Sources of productivity growth in Eastern Europe and Russia after transition

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Abstract

Many East European economies had similar stylized facts of development in the planned economy period, which are extensive growth, distorted economic structure and technological backwardness. At the same time, in two decades after transition country specific origins of growth, such as natural resources abundance or development of institutions, could dominate. To what extent sources of labour productivity growth in former planned economies after transition similar or country specific?

Using World KLEMS data, this paper compares Central and East European economies – early reformers (CEEs) and a late reformer, Russia, in 1995-2007. We show that (i) industrial origins of productivity growth in CEEs and Russia are rather similar and driven by manufacturing industries, which are remote from the technology frontier; (ii) the structural bonus did not materialize; and (iii) MFP growth rates variation both across industries within a country and across countries is explained by variations in the distance to the technology frontier.

JEL: O47; L16, P27.

Key words: transitional economies, productivity, industrial growth accounting, level accounting, structural change

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1. Introduction

Heterogeneity of income per capita across countries is one of the central issues in modern economics of growth (Acemoglu 2009). In the past Socialist economies fell behind the West in levels of income per capita (Bergson 1987). By early 1990-s, when the command-economy experiment was over, perspectives for growth in these economies seemed optimistic. Indeed, they had already passed the first phase of structural change, industrialization, and their labour force was educated and relatively healthy. The level of technological sophistication was high in some sectors of the economy, notably the advanced defense equipment industries with potential for diffusion to other industries (Campos and Coricelli 2002). With the opening for international trade and investments an inflow of state-of-the-art technologies from developed economies was to be expected, and given the seemingly high level of absorptive capacity, these could be quickly put to efficient use. In addition, there was a high potential for structural change to contribute to growth. Given the large differences in efficiency levels across and within industries, a reallocation of labour and capital from less to more productivity activities would boost aggregate growth.

[Table 1 is about here]

These expectations did not materialize. Surprisingly, after two decades of transition in Central-East European economies (CEE) and Russia (CEER) the gap in GDP per capita with Western Europe and the US still exists. Table 1 summarizes GDP per capita and labour productivity levels of nine post-transition economies in 2007, just before the global financial crisis, relative to Germany - the European productivity leader. As can be seen in column 1, even the GDP per capita of the most advanced economy, Slovenia, is below 70% of the level of Germany. Levels of other post-transition economies hover around 40-60% of the German level. Is this due to structural characteristics of the labour force, such as participation rates, unemployment or hours worked, or due to differences in labour productivity, which can be

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2 The paper considers Russia and eight East-European economies: the Czech Republic, Hungary, Slovenia, Estonia, Latvia, Lithuania, Poland and Slovakia. However, because of data availability more attention will be paid to the first four.

3 According to Sonin (2013), this point was made by Daniel Treisman, who noted that convergence of post-transition economies took much longer, than expected, so the level of developed economies has not been achieved yet.
considered as a broad measure of the technology level in the economy? Table 1 reports that the gap in labour productivity, which is defined as real value added per hour worked, explains the major part of the difference in GDP per capita levels. Cross-country variation in the amount of hours worked per worker or the labour force participation rates (the involvement of population in working age in production) does not explain much of the gap. On the contrary, working hours are longer in most countries, and participation rates higher. For example, the level of GDP per capita in Slovenia is 68.6 per cent, whereas the labour productivity level equals 57.3 per cent. The difference of 11.3 percentage points is explained by the effect of working hours (a higher number of hours worked by an average Slovenian worker in comparison with a German one); while a lower employment-population ratio reduces the gap by 3.9 percentage points. As a result, the levels of labour productivity of post-transition economies (column 4) are even lower that corresponding levels of GDP per capita (column 1) in all countries, except Poland. A similar picture emerges when one focuses on that part of the economy that can be well measured, namely the market economy\(^4\), see column (5). I conclude that it is labour productivity backwardness, which causes the gap in incomes between post-transition and developed economies.

In this context, a general question can be raised, to what extent the present backwardness of Eastern Europe and Russia in terms of labour productivity levels is related to shared features of a past of economic planning that resulted in a long period of extensive growth based on heavy investment\(^5\), but with little technological change\(^6\) and distorted economic structure, tilted towards heavy industry sectors\(^7\). At the same time, it is clear, that in past two decades different developments paths have been followed and country-specific features have become more important in driving development. These include differences in the integration into European markets, including the importance of foreign direct investments (Campos and Kinoshita 2002), and in “the great divide” of institutions (Berglof and Bolton 2002). Also differences in resource abundance played an increasing role. From this perspective, the fact that, say, Russia and the Czech Republic were command economies in the past, should nowadays be less relevant for understanding the difference in their current performance than the facts that Russia has abundant natural resources and the Czech Republic is closer to European markets, a member of the European Union and deeply integrated into European production chains. These idiosyncratic factors also influenced the speed of transition and restructuring (Ahrend, de Rosa, and Tompson 2006) and it is useful to split the former planned economies into two groups: early reformers, which include the Central-East European

\(^4\) Market economy aggregates correspond to the total economy ones except public administration, education, real estate activities, and health and social work, as discussed in (Timmer and Voskoboynikov 2013).
\(^5\) (Ofer 1987; Krugman 1994)
\(^6\) (Gomulka and Nove 1984; Holliday 1984)
\(^7\) (Bergson 1964)
countries, and late reformers, which are mostly the former republics of the Soviet Union, including Russia (Roland 2000, Ch. 1).

This paper analyses developments in post-transition economies, and provides a comparison of the drivers of productivity growth in early and late reformers from the industry level perspective. Although economic growth processes in early reformers (CEEs) have already been discussed in the literature, little is known, to what extent these findings are applicable to late reformers. Russia is by far the largest, and arguably the most interesting country among late reformers. Accordingly, the novelty of the present study is adding Russia into the comparative analysis across post-transition economies on the basis of a detailed recently developed dataset (Voskoboynikov 2012) and using a productivity level comparison method to measure technology gaps, using a newly developed set of relative output prices (PPPs).

We analyze the period up to 2007 which is just before the global financial crisis of 2008, in order to focus on the long-run trend after transition. We emphasize three particular issues. First, by means of a detailed industry growth accounting study we show that industrial origins of productivity growth in CEEs and Russia are rather similar. It is driven mainly by industries in the manufacturing sector that appear to be far away from the technology frontier. Second, we analyze the contribution of labour reallocation between industries to aggregate growth. This was expected to be substantial because of elimination of multiple planned economy distortions in the industrial structure. We find that the structural bonus did not materialize in CEEs, and only to a small extent in Russia. The reason for this is that the initial variation of productivity levels across industries in first years of transition was relatively low such that the scope for reallocation gains was only small. Third, we analyze patterns in multi-factor productivity growth since 1995 to see, what industries are driving productivity, and why there is a difference in the speed both across industries within a country as well as across countries. To do so, we relate MFP growth to the initial technological gap, measured by the level of MFP in 1995. We find that the initial technological gap was deep, such that it would take more time to catch up than had been expected before transition. We also find beta-convergence in all aggregated sectors of CEEs to Germany, the benchmark advanced economy, and demonstrate that the speed of convergence is higher in manufacturing than in market services.

The rest of the paper is divided into seven sections, including this introduction. Section 2 presents data sources. Section 3 reviews the approach, which includes industry growth accounting and level or developing accounting, which is used for the measurement of the distance to the technology frontier. Section 4 deals with the labour reallocation effect in

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8. At the same time, micro-level comparisons of productivity sources for manufacturing firms were carried out in (Brown, Earle, and Telegdy 2006; Brown and Earle 2008).

comparison with other sources of labour productivity growth (capital deepening and MFP growth). Section 5 focuses in depth on MFP growth, breaking it down by industries and highlighting the role of manufacturing industries. The sixth section provides novel measures of the distance of industry technologies in CEER to the technology level in advanced countries, represented by Germany. Combining these technology gap measures with MFP growth rates, it uncovers a strong convergence process of CEERs. The last, seventh section summarizes results and concludes the paper.

2. Data

A decade ago, reviewing the literature on economic performance in transition Campos and Coricelli (2002) pointed out to data quality as a limitation in understanding growth in transition. Although there has been no remarkable progress with data in the early transition period yet, for years after 1995 new datasets of macro time series for some economies in transition have become available. These are EU KLEMS (Timmer and others 2010)\(^{10}\) and Russia KLEMS\(^{11}\) (Timmer and Voskoboynikov 2013). Both datasets are consistent and constructed to study the relationship between skills formations, investments, technical change and growth. These databases form the dataset of the present study.

The dataset is based on the consistent accounting framework, which is rooted in the neoclassical production function theory and provides an opportunity to decompose output growth rates into contributions of inputs and multifactor productivity\(^{12}\). Although such assumptions of the neoclassical approach as perfect competition and constant returns to scale, which are essential for multifactor productivity (MFP) measurement, might not hold for certain countries, the neoclassical framework is standard for international comparisons. Growth accounting is implemented not only for developed economies, but also for developing economies, command economies and economies in transition\(^{13}\). This framework could also be considered as the benchmark for alternative calculations invoking different assumptions such as non-constant economies of scale and mark-ups (Barro 1999; Basu and Fernald 2001).

The dataset includes measures of output, inputs and derived variables like MFP. EU/Russia KLEMS dataset covers all EU members, including eight former Socialist economies, the United States, Japan and Russia. The number of industries and years covered varies across countries. It is based on data of the System of National Accounts and other macro indicators, provided by national statistical offices. Time series of economies in transition cover the period from 1995 to

\(^{10}\) Release of November 2009  
\(^{11}\) Russia KLEMS data is available on the World KLEMS website [http://www.worldklems.net/data.htm](http://www.worldklems.net/data.htm). 
\(^{12}\) (Solow 1956; Solow 1957; Jorgenson and Griliches 1967; Jorgenson, Gollop, and Fraumeni 1987; Jorgenson, Ho, and Stiroh 2005) 
\(^{13}\) See, e.g., (Kaplan 1969; Ofer 1987; Krugman 1994).
2007\textsuperscript{14} for industries at the two-digit level of classification NACE 1.0 (see the list in Appendix A1). The number of indicators also varies across countries. While data on labour and output is available for all countries in the dataset, only most advanced economies and four post-transition economies are presented with the series of capital services and multifactor productivity. These economies are the Czech Republic, Hungary, Slovenia and Russia.

The EU KLEMS database has been constructed for many countries with a broad set of input measures, which are capital (K), labour (L), energy (E), materials (M) and services (S). This set is sufficient for construction of gross output-based growth accounting. However, only a more restrictive value added-based growth accounting for the Russian economy is possible. In case of Russia main limitation is the absence of time series of supply and use table of an adequate quality. Another important restriction of the Russian segment of the database is the absence of labour quality indicators. For this reason an increase of labour services because of a higher share of educated workers in total amount of hours worked will lead to the overestimation of multifactor productivity growth\textsuperscript{15}.

Capital input in the EU/Russia KLEMS database is based on the concept of capital services, suggested by Jorgenson (1963). The flow of capital services of a particular type of assets is proportional to its stock. Growth rates of each type of capital are aggregated to the total growth rates of capital with weights, which are calculated on the basis of rental prices of capital. These prices depend on depreciations rates being higher for assets with shorter service life. That is why the contribution in the total flow of capital services of a computer for 1000 dollars per year with service life of five years is higher than a railway sleeper of the same purchasing price with service life 25 years. In case of stock-based capital measures both the computer and the sleeper would generate the same amount of capital input per year. In EU/Russia KLEMS database eight types of assets are taken into account, which are split into two groups. The group of ICT assets includes computing equipment, communication equipment and software, whereas the group of Non-ICT assets is composed from residential structures, non-residential structures, machinery and equipment, transport, and other assets.

Implementation of the concept of capital services mitigates the problem of “communist capital”, which is specific for economies in transition in general and for Russia in particular. The term “communist capital” was suggested by Campos and Coricelli (2002). The communist capital is the set of investment goods, which was put into operation before transition and has been remaining idle after transition. At the same time, this stock is presented in capital stock statistics. Taking into account substantial overinvestment into production of non-competitive

\textsuperscript{14} Data for Poland and Slovenia covers years till 2006.

\textsuperscript{15} The quantitative estimations of the influence of these deviations on growth accounting was discussed by Timmer and others in (2010, ch. 3).
goods before transition and multiple difficulties with measurement of capital after transition\textsuperscript{16}, its share of capital stock could be huge. In case of capital services the problem of overestimated capital stocks is relevant for values of net stocks by the end of 1995 and influences growth rates of capital stocks only for the period of service life of a particular asset. Consequently, for the whole group of ICT capital with service live around three years this bias is negligible. For machinery and equipment, transport and other assets with maximum service life 12 years the influence of the communist capital bias disappears by 2000s. The problem could be serious for buildings and constructions with longer service lives, but the share of idle capital among these assets seems to be lower, because buildings and infrastructure, put into operation before transition, abandoned less frequently than old machinery\textsuperscript{17}.

Another problem of capital estimation, essential for post-transition economies, is remarkable variations in capital utilization. During the post-transition period economies experienced huge changes in capacity utilization. For example, in Russian manufacturing it varied from 87.7 per cent in 1990 to 45.5 per cent in 1998 (Bessonov 2004). Hulten (1986) pointed out to theoretical inconsistencies when using of such measures of capital utilization within the neoclassical growth accounting framework. A standard theory-based solution, which is used in the present paper, is the implementation of ex-post rates of return in calculations of capital services. This way changes in capital utilization are taken into account as they will be reflected in the fall of rental prices of capital instead of physical measures of capital stock. However, there are some limitations in this approach. First, as Schreyer (2004) identified the problems with capital measurement when the set of assets is not full. Second, Balk (2009) showed that such assumptions of the ex-post approach, as a perfect foresight, constant returns to scale and perfectly competitive markets, which mostly do not hold in reality.

Comparing different measures of capital services on data for the US economy, Inklaar (2010) opted for the external rate of return-based approach. However, in case of the US economy in almost thirty years (1977-2005) variations in capacity utilization could be not as important as for post-transition economies in a shorter period. For growth accounting exercise of the present paper we use the internal rate of return approach\textsuperscript{18}.

Finally, following Timmer and others (2010, ch. 3), a number of caveats with possible interpretations of MFP on the basis of the EU/Russia KLEMS dataset need to be noted regarding to the present study. The standard interpretation of MFP growth is an increase of the technology level. However, MFP growth can also reflect reorganization of firms, unmeasured inputs such as R&D, and multiple deviations from the neoclassical assumptions. Industrial MFP...

\textsuperscript{16} See the literature review about capital measurement problem in economies in transition of Timmer and Voskoboynikov (2013).

\textsuperscript{17} We do not consider non-market services.

\textsuperscript{18} Comparison of growth accounting decompositions in cases of ex-post and ex-ante rates of return in case of the Russian economy is provided by Timmer and Voskoboynikov (2013).
captures also the reallocation effects between firms of the same industry and various measurement errors like biases in ICT-investment deflators.

3. Methods

The growth accounting methodology allows a breakdown of output growth rates into a weighted average of growth in various inputs and productivity change. It is based on the neoclassical framework of Solow (1956; 1957), Jorgenson and Griliches (1967), Jorgenson, Gollop and Fraumeni (1987) and Jorgenson, Ho and Stiroh (2005).

Under the assumptions of competitive markets of inputs, full input utilization and constant returns to scale the multifactor productivity of industry \( j \) \( (A_j) \) is defined as a function of real value added \( (Z_j) \), capital services \( (K_j) \) and hours worked \( (H_j) \)

\[
\Delta \ln A_j \equiv \Delta \ln Z_j - \bar{v}_{K,j} \Delta \ln K_j - \bar{v}_{L,j} \Delta \ln H_j,
\]

where \( \bar{v}_{ij} \) is the period-average share of the input in the nominal value added. The value shares of capital and labour are defined as \( v_{K,j} = \frac{p^j K_j}{p^j Z_j} \) \( v_{L,j} = \frac{p^j H_j}{p^j Z_j} \).

As shown by Stiroh (2002), labour productivity growth \( (z = Z/H) \) can be represented as a function of capital intensity \( (k = K/H) \) and the reallocation of hours \( (R) \):

\[
\Delta \ln z = \sum_j \bar{v}_{Z,j}^{GDP} \cdot \Delta \ln z_j + \left( \sum_j \bar{v}_{Z,j}^{GDP} \cdot \Delta \ln H_j - \Delta \ln H_j \right) = \sum_j \bar{v}_{Z,j}^{GDP} \cdot \Delta \ln z_j + R.
\]

The reallocation term \( R \) captures changes in labour productivity growth because of the difference of the share of an industry in value added and in hours worked. It is positive if industries with an above average share of value added show positive growth of hours worked or industries with the below average share of value added have declining hours worked.

Substitution of (2) into (1) for industries gives us the final decomposition of aggregate labour productivity growth rates:

\[
\Delta \ln z = \sum_j \bar{v}_{Z,j}^{GDP} \left( \sum_i \bar{v}_{j,i}^K \Delta \ln k_{j,i} + \Delta \ln A_j \right) + R.
\]
which represents contributions of capital intensities of different types of assets, MFP and labour reallocation to aggregate labour productivity growth rates. From (3) one can derive the contribution of capital intensity as well as the contribution of MFP of industry \( j \) as \( \tilde{v}^Z_{j} \Delta \ln A_j \).

An important limitation of this value-added based approach is that it treats capital and labour in comparison with intermediate inputs asymmetrically. Consequently, the substitution of some labour and capital services with intermediate inputs, which takes place in case of outsourcing, in this set up will be treated as change in multifactor productivity. For a better alternative, the gross-output based version of growth accounting, we need time series of supply and use tables, which are not available for Russia.

The technology gap models such as (Acemoglu, Aghion, and Zilibotti 2006) explain differences in productivity growth with differences in distances to the technological frontier. We compare levels of MFP in industries across countries for the estimation of the distance to the technological frontier. This approach is called “level accounting”. In contrast with growth accounting, in which growth rates of a single country are compared in two different moments in time, level accounting compares two countries at the same moment (Caves, Christensen, and Diewert 1982a). For the first time the bilateral production model was applied by Jorgenson and Nishimizu (1978) for the value-added based cross-country comparison. The approach of the present paper is based on (Timmer and others 2010, Ch. 6).

Taking into account the necessary assumption for producer equilibrium in each country, the corresponding index of MFP for a country \( c \) relative to the level of Germany (GER) is defined as the function of differences in value added \( Z_c \), capital services \( K_c \) and labour services \( L_c \) or, alternatively, labour productivity \( Z_c = \frac{Z_c}{H_c} \) and capital intensity \( k_c = \frac{K_c}{H_c} \):

\[
\ln A^Z_c = \ln \frac{Z_c}{Z_{GER}} - \tilde{w}^Z_K \cdot \ln \frac{K_c}{K_{GER}} - \tilde{w}^Z_L \cdot \ln \frac{H_c}{H_{GER}} = \ln \frac{Z_c}{Z_{GER}} - \tilde{w}^Z_K \cdot \ln \frac{K_c}{K_{GER}}
\]

(4)

where weights \( \tilde{w}^Z \) are shares of each factor in nominal value added averaged over the two countries. We assume constant returns to scale, so the sum of the shares of labour and capital compensation in value added are equal to one.

Equation (4) shows that comparable volume measures of value added, labour and capital are necessary. In case of a single output or input, which can be measured in physical units such as cars or hours worked, the issue is trivial. However, dealing with the measures at the aggregate level corresponding quantity indices should be calculated implicitly with relevant
price deflators. Levels of output and inputs between countries can be matched by means of relative prices to express these values in the same currency. The use of exchange rates for this will probably lead to substantial errors. There are well-known reasons why exchange rates and relative prices differ (Inklaar and Timmer 2008). For non-tradable goods there are no reasons why exchange rates and relative prices should be equal. For tradable goods the two economies or corresponding sectors could have different degrees of monopolistic power. Another reason is the existence of time lags in response of prices to exchange rate movements. At the same time, the movements could be influenced by short-term international capital reallocations, originated from monetary policy, speculations or fluctuations of world prices, and not related with changes in output or productivity.

Necessary comparable volume measures of output and inputs can be obtained with purchasing power parity (PPP) which indicates the price of the volume measure relative to another country. Such value added volume index for country $c$ is defined as

\[
Z_c = \frac{V^Z_c}{PPP^Z_c},
\]

where $V^Z_c$ is nominal value added in country $c$ at national prices and $PPP^Z_c$ the price of value added in country $c$ relative to that in USA. Finally, for capital services in country $c$ we have:

\[
K_c = \frac{V^K_c}{PPP^K_c},
\]

with $V^K_c$ the nominal value of capital compensation in country $c$ and $PPP^K_c$ the relative price of capital services in country $c$ as

\[
PPP_{k,c}^{K,US} = PPP_{k,c}^{I,US} \frac{(p^K_{k,c}/p^I_{k,c})}{(p^K_{k,US}/p^I_{k,US})},
\]
where \( PPP_{k,c}^{US} \) are PPPs for investment goods for types of assets \( k \) relative to US, \( p_{k,c}^l / p_{k,c}^l \) – the ratio of price deflators on capital services and investment goods of assets of type \( k \). This ratio equals

\[
\frac{p_{k,c}^K}{p_{k,c}^l} = r_{c,t} + \delta_k - \frac{p_{k,c,t}^l - p_{k,c,t-1}^l}{p_{k,c,t}^l},
\]

where \( \delta \) is a corresponding depreciation rate and \( r \) is the nominal rate of return\(^{19} \).

In is important to note that \( PPP_c^K \) in (6-8) depends on the method we use for calculations of the rate of return in capital services. Theoretically consistent approach is the implementation of ex-post rates of return. However, Inklaar and Timmer (2008) noted that for cross-countries comparisons its drawbacks could be more substantial than for inter-temporal comparisons within a country. Indeed, deviations from the neoclassical assumptions of perfect markets and constant return to scale change slowly within a country in time, while they may be substantial in cross-countries comparisons. That is why we use ten-years government bond yields from the IMF statics as ex-ante rates of return\(^{20} \), following the approach of (Inklaar and Timmer 2008).

Data on PPP for value added in industries is available in (Inklaar and Timmer 2012), whereas PPPs for investments by type of assets \( (PPP_{k,c}^{US}) \) have been calculated on the basis of detailed Eurostat/OECD PPP data for 2005\(^{21} \).

For extensions of MFP levels from the benchmark year 2005 till 2007 and back to 1995 we use productivity volume growth rates and apply them to the levels in 2005.

### 4. Labour reallocation

The transition from plan to market economy is a process of reallocation of resources on the basis of market incentives (Campos and Coricelli 2002). Many studies have attempted to

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\(^{19} \) \( PPP_c^K \) is calculated in (7) relative to the US price level. However, according to (Caves, Christensen, and Diewert 1982b), to make data comparable across countries, one more step is necessary. \[ \ln PPP_c^K = \sum_k \bar{\sigma}_k^K \left( \ln PPP_c^{US} - \ln PPP_c^K \right), \] where \( \ln PPP_c^K = \frac{1}{N} \sum_{c=1}^{N} \ln PPP_c^{US} \) and \( N \) is the number of countries in the sample. In this exercise for level comparisons of MFP along with CEERs, Germany and US we used EU KLEMS data for all countries, for which growth accounting is possible and the PPPs are available. These countries are Austria, Denmark, Spain, Finland, Italy, the Netherlands, the United Kingdom and France.

\(^{20} \) The only exception is the Russian economy, for which interest rates of government bonds lead to negative rental prices of capital in many cases. In case of the Russian economy we implement the fixed exogenous real interest rate 4 per cent a year, as recommended by the OECD Productivity Manual (Schreyer 2001, p. 133).

\(^{21} \) Detailed PPPs for investments by type of asset were kindly provided by Robert Inklaar.
investigate this structural change process in a three-sectoral framework (agriculture, industry and services) following the tradition of Kaldor, Maddison, Kuznets and Chenery. For example, Döhrn and Heilemann (1996) used the Chenery hypothesis, which links the sectoral structure of an economy with its stage of development, size and the endowment with natural resources. They found that by 1988 manufacturing was oversized in many Socialist countries, while services were small and underdeveloped. The authors projected a substantial shift of economic activity from manufacturing to services after transition. This shift occurred as was later documented and discussed in the literature\(^{22}\).

However, the more recent literature\(^ {23}\) on issues of structural change, productivity and growth in comparative perspective argues for the need for more detailed analysis. Indeed, the three-sectoral framework may be misleading in light of new stylized facts of development, suggested by Jorgenson and Timmer (2011). They showed that reallocations between agriculture and manufacturing in the post-industrialized world are marginal in comparison with reallocations within services, which now account for around 70 per cent of value added and hours worked in developed economies. Timmer and De Vries (2009) found a similar pattern for Latin-America since the 1980s. That is why it is important to consider performance of different groups of sub-industries within Services. In addition, the estimation of the role of reallocation on productivity is more accurate when more detailed industrial level is considered. De Vries and others (2012, tab. 11) have shown that the estimation of the reallocation effect is sensitive to the level of disaggregation. Finally, the use of more detailed data in analysis of structural change provides an opportunity of a more flexible industry grouping depending on country-specific issues. For example, to deal with the transfer pricing and vertical integration of large mining firms in Russia it is necessary to group mining and fuel from manufacturing and Wholesale trade from services into one sector extended mining\(^ {24}\).

[Fig. 1. is about here]

Fig. 1 represents decomposition (2) of labour productivity growth in eight CEERs and Germany. As can be seen, the role of labour reallocation is relatively modest, explaining at best half of a percentage point of growth. The notable exception is Russia where the contribution of labour reallocation is 1.1 out of 4.6 percentage points, or about a quarter of total growth. In

\(^{22}\)See, e.g. (Raiser, Schaffer, and Schuchhardt 2004; Lazarev and Gregory 2007; World Bank 2008; Bah and Brada 2009)

\(^{23}\)See, e.g., (van Ark, O'Mahony, and Timmer 2008; Lin 2010; McMillan and Rodrik 2011).

\(^{24}\)See detailed discussion of this in (Timmer and Voskoboynikov 2013).
this context there is a concern that a substantial labour reallocation during the years of transition, which was expected to eliminate huge planned economy distortions, did not provide a sizable increase in an aggregate labour productivity level. Explaining this, it makes sense to compare post-transition economies with developing economies, in which the reallocation of labour from sectors with a low productivity level to high productive ones, which is referred to as a structural bonus, provides a more substantial contribution.

[Fig. 2 is about here]

Post-transition countries differ in two aspects. The first is a different direction of labour flows. In developing economies labour reallocates from low productive agriculture to high productive manufacturing. This is clearly not the case for CEERs, which passed industrialization no later than in the second half of 20th century. Instead, structural transformation in CEERs is more about labour reallocation from manufacturing to services, confirming the prediction of Döhrn and Heilemann (1996). This is illustrated in fig. 2, which represents shares of hours worked in 1995 and 2007 in the Czech Republic, Hungary, Russia and Slovenia with Germany. In the remainder of the paper we will focus on these 5 countries as the necessary data is not available for the other CEEs. We distinguish four sectors, which are high skill intensive goods (HSI goods), low skill intensive goods (LSI goods), high skill intensive services (HSI services) and low skill intensive services (LSI services)\(^\text{25}\). Indeed, both the low skills intensive and the high skills intensive services in CEERs are expanding, while manufacturing in most cases is contracting not only in CEERs, but also in Germany, leaving the market for China. At the same time, advanced manufacturing in the Czech Republic, Hungary and Slovenia (HSI Goods) gains. Havlik, Leitner and Steher (2012) explained this by pointing at factors like the educated labour force, closeness to the Western Europe, outsourcing, and substantial FDI inflows, in particular from Germany and Austria (Marin 2009)\(^\text{26}\). Another remarkable point is differences in structural change in services between CEEs and Russia. By analogy with the developed economies trend, high skills intensive services (financial intermediation and business services) expanded more in CEEs, while in Russia this effect is almost absent. As for the sector of low skill intensive services,

\(^{25}\)See detailed sectoral composition in Appendix A1.

\(^{26}\)With firm-level data of German and Austrian multinationals, Marin (2009) showed that this outsourcing was caused by the abundance of high skill labour in East European economies, including Russia and Ukraine, in comparison with Germany and Austria. Interestingly, according to this paper, comparative advantages in high skill labour are applicable to all East European economies, but the lion’s share of FDI were hosted by CEEs, rather than Russia, Ukraine and other CIS countries. This can be interpreted as the effect of the EU membership.
which includes such industries as retail and construction, extension is the largest in Russia, being smaller in CEEs and negative in Germany. This could reflect the limited development of this type of services in initial years of transition in Russia.

[Fig. 3 is about here]

The second distinction of post-transition economies compared to other countries is a low variation of productivity levels across industries. As McMillan and Rodrik (2011, fig. 2) showed, there is a negative correlation between the level of inter-industry productivity variation and average labour productivity of a country. For this, they used the variation coefficient of (the logarithm of) industrial labour productivity as a measure of inter-industry productivity variation. Labour productivity here is defined as the GDP PPP converted value added to the number of hours worked. We extend their analysis and perform it for data on thirty nine countries in Figure 3. At the vertical axis is the coefficient of variation of labour productivity levels across 30 industries in a particular country. On the horizontal axis is the (log) level of aggregate labour productivity in the country relative to Germany. Data for forty developed, developing and post-transition economies are plotted and a dotted OLS regression line is given as well. As can be seen in Fig. 3, the observations for the post-transition economies fit the regression line\(^{27}\). It means that in 1995 the level of cross-sectoral variation of productivity in CEERs is much closer to developed, than developing economies, and the potential for the structural change bonus was only limited.

Only the Russian case is exceptional as the contribution of labour reallocation in the Russian economy is comparable with India and higher than in China\(^{28}\). Half of this is explained by the expansion of the extended mining sector, a unique Russian feature\(^{29}\). The other half could be explained by an earlier stage of transition in Russia in comparison with CEEs in 1995.

\(^{27}\)Appendix A2 presents the results of the regression. Negative slope coefficient in Fig. 3 is highly significant and the deviation of the slope coefficient from the trend for CEERs (CEER x the log of the aggregate labour productivity level) is also significant.

\(^{28}\)Of 7.6 % of yearly average labour productivity growth of market economy in China, labour reallocation contributed 0.4%. For India corresponding values are 5.3 and 0.9. These results are consistent with findings of de Vries and others (2012), obtained with a different approach.

\(^{29}\)To show this, I represent the reallocation effect as a sum of industrial components:

\[ R = \sum_j \hat{v}_{Z,j}^{GDP} \cdot \Delta \ln \frac{H_j}{H}. \]

Each sectoral component is, in turn, the product of the change of labour share of this sector \(\Delta \ln \frac{H_j}{H}\) and the average share of its value added \(\hat{v}_{Z,j}^{GDP}\). Grouping industries by sectors I decompose total reallocation \(R\) into contributions of each sector. Of total 1.1 per cent of reallocation the contribution of Extended Mining is 0.62, which is the product of the change of labour share, 2.35 per cent, and the time average VA share 26.4 %.
Indeed, working with the manufacturing firms’ level data of early reformers (Hungary and Lithuania) and late reformers (Georgia, Romania, Russia, and Ukraine), Brown and Earle (2008) showed how the role of reallocation changed in time. Before transition the contribution of labour reallocation to productivity growth was low in all these economies. In first years of transition the contribution of labour reallocation was higher in early reformers, diminishing in the following years. In turn, late reformers demonstrated the peak of the labour reallocation effect with a delay. The authors explained this pattern with a more intensive cleansing of less productive firms in early reformers in first years of transition, which intensified labour reallocation.

A more accurate econometric analysis of the sample in this scatterplot unveils a specific feature of the post-transition economies. Although these economies demonstrate a negative dependence of sectoral productivity variation on the level of aggregate labour productivity, as can be seen in Fig. 3, the slope is flatter. In other words, in comparison with developing economies an increase of labour productivity in post-transition countries is accompanied with a smaller decline in sectoral productivity variation. This finding confirms the idea, broadly discussed in the literature (Döhrn and Heilemann 1996; Gregory and Stuart 2001, ch. 16), that overinvestments in the planned economy period distorted the economy in favor of manufacturing. Transition is accompanied by a negative labour reallocation effect as labour moves from a more productive, but artificially inflated manufacturing sector, to less productive services.

Given the minor role of labour reallocation in driving aggregate growth, we need to analyze in-depth the sources of productivity growth within industries, which is the topic of the next section.

5. Sectoral drivers of aggregate productivity growth

Using the growth accounting formula (3), we can analyze the sources of intra-industry productivity growth and compare it to the role of labour reallocation in driving aggregate growth. According to table 2, which summarizes all components of labour productivity growth in the four CEERs and Germany, MFP growth in sectors is an important source of aggregate labour productivity growth in all countries. Moreover, it is MFP which makes labour productivity growth in CEERs higher than in Germany. Indeed, the Czech Republic, which had the lowest average MFP growth rates among CEERs, still has double the growth rate in Germany, while MFP in the Russian economy exceeds Germany by a factor of more than 3.5.
One step further is identifying key sectors and industries in CEERs, which drive this MFP growth. Table 3 demonstrates sectoral MFP growth rates and contributions in CEERs in comparison with Germany. Growth rates of MFP in sectors are given in the upper panel, while contributions of sectors to aggregate MFP growth are given in the lower panel. The latter are derived as a sector’s MFP growth times its share in value added. As can be seen, the goods sector in the CEERs, including both high skills and low skills intensive goods, outperforms Germany. MFP growth rates of advanced manufacturing industries (HSI Goods) in CEERs demonstrate growth around 6 per cent per year against 3.5 per cent in Germany, while traditional manufacturing (LSI goods) shows 3 – 4 percent versus 1.8 percent in Germany. Further, in all economies, except Russia, it is the goods sector that contributes the lion’s share of multifactor productivity growth. The role of market services in MFP growth is substantial only in Russia with outstanding 9.4 percent growth of high skills intensive services. In Hungary MFP growth in low-skilled intensive services was particularly important.

To get a more detailed view of which sectors are leading, we zoom in at a more detailed industry level, considering 30 industries in total by means of Harberger diagrams for MFP growth rates. These are presented in fig. 4. These diagrams provide an explicit and intuitive way of identifying if aggregate growth rate is produced by a few industries (a mushroom pattern) or widespread (a yeasty pattern). The upper curve in each panel is for a CEER country. In order to compare with Germany, the German curve is included in each panel as well (the lower one in each of the four graphs). As can be seen, in all post transition economies MFP growth is mushroom-like and more concentrated than in Germany. The curvature in CEERs varies from 0.37 in Slovenia, which corresponds to the highest degree of concentration, to 0.61 in the Czech Republic, which is close to the level of Germany, 0.64. The figure also provides

---
30 See (Harberger 1998; Timmer and others 2010) for a more detailed description of the Harberger diagram technique.
evidence that MFP growth rates mostly originate from goods both in CEERs and in Germany (corresponding points are filled). Such industries are mostly on the upward sloping parts of the diagrams. Moreover, the diagrams show that many manufacturing industries in CEERs perform better than in Germany. This can be seen from the slope of the diagram curve to the left of the industry dot. For example, machinery (code 29) in Hungary, the second industry from the origin, grows much faster than in Germany, because the diagram in the vicinity of 29 for Hungary is much steeper, than for Germany. In total, there is a core of five manufacturing industries in CEERs, in which MFP growth exceeds the German. These are paper and publishing (21t22), rubber and plastic (25), non-metal minerals (26), machinery (29), electrical equipment (30t33) and other manufacturing (36t37).

6. Distance to the technology frontier and convergence

As it has been discussed in the previous section, aggregate MFP growth in CEERs is mostly driven by manufacturing subsectors. This raises the question what is driving this rapid growth. In a famous article, Gerschenkron (1962) argued that backward economies could catch up to the productivity level of advanced economies by adapting frontier technologies. He hypothesized that the larger the distance to the technology frontier the more options for improvements are available, driving higher productivity growth. If productivity growth in a country depends positively on its distance to the global technology frontier, a process of convergence will occur in which countries grow towards a similar level of productivity. More recently, Acemoglu, Aghion and Zilibotti (2006) provide a formal model of this idea, in addition arguing that optimal policies need to be adapted, depending on the distance to the frontier. They suggest that the institutional environment relevant for investments- and adaptation-based strategy is not optimal anymore once the economy is close to the technology frontier.

It is well known that CEERs were far remote from the technology frontier in the period before transition. Gregory and Stuart (2001, pp. 187-9) reviewed the literature and summarized that economic development of the Soviet Union took place at the background of a large technology gap, which either increased or, at best, remained constant during the Soviet period, while technology levels in the US and Western countries continued to push out the global frontier. Nevertheless, there were areas, such as metallurgy, aeronautics and military production, in which the Soviet technologies were close to the frontier, at least in the early period. Technological weakness of the Soviet economy in comparison with developed economies is well documented not only by Western scholars, but also by the official Soviet statistics. According to an official survey, by April 1, 1986, only 6.5% of the total amount of machine tools in the Soviet economy was considered as “confirmed to the up-to-date world
technology level”; 45% met the domestic technological requirements and 48.5% were obsolete and were entitled to updates, which never materialized (Rosstat 1989, p. 225). The literature also provides multiple evidences of technological inferiority of CEEs before transition (Gomulka and Nove 1984).

A natural approach to explain variations in productivity growth across industries in CEERs after transition is thus to relate them to the initial distances to the technology frontier at the beginning of transition. We will use comparative industry MFP levels to measure this distance, following (Timmer and others 2010). Such analyses are scarce at the industry level, because the estimation of MFP levels is a sophisticated and data demanding exercise. Timmer and others (2010), which covers OECD countries including three CEEs, is the only example of this approach in the literature built on the basis of a representative macro data. However, other more indirect evidence of technology convergence is also available in the literature. On the basis of high MFP growth rates Havlik and others (2012) showed that catching up in CEEs was faster in skill intensive manufacturing industries because of well-educated workforce, FDI inflows and stimulated outsourcing activities. Fernandes (2009) investigated labour productivity in the service sector of CEEs and confirmed the Gerschenkron hypothesis for productivity growth in services. So far, however, there are no studies of this type which put Russia in a comparative perspective with the early reforming CEEs.

We use equation (4) to derive comparative MFP levels for the market economy and these are shown in Figure 5 for 1995 and 2007. This shows how much of the initial technological gap has been overcome since 1995. As can be seen from Fig. 5, in the most advanced post-transition economy, Slovenia, MFP level has grown from 60 per cent of the German one in 1995 to 74 per cent in 2006, reducing the distance by almost 14 percentage points in eleven years. But large distances remain for the other CEEs. In 1995, Russian productivity levels were only 42

31 The World Bank (2008) also reported data on MFP levels for total economies on the basis of the micro database “Amadeus”.
per cent of Germany, which makes the Russia the weakest economy among the four post-transition countries. By 2007, Russia has closed only 10 percentage points of the gap, which still stands large at about half the German level. On the whole, technology catch up in CEERs was slow and the scope for future growth based on further catch up is still large.

This is true for almost all sectors. Table 4 provides the distance to the global frontier in 1995 for the 4 major sectors. It shows that in 1995 the distance to the frontier for Goods was larger than for Market Services, which might explain higher MFP growth rates in goods than in market services in tab. 3. In turn, the technology gap is larger for HSI Goods in comparison with LSI Goods, which could reflect advantages of developed economies in HSI activities. The technology level in high skills intensive services in Hungary and Slovenia in early transition was surprisingly high. These results predetermine a more formal testing of the convergence hypothesis.

[Tab. 5 is about here]

Using the framework of Bernard and Jones (1996), we have tested the existence of beta-convergence across CEERs both at the aggregate level and in sectors. For each industry it is assumed that MFP growth rates averaged over the period 1995-2007 ($\Delta \ln MFP_i$) depend on the (logarithm of the) distance to the frontier in 1995 ($\ln MFP_i^{1995}$).

$$\Delta \ln MFP_i = \alpha + \beta \ln MFP_i^{1995} + \varepsilon_i.$$  

(9)

If the catch up hypothesis of Gerschenkron holds, the estimator of $\beta$ should be significantly negative. We estimate equation (9) on a set of 26 industries for four countries. Table 5 reports the results. As can be seen, CEERs demonstrate convergence with strong statistical significance in all sectors, which make the post-transition economies different from advanced OECD economies. As Bernard and Jones (1996) demonstrated, in the latter group of countries there was no evidence of convergence in manufacturing. Next, we look at the speed of convergence, which is expected to be higher in industries with higher MFP growth rates. The speed of

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32 The total amount of industries of Market economy in three CEEs is 26. In the Russian economy the number of industries is 24, because the basic set of industries is the same, but the Extended Mining is considered as one industry instead of Mining (code C), Fuel (23) and Wholesale trade(51).
convergence ($\lambda$) is calculated from the estimates of $\beta$ using the following definition:

$$
\beta = -\frac{1-(1-\lambda)^T}{T} \tag{Bernard and Jones 1996}.
$$

Standard errors are obtained from the heteroskedasticity consistent covariance matrix (White) estimator.

Results are also shown in table 5. In this setting the interpretation of the speed of convergence ($\lambda$) is the rate at which the level of productivity is approaching to a worldwide (German) productivity level. As can be seen, Goods are converging with a rate of almost 12 per cent a year, while Market Services – just with the rate slightly higher than 3 per cent. The speed of convergence also differs within Market Services, being 5 percentage points higher in HSI Services in comparison with LSI Services.

The findings of this subsection confirm that technological catch up in CEERs is taking place, but at a slow pace. It will take more time before even the most advanced economies among CEERs achieve the level of developed economies. The productivity gap is persistent, in particular for Russia.

7. Conclusion

The purpose of this paper was to determine the role of technology catching up and labour reallocation in the productivity performance of formerly centrally planned economies in Europe. We compared early reformers (represented by the Czech Republic, Slovenia and Hungary) and a late reformer (Russia). These countries share various characteristics in terms of structural change and drivers of productivity growth. Since 1995 technology catching up was a substantial driver of productivity growth in all these economies, confirming the findings of Havlik, Leitner and Steher (2012) and Fernandes (2009). This conclusion is clearly supported by our finding of MFP dominance among sources of labour productivity growth, and our confirmation of the convergence hypothesis. Compared with technology catching up, the contribution of labour reallocation to aggregate productivity growth is minor. We found that the potential for the structural change bonus stressed in McMillan and Rodrik (2011) was low to begin with as productivity level variation across industries in post-transition economies was already limited in 1995. Another similarity in the growth process of these economies is the role of manufacturing industries. Being initially more remote from the technology frontier than market services, they have demonstrated higher speed of convergence.

This study also revealed important differences in post-transition development between CEEs and Russia. Various skill-intensive manufacturing industries (such as transport equipment manufacturing) perform well in the CEEs, while their progress in Russia is relatively modest. Development of these industries in CEEs is clearly linked to the outsourcing strategies of major European firms (Havlik, Leitner, and Stehrer 2012). The second difference is a remarkable productivity growth performance of high skills intensive services in Russia relative to the CEEs.
But this is a sign of Russia's catching up with the CEEs, rather than forging ahead, given the extremely low levels of productivity in Russia in initial years of transition. Finally, the contribution of labour reallocation in Russia was found to be higher than in CEEs. We found that this should be attributed to the Russia-specific abundance of natural resources, which exploration attracted factor resources. This process cannot be a permanent source of growth.

This research has thrown up many questions in need of further investigation. This study is the first attempt to make productivity comparisons of early and late reformers at the detailed industry level. It would be useful to consider other late reform countries, such as Ukraine, which is similar to Russia in many aspects except abundance of natural resources. One more direction is an attempt to answer the question what factors influence the speed of convergence in post-transition economies. In a broader context this question is linked to the determinants of the speed of transition, which could depend on differences in reform paths and strategies, macroeconomic stability after liberalization, enterprise performance in response in changes in ownership and corporate governance (Roland 2000; Dell'Anno and Villa 2013). Next, while we were able to study structural change in CEERs in more detail than most of the previous studies, it is important to recognize that the industry perspective might still be too restrictive. Micro studies, such as of Bartelsman, Haltiwanger, and Scarpetta (2013), showed, that distortions in the allocation factors at the level of firms can lead to substantial differences in aggregate performance across countries. Early studies of labour reallocation and productivity in manufacturing firms showed this might be also important in understanding of differences of performance in CEERs after transition. Lastly, the current analysis might fail to pick up resource reallocation between formal and informal sectors of the economy. Given the substantial presence of an informal economy in many post-transition countries this might be an additional driver or hindrance to growth.

33 There are some comparative studies of manufacturing (Ahrend, de Rosa, and Tompson 2006; Brown, Earle, and Telegdy 2006; Brown and Earle 2008), but the comprehensive industry-level picture is still absent.
### Tables

**Tab. 1. GDP per capita and labour productivity differences in post-transition economies relative to Germany in 2007**

<table>
<thead>
<tr>
<th>GDP per capita (Germany = 100)</th>
<th>Effect of working hours</th>
<th>Labour productivity (Germany = 100)</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>1</td>
<td>2</td>
</tr>
<tr>
<td>Czech Rep.</td>
<td>64.8</td>
<td>16.3</td>
</tr>
<tr>
<td>Estonia</td>
<td>56.4</td>
<td>15.2</td>
</tr>
<tr>
<td>Germany</td>
<td>100.0</td>
<td>0.0</td>
</tr>
<tr>
<td>Hungary</td>
<td>51.2</td>
<td>17.8</td>
</tr>
<tr>
<td>Latvia</td>
<td>48.5</td>
<td>13.0</td>
</tr>
<tr>
<td>Lithuania</td>
<td>44.9</td>
<td>3.3</td>
</tr>
<tr>
<td>Poland*</td>
<td>40.8</td>
<td>12.6</td>
</tr>
<tr>
<td>Russia</td>
<td>43.0</td>
<td>13.7</td>
</tr>
<tr>
<td>Slovakia</td>
<td>57.0</td>
<td>12.9</td>
</tr>
<tr>
<td>Slovenia*</td>
<td>68.6</td>
<td>15.2</td>
</tr>
</tbody>
</table>

**NOTES:**

1) Converted to USD at 2005 PPPs from (Inklaar and Timmer 2012).

2) 2006 for Poland and Slovenia

Numbers may not sum because of rounding.

**SOURCES:** Author’s calculation; EU KLEMS; Russia KLEMS; Penn World Tables, ver. 7.1. See main text.
Tab. 2. Labour productivity growth decomposition for 1995-2007
(percentage points)

<table>
<thead>
<tr>
<th></th>
<th>Russia</th>
<th>Czech Rep.</th>
<th>Hungary</th>
<th>Slovenia*</th>
<th>Germany</th>
</tr>
</thead>
<tbody>
<tr>
<td>Aggregate labour productivity, due to</td>
<td>4.60</td>
<td>3.63</td>
<td>3.66</td>
<td>4.67</td>
<td>1.70</td>
</tr>
<tr>
<td>Labour reallocation</td>
<td>1.05</td>
<td>0.03</td>
<td>0.21</td>
<td>0.52</td>
<td>0.11</td>
</tr>
<tr>
<td>Labour productivity, due to Multifactor productivity</td>
<td>3.56</td>
<td>3.60</td>
<td>3.45</td>
<td>4.14</td>
<td>1.59</td>
</tr>
<tr>
<td>Capital intensity</td>
<td>0.85</td>
<td>2.08</td>
<td>0.96</td>
<td>1.71</td>
<td>0.84</td>
</tr>
</tbody>
</table>

Notes: (*) – 2006.
Numbers are average productivity growth rates and refer to Market economy
Source: Author’s calculation; see main text.
Tab. 3. Multifactor productivity growth rates and contributions of Market economy for 1995-2007

<table>
<thead>
<tr>
<th></th>
<th>Russia</th>
<th>Czech Rep.</th>
<th>Hungary</th>
<th>Slovenia **</th>
<th>Germany</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Annual growth rates</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Market economy, total</td>
<td>2.71</td>
<td>1.52</td>
<td>2.49</td>
<td>2.43</td>
<td>0.75</td>
</tr>
<tr>
<td>HSI Goods</td>
<td>6.46</td>
<td>5.94</td>
<td>6.22</td>
<td>6.13</td>
<td>3.51</td>
</tr>
<tr>
<td>LSI Goods</td>
<td>3.05</td>
<td>3.19</td>
<td>3.81</td>
<td>4.17</td>
<td>1.83</td>
</tr>
<tr>
<td>HSI Services</td>
<td>9.40</td>
<td>1.00</td>
<td>0.57</td>
<td>0.35</td>
<td>-2.24</td>
</tr>
<tr>
<td>LSI Services</td>
<td>1.60</td>
<td>0.51</td>
<td>2.04</td>
<td>0.91</td>
<td>1.02</td>
</tr>
<tr>
<td>Mining, Fuel*</td>
<td>1.05</td>
<td>-9.24</td>
<td>-3.84</td>
<td>5.15</td>
<td>-1.23</td>
</tr>
<tr>
<td><strong>Contributions, p.p.</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Market economy, of which</td>
<td>2.71</td>
<td>1.52</td>
<td>2.49</td>
<td>2.43</td>
<td>0.75</td>
</tr>
<tr>
<td>HSI Goods</td>
<td>0.25</td>
<td>0.51</td>
<td>0.59</td>
<td>0.61</td>
<td>0.45</td>
</tr>
<tr>
<td>LSI Goods</td>
<td>0.67</td>
<td>0.88</td>
<td>0.98</td>
<td>1.34</td>
<td>0.39</td>
</tr>
<tr>
<td>HSI Services</td>
<td>0.90</td>
<td>0.13</td>
<td>0.10</td>
<td>0.06</td>
<td>-0.51</td>
</tr>
<tr>
<td>LSI Services</td>
<td>0.61</td>
<td>0.24</td>
<td>0.92</td>
<td>0.37</td>
<td>0.43</td>
</tr>
<tr>
<td>Mining, Fuel*</td>
<td>0.28</td>
<td>-0.25</td>
<td>-0.10</td>
<td>0.05</td>
<td>-0.01</td>
</tr>
</tbody>
</table>

NOTES:
* For Russia this is Extended Oil and Gas sector, which includes Mining, Fuel and Wholesale trade. In other countries Wholesale trade is included into LSI Services. A more detailed table, in which these industries are separated, is available by request.
** For Slovenia average growth rates have been calculated in 1995-2006.
Numbers may not sum because of rounding.
Source: Author’s calculation; see main text.
Tab. 4. Multifactor productivity levels relative to Germany in 1995 in PPP USD 2005

(Germany = 100)

<table>
<thead>
<tr>
<th></th>
<th>Russia</th>
<th>Czech Rep.</th>
<th>Hungary</th>
<th>Slovenia*</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Multifactor productivity, Germany = 100</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Market economy, total</td>
<td>42.3</td>
<td>57.6</td>
<td>47.6</td>
<td>59.9</td>
</tr>
<tr>
<td>HSI Goods</td>
<td>12.3</td>
<td>28.4</td>
<td>30.3</td>
<td>36.9</td>
</tr>
<tr>
<td>LSI Goods</td>
<td>23.2</td>
<td>40.2</td>
<td>33.8</td>
<td>44.4</td>
</tr>
<tr>
<td>HSI Services</td>
<td>12.1</td>
<td>53.1</td>
<td>66.4</td>
<td>66.7</td>
</tr>
<tr>
<td>LSI Services</td>
<td>64.4</td>
<td>79.7</td>
<td>54.0</td>
<td>83.4</td>
</tr>
<tr>
<td>Mining, Fuel*</td>
<td>83.0</td>
<td>99.7</td>
<td>95.8</td>
<td>35.5</td>
</tr>
</tbody>
</table>

NOTES:
This table presents results of level accounting, for which PPPs have been aggregated using the multilateral translog price indices (Caves, Christensen, and Diewert 1982a). In line with this approach, an artificial country is formed by averaging across all countries in the dataset. This artificial country is used as a bridge for making binary comparisons between any two countries. Since Estonia, Latvia, Lithuania, Poland and Slovenia are not included into this dataset for level accounting, labour productivity levels in table 1 slightly differ from the corresponding values in the present table.

* For Russia this is Extended Oil and Gas sector, which includes Mining, Fuel and Wholesale trade. In other countries Wholesale trade is included into LSI Services. A more detailed table, in which these industries are separated, is available by request.

Source: Author’s calculation; see main text.
## Tab. 5. Convergence regressions, sectoral multifactor productivity

<table>
<thead>
<tr>
<th></th>
<th>β</th>
<th>SE</th>
<th>λ</th>
<th>R²</th>
<th>Obs.</th>
</tr>
</thead>
<tbody>
<tr>
<td>Market economy, total</td>
<td>-</td>
<td>0.005</td>
<td>0.069</td>
<td>0.52</td>
<td>102</td>
</tr>
<tr>
<td>Goods</td>
<td>-</td>
<td>0.007</td>
<td>0.118</td>
<td>0.63</td>
<td>58</td>
</tr>
<tr>
<td>HSI Goods</td>
<td>-</td>
<td>0.011</td>
<td>0.050</td>
<td>0.57</td>
<td>12</td>
</tr>
<tr>
<td>LSI Goods</td>
<td>-</td>
<td>0.008</td>
<td>0.132</td>
<td>0.64</td>
<td>46</td>
</tr>
<tr>
<td>Market Services</td>
<td>-</td>
<td>0.005</td>
<td>0.034</td>
<td>0.42</td>
<td>43</td>
</tr>
<tr>
<td>HSI Services</td>
<td>-</td>
<td>0.009</td>
<td>0.073</td>
<td>0.84</td>
<td>8</td>
</tr>
<tr>
<td>LSI Services</td>
<td>-</td>
<td>0.006</td>
<td>0.025</td>
<td>0.29</td>
<td>35</td>
</tr>
</tbody>
</table>

**NOTES:**

Coefficients are estimated from the following equation $\Delta \ln MFP_i = \alpha + \beta \ln MFP_{i}^{1995} + \varepsilon_i$ with the speed of convergence ($\lambda$) calculated from $\beta = -\frac{1-(1-\lambda)^T}{T}$ (Bernard and Jones 1996). We use OLS with the heteroskedasticity robust White estimator of the covariance matrix.

The number of industries in the three East-European economies is 26 and 24 for Russia as Mining (code C), Fuel (23) and Wholesale trade(51) are combined.

* - significance at the 10% level; ** - 1% level; *** - 0.1% level.
Figures

Fig. 1. Intra- and inter-industry components of labour productivity growth in post-transition economies and Germany for 1995-2007

Notes:
(*) – 2006.
Decomposition of aggregate labour productivity growth into the intra-industry growth (within) and labour reallocation (reall) components in CEE economies, Russia and Germany in 1995-2007* (percentage)
Source: Author’s calculations based on the EU KLEMS and Russia KLEMS datasets.
Fig. 2. Shares of hours worked in 1995 and 2007
In percentage of Market economy

**a. High skill intensive goods**

**b. High skill intensive services**

**c. Low skill intensive goods**

**d. Low skill intensive services**

*Sources*: Author’s calculations based on the EU KLEMS and Russia KLEMS datasets.

*Notes*: Arranged by the descending difference of the shares

* For Slovenia values are calculated for 1995 – 2006
Fig. 3. Inter-sectoral productivity variation and aggregate labour productivity levels in 1995

SOURCES: Own calculations on the basis of WIOD Social Economic Accounts; EU KLEMS; Russia KLEMS; (Inklaar and Timmer 2012).

NOTE: Post-transition economies are highlighted in green
Coefficient of variation of (the log of) labour productivity in 31 industries is presented on a vertical axis. Labour productivity is defined as value added per hour worked, converted with GDP PPP 2005 from (Inklaar and Timmer 2012) to constant 2005 US dollars and projected to 1995.
Labour productivity levels of market economies (horizontal axis) are value added over hours worked, converted with GDP PPP to constant 2005 US dollars from (Inklaar and Timmer 2012), projected to 1995 and normalized to the level of Germany.
Dash lines correspond to the OLS regression (see estimated parameters in Tab. A2 of the Appendix).
Fig. 4. Harberger diagrams for multifactor productivity growth in 1995-2007

Russia
(curvature 0.41; value added share of industries with positive contribution: 0.69)

The Czech Republic
(curvature 0.61; value added share of industries with positive contribution: 0.47)

Hungary
(curvature 0.45; value added share of industries with positive contribution: 0.73)

Slovenia (MFP growth rates in 1995-2006)
(curvature 0.37; value added share of industries with positive contribution: 0.81)

NOTES: Filled dots correspond to industries, which belong to Goods, empty dots – to Market Services. The upper curve in each curve belongs to a corresponding country, while the lower curve is of Germany
Fig. 5. Multifactor productivity levels relative to Germany.

Notes:

(*) – 2006.

Calculations have been made for Market economy in PPP USD 2005.

Source: Own calculations on the basis of EU KLEMS, Russia KLEMS, and (Inklaar and Timmer 2012).
## Appendix

### A1. List of sectors and industries

<table>
<thead>
<tr>
<th>NACE 1.0 Code</th>
<th>Name of a sector/industry used in the paper</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>TOTAL ECONOMY</td>
</tr>
<tr>
<td></td>
<td>Market Economy</td>
</tr>
<tr>
<td></td>
<td>Goods</td>
</tr>
<tr>
<td></td>
<td><em>High Skill-intensive</em></td>
</tr>
<tr>
<td>24</td>
<td>Chemicals and Chemical Products</td>
</tr>
<tr>
<td>29</td>
<td>Machinery, Nec</td>
</tr>
<tr>
<td>30t33</td>
<td>Electrical and Optical Equipment</td>
</tr>
<tr>
<td></td>
<td><em>Low Skill-intensive</em></td>
</tr>
<tr>
<td>AtB</td>
<td>Agriculture, Hunting, Forestry and Fishing</td>
</tr>
<tr>
<td>15t16</td>
<td>Food, Beverages and Tobacco</td>
</tr>
<tr>
<td>17t18</td>
<td>Textiles and Textile Products</td>
</tr>
<tr>
<td>19</td>
<td>Leather, Leather and Footwear</td>
</tr>
<tr>
<td>20</td>
<td>Wood and Products of Wood and Cork</td>
</tr>
<tr>
<td>21t22</td>
<td>Pulp, Paper, Paper, Printing and Publishing</td>
</tr>
<tr>
<td>25</td>
<td>Rubber and Plastics</td>
</tr>
<tr>
<td>26</td>
<td>Other Non-Metallic Mineral</td>
</tr>
<tr>
<td>27t28</td>
<td>Basic Metals and Fabricated Metal</td>
</tr>
<tr>
<td>34t35</td>
<td>Transport Equipment</td>
</tr>
<tr>
<td>36t37</td>
<td>Manufacturing, Nec; Recycling</td>
</tr>
<tr>
<td></td>
<td>Market Services</td>
</tr>
<tr>
<td></td>
<td><em>High Skill-intensive</em></td>
</tr>
<tr>
<td>J</td>
<td>Financial intermediation</td>
</tr>
<tr>
<td>71t74</td>
<td>Renting of Machinery and Equipment and Other Business Activities</td>
</tr>
<tr>
<td></td>
<td><em>Low Skill-intensive</em></td>
</tr>
<tr>
<td>E</td>
<td>Electricity, Gas and Water Supply</td>
</tr>
<tr>
<td>F</td>
<td>Construction</td>
</tr>
<tr>
<td>H</td>
<td>Hotels and Restaurants</td>
</tr>
<tr>
<td>50</td>
<td>Sale, Maintenance and Repair of Motor Vehicles and Motorcycles</td>
</tr>
<tr>
<td>52</td>
<td>Retail trade, except of motor vehicles and motorcycles; repair of personal and household goods</td>
</tr>
<tr>
<td>60t63</td>
<td>Transport and transport services</td>
</tr>
<tr>
<td>64</td>
<td>Post and Telecommunications</td>
</tr>
<tr>
<td>O</td>
<td>Other Community, Social and Personal Services</td>
</tr>
<tr>
<td></td>
<td>Extended Mining</td>
</tr>
<tr>
<td>23</td>
<td>Fuel</td>
</tr>
<tr>
<td>C</td>
<td>Mining and quarrying</td>
</tr>
<tr>
<td>51</td>
<td>Wholesale trade</td>
</tr>
<tr>
<td></td>
<td>Non-Market Economy</td>
</tr>
<tr>
<td>70</td>
<td>Real Estate Activities</td>
</tr>
<tr>
<td>L</td>
<td>Public administration and defence; compulsory social security</td>
</tr>
<tr>
<td>M</td>
<td>Education</td>
</tr>
<tr>
<td>N</td>
<td>Health and social work</td>
</tr>
</tbody>
</table>
A2. Dependence of the (log of the) variation coefficient of industries labour productivity levels from the (log of the) aggregate labour productivity levels in 1995

<table>
<thead>
<tr>
<th></th>
<th>Coef.</th>
<th>s.e.</th>
<th>t-stat.</th>
<th>p-value</th>
</tr>
</thead>
<tbody>
<tr>
<td>The log of the aggregate labour productivity level</td>
<td>-0.22</td>
<td>0.019</td>
<td>-11.93</td>
<td>0.000</td>
</tr>
<tr>
<td>CEER dummy</td>
<td>0.08</td>
<td>0.058</td>
<td>1.39</td>
<td>0.174</td>
</tr>
<tr>
<td>[CEER]x[The log of the aggregate labour productivity level]</td>
<td>0.12</td>
<td>0.050</td>
<td>2.37</td>
<td>0.023</td>
</tr>
<tr>
<td>Constant</td>
<td>0.22</td>
<td>0.011</td>
<td>19.14</td>
<td>0.000</td>
</tr>
</tbody>
</table>

NOTES:
The sample consists of 39 developed, developing and post-transition economies (see the list in tab. A5 of the Appendix). The determination coefficient $R^2 = 0.89$.
CEER dummy equals one for post-transition economies (see the list in Table A5 of the Appendix) and null for the other countries.
We use OLS with the heteroskedasticity robust White estimator of the covariance matrix.
Bibliography


