, Boehm, Flaaen, and Pandalai-Nayar (2020) argue that MNEs’ offshoring decisions accounted for 41 percent of the decline in U.S. manufacturing employment. By exploiting the bilateral and directional nature of GVC participation and cross-border ownership we can shed light on the nature of this relationship. To the extent that the returns to capital (or profits) of affiliates of MNEs are repatriated to the country of origin, this has implications for labor shares in national income (compared to domestic product) and, therefore, inequality across individuals.

We estimate the following equation in a cross section in 2007:

$$VBY_{od} = \exp \left\{ \begin{array}{c} \beta \cdot \operatorname{arcsinh}(affiliates_{od}) + \delta \cdot \operatorname{arcsinh}(affiliates_{do}) \\ + \gamma' gravity_{od} + \alpha_o + \alpha_d \end{array} \right\} + \varepsilon_{od} , \quad (15)$$

where VBY_{od} is income accruing to primary factors in country o due to exports of intermediate

⁵⁷We drop the top and bottom 1 percent of observations in order to eliminate very extreme outliers, which are driven by observations in which exports of final goods is most of value added, leading to unrealistic values for the modified ΔFWD .

⁵⁸While the increase in the share of value added that crosses borders more than once within total trade is often portrayed as the hallmark of GVC integration (e.g., Antràs (2021)), it is not as important as a share of production value added.

⁵⁹In 2016 MNEs accounted for about a third of global GDP, two thirds of global exports and 40% of imports of intermediate inputs; see Cadestin, De Backer, Desnoyers-James, Miroudot, Ye, and Rigo (2018).

inputs for final goods production in country d . Here $affiliates_{od}$ is the number of affiliates located in o with parents in (owned by) d ; $affiliates_{do}$ is the number of affiliates located in d with parents in (owned by) o ; and $\text{arcsinh}(x) = \ln[x + (1 + x^2)^{1/2}]$ is the inverse hyperbolic sine transformation, which approximates the logarithmic function but allows for zeros (other functional forms, e.g., $\ln(1 + x)$, yield similar results). Thus, β captures the degree to which exports of intermediate inputs are associated with foreign-owned upstream input production; and δ captures the degree to which exports of intermediate inputs are associated with foreign-owned downstream assembly activity. The data on affiliates are averages in 1996–2001, and taken from Ramondo, Rodríguez-Clare, and Tintelnot (2015). Using other measures of cross-border ownership intensities such as FDI stocks and affiliate sales yield, qualitatively, the same results.⁶⁰ The variables in $gravity_{od}$ are described in the previous section, and we include origin and destination fixed effects to control for unobserved factors that may determine overall attractiveness of a location for input or final good production, and overall MNE prowess. We estimate (15) by PPML, and report robust two-way clustered standard errors at the country o and country d level.

We report the results in Table 10. In columns 1–3 we see that cross-border MNE activity in both directions is associated with exports of intermediate inputs, and the difference in the strength of the association is small. This implies that both types of vertical international integration play a role in fostering forward GVC participation: offshoring of upstream input production and of downstream assembly. This dovetails with our results in Table 5, where we did not find significant differences for the effect of forward GVC participation on labor shares once we excluded assembly offshoring.

We also investigate whether these associations are different for direct exports, VB^xY_{od} , compared to indirect exports of intermediate inputs, VB^gY_{od} , where $VB^Y_{od} = VB^xY_{od} + VB^gY_{od}$ for $o \neq d$ (see (4) above). We see in columns 4–6 and 7–9 of Table 10 a stronger association with direct exports of intermediate goods, VB^xY_{od} , which is reasonable given the bilateral nature the cross-border ownership indicators (other gravity variables has similar correlations as in Table 7).⁶¹

6 Concluding remarks

In this paper we studied the evolution of labor shares in a sample of 39 countries, both developed and less developed, in 1995–2014. Our main message is that deepening of forward foreign GVC

⁶⁰These results are available upon request. Many of the the observations on cross-border ownership are imputed ("extrapolated"), as described in the data appendix of Ramondo, Rodríguez-Clare, and Tintelnot (2015). After merging the data on affiliates to our data we are left with 1122 observations, where we lose information on five of the 39 countries in our sample: Cyprus, Estonia, Latvia, Malta, Taiwan ($1122 = 34 \times 33$, compared to $1482 = 39 \times 38$).

⁶¹The results are robust to dropping Ireland and the aggregated Belgium+Luxemburg, which are the only tax havens in our data according to Hines Jr and Rice (1994).

participation caused declines in labor shares, especially in 2001–2007, after China’s accession to the WTO and until the trade collapse of 2008. This is driven by increases in exporting—in value added terms—of intermediate inputs which tend to be more capital intensive. It manifests in an increase in industry upstreamness, and in a shift away from fabrication, as well as marketing and management labor activities. The increase in foreign forward GVC participation and in upstreamness all but halt after 2007, when labor shares cease to decline, on average. Since skilled labor shares increase while the overall labor share declines, this implies that unskilled labor shoulders the decreases in payments to labor, which is consistent with the greater drop in fabrication compared to other labor activities. We show that the rise in forward GVC participation is associated with differences in relative capital abundances, and with cross-border MNE presence. While declines in labor shares started before the period of acceleration in GVC participation, our analysis is confined to the GVC channel. Studying other forces that may have operated before 1995 are beyond the scope of this paper.

Concerns for the social costs of GVC participation, and on labor in particular, have recently been aired by The White House (2021); our paper illustrates that some of these concerns are not misplaced. First, to the extent that inequality is a concern, and given that redistribution of the gains from globalization is far from perfect (and potentially very costly), our findings on the effect of globalization on lowering labor shares raise concerns about the costs of further trade openness. Second, our findings on MNE activity imply that studying the effects of globalization on income distribution through the evolution of labor shares should take into account a national product approach, rather than rely solely on domestic production approach. This is because MNEs own (most of) the capital installed in their foreign affiliates, and thus a large part of the income flows due to GVC participation do not accrue where production takes place, but to the owners of capital involved in production. In contrast, our analysis suggests that if GVCs become shorter, less global and more regional, as some recent indicators seem to suggest, then this may be a force to increase labor shares.

One important caveat of our study is that we do not identify the underlying forces that drive globalization (for example, reductions in man-made trade barriers, or technological forces). Nevertheless, whatever the underlying causal forces, we demonstrate that they operate through forward foreign GVC participation. In addition, since we do not observe firms, we cannot study whether GVC deepening accounts for declines in labor shares by changing the composition of firms—along the lines Harrigan and Reshef (2015), Autor, Dorn, Katz, Patterson, and Van Reenen (2020) and De Loecker and Eeckhout (2018)—or by changes within firms. Understanding the role of firms in GVC participation and changes in labor shares is an important task for future research.

Appendix

A Labor share concepts in the Penn World Tables

The Penn World Tables report five different concepts of labor shares. Here we list them, and we denote their labels in the Figure A1 and in brackets:

1. Compensation of employees [**Employees**].
2. Compensation of employees + all income of self employed (mixed income) [**Employees + All mixed income**].
3. Compensation of employees + labor income of self employed (mixed income), computed by assuming that self-employed workers use labor and capital in the same proportion as the rest of the economy [**Employees + Self empl (K/L)**].
4. Compensation of employees + labor income of self employed (mixed income), computed by using the aggregate average wage of self-employed [**Employees + Self empl (avg wage)**].
5. Compensation of employees + value added in agriculture, assuming that all value added in agriculture is labor compensation (on average, it is 90%). This correction is useful for developing countries, where about half of self-employed workers are in agriculture [**Employees + Self empl (agric)**].

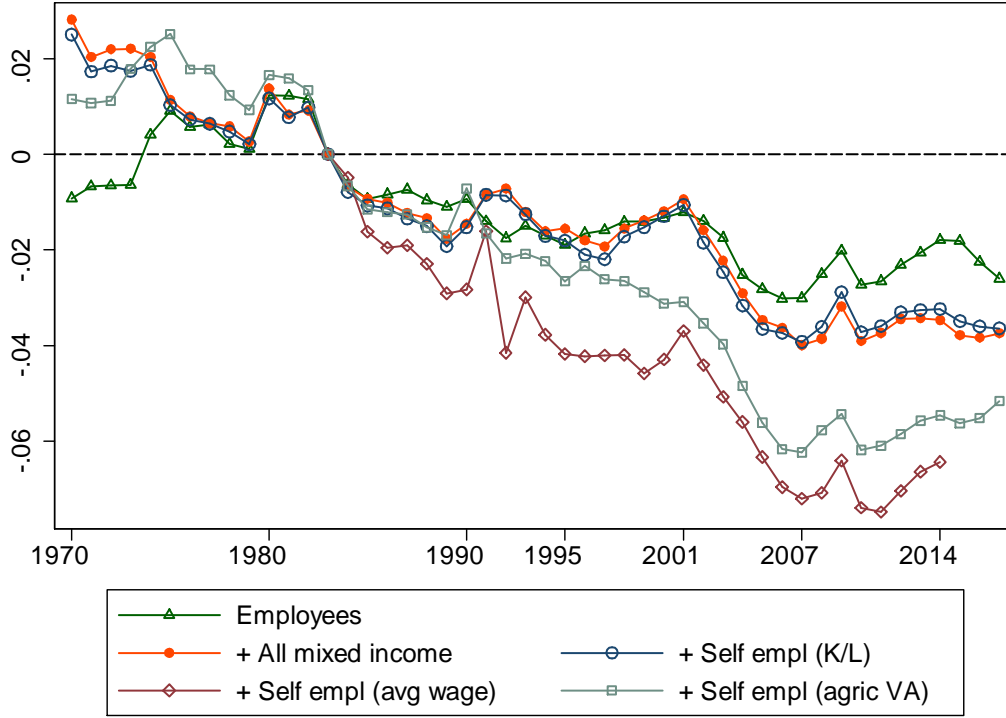
See Feenstra, Inklaar, and Timmer (2015) and their online appendix for fuller details. See also Mućk, McAdam, and Growiec (2018) for an overview of different measures of the labor share.

In Figure A1 we report year fixed effects from regressions of each labor share concept on year fixed effects and country fixed effects, weighted by real GDP. We use data from Penn World Tables 9.1 on all labor shares concepts, dropping all observations that are extrapolated (extrapolated values are set to the first or last observed value, so they have no content for the evolution of labor shares). The year fixed effects are adjusted be equal to zero in 1983, the first year in which the fourth concept is available. The sample includes 39 countries that correspond to the WIOD 2013 release sample of countries.

Figure A1 demonstrates that all measures of the labor share exhibit similar trends, especially the acceleration in the decline in 2001–2007. Mućk, McAdam, and Growiec (2018) also show that for the United States, all measures of the labor share exhibit common trends from 2001 and on. This is important for us, because 2001 is the year after which the association of labor shares declines with forward foreign GVC deepening is strongest.

In Figure 1 in the main text all countries' labor shares are the Employees + Self empl (K/L) series, except for China (compensation of employees) and India and Indonesia (compensation of employees + value added in agriculture).

Figure A1: Different Labor Share Concepts in the Penn World Tables



B WIOD data and computations

B.1 Data structure

The 2013 release of the World Input-Output database (WIOD) covers the period 1995–2011. Along with detailed Input-Output tables for 40 countries and 35 industries (ISIC rev. 3), this release also provides the Socio-Economic Accounts with data on employment, labor compensation and capital stocks, all by country and industry. In addition, the 2013 release reports employment and labor compensation by educational attainment within each country and industry. We also use the more recent WIOD 2016 release, covering 43 countries and 56 sectors (ISIC rev. 4) for the period 2000–2014. The Socio-Economic Accounts in the 2016 release do not include employment breakdown by educational level. WIOD 2013 uses the 1993 System of National Accounts (SNA), while WIOD 2016 uses the 2008 SNA, which renders merging the two impossible.

One major caveat in using these data arises from the proportionality assumptions in constructing WIOD. Value added shares within industry gross output and factor expenditure shares within value added are the same within an industry and country, regardless of the using industry and country or final consumption destination. For example, this means that the WIOD data do not allow the value added intensity in gross output to depend on the use of output (downstream industries or consumption, domestic or foreign). de Gortari (2017) demonstrates that the latter can have significant quantitative implications for measures of economic ties between the U.S. and Mexico. In contrast, Puzzello (2012) finds that similar proportionality assumptions lead to small bias in factor content of trade.

Figure A2 depicts a schematic outline for the structure of the WIOD for the exemplary case of 3 countries and 2 sectors. See <http://www.wiod.org/home> for further details on the country coverage and data availability.

Figure A2: Schematic Outline of a World Input-Output Table

		A						Y			
Cty	Ind	S		R		T		S	R	T	X
		1	2	1	2	1	2				
S	1	A_{11}^{ss}	A_{12}^{ss}	A_{11}^{sr}	A_{12}^{sr}	A_{11}^{st}	A_{12}^{st}	Y_1^{ss}	Y_1^{sr}	Y_1^{st}	$X_1^s = \sum_{c \in \{s,r,t\}} \sum_{i \in \{1,2\}} A_{1i}^{sc} + \sum_{c \in \{s,r,t\}} Y_1^{sc}$
	2	A_{21}^{ss}	A_{22}^{ss}	A_{21}^{sr}	A_{22}^{sr}	A_{21}^{st}	A_{22}^{st}	Y_2^{ss}	Y_2^{sr}	Y_2^{st}	$X_2^s = \sum_{c \in \{s,r,t\}} \sum_{i \in \{1,2\}} A_{2i}^{sc} + \sum_{c \in \{s,r,t\}} Y_2^{sc}$
R	1	A_{11}^{rs}	A_{12}^{rs}	A_{11}^{rr}	A_{12}^{rr}	A_{11}^{rt}	A_{12}^{rt}	Y_1^{rs}	Y_1^{rr}	Y_1^{rt}	$X_1^r = \sum_{c \in \{s,r,t\}} \sum_{i \in \{1,2\}} A_{1i}^{rc} + \sum_{c \in \{s,r,t\}} Y_1^{rc}$
	2	A_{21}^{rs}	A_{22}^{rs}	A_{21}^{rr}	A_{22}^{rr}	A_{21}^{rt}	A_{22}^{rt}	Y_2^{rs}	Y_2^{rr}	Y_2^{rt}	$X_2^r = \sum_{c \in \{s,r,t\}} \sum_{i \in \{1,2\}} A_{2i}^{rc} + \sum_{c \in \{s,r,t\}} Y_2^{rc}$
T	1	A_{11}^{ts}	A_{12}^{ts}	A_{11}^{tr}	A_{12}^{tr}	A_{11}^{tt}	A_{12}^{tt}	Y_1^{ts}	Y_1^{tr}	Y_1^{tt}	$X_1^t = \sum_{c \in \{s,r,t\}} \sum_{i \in \{1,2\}} A_{1i}^{tc} + \sum_{c \in \{s,r,t\}} Y_1^{tc}$
	2	A_{21}^{ts}	A_{22}^{ts}	A_{21}^{tr}	A_{22}^{tr}	A_{21}^{tt}	A_{22}^{tt}	Y_2^{ts}	Y_2^{tr}	Y_2^{tt}	$X_2^t = \sum_{c \in \{s,r,t\}} \sum_{i \in \{1,2\}} A_{2i}^{tc} + \sum_{c \in \{s,r,t\}} Y_2^{tc}$
Total intermediate consumption A_i^c		$A_1^s = \sum_{k,j} A_{j1}^{ks}$	$A_2^s = \sum_{k,j} A_{j2}^{ks}$	$A_1^r = \sum_{k,j} A_{j1}^{kr}$	$A_2^r = \sum_{k,j} A_{j2}^{kr}$	$A_1^t = \sum_{k,j} A_{j1}^{kt}$	$A_2^t = \sum_{k,j} A_{j2}^{kt}$				
Value added V_i^c		$V_1^s = X_1^s - A_1^s$ $- A_1^s$	$V_2^s = X_2^s - A_2^s$ $- A_2^s$	$V_1^r = X_1^r - A_1^r$ $- A_1^r$	$V_2^r = X_2^r - A_2^r$ $- A_2^r$	$V_1^t = X_1^t - A_1^t$ $- A_1^t$	$V_2^t = X_2^t - A_2^t$ $- A_2^t$				

In Figure A2 above the area shaded in light grey includes intermediate value flows, A , among industries (indexed by $i \in \{1, 2\}$) located in countries (indexed by $c \in \{s, r, t\}$). For example, A_{12}^{sr} describes the total value of intermediate use by industry 2 located in country r (indicated by the column) of input from industry 1 located in country s (indicated by the row). The area shaded in dark grey indicates demand for final goods, Y . For example, Y_2^{rt} is total demand for final goods in country t for good 2 sourced from country r . The WIOD distinguishes among five final demand use categories. In order to conserve on space, these five categories are not displayed in the Figure A2 (the categories are: final consumption expenditure by households, final consumption expenditure by non-profit organizations, final consumption expenditure by government, gross fixed capital formation and changes in inventories and valuables). Furthermore, X is a vector of total gross outputs for industries by location (indicated by the row). Total intermediate consumption for an industry i located in a country c (indicated by the column) A_i^c is the sum of all A elements within a column. Value added V_i^c of an industry i located in a country c (indicated by the column) is obtained by deducting A_i^c from the corresponding total gross output entry X_i^c for that industry i and country c (indicated by the row).

Summing all Y elements gives global consumption of final goods. From the expenditure approach to national accounting this is also global GDP.

B.2 Value added computations

Value added computations are based on Timmer, Los, Stehrer, and de Vries (2013), which is rooted in the seminal work of Leontief (1936). The goal is to decompose the value of final goods production (i.e., final demand) according to the industry and location where the value added originated. Conversely, one can also compute the allocation of payments to primary factors (capital and labor) according to the industries where these value added payments originate. Technically, the computation relies on a diagonal matrix of final demand Y , the Leontief inverse matrix B , as well as a diagonal matrix of direct value added coefficients per sector, V . All these are obtained from the values depicted in the Figure A2 above.

The elements of the diagonal matrix of final goods demand Y are obtained by a row-wise summation of the “ Y -area” in Figure A2 across all countries (and use categories; see above for details):

$$Y_i^c = \sum_k Y_i^{ck}.$$

The elements of the diagonal matrix of value added coefficients V are obtained by subtracting the entire intermediate consumption of a sector (column sum in the input-output matrix A) from the

sectoral gross output and dividing this by the gross output of the sector

$$v_i^c = \frac{X_i^c - \sum_{k,j} A_{ji}^{kc}}{X_i^c}.$$

The Leontief inverse matrix is $B = (I - A)^{-1}$, where A is the matrix containing all sub-elements equal to

$$a_{ij}^{sr} = \frac{A_{ij}^{sr}}{X_j^r}$$

and I is the identity matrix. We compute the B matrix in a few steps. In the first, we derive the input-output coefficients, a_{ij}^{sr} . We obtain these coefficients by dividing each cell in the A region in Figure A2 along a column by the gross output X of the respective column sector. This gives the matrix A . A typical element a_{ij}^{sr} of A indicates the amount of output from industry i located in source country s (indicated by the row) that is needed to sustain the production of one unit of output in industry j in destination country r (indicated by the column). In the second step we compute an auxiliary matrix by subtracting the A matrix of input-output coefficients from an identity matrix I . Finally, we invert the auxiliary matrix to obtain the required Leontief matrix B . A typical element b_{ij}^{sr} of B indicates the amount of output from industry i located in source country s (indicated by the row) that is needed to sustain the production of one unit of final demand of product j in destination country r (indicated by the column).

In order to obtain the gross output needed to sustain final demand we multiply BY . In order to get the corresponding concept in value added terms, we pre-multiply BY by the diagonal matrix V with elements V_i^c on the diagonal (appropriately ordered) to get VBY . For illustration, an example of the matrix VBY for the case of three countries and two industries is

$$\begin{aligned} & VBY \\ &= \begin{bmatrix} v_1^s & 0 & 0 & 0 & 0 & 0 \\ 0 & v_2^s & 0 & 0 & 0 & 0 \\ 0 & 0 & v_1^r & 0 & 0 & 0 \\ 0 & 0 & 0 & v_2^r & 0 & 0 \\ 0 & 0 & 0 & 0 & v_1^t & 0 \\ 0 & 0 & 0 & 0 & 0 & v_2^t \end{bmatrix} \begin{bmatrix} b_{11}^{ss} & b_{12}^{ss} & b_{11}^{sr} & b_{12}^{sr} & b_{11}^{st} & b_{12}^{st} \\ b_{21}^{ss} & b_{22}^{ss} & b_{21}^{sr} & b_{22}^{sr} & b_{21}^{st} & b_{22}^{st} \\ b_{11}^{rs} & b_{12}^{rs} & b_{11}^{rr} & b_{12}^{rr} & b_{11}^{rt} & b_{12}^{rt} \\ b_{21}^{rs} & b_{22}^{rs} & b_{21}^{rr} & b_{22}^{rr} & b_{21}^{rt} & b_{22}^{rt} \\ b_{11}^{ts} & b_{12}^{ts} & b_{11}^{tr} & b_{12}^{tr} & b_{11}^{tt} & b_{12}^{tt} \\ b_{21}^{ts} & b_{22}^{ts} & b_{21}^{tr} & b_{22}^{tr} & b_{21}^{tt} & b_{22}^{tt} \end{bmatrix} \begin{bmatrix} y_1^s & 0 & 0 & 0 & 0 & 0 \\ 0 & y_2^s & 0 & 0 & 0 & 0 \\ 0 & 0 & y_1^r & 0 & 0 & 0 \\ 0 & 0 & 0 & y_2^r & 0 & 0 \\ 0 & 0 & 0 & 0 & y_1^t & 0 \\ 0 & 0 & 0 & 0 & 0 & y_2^t \end{bmatrix} \\ &= \begin{bmatrix} v_1^s b_{11}^{ss} y_1^s & v_1^s b_{12}^{ss} y_2^s & v_1^s b_{11}^{sr} y_1^r & v_1^s b_{12}^{sr} y_2^r & v_1^s b_{11}^{st} y_1^t & v_1^s b_{12}^{st} y_2^t \\ v_2^s b_{21}^{ss} y_1^s & v_2^s b_{22}^{ss} y_2^s & v_2^s b_{21}^{sr} y_1^r & v_2^s b_{22}^{sr} y_2^r & v_2^s b_{21}^{st} y_1^t & v_2^s b_{22}^{st} y_2^t \\ v_1^r b_{11}^{rs} y_1^s & v_1^r b_{12}^{rs} y_2^s & v_1^r b_{11}^{rr} y_1^r & v_1^r b_{12}^{rr} y_2^r & v_1^r b_{11}^{rt} y_1^t & v_1^r b_{12}^{rt} y_2^t \\ v_2^r b_{21}^{rs} y_1^s & v_2^r b_{22}^{rs} y_2^s & v_2^r b_{21}^{rr} y_1^r & v_2^r b_{22}^{rr} y_2^r & v_2^r b_{21}^{rt} y_1^t & v_2^r b_{22}^{rt} y_2^t \\ v_1^t b_{11}^{ts} y_1^s & v_1^t b_{12}^{ts} y_2^s & v_1^t b_{11}^{tr} y_1^r & v_1^t b_{12}^{tr} y_2^r & v_1^t b_{11}^{tt} y_1^t & v_1^t b_{12}^{tt} y_2^t \\ v_2^t b_{21}^{ts} y_1^s & v_2^t b_{22}^{ts} y_2^s & v_2^t b_{21}^{tr} y_1^r & v_2^t b_{22}^{tr} y_2^r & v_2^t b_{21}^{tt} y_1^t & v_2^t b_{22}^{tt} y_2^t \end{bmatrix}. \end{aligned} \quad (16)$$

The elements of the VBY matrix can be interpreted in two ways. First, the values of the matrix along a *column* indicate *backward* linkages of production. The sum within a column is the value added that an industry located in a country generates in order to satisfy demand for final goods that it produces. Values within a column denote the value contribution of all industries and countries (given by the row) to the production of another industry located in a country (given by the column). For example, $v_1^r b_{12}^{rs} y_2^s$ indicates the value added of sector 1 located in country r that is supplied in order to produce final goods of industry 2 in country s . By summing across all rows within a column one obtains the total value of final goods production y_2^s , which is also final demand for industry 2 located in country s , no matter where this is sold around the world (i.e., no matter where demands arises from). For example, $\sum_{i,k} v_i^k b_{i2}^{ks} y_2^s = FD_2^s = y_2^s$. Summing all y_j^s across columns j within a country s does not give the GDP of country s because trade may not be balanced (if trade were balanced, then this sum does give GDP of country s). However, summing all y_j^s across all j and s gives global GDP.

The second interpretation considers the values of the VBY matrix within a *row*, indicating the

forward linkages of production. In this interpretation values indicate how payments to primary factors employed in a country-industry (given by the row) are “financed” by production of final goods (in terms of value added) of other industries and countries (given by the columns). Thus, in the context of forward linkages, $v_1^r b_{12}^{rs} y_2^s$ is the part of GDP paid to factors employed in industry 1 in country r by production of final goods in industry 2 of country s . The sum across all columns within a row is thus equal to the country-industry’s value added of the considered row, for example, $\sum_{j,k} v_1^r b_{1j}^{rk} y_j^k = VA_1^r$. Therefore, summing the industry rows for a given country gives GDP of that country, for example $\sum_i VA_i^r = \text{GDP}^r$.

B.3 Foreign value added shares

We compute two foreign value added shares. The first is foreign value added shares in final goods production based on the *backward* perspective. These are payments to factors located in foreign countries. This is calculated by summing within a column entries across rows of all industries located in foreign countries:

$$\text{backward}_i^c = \frac{\sum_{s \neq c} \sum_j v_j^s b_{ji}^{sc} y_i^c}{\sum_s \sum_j v_j^s b_{ji}^{sc} y_i^c} = \frac{\sum_{s \neq c} \sum_j v_j^s b_{ji}^{sc} y_i^c}{y_i^c}.$$

Using the example in (16), the foreign value added (not share thereof) in production of sector 1 in country s , is the sum of $v_1^{sb_{11}^{ss} y_1^s}$, $v_2^{sb_{21}^{ss} y_1^s}$, $v_1^{tb_{11}^{ts} y_1^s}$ and $v_2^{tb_{21}^{ts} y_1^s}$.

The second foreign value added share concept entails shares in factor payments (value added) paid by foreign industries, based on the *forward* perspective. This is calculated by summing within a row entries across columns of all industries located in foreign countries:

$$\text{forward}_i^c = \frac{\sum_{s \neq c} \sum_j v_i^c b_{ij}^{cs} y_j^s}{\sum_s \sum_j v_i^c b_{ij}^{cs} y_j^s} = \frac{\sum_{s \neq c} \sum_j v_i^c b_{ij}^{cs} y_j^s}{va_i^c}$$

Using the example in (16), the foreign value added (not share thereof) in factor payments of sector 1 in country s , is the sum of $v_1^{sb_{11}^{sr} y_1^r}$, $v_1^{sb_{12}^{sr} y_2^r}$, $v_1^{sb_{11}^{st} y_1^t}$ and $v_1^{sb_{12}^{st} y_2^t}$. The denominator in this case is equal to value added of industry 1 in country s , va_1^s .

B.4 Production factors computations

As described in Timmer, Erumban, Los, Stehrer, and de Vries (2014), the methodology described above can also be applied to decompose the value of final goods production according to capital and labor. The only difference consists the use of a different vector of coefficients. The calculations above transform gross outputs $X = BY$ into value added by pre-multiplying by the diagonal matrix V . Instead, we only need to pre-multiply X by a different diagonal matrix, one that transforms gross outputs into factor payments.

In order to derive this it is necessary to divide sector level data on capital and labor compensation by sectoral output

$$v_{f,i}^c = \frac{F_i^c}{X_i^c},$$

where F and f denote payments and the share of payments to a particular factor. Thus, $v_{f,i}^c$ is the gross output share of factor f . Values for F_i^c are given by the Socio-Economic Accounts in the WIOD. Pre-multiplying BY by a diagonal matrix V_f with elements $v_{f,i}^c$ on the diagonal gives a matrix of factor shares in production, $V_f BY$, which can be read like the VBY matrix above, only in terms of payments to factor f . The decomposition of the final goods’ value into to capital, high- and less-skilled labor incomes requires three different matrices.

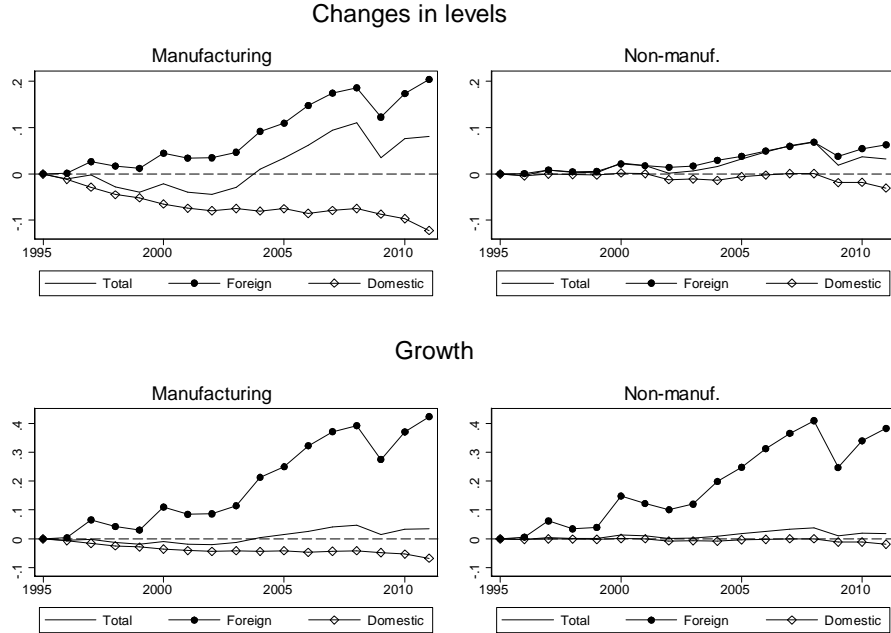
C Backward GVC participation

Each element of the VBY matrix contains all payments to factors that are employed in sector i in origin country o that contribute to the production of sector j in destination country d : $(vby)_{ij}^{od}$. Then $\sum_{o \neq d} \sum_i (vby)_{ij}^{od}$ summarizes payments to foreign primary factors due to imports (in value added terms) of intermediate inputs used in industry j located in country d . Define backward foreign GVC participation for an industry j in country d as the share of imports of intermediate inputs in value added terms in input use: $backward_j^d = \sum_{o \neq d} \sum_i (vby)_{ij}^{od} / \sum_o \sum_i (vby)_{ij}^{od}$. The equivalent concept at the country level is $backward^d = \sum_{o \neq d} \sum_i \sum_j (vby)_{ij}^{od} / \sum_o \sum_i \sum_j (vby)_{ij}^{od}$. Table A3 and Table A4 in the appendix report country-level backward GVC participation using WIOD 2013 and WIOD 2016, respectively, in 1995, 2007 and 2014.

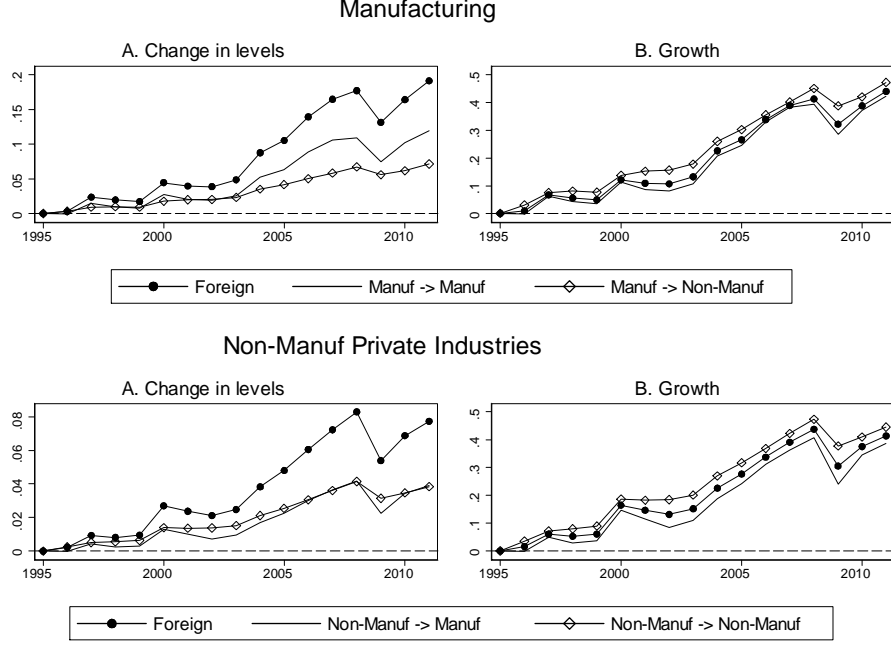
The global trend in backward GVC participation is similar to that of forward GVC participation (exhibited in Figure 1), because the numerators use the same global trade in inputs in value added terms. However, forward and backward differ at the country and industry level, due to differences across industries and countries in the tendency to import inputs or to export inputs from and to other industries, respectively.

D Upstreamness breakdowns

We study the evolution of U , U^D and U^F from equation (2) separately in manufacturing industries and in non-manufacturing industries. The following figure summarizes the analysis. The figure indicates that the level of upstreamness in non-manufacturing industries increases less than in manufacturing, but since it increases from a lower level, the growth rate is very similar to that in manufacturing. Moreover, the increase in both sectors is driven predominantly by foreign transactions.



Notes. The displayed series are year fixed effects from a regression of either country-level upstreamness (U_{ct}), its domestic (U_{ct}^D) and foreign (U_{ct}^F) components, or their natural logarithms ($\ln U_{ct}$, $\ln U_{ct}^D$, $\ln U_{ct}^F$) on year fixed effects and country fixed effects, weighted by GDP in 1995. Country-level upstreamness indices are computed as the weighted average of industry values from (2) (using value added as weights), and $U_{ct} = U_{ct}^D + U_{ct}^F$. The sample includes 39 countries that correspond to the WIOD 2013 release, which is also the source of the data.



We also study whether the increase in upstreamness is accounted for differently depending on the final good using sector: manufacturing and non-manufacturing. Inspection of the way the upstreamness index is constructed in (1) reveals that splitting by final good sector gives a linear, additive split of the original index. Since the results clearly indicate that the foreign part is driving the increase in upstreamness, we focus on that part. This gives us, for each source sector s and final use sector s'

$$U(s)_{ct}^F = U(s \rightarrow s)_{ct}^F + U(s \rightarrow s')_{ct}^F.$$

The following figure summarizes the analysis. Here, we repeat the foreign component of upstreamness from the figure just above, and add the splits by final using industry. The figure indicates that the foreign component of upstreamness in manufacturing is accounted for more by manufacturing than non-manufacturing private sector industries. But since in levels the Manuf→Manuf sub component starts at a higher level, the growth rate of both sub-components (by manufacturing or non-manufacturing at the end of the value chain) is similar. Within nonmanufacturing industries, both levels and growth are similar for both sub-components. Overall, there aren't large differences depending on whether the final step of production is in manufacturing or not.

Notes. The displayed series are year fixed effects from a regression of either country–sector–level foreign component of upstreamness ($U(s)_{ct}^F$), its same-sector component ($U(s \rightarrow s)_{ct}^F$) and cross-sector component ($U(s \rightarrow s')_{ct}^F$), or their natural logarithms ($\ln U(s)_{ct}^F$, $\ln U(s \rightarrow s)_{ct}^F$, $\ln U(s \rightarrow s')_{ct}^F$) on year fixed effects and country fixed effects, weighted by GDP in 1995. Country-level upstreamness indices are computed as the weighted average of industry values from (2) (using value added as weights), and $U(s)_{ct}^F = U(s \rightarrow s)_{ct}^F + U(s \rightarrow s')_{ct}^F$. The sample includes 39 countries that correspond to the WIOD 2013 release, which is also the source of the data.

E Proof of decomposition equation (3)

The change in the product VX (indeed, of any two conformable matrices) can be written as

$$\Delta(VX) = \Delta V X_1 + V_1 \Delta X + \Delta V \Delta X . \quad (17)$$

To see this, start with

$$\Delta V X = V_2 X_2 - V_1 X_1 .$$

Add and subtract $V_2 X_1$ and rearrange to get

$$\begin{aligned} \Delta V X &= V_2 X_2 - V_1 X_1 + (V_2 X_1 - V_2 X_1) \\ &= V_2 (X_2 - X_1) + (V_2 - V_1) X_1 \\ &= \Delta V X_1 + V_2 \Delta X . \end{aligned}$$

Now add and subtract $V_1 \Delta X$ and rearrange to get

$$\begin{aligned} \Delta V X &= \Delta V X_1 + V_2 \Delta X + (V_1 \Delta X - V_1 \Delta X) \\ &= \Delta V X_1 + V_1 \Delta X + V_2 \Delta X - V_1 \Delta X \\ &= \Delta V X_1 + V_1 \Delta X + (V_2 - V_1) \Delta X \\ &= \Delta V X_1 + V_1 \Delta X + \Delta V \Delta X . \end{aligned}$$

Applying the same algebra in (17) to $X = (BY)$ and plugging this back into (17) yields (3).

F Stone's additive decomposition

This is based on Miller and Blair (2009), pages 285–290, originally from Stone (1961).

Consider \tilde{A} , an $n \times n$ matrix. Start with

$$X = AX + Y$$

and subtract $\tilde{A}X$

$$X - \tilde{A}X = AX - \tilde{A}X + Y \implies (I - \tilde{A})X = (A - \tilde{A})X + Y$$

to get

$$X = (I - \tilde{A})^{-1}(A - \tilde{A})X + (I - \tilde{A})^{-1}Y$$

Define

$$A^* \equiv (I - \tilde{A})^{-1}(A - \tilde{A})$$

and write

$$X = A^*X + (I - \tilde{A})^{-1}Y \quad (18)$$

Pre-multiply by A^* to get

$$A^*X = (A^*)^2X + A^*(I - \tilde{A})^{-1}Y \quad (19)$$

and use (19) in (18) to get

$$\begin{aligned} X &= (A^*)^2X + A^*(I - \tilde{A})^{-1}Y + (I - \tilde{A})^{-1}Y \\ &= (A^*)^2X + (I + A^*)(I - \tilde{A})^{-1}Y \end{aligned}$$

Now solve again for X to get

$$X = \underbrace{[I - (A^*)^2]^{-1}}_{M_3} \cdot \underbrace{(I + A^*)}_{M_2} \cdot \underbrace{(I - \tilde{A})^{-1}}_{M_1} \cdot Y \quad (20)$$

Stone's additive decomposition starts with $X = M_3 M_2 M_1 Y$ in (20) and arrives at:

$$X = IY + \underbrace{(M_1 - I) Y}_{\widetilde{M}_1} + \underbrace{(M_2 - I) M_1 Y}_{\widetilde{M}_2} + \underbrace{(M_3 - I) M_2 M_1 Y}_{\widetilde{M}_3} \quad (21)$$

Here is the derivation of (21) starting with (20):

$$\begin{aligned} B &= M_1 M_2 M_3 \\ &= M_2 M_1 + M_1 M_2 M_3 - M_2 M_1 \\ &= M_2 M_1 + (M_3 - I) M_2 M_1 \\ &= M_1 + M_2 M_1 - M_1 + (M_3 - I) M_2 M_1 \\ &= M_1 + (M_2 - I) M_1 + (M_3 - I) M_2 M_1 \\ &= I + (M_1 - I) + (M_2 - I) M_1 + (M_3 - I) M_2 M_1 \end{aligned}$$

In the context of international analysis, $\widetilde{A} = A^d$ is the matrix of diagonal (or block-diagonal, if industries are not aggregated) elements such that

$$A^d = \begin{bmatrix} A_{11} & 0 & 0 & 0 \\ 0 & A_{22} & 0 & 0 \\ 0 & 0 & \ddots & 0 \\ 0 & 0 & 0 & A_{nn} \end{bmatrix}$$

and

$$A^f = \begin{bmatrix} 0 & A_{12} & \cdots & A_{1n} \\ A_{21} & 0 & & \vdots \\ \vdots & & \ddots & A_{n-1,n} \\ A_{n1} & \cdots & A_{n,n-1} & 0 \end{bmatrix}.$$

Then

$$B^d = (I - A^d)^{-1} = \begin{bmatrix} (I - A_{11})^{-1} & 0 & 0 & 0 \\ 0 & (I - A_{22})^{-1} & 0 & 0 \\ 0 & 0 & \ddots & 0 \\ 0 & 0 & 0 & (I - A_{nn})^{-1} \end{bmatrix}$$

Using these in A^* gives

$$\begin{aligned} A^* &= (I - A^d)^{-1} A^f = B^d A^f \\ &= \begin{bmatrix} (I - A_{11})^{-1} & 0 & 0 & 0 \\ 0 & (I - A_{22})^{-1} & 0 & 0 \\ 0 & 0 & \ddots & 0 \\ 0 & 0 & 0 & (I - A_{nn})^{-1} \end{bmatrix} \begin{bmatrix} 0 & A_{12} & \cdots & A_{1n} \\ A_{21} & 0 & & \vdots \\ \vdots & & \ddots & A_{n-1,n} \\ A_{n1} & \cdots & A_{n,n-1} & 0 \end{bmatrix} \\ &= \begin{bmatrix} 0 & (I - A_{11})^{-1} A_{12} & \cdots & (I - A_{11})^{-1} A_{1n} \\ (I - A_{22})^{-1} A_{21} & 0 & & \vdots \\ \vdots & & \ddots & (I - A_{n-1,n-1})^{-1} A_{n-1,n} \\ (I - A_{nn})^{-1} A_{n1} & \cdots & (I - A_{nn})^{-1} A_{n,n-1} & 0 \end{bmatrix} \end{aligned}$$

and also the related M_2 matrix $M_2 = I + A^*$. The typical off-diagonal $(i, j)_{i \neq j}$ element of A^* (and also of M_2) is $(I - A_{ii})^{-1} A_{ij}$; it captures demand for factors in i that originate from intermediate

inputs demand in production in j that cross borders from j to i *once*.

Now consider

$$(A^*)^2 = \left[\sum_{l \neq i, j} (I - A_{ii})^{-1} A_{il} (I - A_{ll})^{-1} A_{lj} \right]_{i, j}$$

which has a typical (i, j) element $\sum_{l \neq i, j} (I - A_{ii})^{-1} A_{il} (I - A_{ll})^{-1} A_{lj}$. This captures demand for factors in i that originate from intermediate inputs demand in production in j that cross borders *twice* from j to i . The first matrix on the right A_{lj} gives demand from j 's industries in l . The second matrix $(I - A_{ll})^{-1}$ calculates the output that needs to be produced in l in order to satisfy the demand from j . The third matrix A_{il} gives the implication of this for demand from i 's industries. And the fourth matrix $(I - A_{ii})^{-1}$ calculates the output that needs to be produced in i in order to satisfy the demand from l .

Applying the above to Stone's additive decomposition gives

$$X = IY + \underbrace{[B^d - I] Y}_{\widetilde{M}_1} + \underbrace{B^d A^f B^d Y}_{\widetilde{M}_2} + \underbrace{(B - B^d - B^d A^f B^d) Y}_{\widetilde{M}_3}.$$

If we consider $\widetilde{A} = A^f$, we have

$$M_1 = B^f = (I - A^f)^{-1}$$

Here B^f captures total demand for output (including the initial injection of direct demand from Y) due to value chains that always cross borders. For example, B^f includes chains like $A_{ij}A_{jk}A_{kl}A_{lm}\dots$, where $i \neq j, j \neq k, k \neq l, l \neq m\dots$, but it is possible to have, for example, $i = k$. Thus domestic feedbacks are possible in B^f . Here $\widetilde{M}_1 = B^f - I$ in Stone's additive decomposition nets out the direct effect of the initial injection by deducting I . However, $\widetilde{M}_2 = B^f A^d B^f$ does not have a clear interpretation, despite clearly capturing some of the possible value chains. Similarly for \widetilde{M}_3 . However, we can say that $\widetilde{M}_2 + \widetilde{M}_3$ gives the remainder of output that is induced by demand after taking into account the direct injection and \widetilde{M}_1 .

Stone's additive decomposition reads

$$B = I + (B^d - I) + \underbrace{B^d A^f B^d}_{B^x} + \underbrace{(B - B^d - B^d A^f B^d)}_{B^g}.$$

Here I captures the direct effect of demand on output. Next, $B^d - I$ captures output that is induced by all strictly *domestic* indirect linkages. To see this, note that $B^d = (I - A^d)^{-1}$. To see this, note that $B^d = (I - A^d)^{-1}$, where A^d is the matrix of block diagonal elements from A , capturing only domestic linkages. Next, B^x captures output that is induced by all strictly *bilateral trade* in intermediate inputs *that cross borders only once* (exports from the standpoint of the producing country). To see this, note that $A^f = A - A^d$, i.e. the off-block-diagonal elements of A . B^x takes all domestic output requirements (the first B^d on the right), computes the implied international demand for intermediate inputs captured in A^f , and then the implied total domestic requirements in the producing country (the second B^d on the left). Finally, B^g captures all other types of linkages, essentially net interregional feedback effects (net of strictly direct intra- and direct international effects captured B^d and B^x , respectively). I.e., B^g captures the effect of *complex global value chains*: output that is induced by combining both domestic and foreign linkages, that may cross borders more than once, and that may include return effects.

Taking changes of the last equation delivers the decomposition in the main text.

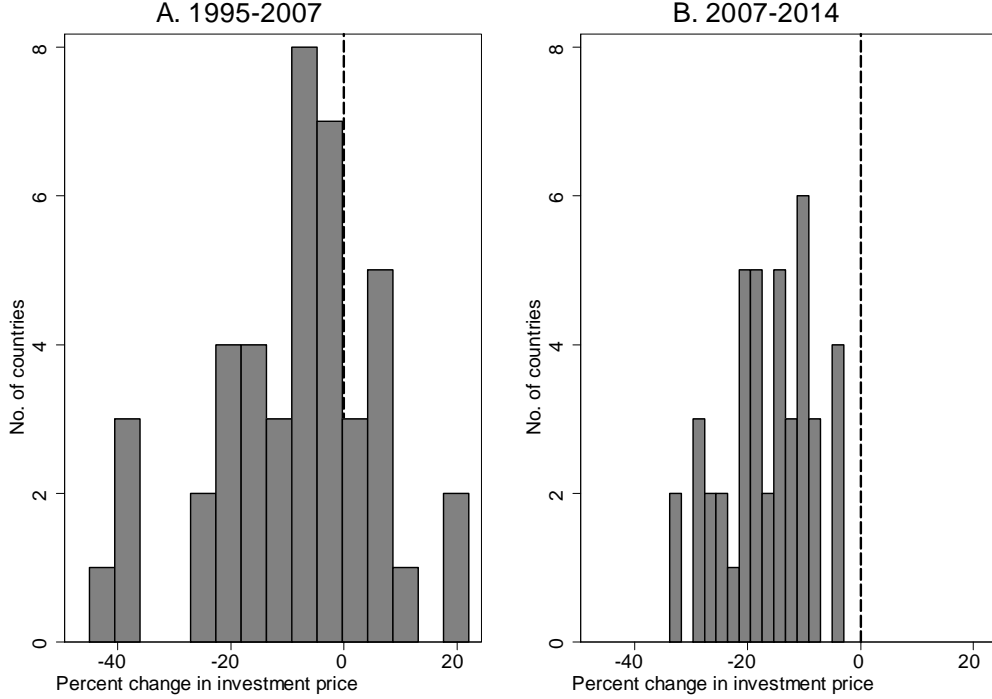
G Price of investment in capital equipment

We use data from the Penn World Tables mark 9.0 (PWT, see Feenstra, Inklaar, and Timmer (2015)) and from the United States' Bureau of Economic Analysis (BEA), applying the same methodology as in Karabarbounis and Neiman (2014) with updated data (their sample ends in 2010).

For each country c in year t we divide the investment price index (P_{ct}^{inv}) by the consumption price index (P_{ct}^{con}), both in terms of their corresponding PPP US prices (from PWT data). This means that $P_{ct}^{inv}/P_{ct}^{con}$ is the relative price of investment in terms of that of the United States' ratio in PPP. In order to convert this to the relative price from the domestic standpoint we divide $P_{ct}^{inv}/P_{ct}^{con}$ by $P_{USA,t}^{inv}/P_{USA,t}^{con}$ and then multiply by the ratio of the price index for private fixed investment ($P_{USA,t}^{pfi}$) to the personal consumption expenditures price index ($P_{USA,t}^{pce}$) (the latter from BEA data). In the regressions we use log differences in the resulting $q_{ct} = (P_{ct}^{inv}/P_{ct}^{con})/(P_{USA,t}^{inv}/P_{USA,t}^{con}) \cdot (P_{USA,t}^{pfi}/P_{USA,t}^{pce})$.

The weighted average decline in q_c in 1995–2007 is roughly 12%, or 1% per year, while the weighted average decline in 2007–2014 is roughly 15%, or 2% per year—twice the annual rate in the previous period. Figure A3 displays histograms of percent changes in the investment price in 1995–2007 and in 2007–2014.

Figure A3: Percent Changes in the Investment Price



H Testing for spatial correlation in source competitiveness

Here we explain how we address the concern for correlation between the source/producer fixed effects across space. Before moving ahead, it important to note that even if these correlations were detected in the cross section, the threat to the validity of the instrument is only if these the changes in these fixed effects over time are correlated.

Here is how we address these concerns. The procedure is as follows. (1) We collect all origin $\hat{\alpha}(t)_{ci}^j$ fixed effects from the PPML estimation. Recall that these have the interpretation of log values. (2) We then match each of these fixed effects with the corresponding ijt fixed effects of all other countries (“one to many” in Stata jargon). This results in two vectors of length $(\#countries \times \#industries^2) \times \#countries$. We call the first vector TC_{cijt} and the other one TC_{dijt} , where d should not be confused with destinations from the bilateral gravity regression. (3) We drop the observations for which $c = d$ (not doing this changes virtually nothing), and then fit two types of regressions:

$$(TC_{cij,2007} - TC_{dij,2007}) = \rho \ln(\text{distance}_{cd}) + \text{fixed effects} + \varepsilon_{cdij,2007}$$

and

$$(TC_{cij,2007} - TC_{cij,2001}) - (TC_{dij,2007} - TC_{dij,2001}) = \rho \ln(\text{distance}_{cd}) + \text{fixed effects} + \varepsilon_{cdij}$$

The first regression tests whether competitiveness (productivity \times production costs) is more similar across countries that are closer. The second regression tests whether changes (growth) in competitiveness is more similar across countries that are closer. We estimate these regressions with three “saturation” levels of fixed effects: none, ij (supplying \times using industry pairs), and ij, ci (supplying country and industry) and dj (destination country and using industry). We cluster standard errors at the bilateral cd level (Moulton (1990)).

Our estimates of ρ are all tiny and not statistically significant, except in the most saturated regression in levels, where the coefficient is 0.004 with a standard error of 0.001—a tiny elasticity, albeit precisely estimated. Importantly, the estimate of ρ in all of the regressions for changes in

competitiveness is “precisely estimated zero” (with values 0.004 without or with ij fixed effects and standard error of 0.007, or -0.001 with standard error of 0.001 for the most saturated specification with ij , ci and dj fixed effects).

This implies that the threat to the validity of our instrument due to spatial correlation is not important in practice.

I Identifying 2001 as the split year in 1995–2007

We identify 2001 as a natural split for the 1995–2007 sub-sample. We estimate a series of regressions of the type

$$\begin{aligned}\Delta_{1995-\tau}LS_{ic} = & \gamma_1\Delta_{1995-\tau}FWD_{ic} \\ & +\gamma_2\Delta_{1995-\tau}BACK_{ic} + \gamma_3\Delta_{1995-\tau}EXP_{ic} + \gamma_4\Delta_{1995-\tau}IMP_{ic} \\ & +\kappa_1\Delta_{1995-\tau}\ln q_c + \text{fixed effects} + \varepsilon_{ic}\end{aligned}\quad (22)$$

and

$$\begin{aligned}\Delta_{\tau-2007}LS_{ic} = & \gamma_1\Delta_{\tau-2007}FWD_{ic} \\ & +\gamma_2\Delta_{\tau-2007}BACK_{ic} + \gamma_3\Delta_{\tau-2007}EXP_{ic} + \gamma_4\Delta_{\tau-2007}IMP_{ic} \\ & +\kappa_1\Delta_{\tau-2007}\ln q_c + \text{fixed effects} + \varepsilon_{ic},\end{aligned}\quad (23)$$

where $\Delta_{1995-\tau}$ denotes changes from 1995 to year τ , and $\Delta_{\tau-2007}$ denotes changes from year τ to 2007, for $\tau = 1996, 1997, \dots, 2006$. The results (available upon request) imply that the coefficient to FWD is small and imprecisely estimated in (22) unless $\tau > 2001$; the coefficient to FWD is large and precisely estimated in (23) when $\tau \geq 2001$. Regressions like (22) and (23) using data aggregated to the country level deliver a similar message.

J Outlier analysis

In order to check for the influence of outliers we repeatedly estimated (6), each time dropping a country, an industry, or a country-industry pair. The point estimates are not very sensitive to these sample restrictions. We report in Figure A4 and Figure A5 the coefficient to ΔFWD when we estimate (6) while dropping one country-industry observation at a time, together with the 95% confidence intervals. In manufacturing industries (Figure A4) influential observations are: USA in “Food, beverages and tobacco” (15t16), “Electrical and optical equipment” (30t33), “Transport equipment” (34t35); Japan in “Basic metals and fabricated metals” (27t28), “Electrical and optical equipment” (30t33), “Transport equipment” (34t35); Germany in “Transport equipment” (34t35). In all private sector industries (Figure A5) influential observations are: Japan in “Transport equipment” (34t35), China in “Agriculture, hunting, forestry and fishing”, (AtB), Great Britain in “Financial intermediation” (J).

Figure A4: Manufacturing Industries

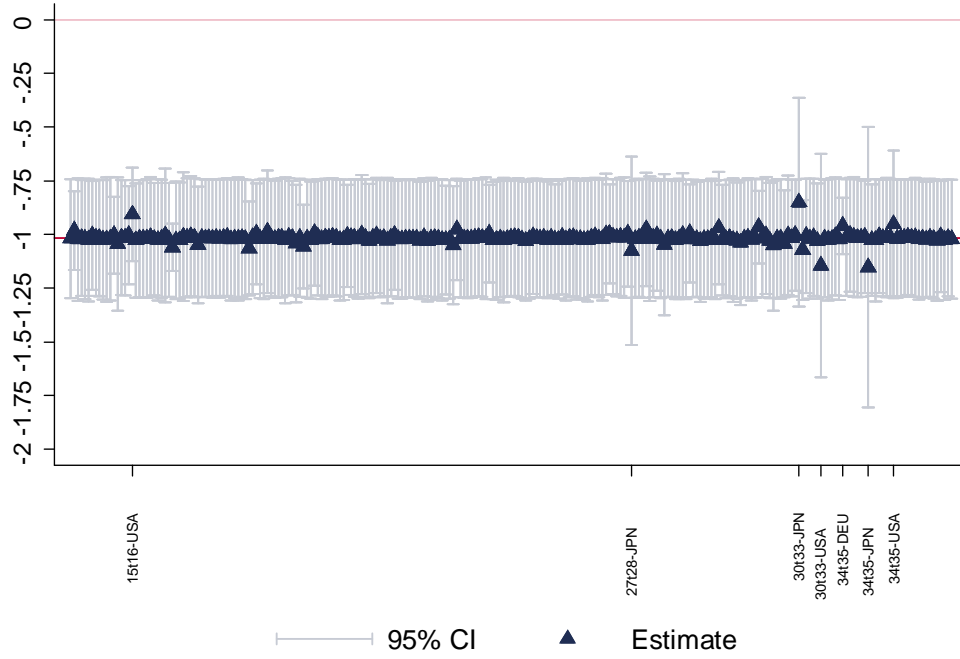
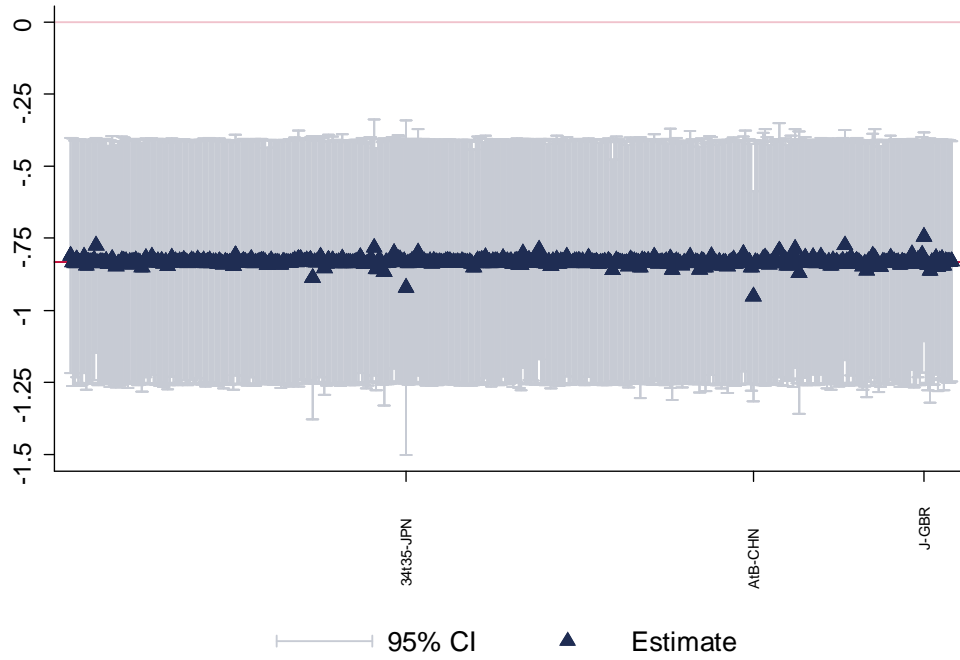


Figure A5: Private Sector Industries



K Testing for differences in coefficients across business functions

We test whether the foreign component of upstreamness ΔU^F and forward GVC participation ΔFWD have different effects on expenditure on fabrication ΔFAB versus other business functions.

We pool all functions and fit two types of regressions that are similar to (13) and (14). In the first, we allow for a full set of interactions by function as follows

$$\Delta FUNCTION_{ic}^j = \sum_{j \in \{MGT, R\&D, FAB, MKT\}} I_j \left[\begin{array}{c} \beta_j^F \Delta U_{ic}^F + \beta_j^D \Delta U_{ic}^D + \delta_j \Delta D_{ic} \\ + \text{fixed effects} \end{array} \right] + \varepsilon_{ic}^j \quad (24)$$

and

$$\Delta FUNCTION_{ic}^j = \sum_{j \in \{MGT, R\&D, FAB, MKT\}} I_j \left[\begin{array}{c} \gamma_{1j} \Delta FWD_{ic} + \gamma_{2j} \Delta OFF_{ic} \\ + \gamma_{3j} \Delta EXP_{ic} + \gamma_{4j} \Delta IMP_{ic} \\ + \kappa_{1j} \Delta \ln q_c + \text{fixed effects} \end{array} \right] + \varepsilon_{ic}^j, \quad (25)$$

where I_j is equal to 1 when the function is equal to j and zero otherwise. This specification allows all coefficients and fixed effects, including to ΔFWD_{ic} to vary over functions j . It recovers the same coefficients as in Table 9. The fixed effects are: none, industries, countries, as in Table 9. Importantly, we keep clustering standard errors in two dimensions: by country and industry. This allows unrestricted covariance structures across functions within these units, so the standard errors we get are conservative, and slightly larger than those reported in Table 9 (since in Table 9 we estimate regressions for each function independently). This allows us to test for whether the coefficient to changes in the foreign component of upstreamness (ΔU_{ic}^F) and to ΔFWD_{ic} are statistically significantly different across functions.

In particular, we test whether the coefficients for ΔFAB (labor in fabrication) are different than for other business functions. The short answer is yes: ΔU_{ic}^F and ΔFWD_{ic} have statistically significantly different coefficients for ΔFAB than for other functions. This is more clearly evident within manufacturing industries, as are most of the results in the paper. The p-values for these two-sided tests are reported in Table 9. Note that the p-values for having a greater coefficient in absolute value (a one-sided test) would be half of those reported (unless the rank of magnitudes switch), so we are being conservative here. There are, however, some pair comparisons in which we do not reject the equality of coefficients at conventional levels of significance. But even when considering all private sector industries, we reject the null hypothesis of equal coefficient in the presence of country-by function fixed effects.

We also estimated less-demanding versions of (24) and (25), where we do not interact the fixed effects with the function dummy, and instead include a separate function dummy, which restricts industry and country trends to be the same for all functions:

$$\Delta FUNCTION_{ic}^j = \sum_{j \in \{MGT, R\&D, FAB, MKT\}} I_j [\beta_j^F \Delta U_{ic}^F + \beta_j^D \Delta U_{ic}^D + \delta_j \Delta D_{ic}] + \text{fixed effects} + \varphi_j + \varepsilon_{ic}^j \quad (26)$$

and

$$\Delta FUNCTION_{ic}^j = \sum_{j \in \{MGT, R\&D, FAB, MKT\}} I_j \left[\begin{array}{c} \gamma_{1j} \Delta FWD_{ic} + \gamma_{2j} \Delta OFF_{ic} \\ + \gamma_{3j} \Delta EXP_{ic} + \gamma_{4j} \Delta IMP_{ic} \\ + \kappa_{1j} \Delta \ln q_c \end{array} \right] + \text{fixed effects} + \varphi_j + \varepsilon_{ic}^j, \quad (27)$$

where φ_j is a dummy for function j . We obtain results that are consistent with Table 9, with larger magnitudes for the coefficients to the interactions with FAB , which are also more statistical significant. In these specifications we reject equality of the coefficients to the interactions with FAB in all specifications in manufacturing, while statistical significance in these differences is somewhat lower when we estimate these specifications in the sample including all industries. All other interaction coefficients (to ΔU_{ic}^D , ΔD_{ic} , ΔOFF_{ic} , etc.) are not statistically significant. We report these results in Table A16.

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Table 1: Decomposition of Changes in Labor Shares in GDP and in Manufacturing Value Added, 1995-2007

A. All sectors, percent GDP

	K income from domestic industries (1)	L income from domestic industries (2)	K income from foreign industries (3)	L income from foreign industries (4)	L income (domestic + foreign) (5)	Income from foreign industries (6)
Levels						
VfBY 1995	35.01	56.11	3.48	5.40	61.51	8.88
VfBY 2007	35.88	52.39	5.06	6.67	59.06	11.73
Changes						
Vf2007*B1995*Y1995 - VfBY 1995	0.95	-0.77	0.11	-0.29	-1.06	-0.18
Vf1995* B2007*Y1995 - VfBY 1995	-0.66	-1.53	1.13	1.07	-0.47	2.19
Vf1995*B1995* Y2007 - VfBY 1995	-0.12	-1.65	0.56	1.21	-0.44	1.77
Vf1995* B2007*Y2007 - VfBY 1995	-0.42	-2.71	1.28	1.85	-0.86	3.13
VfBY 2007 - VfBY 1995	0.87	-3.72	1.57	1.27	-2.45	2.84

B. Manufacturing industries, percent in value added

	K income from domestic industries (1)	L income from domestic industries (2)	K income from foreign industries (3)	L income from foreign industries (4)	L income (domestic + foreign) (5)	Income from foreign industries (6)
Levels						
VBY 1995	31.84	49.30	6.89	11.96	61.27	18.86
VBY 2007	32.06	41.94	10.70	15.29	57.23	25.99
Changes						
Vf2007*B1995*Y1995 - VfBY 1995	1.15	-1.02	0.36	-0.49	-1.51	-0.13
Vf1995* B2007*Y1995 - VfBY 1995	-0.88	-2.49	1.79	1.57	-0.92	3.36
Vf1995*B1995* Y2007 - VfBY 1995	-1.22	-4.12	1.82	3.52	-0.60	5.34
Vf1995* B2007*Y2007 - VfBY 1995	-1.34	-6.00	2.97	4.37	-1.63	7.34
VfBY 2007 - VfBY 1995	0.22	-7.36	3.81	3.32	-4.04	7.13

Notes. Panel A reports decompositions of changes in factor shares in GDP, while Panel B reports decompositions of changes in factor shares within manufacturing industries' value added. Columns 1-4 report the shares of income accruing to capital and labor from domestic industries and from foreign industries. Column 5 reports the overall labor share (columns 2 + 4). Column 6 reports the share of income accruing from serving foreign industries. The split between domestic and foreign industries is given by different entries within rows in VfBY. The contribution of foreign industries to factor shares is given by the forward concept defined in the text. The contribution of domestic industries is given by the complement of the forward concept. The rows labeled "Levels" report levels in 1995 and in 2007. Rows labeled as "Changes" report true and counterfactual changes. All numbers are weighted averages using GDP in 1995 as weights. Source: authors' calculations based on WIOD 2013 release.

Table 2: Sources of Compositional Changes in Payments to Labor

A. All sectors

	1995, Percent in GDP			$\Delta 1995-2007$		
	Income from domestic industries (1)	Income from foreign industries (2)	Domestic + foreign (3)	Income from domestic industries (4)	Income from foreign industries (5)	Domestic + foreign (6)
Value chains						
Domestic (Bd)	55.99	0	55.99	-1.59	0	-1.59
Bilateral trade (Bx)	0	4.52	4.52	0	0.42	0.42
Complex GVCs (Bg)	0.12	0.88	1.00	0.05	0.64	0.70
Total (B)	56.11	5.40	61.51	-1.53	1.07	-0.47
Sources of demand						
Domestic (Yd)	52.55	4.48	57.03	-2.45	0.75	-1.71
Foreign (Yf)	3.56	0.92	4.48	0.80	0.46	1.27
Total (Y)	56.11	5.40	61.51	-1.65	1.21	-0.44

B. Manufacturing

	1995, Percent in GDP			$\Delta 1995-2007$		
	Income from domestic industries (1)	Income from foreign industries (2)	Domestic + foreign (3)	Income from domestic industries (4)	Income from foreign industries (5)	Domestic + foreign (6)
Value chains						
Domestic (Bd)	48.99	0	48.99	-2.64	0	-2.64
Bilateral trade (Bx)	0	9.91	9.91	0	0.30	0.30
Complex GVCs (Bg)	0.31	2.06	2.37	0.15	1.27	1.42
Total (B)	49.30	11.96	61.27	-2.49	1.57	-0.92
Sources of demand						
Domestic (Yd)	39.51	9.61	49.11	-6.43	2.24	-4.19
Foreign (Yf)	9.80	2.36	12.15	2.31	1.28	3.59
Total (Y)	49.30	11.96	61.27	-4.12	3.52	-0.60

Notes. Panel A reports decompositions of levels and changes in labor shares in GDP, while Panel B reports decomposition of levels and changes in labor shares within manufacturing industries' value added. The four "Total" rows in columns 1-3 report labor shares in GDP that are paid by domestic industries, foreign industries, and overall in 1995; these are the same numbers in columns 2, 4 and 5 in Table 1. The "Total" rows in columns 4-6 report the changes in the same concepts. The rows above the "Total" rows indicate the contributions of sub-components of either B or Y to levels in columns 1-3 or to changes in columns 4-6. All numbers are weighted averages using GDP in 1995 as weights. Source: authors' calculations based on WIOD 2013 release.

Table 3. Changes in Labor Shares and Forward GVC Integration in 1995-2014

	Dependent variable: Δ labor share								
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
A. Manufacturing industries									
Δ forward intensity	-0.276** (0.118)	-0.224* (0.105)	-0.313** (0.110)	-0.321** (0.123)	-0.365** (0.123)	-0.296*** (0.053)	-0.429*** (0.080)	-0.151** (0.051)	-0.134* (0.068)
Δ backward intensity	-0.159 (0.179)	-0.137 (0.171)	-0.228 (0.198)	-0.179 (0.190)	-0.259 (0.225)	-0.217 (0.206)	-0.287 (0.298)	-0.064 (0.173)	-0.164 (0.189)
Δ exports of final goods intensity	0.031 (0.090)	0.049 (0.068)	0.066 (0.063)	0.077 (0.096)	0.115 (0.087)	0.130* (0.067)	0.135 (0.096)	0.135* (0.067)	0.034 (0.058)
Δ imports of final goods intensity	0.015 (0.015)	0.015 (0.013)	0.013 (0.013)	0.013 (0.015)	0.012 (0.012)	0.013 (0.010)	0.022 (0.013)	0.006 (0.005)	0.006 (0.008)
Δ log relative price of investment	-0.006 (0.050)	0.006 (0.057)	-0.017 (0.061)	-0.006 (0.051)	-0.017 (0.061)	0.002 (0.075)	0.012 (0.075)		-0.005 (0.076)
Fixed effects	-	Per	Cty	Ind	Cty, Ind	Cty, Ind, Per	Cty-Ind, Per	Cty-Per, Ind	Cty, Ind-Per
Observations	1,532	1,532	1,532	1,532	1,532	1,532	1,532	1,532	1,532
R-squared	0.026	0.079	0.117	0.044	0.135	0.185	0.301	0.454	0.330
B. Private sector industries									
Δ forward intensity	-0.167*** (0.060)	-0.134*** (0.047)	-0.201** (0.089)	-0.192** (0.082)	-0.230** (0.098)	-0.192** (0.073)	-0.283** (0.107)	-0.055 (0.033)	-0.117 (0.077)
Δ backward intensity	-0.132 (0.113)	-0.097 (0.096)	-0.163 (0.140)	-0.122 (0.136)	-0.145 (0.158)	-0.104 (0.133)	-0.153 (0.149)	0.007 (0.080)	-0.087 (0.140)
Δ exports of final goods intensity	0.070 (0.094)	0.058 (0.083)	0.065 (0.097)	0.093 (0.118)	0.090 (0.119)	0.079 (0.104)	0.126 (0.129)	0.094 (0.095)	0.012 (0.106)
Δ imports of final goods intensity	0.011 (0.008)	0.011 (0.007)	0.010 (0.007)	0.010 (0.006)	0.008 (0.006)	0.009 (0.005)	0.013* (0.007)	0.005 (0.004)	0.006 (0.004)
Δ log relative price of investment	-0.008 (0.027)	0.001 (0.033)	-0.002 (0.033)	-0.008 (0.027)	-0.001 (0.032)	0.013 (0.043)	0.019 (0.044)		0.011 (0.044)
Fixed effects	-	Per	Cty	Ind	Cty, Ind	Cty, Ind, Per	Cty-Ind, Per	Cty-Per, Ind	Cty, Ind-Per
Observations	3,270	3,270	3,270	3,270	3,270	3,270	3,270	3,270	3,270
R-squared	0.010	0.041	0.049	0.047	0.086	0.116	0.310	0.227	0.218

Notes. The dependent variable is changes in country-industry specific labor shares in three periods: 1995-2001, 2001-2007, 2007-2014. Δ forward is the change in intermediate inputs exports intensity in value added (forward GVC integration), Δ backward is the change in importing of intermediate inputs intensity in total input purchases, Δ exports of final goods is the change in export intensity of final goods in value added, and Δ imports of final goods is the change in import intensity of final goods in domestic absorption. Δ log relative price of investment is the country-level log change in investment prices. All regressions are weighted by value added in 1995 as weights. A constant is always included but not reported. Robust standard errors in parentheses are computed by two-way clustering by country and by industry. *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$.

Table 4. Changes in Labor Shares and Forward GVC Integration in 1995-2014: W2SLS

	Dependent variable: Δ labor share								
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
A. Manufacturing industries									
Δ forward intensity	-1.485*** (0.313)	-1.508*** (0.357)	-1.191*** (0.295)	-1.584*** (0.336)	-1.277*** (0.293)	-1.052*** (0.233)	-1.208*** (0.084)	-1.133*** (0.188)	-0.482* (0.243)
Δ backward intensity	-0.546 (0.337)	-0.644** (0.283)	-0.545* (0.288)	-0.652 (0.389)	-0.626* (0.332)	-0.570** (0.229)	-0.642 (0.378)	-0.600** (0.213)	-0.339 (0.208)
Δ exports of final goods intensity	0.211 (0.154)	0.282* (0.138)	0.145 (0.107)	0.337* (0.167)	0.240* (0.129)	0.256* (0.141)	0.304* (0.162)	0.202* (0.098)	0.104 (0.102)
Δ imports of final goods intensity	0.029 (0.025)	0.031 (0.025)	0.018 (0.017)	0.027 (0.025)	0.016 (0.015)	0.017 (0.010)	0.028 (0.016)	0.011 (0.009)	0.009 (0.009)
Δ log relative price of investment	0.045 (0.061)	0.108 (0.069)	0.006 (0.068)	0.040 (0.063)	0.005 (0.068)	0.053 (0.082)	0.063 (0.077)		0.018 (0.087)
Fixed effects	-	Per	Cty	Ind	Cty, Ind	Cty, Ind, Per	Cty-Ind, Per	Cty-Per, Ind	Cty, Ind-Per
Observations	1,471	1,471	1,471	1,471	1,471	1,471	1,454	1,471	1,471
Kleibergen-Paap F statistic	44.96	16.04	42.29	43	41.42	31.99	30.76	17.24	23.09
B. Private sector industries									
Δ forward intensity	-1.369*** (0.240)	-1.074*** (0.349)	-1.339*** (0.288)	-1.483*** (0.309)	-1.452*** (0.321)	-1.426*** (0.501)	-1.424** (0.606)	-2.147** (0.859)	-1.144*** (0.372)
Δ backward intensity	-0.495*** (0.167)	-0.432*** (0.136)	-0.572*** (0.184)	-0.485** (0.196)	-0.561** (0.209)	-0.593** (0.222)	-0.522** (0.230)	-1.145** (0.449)	-0.498** (0.203)
Δ exports of final goods intensity	0.469** (0.197)	0.385 (0.229)	0.414** (0.184)	0.491* (0.250)	0.438* (0.229)	0.452 (0.268)	0.464 (0.304)	0.696* (0.380)	0.366 (0.247)
Δ imports of final goods intensity	0.027 (0.018)	0.024 (0.016)	0.022 (0.016)	0.021 (0.015)	0.017 (0.012)	0.017 (0.012)	0.022* (0.012)	0.018 (0.013)	0.014 (0.010)
Δ log relative price of investment	0.028 (0.033)	0.048 (0.036)	0.030 (0.040)	0.031 (0.035)	0.032 (0.041)	0.077 (0.049)	0.075 (0.056)		0.065 (0.050)
Fixed effects	-	Per	Cty	Ind	Cty, Ind	Cty, Ind, Per	Cty-Ind, Per	Cty-Per, Ind	Cty, Ind-Per
Observations	3,129	3,129	3,129	3,129	3,129	3,129	3,078	3,129	3,129
Kleibergen-Paap F statistic	20.69	12.48	17.10	19.93	17.04	10.47	9	9.22	20.22

Notes. The dependent variable is changes in country-industry specific labor shares in three periods: 1995-2001, 2001-2007, 2007-2014. Δ forward is the change in intermediate inputs exports intensity in value added (forward GVC integration). The instrument ΔZ is constructed by eliminating home-country and industry sources of variation from Δ forward; see text for complete details. Δ backward is the change in importing of intermediate inputs intensity in total input purchases, Δ exports of final goods is the change in export intensity of final goods in value added, and Δ imports of final goods is the change in import intensity of final goods in domestic absorption. Δ log relative price of investment is the country-level log change in investment prices. All regressions are weighted by value added in 1995 as weights. A constant is always included but not reported. Robust standard errors in parentheses are computed by two-way clustering by country and by industry. *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$.

Table 5. Changes in Labor Shares and Forward GVC Integration in 1995-2014: period splits, China and assembly offshoring

	Dependent variable: Δ labor share									
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)
	Manufacturing industries					Private sector industries				
Δ forward intensity	-0.429*** (0.080)					-0.283** (0.107)				
1995-2001		-0.160 (0.247)					-0.152 (0.239)			
2001-2007		-1.038*** (0.271)					-0.815*** (0.232)			
2007-2014		0.479 (0.391)					0.181 (0.183)			
Δ forward intensity to China			-1.582** (0.698)					-1.266** (0.551)		
1995-2001				-0.616 (0.442)					-1.625*** (0.454)	
2001-2007				-1.744** (0.679)					-1.624*** (0.490)	
2007-2014				2.154 (1.519)					0.568 (0.447)	
Δ forward intensity to RoW			-0.220 (0.196)					-0.183 (0.114)		
1995-2001				-0.082 (0.279)					-0.055 (0.235)	
2001-2007				-0.769*** (0.234)					-0.649*** (0.177)	
2007-2014				0.310 (0.338)					0.087 (0.155)	
Δ forward intensity, net of assembly offshoring					-0.379*** (0.051)					-0.308*** (0.094)
Δ backward intensity	-0.287 (0.298)	-0.201 (0.301)	-0.276 (0.312)	-0.200 (0.286)	-0.305 (0.317)	-0.153 (0.149)	-0.112 (0.162)	-0.156 (0.153)	-0.123 (0.166)	-0.180 (0.148)
Δ exports of final goods intensity	0.135 (0.096)	0.017 (0.046)	0.074 (0.091)	-0.016 (0.069)	0.381*** (0.084)	0.126 (0.129)	0.029 (0.099)	0.088 (0.124)	0.038 (0.106)	0.347** (0.160)
Δ imports of final goods intensity	0.022 (0.013)	0.019* (0.010)	0.022* (0.012)	0.017 (0.010)	0.027* (0.015)	0.013* (0.007)	0.012** (0.006)	0.012* (0.007)	0.012** (0.006)	0.015* (0.008)
Δ log relative price of investment	0.012 (0.075)	0.006 (0.066)	0.012 (0.073)	0.009 (0.068)	0.009 (0.072)	0.019 (0.044)	0.022 (0.042)	0.017 (0.043)	0.020 (0.041)	0.021 (0.044)
Fixed effects: Cty-Ind, Period	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Observations	1,522	1,522	1,522	1,522	1,483	3,241	3,241	3,241	3,241	3,160
R-squared	0.301	0.395	0.323	0.412	0.313	0.310	0.346	0.319	0.354	0.320

Notes. The dependent variable is changes in country-industry specific labor shares in three periods: 1995-2001, 2001-2007, 2007-2014. Δ forward is the change in intermediate inputs exports intensity in value added (forward GVC integration). In columns 2 and 7 we split Δ forward by period; in columns 3 and 8 we split by shipments to China and the Rest of the World (RoW); and in columns 4 and 9 we split by both period and destination of shipments (China, RoW). In columns 5 and 10 Δ forward net of assembly offshoring is the change in intermediate inputs exports intensity in value added that excludes exports of final goods; we drop the bottom and top percentiles of this variable to avoid extreme outliers that are driven by deducting exports of final goods from value added in the denominator. Δ backward is the change in offshoring of intermediate inputs intensity in total input purchases, Δ exports of final goods is the change in export intensity of final goods in value added, and Δ imports of final goods is the change in import intensity of final goods in domestic absorption. Δ log relative price of investment is the country-level log change in investment prices. All regressions are weighted by value added in 1995 as weights. A constant is always included but not reported. Robust standard errors in parentheses are computed by two-way clustering by country and by industry. *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$.

Table 6. Changes in Labor Shares and Forward GVC Integration in 1995-2014: Country level regressions

Dependent variable:	Δ labor share					Δ forward intensity
	(1)	(2)	(3)	(4)	(5)	(6)
A. Manufacturing industries						
	WLS				W2SLS	1st stage
Δ forward intensity	-0.652*** (0.193)	-0.560*** (0.202)	-0.802*** (0.261)	-0.679** (0.314)	-1.147*** (0.287)	
Δ backward intensity	-0.541 (0.538)	-0.595 (0.440)	-0.562 (0.624)	-0.616 (0.610)	-0.502 (0.674)	-0.323 (0.317)
Δ exports of final goods intensity	-0.085 (0.455)	-0.055 (0.453)	-0.171 (0.490)	-0.122 (0.522)	-0.088 (0.562)	0.541** (0.233)
Δ imports of final goods intensity	0.357 (0.246)	0.395* (0.223)	0.268 (0.246)	0.378* (0.196)	0.277 (0.243)	0.004 (0.089)
Δ log relative price of investment	-0.014 (0.054)	0.013 (0.063)	0.005 (0.071)	0.042 (0.079)	0.008 (0.074)	0.042* (0.023)
Δ Z						0.045*** (0.007)
Fixed effects	-	Per	Cty	Cty, Per	Cty	Cty
Observations	114	114	114	114	114	114
R-squared	0.208	0.319	0.448	0.546		0.706
Kleibergen-Paap F statistic					42.46	
B. Private sector industries						
	WLS				W2SLS	1st stage
Δ forward intensity	-0.449** (0.179)	-0.372* (0.184)	-0.498* (0.264)	-0.387 (0.249)	-1.008 (0.704)	
Δ backward intensity	0.024 (0.252)	-0.052 (0.284)	0.120 (0.412)	0.047 (0.410)	0.218 (0.400)	-0.048 (0.179)
Δ exports of final goods intensity	-0.591 (0.389)	-0.649 (0.393)	-0.975* (0.543)	-1.019** (0.487)	-0.589 (0.714)	0.960*** (0.272)
Δ imports of final goods intensity	0.117 (0.324)	0.329 (0.329)	-0.175 (0.351)	0.099 (0.306)	-0.070 (0.390)	0.185 (0.269)
Δ log relative price of investment	0.011 (0.024)	0.020 (0.033)	0.038 (0.032)	0.044 (0.041)	0.039 (0.032)	0.016 (0.015)
Δ Z						0.014*** (0.004)
Fixed effects	-	Per	Cty	Cty, Per	Cty	Cty
Observations	113	113	113	113	113	113
R-squared	0.169	0.262	0.396	0.471		0.627
Kleibergen-Paap F statistic					11.70	

Notes. The dependent variable is changes in country-level labor shares in three periods: 1995-2001, 2001-2007, 2007-2014. Δ forward is the change in intermediate inputs exports intensity in value added (forward GVC integration). The instrument Δ Z is constructed by eliminating home-country and industry sources of variation from Δ forward in a based on a gravity relationship; see text for complete details. Δ backward is the change in importing of intermediate inputs intensity in total input purchases, Δ exports of final goods is the change in export intensity of final goods in GDP, and Δ imports of final goods is the change in import intensity of final goods in domestic absorption. Δ log relative price of investment is the country-level log change in investment prices. All regressions are weighted by value added in 1995 as weights. A constant is always included but not reported. Robust standard errors in parentheses are clustered by country.

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

Table 7. Factor endowments and forward GVC integration

	(1)	(2)	(3)	(4)	(5)	(6)
Estimator:	OLS	OLS	PPML	PPML	PPML	PPML
Dependent variable:	(VKBY-VLBY)/GDP, 2001	(VKBY-VLBY)/GDP, 2007	VBY, 2001	VBY, 2007	V(Bx)Y, 2007	V(Bg)Y, 2007
Relative capital abundance in 2001	0.268** (0.113)		0.522** (0.229)			
Relative capital abundance in 2007		0.317** (0.149)		0.863*** (0.208)	1.067*** (0.287)	0.293*** (0.101)
Log distance	0.017 (0.022)	0.015 (0.018)	-0.484*** (0.071)	-0.526*** (0.061)	-0.635*** (0.075)	-0.175*** (0.032)
Common border	-0.093 (0.057)	-0.023 (0.073)	0.155 (0.108)	0.143 (0.107)	0.190 (0.124)	-0.302*** (0.070)
Colonial ties	-0.011 (0.057)	-0.045 (0.060)	0.174 (0.115)	0.153 (0.107)	0.213 (0.135)	0.018 (0.032)
Common language	0.038 (0.067)	0.053 (0.070)	0.242** (0.117)	0.315*** (0.116)	0.399** (0.157)	0.016 (0.038)
Free trade agreement	0.017 (0.024)	0.080* (0.043)	0.259 (0.163)	0.099 (0.121)	0.084 (0.151)	0.127*** (0.047)
Common currency	-0.032 (0.029)	-0.056* (0.028)	0.084 (0.137)	0.044 (0.145)	0.077 (0.187)	0.198*** (0.065)

Notes. The dependent variables in all columns pertain to (value added) income accruing to primary factors located in origin country *o* due to exports of intermediate inputs that are part of GVCs that end in final good production in destination country *d*. In columns 1-2 it is income accruing to capital located in *o* minus income accruing to labor in *o*, as a share of GDP in *o*; in columns 3-4 it is the total income flow of this type; in column 5 it is income due to direct bilateral exports of intermediate inputs [V(Bx)Y]; in column 6 it is income due to flows that pass through at least one other country, not *o* and not *d*, i.e., complex GVCs [V(Bg)Y]. Income flows calculated from WIOD 2013. Relative capital abundance of *o* relative to *d* is described in the text. All regressions estimated on 1,482 observations (=39*38, where 39 is the number of countries in our sample), and include origin and destination fixed effects. Robust standard errors in parentheses are computed by two-way clustering by origin and by destination. *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$.

Table 8. Upstreamness and Labor Shares, 2001-2007

	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
A. Manufacturing industries								
	Dependent variable: labor share						Dependent variable: Δ labor share	
Upstreamness	-0.041 (0.047) [-0.132]	-0.136 (0.081) [-0.438]	-0.125* (0.069) [-0.402]				-0.128* (0.064) [-0.210]	
Domestic				-0.085** (0.038) [-0.217]	-0.020 (0.042) [-0.052]	-0.028 (0.049) [-0.071]		0.036 (0.041) [0.055]
Foreign				0.059 (0.065) [0.141]	-0.240*** (0.060) [-0.575]	-0.232*** (0.064) [-0.555]		-0.227*** (0.043) [-0.347]
Downstreamness	-0.109 (0.100) [-0.153]	-0.007 (0.083) [-0.010]	0.042 (0.089) [0.058]	-0.091 (0.089) [-0.126]	0.029 (0.070) [0.041]	0.035 (0.078) [0.049]	-0.035 (0.078) [-0.046]	-0.059 (0.065) [-0.077]
Fixed effects	Year	Cty X Ind	Cty X Ind, Yr	Year	Cty X Ind	Cty X Ind, Yr	-	-
Observations	3,727	3,725	3,725	3,727	3,725	3,725	530	530
R-squared	0.069	0.952	0.954	0.130	0.956	0.956	0.081	0.182
B. Private sector industries								
	Dependent variable: labor share						Dependent variable: Δ labor share	
Upstreamness	0.012 (0.082) [0.033]	-0.037 (0.056) [-0.100]	-0.024 (0.053) [-0.066]				-0.002 (0.054) [-0.003]	
Domestic				0.019 (0.096) [0.040]	0.050** (0.023) [0.107]	0.040 (0.027) [0.086]		0.105*** (0.021) [0.151]
Foreign				-0.005 (0.075) [-0.009]	-0.205*** (0.038) [-0.416]	-0.168*** (0.045) [-0.341]		-0.193*** (0.031) [-0.264]
Downstreamness	0.169 (0.128) [0.309]	0.005 (0.061) [0.010]	0.064 (0.059) [0.117]	0.172 (0.133) [0.314]	0.038 (0.056) [0.070]	0.067 (0.052) [0.123]	0.020 (0.064) [0.028]	0.018 (0.049) [0.026]
Fixed effects	Year	Cty X Ind	Cty X Ind, Yr	Year	Cty X Ind	Cty X Ind, Yr	-	-
Observations	7,926	7,922	7,922	7,926	7,922	7,922	1,117	1,117
R-squared	0.089	0.985	0.986	0.090	0.986	0.986	0.001	0.106

Notes. Dependent variable is always the labor share. Columns (1)-(6) report country by industry by year panel regressions in 2001-2007. Columns (7)-(8) report long difference regressions, where the dependent and explanatory variables are changes between 2001 and 2007. The explanatory variables are upstreamness and its domestic and foreign components, with variation that is commensurate with the dependent variable. All regressions estimated by weighted least squares with value added in 2001 as weights. Robust standard errors in parentheses are computed by two-way clustering by country and by industry. *** p<0.01, ** p<0.05, * p<0.1. Standardized, "beta" coefficients reported in brackets.

Table 9. Functional Specialization, Upstreamness and Forward Foreign GVC Integration, 2001-2007

	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)	(11)	(12)
A. Manufacturing												
Dependent variable:	Δ MGT	Δ R&D	Δ FAB	Δ MAR	Δ MGT	Δ R&D	Δ FAB	Δ MAR	Δ MGT	Δ R&D	Δ FAB	Δ MAR
Δ upstreamness, foreign	-0.043*** (0.014)	-0.009 (0.015)	-0.140*** (0.038)	-0.035*** (0.010)	-0.042** (0.016)	0.006 (0.017)	-0.112*** (0.035)	-0.020** (0.008)	-0.038*** (0.009)	-0.018 (0.015)	-0.143** (0.057)	-0.029** (0.012)
p-value for equality with Δ FAB	0.04	0.01	-	0.03	0.05	0.01	-	0.03	0.12	0.08	-	0.12
Δ upstreamness, domestic	0.003 (0.010)	-0.006 (0.014)	0.029 (0.029)	0.010 (0.010)	0.009 (0.011)	0.004 (0.013)	-0.029 (0.029)	-0.002 (0.010)	-0.003 (0.009)	-0.010 (0.013)	0.026 (0.037)	0.013 (0.014)
Δ downstreamness	-0.040* (0.022)	0.016 (0.014)	-0.000 (0.043)	-0.034** (0.015)	-0.041 (0.025)	0.009 (0.017)	0.041 (0.045)	-0.024 (0.017)	-0.011* (0.006)	0.027*** (0.008)	0.026 (0.054)	-0.027 (0.028)
Δ forward intensity	-0.167*** (0.049)	-0.019 (0.047)	-0.398*** (0.101)	-0.141*** (0.038)	-0.181** (0.066)	-0.007 (0.058)	-0.256*** (0.071)	-0.102*** (0.029)	-0.091*** (0.026)	-0.018 (0.032)	-0.368* (0.194)	-0.093* (0.045)
p-value for equality with Δ FAB	0.04	0.01	-	0.02	0.38	0.00	-	0.04	0.17	0.11	-	0.18
Fixed effects	-	-	-	-	Ind	Ind	Ind	Ind	Cty	Cty	Cty	Cty
B. All industries												
Dependent variable:	Δ MGT	Δ R&D	Δ FAB	Δ MAR	Δ MGT	Δ R&D	Δ FAB	Δ MAR	Δ MGT	Δ R&D	Δ FAB	Δ MAR
Δ upstreamness, foreign	-0.058*** (0.015)	0.008 (0.015)	-0.095*** (0.028)	-0.062** (0.023)	-0.062*** (0.022)	0.011 (0.017)	-0.038 (0.026)	-0.046 (0.030)	-0.033*** (0.012)	0.013 (0.022)	-0.106** (0.043)	-0.065** (0.031)
p-value for equality with Δ FAB	0.35	0.01		0.45	0.42	0.07		0.83	0.15	0.04	-	0.40
Δ upstreamness, domestic	-0.005 (0.011)	-0.004 (0.003)	0.073*** (0.026)	0.021 (0.020)	-0.006 (0.017)	-0.000 (0.006)	0.037 (0.040)	0.009 (0.017)	0.014 (0.010)	-0.002 (0.005)	0.082* (0.041)	0.042** (0.019)
Δ downstreamness	0.017 (0.014)	0.021*** (0.006)	-0.050 (0.030)	0.024 (0.021)	0.013 (0.019)	0.010 (0.007)	-0.011 (0.032)	0.035 (0.031)	0.025* (0.013)	0.027*** (0.007)	-0.048* (0.028)	0.034 (0.026)
Δ forward intensity	-0.145*** (0.037)	0.014 (0.036)	-0.281*** (0.089)	-0.134** (0.044)	-0.152*** (0.049)	0.002 (0.036)	-0.113** (0.046)	-0.101* (0.057)	-0.062*** (0.019)	0.037 (0.034)	-0.379*** (0.137)	-0.139*** (0.049)
p-value for equality with Δ FAB	0.20	0.01	-	0.18	0.52	0.02	-	0.88	0.03	0.01	-	0.09
Fixed effects	-	-	-	-	Ind	Ind	Ind	Ind	Cty	Cty	Cty	Cty

Notes. The dependent variables are changes in labor income shares in value added of four categories of "functional specialization": management (MGT), R&D, fabrication (FAB), and marketing (MKT). Overall labor shares are the sum of shares over these four categories. Changes for all variables are computed for 2001-2007. The first three lines in each panel report coefficients from regressions where foreign and domestic upstreamness are included, as well as downstreamness. The fourth line reports the coefficient to Δ forward, the change in intermediate inputs export and GVC intensity, from regressions where the other explanatory variables are the change in offshoring of intermediate inputs intensity in total input purchases (Δ backward), the change in export intensity of final goods in value added (Δ exports of final goods), the change in import intensity of final goods in total absorption (Δ imports of final goods), and the country-level log change in investment prices (Δ log relative price of investment). Regressions estimated by weighted least squares with value added in 2001 as weights. Robust standard errors in parentheses are computed by two-way clustering by country and by industry. *** p<0.01, ** p<0.05, * p<0.1. The rows "p-value for equality with Δ FAB" report the p-value for testing the (two-sided) null hypothesis of equality of the coefficient of the corresponding variable in that column to the coefficient in the Δ FAB column. In order to compute these p-values we pool all four functions and allow for full interactions for all variables, including fixed effects, with indicators for each function. When estimating these pooled regressions we keep the same two-way clustering by country and by industry. This allows unrestricted covariance structures across functions within these units, so the standard errors we get in the pooled regressions are conservative, and slightly larger than those reported above (in the above we estimate regressions for each function independently).

Table 10. Multinationals and forward GVC integration

	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
	Forward GVC integration			Direct bilateral exports of intermediate inputs			Complex global value chains		
Dependent variable:	VBY(o,d) = V(Bx)Y(o,d) + V[Bg]Y(o,d)			V(Bx)Y(o,d)			V(Bg)Y(o,d)		
arcsinh no. of affiliates in o with parents in d	0.125*** (0.028)		0.099*** (0.024)	0.161*** (0.037)		0.127*** (0.031)	0.043*** (0.015)		0.035** (0.014)
arcsinh no. of affiliates in d with parents in o		0.135*** (0.038)	0.112*** (0.038)		0.172*** (0.050)	0.141*** (0.050)		0.049*** (0.015)	0.043*** (0.014)
Log distance	-0.423*** (0.062)	-0.407*** (0.062)	-0.357*** (0.057)	-0.500*** (0.075)	-0.482*** (0.077)	-0.419*** (0.071)	-0.142*** (0.035)	-0.133*** (0.032)	-0.115*** (0.033)
Common border	0.195* (0.106)	0.213** (0.092)	0.206** (0.091)	0.257** (0.123)	0.277** (0.108)	0.267** (0.108)	-0.274*** (0.070)	-0.264*** (0.063)	-0.267*** (0.063)
Colonial ties	0.077 (0.104)	0.082 (0.098)	0.041 (0.100)	0.111 (0.128)	0.119 (0.119)	0.068 (0.124)	0.016 (0.036)	0.016 (0.037)	0.000 (0.033)
Common language	0.244** (0.117)	0.237** (0.108)	0.208** (0.106)	0.309* (0.158)	0.301** (0.147)	0.263* (0.144)	-0.008 (0.044)	-0.011 (0.042)	-0.021 (0.041)
Free trade agreement	0.227* (0.126)	0.231* (0.125)	0.242** (0.119)	0.246 (0.155)	0.252 (0.154)	0.270* (0.148)	0.184*** (0.054)	0.185*** (0.055)	0.184*** (0.050)
Common currency	-0.015 (0.148)	-0.013 (0.139)	-0.023 (0.139)	-0.001 (0.190)	0.001 (0.179)	-0.014 (0.179)	0.160** (0.067)	0.159** (0.065)	0.159** (0.065)

Notes. The dependent variables are all (value added) income accruing to primary factors located in origin country o due to exports of intermediate inputs that are part of GVCs that end in final good production in destination country d. In columns 1-3 it is the total income flow of this type; in columns 4-6 it is income due to direct bilateral exports of intermediate inputs $[V(Bx)Y]$; in columns 7-9 it is income due to flows that pass through at least one other country, not o and not d, i.e., complex GVCs $[V(Bg)Y]$. Income flows calculated from WIOD 2013 release in 2007. The number of affiliates are averages in 1996-2001 from Ramondo, Rodriguez-Clare and Tintelnot (2015), and other variables from the CEPII gravity dataset. The $\text{arcsinh}(x) = \ln[x + (1+x^2)^{0.5}]$ is the inverse hyperbolic sine function that approximates $\ln(x)$. All regressions are estimated on 1122 observations ($=34 \times 33$, where 34 is the number of countries in this sample after merging the data on the number of affiliates) using PPML, and include origin and destination fixed effects. Robust standard errors in parentheses are computed by two-way clustering by origin and by destination. *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$.

APPENDIX TABLES

Table A1: Labor Shares and High Skill Labor Shares, 1995-2007

Country	Labor Shares in GDP			High Skill Labor Shares in GDP		
	1995	2007	Change	1995	2007	Change
AUS	62.8	60.1	-2.7	14.1	17.9	3.9
AUT	69.7	63.6	-6.0	14.3	18.0	3.7
BEL	67.3	65.6	-1.7	14.1	17.2	3.1
BGR	53.7	50.1	-3.5	8.1	11.1	3.0
BRA	53.1	59.1	6.0	20.1	24.6	4.5
CAN	58.8	57.5	-1.3	14.1	18.1	3.9
CHN	54.7	42.0	-12.6	2.1	5.2	3.1
CYP	62.5	64.4	1.9	28.2	30.7	2.5
CZE	43.6	59.6	16.0	8.6	15.3	6.6
DEU	68.2	62.8	-5.5	22.1	24.1	2.1
DNK	65.8	69.3	3.5	19.0	25.3	6.2
ESP	65.0	61.2	-3.8	22.9	26.8	3.9
EST	65.0	59.0	-6.0	31.1	25.7	-5.4
FIN	67.0	62.8	-4.3	25.1	28.3	3.1
FRA	63.6	62.1	-1.6	21.7	25.4	3.8
GBR	67.3	68.5	1.2	21.9	30.3	8.4
GRC	50.2	57.4	7.2	14.3	20.8	6.5
HUN	64.3	60.4	-3.9	18.2	23.8	5.6
IDN	50.6	46.4	-4.3	5.8	11.6	5.8
IND	56.6	50.9	-5.7	10.0	13.8	3.8
IRL	62.3	57.1	-5.2	17.6	26.6	9.0
ITA	67.0	64.4	-2.5	10.1	13.8	3.7
JPN	60.3	56.6	-3.8	17.7	21.7	4.0
KOR	81.1	72.5	-8.6	36.5	44.3	7.8
LTU	48.7	54.3	5.6	20.3	24.0	3.7
LUX	56.1	50.3	-5.8	14.4	20.0	5.6
LVA	55.8	58.1	2.3	21.1	21.9	0.8
MEX	35.0	32.2	-2.7	9.9	8.0	-1.9
MLT	57.5	58.2	0.7	12.1	16.6	4.5
NLD	67.3	64.6	-2.7	17.9	25.6	7.6
PRT	65.0	64.9	-0.1	13.5	17.2	3.7
ROU	58.4	62.4	4.0	6.8	10.2	3.4
RUS	58.0	58.9	0.9	11.6	16.0	4.4
SVK	37.4	37.7	0.4	7.6	9.9	2.2
SVN	84.0	69.5	-14.5	22.6	26.0	3.5
SWE	64.8	65.3	0.5	15.8	22.5	6.7
TUR	33.3	37.7	4.4	5.9	10.3	4.5
TWN	65.2	56.5	-8.7	20.3	23.9	3.6
USA	60.2	59.3	-0.9	22.0	26.6	4.6
Average	59.67	58.03	-1.64	16.40	20.49	4.09
Weighted average	61.51	59.06	-2.45	18.83	23.09	4.26

Notes. Weighted averages using GDP in 1995 as weights. Source: authors' calculations based on WIOD 2013 release.

Table A2: Labor Shares and High Skill Labor Shares, 2007-2014

Country	Labor Shares in GDP			High Skill Labor Shares in GDP		
	2007	2014	Change	2008	2014	Change
AUS	58.1	57.7	-0.4			
AUT	57.3	60.9	3.6	15.2	17.7	2.4
BEL	61.8	64.1	2.2			
BGR	47.9	63.3	15.5	14.4	19.5	5.1
BRA	48.6	55.1	6.5			
CAN	59.4	58.2	-1.3			
CHN	45.4	55.1	9.6			
CYP	54.9	54.6	-0.3	25.8	30.0	4.1
CZE	50.4	51.3	0.9	10.7	11.5	0.9
DEU	58.9	62.4	3.5	21.4	18.9	-2.5
DNK	65.3	64.5	-0.8	23.8	25.6	1.8
ESP	60.7	58.7	-2.0	23.8	28.5	4.7
EST	53.8	54.8	1.0	24.3	22.3	-2.0
FIN	58.5	64.4	5.8	24.8	31.1	6.4
FRA	60.9	65.0	4.1	23.1	27.5	4.4
GBR	67.1	64.9	-2.3	19.8	24.4	4.5
GRC	53.6	49.6	-4.0	14.3	18.1	3.7
HUN	56.7	53.7	-3.0	21.4	22.1	0.8
IDN	48.3	48.5	0.2			
IND	47.6	49.7	2.1			
IRL	53.9	48.8	-5.0	22.4	24.3	1.9
ITA	56.5	58.7	2.2	10.1	9.5	-0.6
JPN	58.0	58.3	0.4			
KOR	64.1	63.8	-0.3			
LTU	54.4	48.8	-5.6	24.5	27.4	2.9
LUX	54.9	59.4	4.5	23.9	30.6	6.7
LVA	54.7	53.0	-1.7	21.2	23.0	1.8
MEX	33.6	33.0	-0.7			
MLT	60.6	58.4	-2.2			
NLD	59.9	62.1	2.2	21.6	20.5	-1.1
POL	49.4	49.8	0.3	15.1	16.5	1.4
PRT	60.6	56.7	-3.9	13.1	18.9	5.9
ROU	45.4	43.5	-1.9	14.4	12.9	-1.5
RUS	56.9	63.2	6.3			
SVK	46.0	48.7	2.7	9.8	10.2	0.4
SVN	63.2	65.6	2.5	26.9	26.1	-0.8
SWE	54.0	57.0	3.0	19.3	25.9	6.6
TUR	37.3	37.9	0.6			
USA	57.8	56.3	-1.5			
Average	54.8	55.9	1.1	19.4	21.7	2.3
Weighted average	56.4	57.4	1.0	19.4	21.3	1.9

Notes. Weighted averages using GDP in 2007 as weights. Source: authors' calculations based on WIOD 2016 release (labor shares) and EU KLEMS 2017 release (high skill labor shares).

Table A3: Forward and Backward Linkages, 1995-2007

Country	Forward Linkages: Foreign Value added Share in GDP			Backward Linkages: Foreign Value Added Share in Domestic Industries VA		
	1995	2007	Change	1995	2007	Change
AUS	13.2	15.0	1.8	10.5	10.3	-0.2
AUT	16.0	23.3	7.3	14.9	20.3	5.4
BEL	24.3	26.8	2.5	23.2	26.1	2.8
BGR	15.8	23.2	7.5	21.1	32.0	11.0
BRA	5.2	8.5	3.3	5.1	7.6	2.5
CAN	18.7	18.7	0.0	14.6	13.9	-0.7
CHN	8.7	14.2	5.5	11.2	16.5	5.3
CYP	7.3	10.1	2.8	17.8	17.9	0.1
CZE	22.3	26.2	3.8	23.7	30.0	6.3
DEU	11.0	19.6	8.6	9.2	15.1	5.9
DNK	13.2	19.2	6.0	13.6	19.8	6.2
ESP	7.7	9.7	2.0	11.0	15.1	4.1
EST	23.3	24.6	1.3	28.7	26.0	-2.8
FIN	20.3	22.7	2.4	14.8	18.7	3.9
FRA	10.2	10.3	0.1	9.7	12.1	2.4
GBR	13.4	14.5	1.2	12.3	12.3	0.0
GRC	3.4	8.7	5.4	10.9	15.3	4.4
HUN	16.5	24.8	8.3	22.8	32.3	9.5
IDN	13.5	19.1	5.6	13.1	14.3	1.2
IND	5.8	9.5	3.7	7.9	14.5	6.6
IRL	23.0	31.5	8.5	28.4	32.1	3.7
ITA	9.9	11.4	1.5	11.3	14.4	3.1
JPN	5.2	9.6	4.4	3.7	8.1	4.4
KOR	13.4	18.1	4.7	15.3	19.3	4.0
LTU	18.0	20.9	2.9	23.6	21.5	-2.1
LUX	43.1	48.1	5.0	25.0	40.0	15.0
LVA	22.0	17.8	-4.1	20.7	22.1	1.4
MEX	12.1	12.7	0.6	13.2	13.7	0.5
MLT	19.5	28.2	8.6	28.9	30.1	1.2
NLD	21.5	23.3	1.8	20.3	21.8	1.5
PRT	9.4	12.3	2.9	16.5	17.0	0.5
ROU	11.9	14.5	2.6	15.9	19.5	3.7
RUS	19.8	23.1	3.4	7.5	8.0	0.6
SVK	27.0	26.9	0.0	23.4	32.1	8.7
SVN	16.8	22.3	5.5	21.8	26.6	4.8
SWE	18.8	22.8	4.0	15.8	19.4	3.6
TUR	4.5	6.4	1.9	9.4	14.5	5.1
TWN	17.0	28.7	11.7	20.1	22.2	2.1
USA	6.0	6.2	0.2	5.1	7.1	2.1
Average	15.1	18.8	3.7	15.9	19.5	3.5
Weighted average	8.9	11.7	2.8	8.1	11.3	3.1

Notes. Weighted averages using GDP in 1995 as weights. Source: authors' calculations based on WIOD 2013 release.

Table A4: Forward and Backward Linkages, 2007-2014

Country	Forward Linkages: Foreign Value added Share in GDP			Backward Linkages: Foreign Value Added Share in Domestic Industries VA			GDP in 2007
	2007	2014	Change	2007	2014	Change	
AUS	14.0	14.8	0.9	9.6	9.6	0.0	912442
AUT	20.6	21.0	0.4	18.1	19.0	0.9	345266
BEL	24.8	26.1	1.3	22.4	27.4	5.0	422059
BGR	17.7	25.1	7.4	28.4	26.8	-1.6	38093
BRA	7.8	7.5	-0.3	7.8	8.9	1.0	1204191
CAN	17.7	17.9	0.2	13.6	15.0	1.5	1372537
CHN	11.9	9.2	-2.8	15.2	10.5	-4.8	3495060
CYP	18.6	21.1	2.6	17.9	17.3	-0.6	21436
CZE	23.0	26.8	3.8	25.8	29.6	3.8	171753
DEU	17.6	18.2	0.7	13.9	15.0	1.1	3099194
DNK	18.4	18.6	0.2	21.1	21.1	0.0	271418
ESP	9.7	10.5	0.9	14.5	13.2	-1.3	1333298
EST	24.5	29.2	4.7	24.1	27.3	3.2	19507
FIN	19.2	17.7	-1.5	17.3	18.1	0.8	224288
FRA	10.6	11.7	1.1	12.3	13.8	1.6	2394018
GBR	12.9	13.8	0.9	12.2	12.6	0.4	2664476
GRC	9.1	11.6	2.5	13.1	12.6	-0.6	281318
HUN	22.1	26.5	4.3	32.0	34.2	2.2	119649
IDN	18.1	14.8	-3.3	13.8	14.3	0.5	455190
IND	10.1	7.3	-2.8	13.9	11.7	-2.3	1135324
IRL	27.7	33.4	5.7	30.0	37.1	7.1	239541
ITA	10.9	11.4	0.5	12.9	12.4	-0.4	1982454
JPN	8.0	7.8	-0.2	8.3	10.3	2.0	4310742
KOR	15.2	19.5	4.3	17.9	20.9	3.0	1013652
LTU	20.0	27.1	7.1	19.5	21.8	2.3	35738
LUX	45.0	45.3	0.3	47.3	51.6	4.3	45275
LVA	17.7	24.0	6.2	20.5	21.2	0.6	27594
MEX	9.7	10.9	1.2	14.1	15.2	1.1	1003194
MLT	26.8	22.9	-3.9	41.2	44.7	3.6	6910
NLD	23.0	31.8	8.9	17.8	22.0	4.1	750373
POL	16.2	20.2	3.9	19.3	20.2	0.9	375515
PRT	11.9	15.1	3.2	15.7	16.7	1.0	208568
ROU	14.0	20.2	6.2	17.6	20.5	2.9	151950
RUS	23.8	24.1	0.2	7.7	8.7	1.0	1114179
SVK	23.1	25.8	2.7	31.1	31.8	0.7	69462
SVN	22.2	26.0	3.8	23.6	22.8	-0.7	42223
SWE	20.8	20.0	-0.8	17.5	15.7	-1.8	430726
TUR	10.3	12.3	2.0	14.4	15.4	1.0	581365
USA	5.1	5.9	0.8	6.2	6.4	0.3	14477638
Average	17.4	19.3	1.9	18.7	19.8	1.1	
Weighted average	10.9	11.5	0.6	11.2	11.5	0.4	

Notes. Weighted averages using GDP in 2007 as weights. Source: authors' calculations based on WIOD 2017 release.

Table A5: Decomposition of Changes in Labor Shares in GDP and in Manufacturing Value Added, 2007-2014

A. All sectors, percent in GDP

	K income from domestic industries (1)	L income from domestic industries (2)	K income from foreign industries (3)	L income from foreign industries (4)	L income (domestic + foreign) (5)	Income from foreign industries (6)
Levels						
VBV 2007	38.4	50.7	5.3	5.7	56.38	10.96
VBV 2014	37.4	51.1	5.2	6.3	57.41	11.56
Changes						
V2014*B2007*Y2007 - VBV 2007	-1.04	1.03	-0.29	0.30	1.33	0.01
V2007*B2014*Y2007 - VBV 2007	-0.26	-0.08	0.28	0.06	-0.02	0.34
V2007*B2007*Y2014 - VBV 2007	-0.19	-0.84	0.34	0.68	-0.16	1.03
V2007*B2014*Y2014 - VBV 2007	-0.11	-0.50	0.25	0.35	-0.15	0.60
VBV 2014 - VBV 2007	-1.01	0.41	-0.02	0.62	1.03	0.60

B. Manufacturing industries, percent in value added

	K income from domestic industries (1)	L income from domestic industries (2)	K income from foreign industries (3)	L income from foreign industries (4)	L income (domestic + foreign) (5)	Income from foreign industries (6)
Levels						
VBV 2007	37.5	39.4	11.0	12.1	51.51	23.07
VBV 2014	35.5	39.6	11.0	13.9	53.52	24.84
Changes						
V2014*B2007*Y2007 - VBV 2007	-1.87	1.84	-0.84	0.87	2.71	0.03
V2007*B2014*Y2007 - VBV 2007	-0.64	-0.16	0.87	-0.08	-0.24	0.79
V2007*B2007*Y2014 - VBV 2007	-0.59	-2.01	0.82	1.78	-0.23	2.61
V2007*B2014*Y2014 - VBV 2007	-0.42	-1.38	0.89	0.90	-0.48	1.80
VBV 2014 - VBV 2007	-2.03	0.25	0.01	1.76	2.01	1.77

Notes. Panel A reports decompositions of changes in factor shares in GDP, while Panel B reports decompositions of changes in factor shares within manufacturing industries' value added. Columns 1-4 report the shares of income accruing to capital and labor from domestic industries and from foreign industries. Column 5 reports the overall labor share (columns 2 + 4). Column 6 reports the share of income accruing from serving foreign industries. The split between domestic and foreign industries is given by different entries within rows in VbY. The contribution of foreign industries to factor shares is given by the forward concept defined in the text. The contribution of domestic industries is given by the complement of the forward concept. The rows labeled "Levels" report levels in 2007 and in 2014. Rows labeled as "Changes" report true and counterfactual changes. All numbers are weighted averages using GDP in 2007 as weights. Source: authors' calculations based on WIOD 2016 release.

Table A6: Sources of Compositional Changes in Payments to Labor

A. All sectors

	2007, Percent in GDP			$\Delta 2007-2014$		
	Income from domestic industries (1)	Income from foreign industries (2)	Domestic + foreign (3)	Income from domestic industries (4)	Income from foreign industries (5)	Domestic + foreign (6)
Value chains						
Domestic (Bd)	50.54	0	50.54	-0.09	0	-0.09
Bilateral trade (Bx)	0	4.43	4.43	0	0.01	0.01
Complex GVCs (Bg)	0.14	1.27	1.41	0.01	0.05	0.06
Total (B)	50.68	5.71	56.38	-0.08	0.06	-0.02
Sources of demand						
Domestic (Yd)	46.60	4.46	51.06	-0.71	0.69	-0.02
Foreign (Yf)	4.08	1.24	5.32	-0.13	-0.01	-0.14
Total (Y)	50.68	5.71	56.38	-0.84	0.68	-0.16

B. Manufacturing

	2007, Percent in GDP			$\Delta 2007-2014$		
	Income from domestic industries (1)	Income from foreign industries (2)	Domestic + foreign (3)	Income from domestic industries (4)	Income from foreign industries (5)	Domestic + foreign (6)
Value chains						
Domestic (Bd)	39.04	0	39.04	-0.17	0	-0.17
Bilateral trade (Bx)	0	9.34	9.34	0	-0.06	-0.06
Complex GVCs (Bg)	0.35	2.78	3.13	0.01	-0.02	-0.01
Total (B)	39.39	12.12	51.51	-0.16	-0.08	-0.24
Sources of demand						
Domestic (Yd)	28.63	9.02	37.65	-1.27	1.74	0.47
Foreign (Yf)	10.76	3.09	13.86	-0.74	0.04	-0.70
Total (Y)	39.39	12.12	51.51	-2.01	1.78	-0.23

Notes. Panel A reports decompositions of levels and changes in labor shares in GDP, while Panel B reports decomposition of levels and changes in labor shares within manufacturing industries' value added. The four "Total" rows in columns 1-3 report labor shares in GDP that are paid by domestic industries, foreign industries, and overall in 2007; these are the same numbers in columns 2, 4 and 5 in Table A5. The "Total" rows in columns 4-6 report the changes in the same concepts. The rows above the "Total" rows indicate the contributions of sub-components of either B or Y to levels in columns 1-3 or to changes in columns 4-6. Source: authors' calculations based on WIOD 2013 release.

Table A7. Descriptive statistics, 1995-2014

	mean	Std.Err.	p25	p50	p75	IQR
A. Manufacturing industries (N = 1532)						
Δ labor share	0.001	0.084	-0.052	-0.002	0.049	0.100
Δ forward intensity	0.031	0.048	0.002	0.025	0.050	0.048
- to China	0.010	0.016	0.001	0.005	0.013	0.012
- to RoW	0.021	0.041	0.001	0.017	0.040	0.039
- net of assembly offshoring	0.044	0.069	0.006	0.032	0.078	0.071
- absorbed in destination	0.032	0.053	0.002	0.022	0.063	0.062
- re-exported	0.012	0.023	0.001	0.009	0.022	0.021
Δ backward intensity	0.011	0.026	-0.001	0.008	0.020	0.021
Δ exports of final goods intensity	0.009	0.041	-0.001	0.004	0.018	0.019
Δ imports of final goods intensity	0.046	0.234	0.006	0.036	0.085	0.080
Δ log relative price of investment	-0.067	0.143	-0.158	-0.072	0.004	0.163
B. All private sector industries (N = 3270)						
Δ labor share	0.000	0.070	-0.033	-0.001	0.035	0.068
Δ forward intensity	0.016	0.038	-0.001	0.007	0.026	0.027
- to China	0.005	0.010	0.000	0.002	0.005	0.005
- to RoW	0.011	0.034	-0.002	0.004	0.019	0.021
- net of assembly offshoring	0.020	0.048	-0.001	0.009	0.029	0.030
- absorbed in destination	0.014	0.038	-0.001	0.005	0.021	0.022
- re-exported	0.005	0.014	0.000	0.002	0.008	0.008
Δ backward intensity	0.011	0.023	0.000	0.008	0.020	0.020
Δ exports of final goods intensity	0.003	0.023	0.000	0.000	0.004	0.004
Δ imports of final goods intensity	0.015	0.179	0.000	0.001	0.014	0.014
Δ log relative price of investment	-0.064	0.139	-0.116	-0.080	0.004	0.120
C. Manufacturing (N = 117)						
Δ labor share	-0.006	0.057	-0.052	-0.005	0.031	0.082
Δ forward intensity	0.031	0.035	0.014	0.027	0.049	0.036
Δ backward intensity	0.012	0.015	0.008	0.013	0.023	0.015
Δ exports of final goods intensity	0.009	0.024	0.003	0.009	0.015	0.012
Δ imports of final goods intensity	0.047	0.045	0.032	0.051	0.073	0.041
Δ log relative price of investment	-0.067	0.143	-0.158	-0.072	0.004	0.163
D. Entire economy (N = 117)						
Δ labor share	-0.006	0.028	-0.021	-0.013	0.012	0.033
Δ forward intensity	0.012	0.020	0.002	0.008	0.014	0.013
Δ backward intensity	0.012	0.014	0.004	0.012	0.023	0.020
Δ exports of final goods intensity	0.002	0.010	-0.002	0.001	0.008	0.010
Δ imports of final goods intensity	0.006	0.012	-0.005	0.007	0.012	0.016
Δ log relative price of investment	-0.064	0.139	-0.116	-0.080	0.004	0.120

Notes. Statistics are for changes in country-industry specific variables (panels A and B) or in country specific variables (panels C and D), in three periods: 1995-2001, 2001-2007, 2007-2014. Δ labor shares is the change in labor compensation ratio in value added. Δ forward is the change in intermediate inputs exports intensity in value added (forward GVC integration). Δ forward net of assembly offshoring is the change in intermediate inputs exports intensity in value added that excludes exports of final goods. Δ forward absorbed in destination is the part of Δ forward net of assembly offshoring value added that cross borders only once and is absorbed in the direct destination country. Δ forward re-exported is the part of Δ forward net of assembly offshoring value added that is embodied in exports of the destination country, i.e. complex GVCs that cross borders more than once. Δ backward is the change in importing of intermediate inputs intensity in total input purchases, Δ exports of final goods is the change in export intensity of final goods in value added, and Δ imports of final goods is the change in import intensity of final goods in domestic absorption. Δ log relative price of investment is the country-level log change in investment prices. IQR = p75-p25 is the inter-quartile range.

Table A8. Changes in Labor Shares and Forward GVC Integration in 1995-2014, W2SLS

	Dependent variable in W2SLS: Δ labor share																							
A. Manufacturing industries	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)	(11)	(12)	(13)	(14)	(15)	(16)	(17)	(18)	(19)	(20)	(21)	(22)	(23)	(24)
	W2SLS	1st stage	W2SLS	1st stage	W2SLS	1st stage	W2SLS	1st stage	W2SLS	1st stage	W2SLS	1st stage	W2SLS	1st stage	W2SLS	1st stage	W2SLS	1st stage	W2SLS	1st stage	W2SLS	1st stage	W2SLS	1st stage
Δ forward intensity	-1.485*** (0.313)		-1.508*** (0.357)		-1.191*** (0.295)		-1.584*** (0.336)		-1.277*** (0.293)		-1.052*** (0.233)		-1.208*** (0.084)		-0.852** (0.298)		-1.133*** (0.188)		-1.262** (0.468)		-0.482* (0.243)		-0.266 (0.434)	
Δ backward intensity	-0.546 (0.337)	-0.423*** (0.134)	-0.644** (0.283)	-0.446*** (0.134)	-0.545* (0.288)	-0.492*** (0.096)	-0.652 (0.389)	-0.459*** (0.135)	-0.626* (0.332)	-0.510*** (0.092)	-0.570** (0.229)	-0.532*** (0.090)	-0.642 (0.378)	-0.541*** (0.107)	-0.409 (0.234)	-0.579*** (0.094)	-0.600** (0.213)	-0.603*** (0.090)	-0.631* (0.355)	-0.480*** (0.132)	-0.339 (0.208)	-0.522*** (0.087)	-0.168 (0.250)	-0.589*** (0.087)
Δ exports of final goods intensity	0.211 (0.154)	0.159** (0.063)	0.282* (0.138)	0.177*** (0.058)	0.145 (0.107)	0.113* (0.058)	0.337* (0.167)	0.210*** (0.063)	0.240* (0.129)	0.151** (0.059)	0.256* (0.141)	0.159*** (0.055)	0.304* (0.162)	0.222** (0.074)	0.106 (0.069)	0.021 (0.057)	0.202* (0.098)	0.061 (0.053)	0.299 (0.188)	0.256*** (0.064)	0.104 (0.102)	0.177** (0.064)	0.069 (0.060)	0.074 (0.061)
Δ imports of final goods intensity	0.029 (0.025)	0.012 (0.009)	0.031 (0.025)	0.012 (0.009)	0.018 (0.017)	0.005 (0.005)	0.027 (0.025)	0.011 (0.008)	0.016 (0.015)	0.005 (0.005)	0.017 (0.010)	0.004 (0.005)	0.028 (0.016)	0.005 (0.004)	0.012 (0.008)	0.005 (0.005)	0.011 (0.009)	0.004 (0.004)	0.021 (0.023)	0.012 (0.008)	0.009 (0.009)	0.006 (0.005)	0.001 (0.005)	0.005 (0.004)
Δ log relative price of investment	0.045 (0.061)	0.075** (0.030)	0.108 (0.069)	0.091*** (0.030)	0.006 (0.068)	0.057** (0.025)	0.040 (0.063)	0.069** (0.025)	0.005 (0.068)	0.055** (0.025)	0.053 (0.082)	0.061*** (0.019)	0.063 (0.077)	0.059** (0.026)					0.084 (0.076)	0.080** (0.027)	0.018 (0.087)	0.060** (0.025)		
Δ Z		0.031*** (0.005)		0.030*** (0.008)		0.036*** (0.006)		0.030*** (0.005)		0.035*** (0.006)		0.047*** (0.008)		0.047*** (0.008)		0.040*** (0.008)		0.037*** (0.009)		0.021** (0.009)		0.042*** (0.009)		0.032*** (0.010)
Fixed effects	-	-	Per	Per	Cty	Cty	Ind	Ind	Cty, Ind	Cty, Ind	Cty, Ind, Per	Cty, Ind, Per	Cty-Ind, Per	Cty-Ind, Per	Cty-Per	Cty-Per	Cty-Per, Ind	Cty-Per, Ind	Ind-Per	Ind-Per	Cty, Ind-Per	Cty, Ind-Per	Cty-Per, Ind-Per	Cty-Per, Ind-Per
Observations	1,532	1,532	1,532	1,532	1,532	1,532	1,532	1,532	1,532	1,532	1,532	1,532	1,532	1,532	1,532	1,532	1,532	1,532	1,532	1,532	1,532	1,532	1,532	1,532
R-squared		0.151		0.161		0.361		0.233		0.422		0.434		0.553		0.549		0.606		0.285		0.465		0.630
Kleibergen-Paap F statistic	44.96		16.04		42.29		43		41.42		31.99		30.76		24.77		17.24		5.790		23.09		10.27	
B. All private sector industries																								
	W2SLS	1st stage	W2SLS	1st stage	W2SLS	1st stage	W2SLS	1st stage	W2SLS	1st stage	W2SLS	1st stage	W2SLS	1st stage	W2SLS	1st stage	W2SLS	1st stage	W2SLS	1st stage	W2SLS	1st stage	W2SLS	1st stage
Δ forward intensity	-1.369*** (0.240)		-1.074*** (0.349)		-1.339*** (0.288)		-1.483*** (0.309)		-1.452*** (0.321)		-1.426*** (0.501)		-1.424** (0.606)		-1.619*** (0.560)		-2.147** (0.859)		-1.051** (0.412)		-1.144*** (0.372)		-1.699** (0.621)	
Δ backward intensity	-0.495*** (0.167)	-0.347*** (0.092)	-0.432*** (0.136)	-0.348*** (0.092)	-0.572*** (0.184)	-0.418*** (0.089)	-0.485** (0.196)	-0.329*** (0.087)	-0.561** (0.209)	-0.401*** (0.079)	-0.593*** (0.222)	-0.403*** (0.079)	-0.522** (0.230)	-0.332*** (0.085)	-0.894*** (0.262)	-0.540*** (0.103)	-1.145** (0.449)	-0.538*** (0.086)	-0.403* (0.200)	-0.332*** (0.094)	-0.498** (0.203)	-0.409*** (0.083)	-0.949** (0.347)	-0.547*** (0.086)
Δ exports of final goods intensity	0.469** (0.197)	0.330*** (0.088)	0.385 (0.229)	0.327*** (0.084)	0.414** (0.184)	0.308*** (0.079)	0.491* (0.250)	0.308*** (0.095)	0.438* (0.229)	0.287*** (0.087)	0.452 (0.268)	0.282*** (0.081)	0.464 (0.304)	0.280*** (0.098)	0.554* (0.281)	0.289*** (0.069)	0.696* (0.380)	0.270*** (0.078)	0.369 (0.264)	0.340*** (0.099)	0.366 (0.247)	0.312*** (0.088)	0.590* (0.341)	0.303*** (0.084)
Δ imports of final goods intensity	0.027 (0.018)	0.013 (0.009)	0.024 (0.016)	0.013 (0.008)	0.022 (0.016)	0.011 (0.007)	0.021 (0.015)	0.009 (0.006)	0.017 (0.012)	0.007 (0.004)	0.017 (0.012)	0.007* (0.004)	0.022* (0.012)	0.008** (0.003)	0.022 (0.016)	0.011 (0.007)	0.018 (0.013)	0.007* (0.004)	0.017 (0.012)	0.010* (0.006)	0.014 (0.010)	0.008** (0.004)	0.014 (0.010)	0.008* (0.004)
Δ log relative price of investment	0.028 (0.033)	0.043** (0.016)	0.048 (0.036)	0.048*** (0.016)	0.030 (0.040)	0.040** (0.015)	0.031 (0.035)	0.043*** (0.016)	0.032 (0.041)	0.039** (0.015)	0.077 (0.049)	0.041** (0.018)	0.075 (0.056)	0.038** (0.018)					0.048 (0.042)	0.050*** (0.017)	0.065 (0.050)	0.040** (0.018)		
Δ Z		0.015*** (0.003)		0.021*** (0.006)		0.016*** (0.004)		0.014*** (0.003)		0.015*** (0.004)		0.024*** (0.007)		0.024*** (0.008)		0.020*** (0.005)		0.017*** (0.006)		0.021*** (0.006)		0.028*** (0.006)		0.020*** (0.004)
Fixed effects	-	-	Per	Per	Cty	Cty	Ind	Ind	Cty, Ind	Cty, Ind	Cty, Ind, Per	Cty, Ind, Per	Cty-Ind, Per	Cty-Ind, Per	Cty-Per	Cty-Per	Cty-Per, Ind	Cty-Per, Ind	Ind-Per	Ind-Per	Cty, Ind-Per	Cty, Ind-Per	Cty-Per, Ind-Per	Cty-Per, Ind-Per
Observations	3,270	3,270	3,270	3,270	3,270	3,270	3,270	3,270	3,270	3,270	3,270	3,270	3,270	3,270	3,270	3,270	3,270	3,270	3,270	3,270	3,270	3,270	3,270	3,270
R-squared		0.131		0.139		0.239		0.231		0.327		0.336		0.477		0.425		0.511		0.284		0.384		0.553
Kleibergen-Paap F statistic	20.69		12.48		17.10		19.93		17.04		10.47		9.004		17.65		9.217		14.59		20.22		21.17	

Notes. The dependent variable is changes in country-industry specific labor shares in three periods: 1995-2001, 2001-2007, 2007-2014. Δ forward is the change in intermediate inputs exports intensity in value added (forward GVC integration). The instrument Δ Z is constructed by eliminating home-country and industry sources of variation from Δ forward in a gravity relationship; see text for complete details. Δ backward is the change in importing of intermediate inputs intensity in total input purchases, Δ exports of final goods is the change in export intensity of final goods in value added, and Δ imports of final goods is the change in import intensity of final goods in domestic absorption. Δ log relative price of investment is the country-level log change in investment prices. All regressions are weighted by value added in 1995 as weights. A constant is always included but not reported. Robust standard errors in parentheses are computed by two-way clustering by country and by industry. *** p<0.01, ** p<0.05, * p<0.1.

Table A9. Changes in Labor Shares in 1995-2014, W2SLS: instrumenting for both Δ forward intensity and Δ backward intensity

	Dependent variable in W2SLS: Δ labor share											
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)	(11)	(12)
A. Manufacturing industries												
Δ forward intensity	-3.191 (1.986)	-4.388 (5.184)	-1.716** (0.750)	-2.721** (1.168)	-1.684*** (0.541)	-1.412*** (0.351)	-1.603*** (0.480)	-1.688* (0.994)	-1.683*** (0.486)	-2.653 (2.717)	-0.687 (0.474)	-0.778 (0.668)
Δ backward intensity	6.153 (5.822)	7.936 (11.881)	1.656 (2.463)	4.614 (3.442)	1.388 (1.719)	0.944 (1.399)	0.818 (1.673)	1.580 (2.517)	1.217 (1.536)	4.221 (5.280)	0.438 (1.212)	1.293 (1.015)
Δ exports of final goods intensity	-1.653 (1.712)	-2.088 (3.429)	-0.487 (0.812)	-1.073 (0.963)	-0.313 (0.550)	-0.161 (0.452)	-0.062 (0.541)	-0.550 (0.879)	-0.339 (0.483)	-1.025 (1.423)	-0.118 (0.399)	-0.392 (0.343)
Δ imports of final goods intensity	0.104 (0.114)	0.136 (0.211)	0.045 (0.047)	0.081 (0.076)	0.040 (0.034)	0.035 (0.026)	0.044 (0.028)	0.039 (0.046)	0.033 (0.028)	0.075 (0.097)	0.018 (0.018)	0.020 (0.019)
Δ log relative price of investment	0.045 (0.074)	0.129 (0.155)	0.027 (0.073)	0.032 (0.068)	0.021 (0.072)	0.064 (0.089)	0.075 (0.091)			0.079 (0.097)	0.025 (0.093)	
Fixed effects	-	Per	Cty	Ind	Cty, Ind	Cty, Ind, Per	Cty-Ind, Per	Cty-Per	Cty-Per, Ind	Ind-Per	Cty, Ind- Per	Cty-Per, Ind-Per
Observations	1,532	1,532	1,532	1,532	1,532	1,532	1,532	1,532	1,532	1,532	1,532	1,532
Kleibergen-Paap F statistic	0.968	0.320	3.938	2.237	6.881	4.575	3.223	1.303	3.282	0.375	3.593	2.903
B. All private sector industries												
Δ forward intensity	-2.246** (0.905)	-0.932*** (0.278)	-2.370** (0.872)	-2.397** (1.115)	-2.501** (1.050)	-1.631** (0.657)	-1.659** (0.726)	-1.210** (0.458)	-1.545** (0.708)	-0.921*** (0.324)	-1.357*** (0.443)	-1.208** (0.517)
Δ backward intensity	2.447* (1.250)	1.192** (0.459)	2.407* (1.217)	2.460 (1.462)	2.374* (1.303)	1.622* (0.812)	1.395* (0.780)	1.751** (0.742)	1.885** (0.845)	1.201* (0.607)	1.451* (0.783)	1.883* (0.950)
Δ exports of final goods intensity	-0.030 (0.405)	-0.055 (0.209)	-0.037 (0.428)	-0.041 (0.444)	-0.045 (0.453)	-0.036 (0.314)	0.046 (0.347)	-0.063 (0.246)	-0.041 (0.302)	-0.104 (0.250)	-0.080 (0.309)	-0.148 (0.308)
Δ imports of final goods intensity	0.057 (0.039)	0.030 (0.018)	0.056 (0.037)	0.045 (0.034)	0.044 (0.031)	0.032 (0.021)	0.039** (0.019)	0.037 (0.025)	0.033 (0.023)	0.023 (0.015)	0.027 (0.018)	0.029 (0.020)
Δ log relative price of investment	-0.023 (0.042)	-0.017 (0.034)	0.013 (0.043)	-0.014 (0.042)	0.016 (0.044)	0.027 (0.050)	0.035 (0.053)			-0.014 (0.039)	0.022 (0.052)	
Fixed effects	-	Per	Cty	Ind	Cty, Ind	Cty, Ind, Per	Cty-Ind, Per	Cty-Per	Cty-Per, Ind	Ind-Per	Cty, Ind- Per	Cty-Per, Ind-Per
Observations	3,270	3,270	3,270	3,270	3,270	3,270	3,270	3,270	3,270	3,270	3,270	3,270
Kleibergen-Paap F statistic	3.361	4.316	3.067	2.638	2.657	2.114	2.582	3.259	2.483	4.728	3.283	3.502

Notes. The dependent variable is changes in country-industry specific labor shares in three periods: 1995-2001, 2001-2007, 2007-2014. Δ forward is the change in intermediate inputs exports intensity in value added (forward GVC integration). Δ backward is the change in importing of intermediate inputs intensity in total input purchases, Δ exports of final goods is the change in export intensity of final goods in value added, and Δ imports of final goods is the change in import intensity of final goods in domestic absorption. Δ log relative price of investment is the country-level log change in investment prices. The instruments for Δ forward and Δ backward are constructed by eliminating home-country and industry sources of variation in a gravity relationship; see text for complete details. All regressions are weighted by value added in 1995 as weights. A constant is always included but not reported. Robust standard errors in parentheses are computed by two-way clustering by country and by industry. *** p<0.01, ** p<0.05, * p<0.1.

Table A10. Changes in Labor Shares in 1995-2014, W2SLS: instrumenting for both Δ forward intensity and Δ backward intensity

	(1)	(2)	(3)	(4)	(5)	(6)
	A. Manufacturing industries			B. Private sector industries		
	W2SLS	1st stage		W2SLS	1st stage	
Dependent variable:	Δ labor share	Δ forward	Δ backward	Δ labor share	Δ forward	Δ backward
Δ forward intensity	-1.603*** (0.480)			-1.659** (0.726)		
Δ backward intensity	0.818 (1.673)			1.395* (0.780)		
Δ exports of final goods intensity	-0.062 (0.541)	0.050 (0.079)	0.251*** (0.032)	0.046 (0.347)	0.170 (0.112)	0.215*** (0.035)
Δ imports of final goods intensity	0.044 (0.028)	0.009 (0.006)	-0.010* (0.005)	0.039** (0.019)	0.009** (0.003)	-0.009*** (0.003)
Δ log relative price of investment	0.075 (0.091)	0.034* (0.020)	-0.010 (0.010)	0.035 (0.053)	0.013 (0.016)	0.008 (0.010)
IV for Δ forward intensity		0.035*** (0.007)	0.006** (0.003)		0.021*** (0.007)	0.001 (0.004)
IV for Δ backward intensity		0.047*** (0.011)	0.037*** (0.009)		0.042** (0.011)	0.041*** (0.009)
Fixed effects	Cty-Ind, Per	Cty-Ind, Per	Cty-Ind, Per	Cty-Ind, Per	Cty-Ind, Per	Cty-Ind, Per
Observations	1,454	1,454	1,454	3,078	3,078	3,078
Cluster	cty	cty	cty	cty ind	cty ind	cty ind
Kleibergen-Paap F statistic	3.223			2.582		
F test of excluded instruments		23.44	9.99		9.75	10.73
Sanderson-Windmeijer multivariate F test		8.65	6.81		5.26	5.88

Notes. The dependent variable is changes in country-industry specific labor shares in three periods: 1995-2001, 2001-2007, 2007-2014. Δ forward is the change in intermediate inputs exports intensity in value added (forward GVC integration). Δ backward is the change in importing of intermediate inputs intensity in total input purchases, Δ exports of final goods is the change in export intensity of final goods in value added, and Δ imports of final goods is the change in import intensity of final goods in domestic absorption. Δ log relative price of investment is the country-level log change in investment prices. The instruments for Δ forward and Δ backward are constructed by eliminating home-country and industry sources of variation in a gravity relationship; see text for complete details. All regressions are weighted by value added in 1995 as weights. A constant is always included but not reported. Robust standard errors in parentheses are computed by two-way clustering by country and by industry. *** p<0.01, ** p<0.05, * p<0.1.

Table A11. Changes in Labor Shares in 1995-2014, W2SLS: instrumenting for all globalization variables

	Dependent variable in W2SLS: Δ labor share											
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)	(11)	(12)
A. Manufacturing industries												
Δ forward intensity	-2.964** (1.323)	-4.863 (5.664)	-1.468* (0.848)	-2.740** (1.026)	-1.607* (0.891)	-1.760 (1.166)	-1.369 (1.638)	-1.782 (1.331)	-2.287 (1.393)	-12.899 (64.016)	0.373 (0.719)	-2.754 (2.794)
Δ backward intensity	2.895 (3.015)	4.802 (9.090)	-1.087 (2.827)	1.506 (2.341)	-1.457 (3.387)	-1.236 (2.193)	-2.063 (3.039)	-0.171 (2.323)	-1.027 (2.529)	9.075 (48.604)	-1.465 (1.211)	-0.062 (1.236)
Δ exports of final goods intensity	3.409 (3.819)	2.838 (3.889)	4.606 (5.280)	4.795 (5.200)	5.819 (6.908)	4.604 (4.103)	5.882 (5.382)	2.856 (3.417)	4.438 (5.249)	10.950 (55.203)	0.394 (0.399)	3.451 (4.539)
Δ imports of final goods intensity	0.212 (0.301)	0.514 (0.479)	-0.050 (0.263)	0.032 (0.376)	-0.185 (0.407)	-0.027 (0.248)	-0.163 (0.335)	0.134 (0.287)	0.005 (0.365)	0.893 (2.952)	-0.011 (0.018)	0.185 (0.273)
Δ log relative price of investment	-0.133 (0.153)	0.079 (0.180)	-0.199 (0.254)	-0.183 (0.204)	-0.245 (0.321)	-0.045 (0.139)	-0.102 (0.188)			0.137 (0.757)	-0.027 (0.090)	
Fixed effects	-	Per	Cty	Ind	Cty, Ind	Cty, Ind, Per	Cty-Ind, Per	Cty-Per	Cty-Per, Ind	Ind-Per	Cty, Ind- Per	Cty-Per, Ind-Per
Observations	1,532	1,532	1,532	1,532	1,532	1,532	1,532	1,532	1,532	1,532	1,532	1,532
Kleibergen-Paap F statistic	0.799	0.210	0.539	0.555	0.464	1.121	0.677	0.344	0.370	0.0101	3.870	0.363
B. All private sector industries												
Δ forward intensity	-1.925** (0.752)	-0.915* (0.498)	-2.056*** (0.716)	-2.402** (1.032)	-2.566** (1.008)	-2.036*** (0.654)	-2.157** (0.926)	-1.216** (0.583)	-1.960*** (0.683)	-1.136 (0.769)	-0.849*** (0.306)	-1.773** (0.729)
Δ backward intensity	1.484* (0.791)	0.752 (0.492)	1.517* (0.828)	1.944 (1.311)	1.994 (1.333)	1.509 (0.957)	1.549 (1.383)	1.080* (0.553)	1.619* (0.823)	0.883 (0.815)	0.845 (0.578)	1.425 (0.912)
Δ exports of final goods intensity	-0.339 (2.975)	-0.429 (1.920)	0.083 (2.797)	2.089 (3.744)	2.505 (3.107)	1.864 (2.298)	3.571 (3.441)	-0.095 (1.871)	1.697 (1.786)	0.724 (2.778)	-0.048 (0.237)	1.584 (1.793)
Δ imports of final goods intensity	0.621 (0.625)	0.422 (0.420)	0.556 (0.598)	0.008 (0.838)	-0.063 (0.723)	0.059 (0.542)	-0.465 (0.826)	0.492 (0.421)	0.125 (0.397)	0.213 (0.453)	0.019 (0.012)	0.248 (0.309)
Δ log relative price of investment	0.001 (0.048)	0.007 (0.044)	0.024 (0.051)	-0.027 (0.073)	-0.006 (0.073)	0.031 (0.064)	0.010 (0.085)			-0.001 (0.048)	0.016 (0.049)	
Fixed effects	-	Per	Cty	Ind	Cty, Ind	Cty, Ind, Per	Cty-Ind, Per	Cty-Per	Cty-Per, Ind	Ind-Per	Cty, Ind- Per	Cty-Per, Ind-Per
Observations	3,270	3,270	3,270	3,270	3,270	3,270	3,270	3,270	3,270	3,270	3,270	3,270
Kleibergen-Paap F statistic	1.220	0.967	1.125	0.868	1.079	0.840	0.718	1.271	1.425	0.454	3.153	2.043

Notes. The dependent variable is changes in country-industry specific labor shares in three periods: 1995-2001, 2001-2007, 2007-2014. Δ forward is the change in intermediate inputs exports intensity in value added (forward GVC integration). Δ backward is the change in importing of intermediate inputs intensity in total input purchases, Δ exports of final goods is the change in export intensity of final goods in value added, and Δ imports of final goods is the change in import intensity of final goods in domestic absorption. Δ log relative price of investment is the country-level log change in investment prices. The instruments for Δ forward, Δ backward, Δ exports of final goods and Δ imports of final goods are constructed by eliminating home-country and industry sources of variation in a gravity relationship; see text for complete details. All regressions are weighted by value added in 1995 as weights. A constant is always included but not reported. Robust standard errors in parentheses are computed by two-way clustering by country and by industry. *** p<0.01, ** p<0.05, * p<0.1.

Table A12. Changes in Labor Shares and Forward GVC Integration in 1995-2014: Country level regressions, W2SLS

	Dependent variable in W2SLS: Δ labor share							
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
A. Manufacturing industries								
	W2SLS	1st stage	W2SLS	1st stage	W2SLS	1st stage	W2SLS	1st stage
Δ forward intensity	-1.342*** (0.300)		-2.608 (2.315)		-1.147*** (0.287)		-0.932** (0.441)	
Δ backward intensity	-0.571 (0.662)	-0.344 (0.402)	-1.601 (1.344)	-0.492 (0.399)	-0.502 (0.674)	-0.323 (0.317)	-0.693 (0.631)	-0.388 (0.304)
Δ exports of final goods intensity	0.180 (0.570)	0.521** (0.225)	1.381 (1.812)	0.669*** (0.224)	-0.088 (0.562)	0.541** (0.233)	0.027 (0.632)	0.554** (0.225)
Δ imports of final goods intensity	0.462* (0.249)	0.170** (0.069)	0.715 (0.492)	0.170** (0.065)	0.277 (0.243)	0.004 (0.089)	0.390** (0.186)	-0.012 (0.078)
Δ log relative price of investment	-0.008 (0.055)	0.046 (0.028)	0.122 (0.143)	0.059** (0.026)	0.008 (0.074)	0.042* (0.023)	0.055 (0.083)	0.047* (0.024)
ΔZ		0.039*** (0.007)		0.017 (0.013)		0.045*** (0.007)		0.075*** (0.016)
Fixed effects	-	-	Per	Per	Cty	Cty	Cty, Per	Cty, Per
Observations	114	114	114	114	114	114	114	114
R-squared		0.382		0.427		0.706		0.728
Kleibergen-Paap F statistic	33.97		1.843		42.46		22.87	
B. All industries								
	W2SLS	1st stage	W2SLS	1st stage	W2SLS	1st stage	W2SLS	1st stage
Δ forward intensity	-0.845 (0.602)		1.014 (1.263)		-1.008 (0.704)		0.628 (1.080)	
Δ backward intensity	0.020 (0.222)	-0.139 (0.135)	0.202 (0.539)	-0.209 (0.189)	0.218 (0.400)	-0.048 (0.179)	0.042 (0.525)	-0.220 (0.243)
Δ exports of final goods intensity	-0.308 (0.560)	0.880*** (0.267)	-1.953 (1.273)	0.904*** (0.270)	-0.589 (0.714)	0.960*** (0.272)	-2.032* (1.165)	1.041*** (0.277)
Δ imports of final goods intensity	0.284 (0.398)	0.371* (0.215)	-0.272 (0.752)	0.464* (0.262)	-0.070 (0.390)	0.185 (0.269)	-0.164 (0.487)	0.307 (0.308)
Δ log relative price of investment	0.009 (0.024)	0.011 (0.013)	-0.008 (0.049)	0.017 (0.013)	0.039 (0.032)	0.016 (0.015)	0.014 (0.053)	0.015 (0.016)
ΔZ		0.014*** (0.004)		0.019* (0.011)		0.014*** (0.004)		0.032* (0.018)
FEs	-	-	Per	Per	Cty	Cty	Cty, Per	Cty, Per
Observations	113	113	113	113	113	113	113	113
R-squared		0.448		0.466		0.627		0.648
Kleibergen-Paap F statistic	16.62		3.033		11.70		3.120	

Notes. The dependent variable is changes in country-level labor shares in three periods: 1995-2001, 2001-2007, 2007-2014. Δ forward is the change in intermediate inputs exports intensity in value added (forward GVC integration). The instrument ΔZ is constructed by eliminating home-country and industry sources of variation from Δ forward in a based on a gravity relationship; see text for complete details. Δ backward is the change in importing of intermediate inputs intensity in total input purchases, Δ exports of final goods is the change in export intensity of final goods in GDP, and Δ imports of final goods is the change in import intensity of final goods in domestic absorption. Δ log relative price of investment is the country-level log change in investment prices. All regressions are weighted by value added in 1995 as weights. A constant is always included but not reported. Robust standard errors in parentheses are clustered by country. *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$.

Table A13. "Structural equation" used for the "local to zero" approximation of Conley, Hansen and Rossi (2012)

	Dependent variable: Δ labor share	
	Manufacturing industries	Private sector industries
	WLS	WLS
Δ forward intensity	-0.324*** (0.087)	-0.229** (0.107)
Δ backward intensity	-0.164 (0.338)	-0.125 (0.144)
Δ exports of final goods intensity	0.108 (0.088)	0.130 (0.124)
Δ imports of final goods intensity	0.024 (0.014)	0.013 (0.008)
Δ log relative price of investment	0.011 (0.078)	0.030 (0.047)
ΔZ	-0.041*** (0.005)	-0.029** (0.011)
Fixed effects: Cty-Ind, Period	Yes	Yes
Observations	1,459	3,086
R-squared	0.311	0.315

Notes. The dependent variable is changes in country-industry specific labor shares in three periods: 1995-2001, 2001-2007, 2007-2014. Δ forward is the change in intermediate inputs exports intensity in value added (forward GVC integration). ΔZ is the instrument, here added to the second stage equation, constructed by eliminating home-country and industry sources of variation from Δ forward in a gravity relationship; see text for complete details. Δ backward is the change in importing of intermediate inputs intensity in total input purchases, Δ exports of final goods is the change in export intensity of final goods in value added, and Δ imports of final goods is the change in import intensity of final goods in domestic absorption. Δ log relative price of investment is the country-level log change in investment prices. All regressions are weighted by value added in 1995 as weights. A constant is always included but not reported. Robust standard errors in parentheses are computed by two-way clustering by country and by industry. *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$.

Table A14. Confidence intervals for ΔFWD obtained using the "local to zero" approximation

Sample	Distribution		Multiples of the variance of the coefficient to ΔZ in Table A10			
			1	10	25	50
Manufacturing	Uniform	Lower Bound	-1.644	-1.650	-1.655	-1.666
		Upper Bound	-0.738	-0.759	-0.776	-0.779
	Normal	Lower Bound	-1.641	-1.647	-1.650	-1.637
		Upper Bound	-0.761	-0.758	-0.772	-0.770
Private sector	Uniform	Lower Bound	-2.373	-2.387	-2.441	-2.478
		Upper Bound	-0.497	-0.469	-0.518	-0.555
	Normal	Lower Bound	-2.349	-2.333	-2.349	-2.438
		Upper Bound	-0.461	-0.468	-0.494	-0.410

Notes. Methodology follows Conley, Hansen and Rossi (2012). The Null hypothesis is that the instrument ΔZ is excluded from the second stage, structural equation, i.e., it's coefficient in this equation is zero. The "local to zero" (LTZ) approximation takes into account uncertainty about whether the true value of this coefficient is zero. The table reports confidence intervals for the endogenous variable ΔFWD in the second stage regression that apply the LTZ approximation with different distributional assumptions and different variances of the value of the coefficient to ΔZ in the structural equation. These variances are multiples of the variance of the coefficient to ΔZ in Table A13.

Table A15. Functional Specialization, Upstreamness and Forward Foreign GVC Integration, 2001-2007

	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)	(11)	(12)
A. Manufacturing												
Dependent variable:	Δ MGT	Δ R&D	Δ FAB	Δ MAR	Δ MGT	Δ R&D	Δ FAB	Δ MAR	Δ MGT	Δ R&D	Δ FAB	Δ MAR
Δ forward intensity	-0.167*** (0.049)	-0.019 (0.047)	-0.398*** (0.101)	-0.141*** (0.038)	-0.181** (0.066)	-0.007 (0.058)	-0.256*** (0.071)	-0.102*** (0.029)	-0.091*** (0.026)	-0.018 (0.032)	-0.368* (0.194)	-0.093* (0.045)
Δ backward intensity	-0.153 (0.088)	-0.125 (0.078)	-0.283* (0.147)	-0.079 (0.050)	-0.181* (0.096)	-0.128 (0.086)	-0.238 (0.154)	-0.056 (0.068)	-0.024 (0.030)	-0.077 (0.046)	-0.199* (0.100)	-0.036 (0.030)
Δ exports of final goods intensity	0.001 (0.006)	0.014 (0.009)	0.016 (0.014)	0.006 (0.010)	0.004 (0.008)	0.011 (0.007)	0.012 (0.011)	0.006 (0.011)	0.003 (0.005)	0.021*** (0.007)	0.027* (0.014)	0.008 (0.008)
Δ imports of final goods intensity	0.002 (0.012)	-0.021 (0.018)	-0.010 (0.029)	0.013 (0.021)	0.013 (0.019)	-0.016 (0.015)	-0.065** (0.026)	0.007 (0.018)	-0.010 (0.009)	-0.013 (0.010)	0.002 (0.041)	0.017 (0.019)
Δ log relative price of investment	-0.001 (0.022)	-0.023 (0.035)	0.018 (0.025)	0.020 (0.015)	-0.001 (0.023)	-0.025 (0.035)	0.006 (0.029)	0.018 (0.017)				
Fixed effects	-	-	-	-	Ind	Ind	Ind	Ind	Cty	Cty	Cty	Cty
Observations	531	531	531	531	531	531	531	531	531	531	531	531
R-squared	0.090	0.047	0.177	0.106	0.126	0.093	0.373	0.198	0.659	0.491	0.319	0.265
B. All industries												
Dependent variable:	Δ MGT	Δ R&D	Δ FAB	Δ MAR	Δ MGT	Δ R&D	Δ FAB	Δ MAR	Δ MGT	Δ R&D	Δ FAB	Δ MAR
Δ forward intensity	-0.147*** (0.040)	0.014 (0.034)	-0.296*** (0.088)	-0.106** (0.042)	-0.156*** (0.048)	0.015 (0.038)	-0.134*** (0.043)	-0.116* (0.061)	-0.075*** (0.014)	0.001 (0.028)	-0.386*** (0.127)	-0.108*** (0.028)
Δ backward intensity	-0.045 (0.080)	-0.014 (0.059)	-0.212*** (0.059)	-0.052 (0.106)	-0.127 (0.076)	-0.061 (0.049)	-0.116* (0.059)	-0.133 (0.091)	0.022 (0.055)	0.015 (0.053)	-0.305*** (0.099)	-0.036 (0.073)
Δ exports of final goods intensity	-0.001 (0.009)	0.013 (0.010)	-0.006 (0.017)	0.017 (0.013)	0.002 (0.007)	0.007 (0.006)	0.001 (0.008)	0.012 (0.007)	0.001 (0.006)	0.017* (0.010)	-0.005 (0.019)	0.018 (0.014)
Δ imports of final goods intensity	0.001 (0.001)	0.002 (0.002)	0.001 (0.004)	0.002 (0.004)	0.001 (0.001)	0.002 (0.002)	0.002 (0.004)	0.000 (0.002)	-0.002 (0.001)	0.003** (0.001)	0.001 (0.003)	0.003 (0.003)
Δ log relative price of investment	-0.027 (0.026)	-0.017 (0.015)	0.016 (0.010)	0.028* (0.016)	-0.021 (0.025)	-0.014 (0.015)	0.006 (0.012)	0.040** (0.018)				
Fixed effects	-	-	-	-	Ind	Ind	Ind	Ind	Cty	Cty	Cty	Cty
Observations	1,118	1,118	1,118	1,118	1,118	1,118	1,118	1,118	1,118	1,118	1,118	1,118
R-squared	0.031	0.023	0.090	0.016	0.064	0.134	0.302	0.115	0.257	0.188	0.134	0.109

Notes. The dependent variables are changes in labor income shares in value added of four categories of "functional specialization": management (MGT), R&D, fabrication (FAB), and marketing (MKT). Overall labor shares are the sum of shares over these four categories. Changes for all variables are computed for 2001-2007. The first two lines in each panel report coefficients from regressions where foreign and domestic upstreamness are included. The third line reports the coefficient to Δ forward, the change in intermediate inputs export and GVC intensity, from regressions where the other explanatory variables are the change in offshoring of intermediate inputs intensity in total input purchases (Δ backward), the change in export intensity of final goods in value added (Δ exports of final goods), the change in import intensity of final goods in total absorption (Δ imports of final goods), and the country-level log change in investment prices (Δ log relative price of investment, normalized: mean=0, S.D.=1). Regressions estimated by weighted least squares with value added in 2001 as weights. Robust standard errors in parentheses are computed by two-way clustering by country and by industry. *** p<0.01, ** p<0.05, * p<0.1.

Table A16. Functional Specialization and Forward Foreign GVC Integration, 2001-2007---stacked regression over functions

		Dependent variable: Δ Function					
		Manufacturing			All industries		
		(1)	(2)	(3)	(4)	(5)	(6)
Δ forward intensity	x MGT	-0.167*** (0.049)	-0.122* (0.058)	-0.128** (0.050)	-0.145*** (0.038)	-0.100** (0.049)	-0.145*** (0.033)
	x R&D	-0.019 (0.049)	0.026 (0.068)	0.020 (0.026)	0.014 (0.037)	0.060 (0.047)	0.015 (0.030)
	x FAB	-0.398*** (0.102)	-0.353*** (0.085)	-0.359** (0.121)	-0.281*** (0.089)	-0.236*** (0.079)	-0.280*** (0.101)
	x MAR	-0.141*** (0.042)	-0.096** (0.044)	-0.102* (0.055)	-0.134*** (0.045)	-0.089 (0.054)	-0.133*** (0.042)
p-value for Ho: FAB =	MGT	0.04	0.04	0.03	0.20	0.20	0.20
	R&D	0.01	0.01	0.01	0.01	0.01	0.01
	MAR	0.02	0.02	0.02	0.18	0.18	0.18
Δ upstreamness, foreign	x MGT	-0.043*** (0.014)	-0.028 (0.018)	-0.043*** (0.011)	-0.058*** (0.016)	-0.040* (0.022)	-0.054*** (0.014)
	x R&D	-0.009 (0.015)	0.005 (0.019)	-0.010 (0.009)	0.008 (0.015)	0.026 (0.018)	0.012 (0.015)
	x FAB	-0.140*** (0.038)	-0.126*** (0.038)	-0.141*** (0.039)	-0.095*** (0.029)	-0.077** (0.030)	-0.091** (0.035)
	x MAR	-0.035** (0.013)	-0.020 (0.012)	-0.035** (0.015)	-0.062** (0.023)	-0.045* (0.026)	-0.059** (0.022)
p-value for Ho: FAB =	MGT	0.04	0.04	0.04	0.35	0.35	0.35
	R&D	0.01	0.01	0.01	0.01	0.01	0.01
	MAR	0.03	0.03	0.03	0.45	0.45	0.45
Fixed effects		Func	Func, Cty	Func, Ind	Func	Func, Cty	Func, Ind
Observations		2,124	2,124	2,124	5,060	5,060	5,060

Notes. The dependent variable is changes in labor income shares in value added of four categories of "functional specialization": management (MGT), R&D, fabrication (FAB), and marketing (MKT). Changes for all variables are computed for 2001-2007. In the top panel the coefficient to Δ forward, the change in intermediate inputs export and GVC intensity, is allowed to change depending on the function on the left hand side. Other explanatory variables in the top panel (not reported here) are the change in offshoring of intermediate inputs intensity in total input purchases (Δ backward), the change in export intensity of final goods in value added (Δ exports of final goods), the change in import intensity of final goods in total absorption (Δ imports of final goods), and the country-level log change in investment prices (Δ log relative price of investment)---all of which are allowed to change depending on the function on the left hand side. In the lower panel Δ upstreamness, foreign, the change in the foreign component of upstreamness, Δ upstreamness, domestic, the domestic component of upstreamness, and Δ downstreamness (the latter two not reported here)---are all allowed to have different coefficients depending on the function on the left hand side. The p-values are for (two-sided) pairwise tests of equality of the coefficient to Δ forward x FAB or to Δ upstreamness, foreign x FAB with the other, respective, functions. Fixed effects are not interacted with indicators for business functions. Regressions estimated by weighted least squares with value added in 2001 as weights. Robust standard errors in parentheses are computed by two-way clustering by country and by industry. *** p<0.01, ** p<0.05, * p<0.1.