



HAL
open science

FDI and productivity convergence in Central and Eastern Europe: an industry-level investigation

Martin Bijsterbosch, Marcin Kolasa

► **To cite this version:**

Martin Bijsterbosch, Marcin Kolasa. FDI and productivity convergence in Central and Eastern Europe: an industry-level investigation. *Review of World Economics*, 2010, 145 (4), pp.689-712. 10.1007/s10290-009-0036-z . hal-00555616

HAL Id: hal-00555616

<https://hal.science/hal-00555616>

Submitted on 14 Jan 2011

HAL is a multi-disciplinary open access archive for the deposit and dissemination of scientific research documents, whether they are published or not. The documents may come from teaching and research institutions in France or abroad, or from public or private research centers.

L'archive ouverte pluridisciplinaire **HAL**, est destinée au dépôt et à la diffusion de documents scientifiques de niveau recherche, publiés ou non, émanant des établissements d'enseignement et de recherche français ou étrangers, des laboratoires publics ou privés.

FDI and productivity convergence in Central and Eastern Europe: an industry-level investigation

Martin Bijsterbosch · Marcin Kolasa

Published online: 14 January 2010
© Kiel Institute 2010

Abstract This paper presents empirical evidence of the effect of FDI inflows on productivity convergence in Central and Eastern Europe, using a new and harmonized industry-level data set. Four conclusions stand out. First, there is a strong convergence effect in productivity, both at the country and at the industry level. Second, FDI inflow plays an important role in accounting for productivity growth. Third, the impact of FDI on productivity critically depends on the absorptive capacity of recipient countries and industries. Fourth, there is important heterogeneity across countries, industries and time with respect to some of the main findings.

Keywords Productivity convergence · FDI · Transition economies · Absorptive capacity · Panel data

JEL Classification C23 · F21 · O33

1 Introduction

After more than 15 years of transition and despite an impressive catching-up process, productivity levels in central and eastern European EU countries remain substantially below those in the rest of the EU. In 2005, for example, average GDP

M. Bijsterbosch
European Central Bank, DG Economics, Kaiserstr. 29, 60311 Frankfurt, Germany
e-mail: Martin.Bijsterbosch@ecb.int

M. Kolasa (✉)
National Bank of Poland, Economic Institute, ul. Swietokrzyska 11/21, 00-919 Warsaw, Poland
e-mail: Marcin.Kolasa@nbp.pl

M. Kolasa
Warsaw School of Economics, Department of Applied and Theoretical Economics,
Al. Niepodleglosci 162, 02-554 Warsaw, Poland

per capita in the region stood at around 55% of the euro area (Arratibel et al. 2007).¹ Further raising productivity levels, therefore, remains a key priority for economic policies in these countries.

The catching-up process in Central and Eastern Europe has coincided with large inflows of foreign direct investment (FDI). A key question arising from this phenomenon is how important FDI inflows have been for the convergence process in general and for productivity gains in particular. If FDI has a consistent positive impact on productivity, this would imply that countries should continue to pursue policies aimed at attracting FDI. In addition, it is important to understand whether and which economic conditions affect the size of the benefits associated with FDI inflows.

The existing cross-country studies on growth and productivity in Central and Eastern Europe largely have a stocktaking or growth-accounting character and concentrate mostly on the macro-level (e.g. Campos and Coricelli 2002; Doyle et al. 2001; European Commission 2004; Lenain and Rawdanowicz 2004). The main focus of this literature is on the pace and nature of the growth process, concentrating on the period since the start of the transition to a market economy. These studies mostly underline the importance of economic policies (including institutions) for growth and convergence.

There have been a number of attempts to investigate the link between FDI and economic growth in a more formal way. Only very few of them, however, take a cross-country or cross-industry perspective, mainly due to the lack of comparable data. For instance, Holland and Pain (1998) examine the early stages of transition in Central and Eastern Europe (1992–1996). They estimate a labour demand function using aggregate data for eight countries and find that the stock of inward foreign investment has a positive impact on productivity, with the beneficial effects being higher in the more market-orientated economies. These results were broadly confirmed by a related study of Barrell and Holland (2000), based on industry-level data covering eleven manufacturing sectors in the Czech Republic, Hungary and Poland.

Other empirical evidence on the role of FDI in the catching-up process of transition economies has been less conclusive. For example, Mencinger (2003) applies a Granger causality test to aggregate data covering eight central and eastern European economies in the period 1994–2001. The main finding is that the relationship between FDI and GDP growth is negative, which is attributed to the character of FDI during this period (mostly privatisation-related rather than greenfield investment). Hunya (1997) estimates that foreign-owned enterprises operating in the region have on average higher labour productivity than domestic firms, but notes that this may be related to the concentration of the former in more capital intensive industries. This hypothesis was confirmed by Djankov and Hoekman (2000), who use firm-level data from the Czech Republic and conclude that, after controlling for various kinds of selection biases, FDI does not seem to have a significant effect on productivity growth.

¹ Weighted average of the Czech Republic, Estonia, Hungary, Latvia, Lithuania, Poland, Slovakia and Slovenia. In this paper, Central and Eastern Europe refers to these eight EU countries.

Following an improved availability of firm-level data, an increasing number of papers on the link between FDI and growth have focused on productivity spillovers from foreign-owned companies to other firms in the economy.² A key conclusion emanating from these studies is that spillovers vary by country, sector and type of firm, depending also on the nature of FDI and absorptive capacity of domestic enterprises.

The approach taken by the most recent spillover literature has many advantages over the earlier studies. While it is certainly of interest whether there are any externalities associated with FDI inflows, an important question remains about their total impact on aggregate productivity. More specifically, since multinational companies are among the most technologically advanced firms, their presence may be beneficial for a recipient country even if their superior knowledge does not spill over to domestic firms. Therefore, it may still be useful to take a less disaggregated perspective for assessing the macroeconomic impact of FDI.

A relatively detailed industry-level analysis has recently become possible thanks to the EU KLEMS database. It covers a wide range of sectors in an internationally comparable way and can be considered as a state-of-the-art source for cross-country and cross-industry comparisons. By merging this relatively new and to a large extent still unexploited database with the WIIW database on Foreign Direct Investment in Central, East and Southeast Europe, we can carry out an econometric analysis, using data with both a country, industry and time dimension.³

There are two other important features of our paper that distinguish it from the previous literature. First, we investigate formally how the size of overall benefits associated with FDI depends on the absorptive capacity of the recipient country, which allows us to shed more light on the cross-country variation in the estimated effects of FDI found in previous studies. The concept of absorptive capacity is based on the idea that the potentially positive impact of FDI on the receiving economy may fail to materialise if domestic companies lack sufficient abilities to imitate or adopt superior technologies used by foreign firms. Absorptive capacity can be measured in various ways, for example on the basis of human capital indicators or using the relative productivity level (see Sect. 3 for details). Second, in the empirical part we employ two alternative econometric approaches, differing in the extent to which they exploit the industry versus the time dimension of the data. This makes our main conclusions more robust compared to earlier studies that rely only on one single method.

The rest of the paper is structured as follows. Section 2 provides stylized facts on the catching-up process and foreign capital inflow to Central and Eastern Europe. In Sect. 3, we discuss some theoretical considerations underlying our empirical investigation. Section 4 presents the econometric strategy. Section 5 describes data sources and definitions of variables. The main results and robustness checks are discussed in Sect. 6. Section 7 concludes.

² See a meta-analysis by Görg and Greenaway (2004) or an investigation in a cross-country setup by Damijan et al. (2003). More recent contributions to the spillover literature focusing on Central and Eastern Europe include Gersl et al. (2007), Gorodnichenko et al. (2007) and Kolasa (2008).

³ Only some very recent studies have used the EU KLEMS database for research on transition economies. See, for example, Baas et al. (2009), Persyn (2008) or Polgar and Wörz (2009).

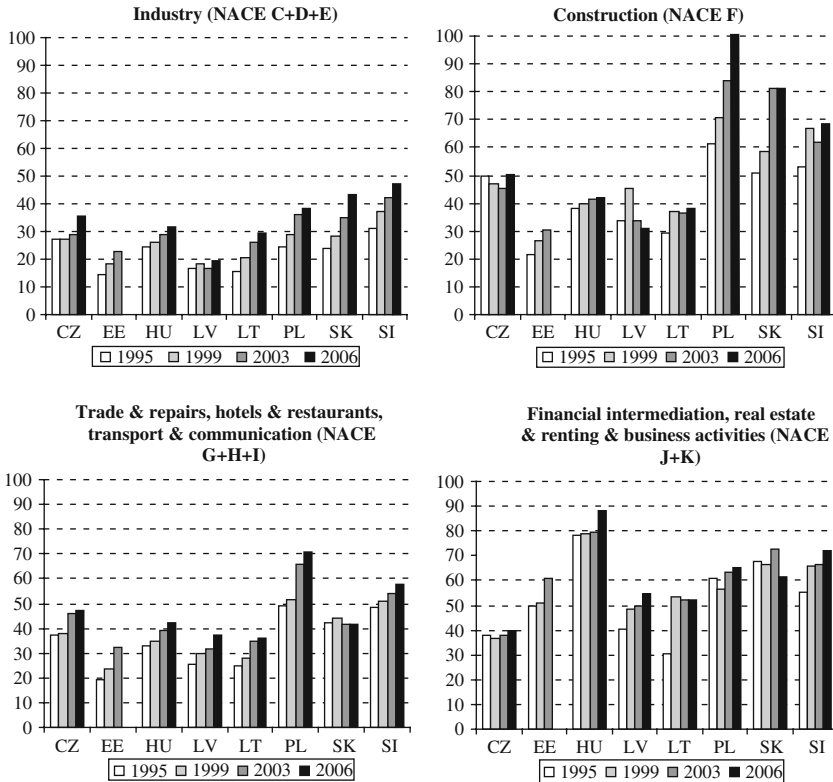


Fig. 1 Labour productivity levels relative to the euro area (euro area = 100). *Source:* EU KLEMS and Eurostat. Labour productivity is calculated as gross value added per hour worked. Level comparisons based on industry-specific PPPs

2 Stylized facts on productivity convergence and FDI inflows in Central and Eastern Europe

Taking the euro area as a benchmark, Figs. 1 and 2 provide an overview of relative labour productivity (calculated as value added per hour worked, see Sect. 5 for more details) in central and eastern European countries.⁴

As Fig. 1 reveals, there is substantial heterogeneity in productivity levels across sectors. Whereas productivity gaps vis-à-vis the euro area in the mid-1990s were relatively large in industry, they were substantially less pronounced in construction

⁴ For presentational reasons, the individual industries for which data are available have been lumped together in this section into four broadly defined sectors. Industry, in the first panel, mainly consists of manufacturing, together with mining and quarrying and electricity, water and gas supply (NACE categories D, C and E, respectively). The second sector is construction (NACE F). The third and fourth sectors are (market) services, with the former covering the more “traditional” services, such as trade and repairs, hotels and restaurants as well as transport and communication (NACE G, H and I), while the latter comprises financial and business-related services (NACE J and K). These four sectors together cover all economic activities except agriculture (and related branches) and non-market services.

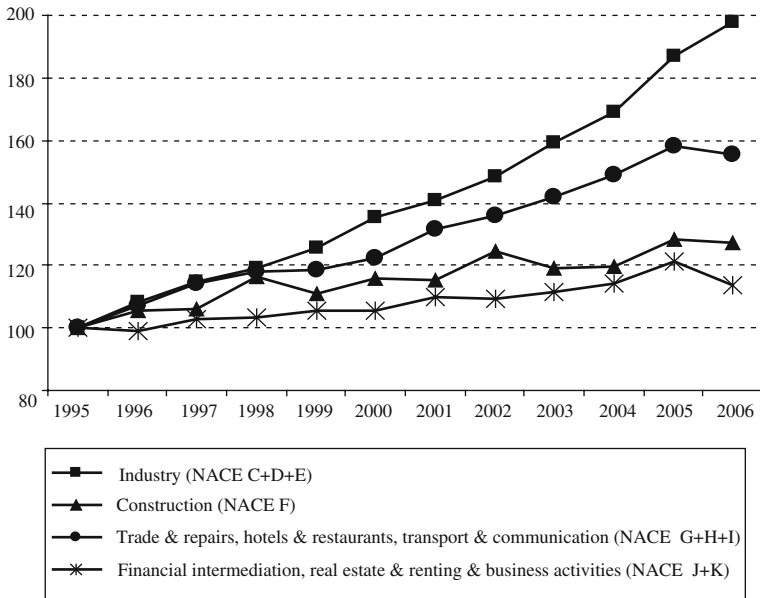


Fig. 2 Cumulative labour productivity growth by sector (1995 = 100, unweighted average of CEE-8 countries). *Source:* EU KLEMS and Eurostat. CEE-8: CZ, EE, HU, LV, LT, PL, SK, SI

and market services. In particular, output per hour in financial and business-related services was relatively close to that in the euro area already in the mid-1990s. Overall, despite marked increases over the past decade, labour productivity levels in almost all sectors remain consistently lower than in the euro area.

Focusing on the dynamics since 1995, Fig. 2 shows that cumulative productivity gains in industry since the mid-1990s have outpaced those in the other sectors. Whereas there seems to be a strong convergence effect present in industry, it was less clearly visible in construction and services. Within the service sector, however, the patterns are not homogenous. In the more traditional services, most countries in the group managed to raise productivity significantly, with cumulative gains between 1995 and 2006 mostly ranging from 20 to 70% (apart from Estonia, where productivity increases were much higher). In the financial and business-related services, by contrast, productivity gains remained more limited, particularly in the central European countries (the Czech Republic, Hungary, Poland, Slovakia and Slovenia). The industrial sector thus seems to be the main driver of labour productivity convergence vis-à-vis the euro area.

A similar convergence effect seems to be present at the macro-level across the countries considered, as the economies with the lowest initial productivity levels have been catching up relatively rapidly. This applies in particular to the Baltic countries (Estonia, Latvia and Lithuania), where labour productivity levels increased from around a quarter of those in the euro area in the mid-1990s to around 30–40% in 2006. In the central European countries, where output per hour was on average around 40% of the euro area level in 1995, productivity rose to around 50% of the euro area in 2006.

Productivity convergence in Central and Eastern Europe has often been associated with FDI inflows, which are considered to be the main vehicle for economic restructuring and technology diffusion (EBRD 1994; Damijan and Rojec 2007). Central and eastern European countries have been quite successful in attracting FDI, also relative to other emerging market economies (Fillat Castejón and Wörz 2006). Annual changes in FDI stocks have averaged around 5% of GDP in the eight central and eastern European countries considered in this paper, though there were large fluctuations from year to year. Looking at the allocation of FDI across countries, Fig. 3 shows that Estonia stood out in receiving the largest inflows relative to its size, with the FDI stock increasing from around a quarter to almost 100% of GDP between 1997 and 2005. The Czech Republic and Hungary also recorded sizeable cumulated inflows and the FDI stock to GDP ratio was slightly above 50% in both economies in 2005.

As regards the allocation of FDI across sectors, most inflows have gone to financial and business-related services and industry. Figure 4 shows that in these sectors FDI stocks relative to value added increased substantially over the past decade. The country data show that the high FDI intensity in Estonia, the Czech Republic and Hungary seems to be broad-based, with these countries consistently having the highest FDI to value added ratios in industry and in both broadly defined service sectors.

At a more disaggregated level, by far the largest recipient of FDI in services was financial intermediation, followed by business-related services (i.e. real estate, renting and business activities) and trade. The FDI stock to value added ratio in the transport, storage and communication also increased very strongly, but reached a peak already around the turn of the century (whereas the FDI intensity in financial and business-related services exhibits a consistent upward trend). Within industry,

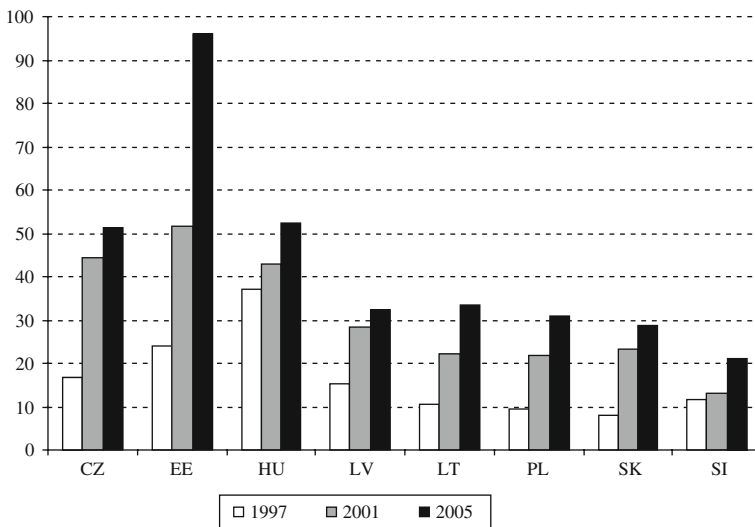


Fig. 3 FDI stock to GDP ratio (in %). *Source:* WIIW and Eurostat. Hungary: 1998 instead of 1997

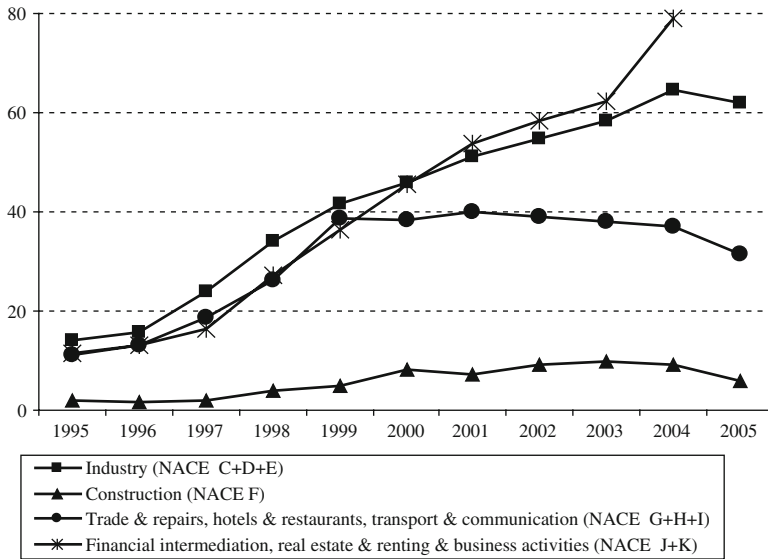


Fig. 4 FDI stocks as a share of value added by sector (in %, unweighted average of CEE-8 countries). *Source:* WIIW, EU KLEMS and Eurostat. CEE-8: CZ, EE, HU, LV, LT, PL, SK, SI. Country composition changes due to differences in data availability

FDI inflows were concentrated in transport equipment, food, as well as electrical and optical equipment.

To conclude, this overview of the data can be summarised by three main observations. First, the initial level of productivity seems to matter for the subsequent speed of convergence towards the euro area. Second, FDI inflows have mostly been concentrated in financial and business-related services and, to a lesser extent, in industry. At the country level, Estonia, the Czech Republic and Hungary have been the main FDI recipients over the past decade relative to their economic size. Third and finally, considerable differences exist across countries and sectors both as regards productivity developments and FDI inflows, particularly at a more disaggregated level.

3 Theoretical considerations

While FDI is definitely not the only channel through which international technological diffusion may occur, it is widely considered to be the most important one. This is because multinational corporations are among the most technologically advanced firms, spending relatively big amounts on research and development and using better managerial practices. This implies that inward FDI may involve the transfer of superior technologies, which can then spread over the entire economy leading to productivity gains in domestic firms (Findlay 1978; Romer 1993).⁵

⁵ In transition economies FDI inflows may also play an important role in the process of restructuring of formerly state-owned companies (see, e.g. Blanchard 1997).

The link between inward FDI and economic growth in developing countries has firm theoretical foundations. As demonstrated by Borensztein et al. (1998), this relationship can be derived using the framework of international technology diffusion developed by Barro and Sala-i-Martin (1997) and drawing on seminal contributions to the theory of endogenous growth by Romer (1990) or Grossman and Helpman (1991).

According to this setup, per capita (or labour productivity) growth occurs via accumulation of human capital and the expansion in the number of varieties of capital goods used in production of final goods. These varieties are produced by domestic and foreign firms that have undertaken a direct investment in the economy. An increase in the number of capital varieties requires a fixed cost of adapting the technology available in more advanced economies. This cost decreases with the share of foreign firms operating in the host economy and is negatively related to the technological gap vis-à-vis developed countries, which reflects decreasing imitation possibilities over the catching-up process.

Similar ideas can also be incorporated into neoclassical growth models. This was done, e.g. by Wang (1990), who assumes that the increase in effective knowledge applied to production can be written as a function of FDI. A description of technology diffusion involving decreasing imitation possibilities during the convergence process with an important role of human capital is owed to Nelson and Phelps (1966). Duczynski (2003) incorporates the concept of international technology diffusion into a Ramsey framework with capital mobility and discusses the implications of his model in the context of transition economies.

On the basis of these theoretical considerations, one can write a simple model of productivity growth in a catching-up economy using inward FDI, the relative productivity level vis-à-vis developed economies and human capital as the main explanatory variables.

However, FDI inflows as such may not necessarily be sufficient to ensure an increase in productivity. The extent to which these flows are translated into technological progress and productivity growth depends on the absorptive capacity of the sector and the country. This, in turn, hinges on the levels of basic technological literacy as well as on more advanced skills in the host country or sector (see, e.g. World Bank 2008).

The absorptive capacity concept can be implemented empirically by extending the simple model sketched out above to include interactions between the main explanatory variables. For instance, by interacting FDI with the relative productivity level we can examine to what extent gains from foreign capital inflows depend on the absorptive capacity, measured as the distance to the technological frontier. In particular, we could test the hypothesis of e.g. Glass and Saggi (1998),⁶ according to which a larger development gap implies a lower quality of technology transferred via FDI and more limited capabilities of domestic firms to benefit from potential spillovers of foreign presence (implying a negative coefficient of the interaction term). On the other hand, a positive estimate would be consistent with an alternative hypothesis provided by Findlay (1978), who emphasises the larger pool of available

⁶ See also Kokko (1994).

technological opportunities and a stronger pressure for change in relatively backward economies.

Absorptive capacity considerations can also be taken into account by interacting human capital with both the relative productivity level (used as a proxy for potential technology transfer) and FDI inflow. A classical reference stressing the role of human capital in technological diffusion is Nelson and Phelps (1966), who interact measures of human capital quality with the productivity gap vis-à-vis the technological frontier in their growth regressions.⁷ Borensztein et al. (1998) and Balasubramanyam et al. (1999) confirm the link between the impact of FDI and the quality of human capital.

It has to be noted that human capital and relative productivity are not the only proxies for absorptive capacity advocated in the literature. In particular, local firms' capabilities to absorb knowledge from abroad can be dependent on their own innovation effort (Cohen and Levinthal 1989). Also, a wide set of other characteristics (like competitive pressure, financial market development, regulations) can affect the speed of the catching-up process and the size of potential spillovers from FDI.

4 Econometric strategy

Having defined a set of potential explanatory variables, the choice of an appropriate econometric strategy is far from straightforward. In general, the most popular approaches followed in the empirical growth literature can be classified into two groups, which we will refer to as cross-section and time series studies.

The first group comprises a vast literature exploiting mainly cross-country or (less frequently) cross-industry correlation between growth and a wide set of explanatory variables. The variables used in regressions are averaged over relatively long time spans covering the whole sample (e.g. Barro 1991; Mankiw et al. 1992) or form a set of non-overlapping averages (e.g. Borensztein et al. 1998; Schadler et al. 2006).

The main advantage of cross-section studies is that their results are less likely to be driven by cyclical movements. Moreover, by exploiting cross-sectional information, they are potentially better suited for addressing questions about the sources of differences in performance across countries or industries. In practice, however, the latter advantage may be undermined by the omitted variables problem and endogeneity, leading to potentially serious biases in the estimates of the coefficients of interest.⁸

The second group of approaches, time series studies, aims at testing relationships of interest within rather than across countries or industries. This type of approach

⁷ See also Benhabib and Spiegel (2005). A confirmation of the Nelson-Phelps hypothesis, using a panel of OECD countries, can be found in Griffith et al. (2004).

⁸ In principle, this kind of problems can be mitigated by using instrumental variable techniques. However, lack of good instruments makes this option rather impractical or can even do more harm than good (see Nelson and Startz 1990 or Bound et al. 1995).

relies mainly on yearly observations and uses panel-data methods (Islam 1995; Griffith et al. 2004; Carkovic and Levine 2005).

The biggest advantage of the time series approach is that it is less vulnerable to the sources of biases that may affect purely cross-section regressions. This is because the inclusion of fixed effects in the panel helps to control for unobservable heterogeneity between objects considered, making the omitted bias problem less severe. Additionally, more sophisticated panel data techniques that rely on generalized method of moments (GMM) attempt to address the endogeneity issue, although in a rather mechanistic fashion. The major weakness of the time series approach is, however, that it does not exploit cross-section variation in the data and that it may not sufficiently emphasise medium and long-run relationships by using data of relatively high frequency.

Although there seems to be a tendency in the empirical growth literature towards using the time series approach, it might be useful to check (at least as a matter of robustness) whether the results obtained using the other approach are at least qualitatively similar. Any striking discrepancy between the time series and cross-section evidence would then call for caution in interpreting the results. Given the above considerations, our empirical investigation will rely on both approaches, the details of which are summarized below.

In the time series approach, we employ the system GMM estimation framework developed by Arellano and Bond (1991) and then extended by Arellano and Bover (1995) and Blundell and Bond (1998).⁹ More specifically, we regress the annual growth rate in labour productivity on the set of explanatory variables lagged one period, with a full set of time dummies.

The use of the system GMM method is motivated by the fact that our specification can be rewritten so that the level of productivity in central and eastern European Member States is expressed as a function of its own lag and the lagged level of productivity in the euro area. The presence of the lagged dependent variable implies that standard methods used for estimating panel data models, like the fixed effects estimator, produce biased results if the number of time periods in the sample is small (Nickell 1981). Lagging other explanatory variables, and FDI in particular, is aimed at avoiding a simultaneity bias,¹⁰ while including time dummies is expected to capture possible cyclical movements between productivity growth and right hand side variables, common across countries and industries.

The cross-section approach is pursued by splitting the sample into two 5-year periods and applying the seemingly unrelated regression (SUR) technique to a system of two equations.¹¹ All variables in these equations are expressed as 5-year averages, except for the relative productivity level, which is measured in the year preceding the beginning of the relevant period. It has to be noted that our sample is different from standard cross-section studies in that it has both a country and an

⁹ We use the `xtabond2` procedure for Stata. See Roodman (2006).

¹⁰ We treat all lagged explanatory variables as predetermined, which means that they are assumed to be uncorrelated with present and future errors. This assumption might be violated, e.g. if FDI inflow is motivated by expectations of future shocks, which seems rather unlikely.

¹¹ This means that our cross-section approach also exploits some time series variation in the data, although to a much lesser extent than the system GMM technique applied to yearly data.

industry dimension, which makes it possible to do the estimations with a full set of country and industry dummies. Naturally, this is not equivalent to fully controlling for unobservable heterogeneity across objects (like in the time series approach with a full set of country-industry specific effects). However, it is reasonable to expect that this strategy will at least attenuate the possible bias afflicting traditional cross-section estimations. An additional advantage of including country dummies is the fact that they can be regarded as (imperfect) substitutes for country-wide indicators usually used in the empirical growth literature (quality of institutions, size of the government, macroeconomic stability, financial market development etc.).

5 Data sources and definitions of variables

The main data source of which this paper makes use is the new EU KLEMS database. It is the result of a project carried out by a consortium of research institutes and financed by the European Commission in order to facilitate productivity analyses in the EU at the industry level (Timmer et al. 2007).¹² The main adjustments to the official statistical sources made in the database relate to filling gaps in industry-level data (using industry statistics) and to linking series over time.¹³

A key advantage of the EU KLEMS database is that it covers a wide range of industries (up to 72 per country, including a breakdown of services) in an internationally comparable way, with the key variables anchored in official statistics. The database covers the EU Member States in Central and Eastern Europe from 1995 onwards. In addition, it includes a large number of variables that are potentially relevant for understanding productivity developments. These features make the database probably the state-of-the-art source for cross-country and cross-industry comparisons.

Another important data source used in the paper is the WIIW database on Foreign Direct Investment in Central, East and Southeast Europe (Hunya and Schwarzhappel 2007). It contains industry-level FDI data as reported by the national central banks of the countries in the region. A key advantage of the database is that the industry breakdown is consistent with the one in the EU KLEMS database. In addition, the data are harmonised in the sense that they are in line with standard IMF definitions and methodological guidelines (although some methodological changes over time have taken place).

Table 1 provides an overview of the variables used in this study. Our total sample covers nineteen sectors of eight central and eastern European EU Member States and spans the period 1995–2005. The countries considered are: the Czech Republic,

¹² EU KLEMS stands for EU analysis of capital (K), labour (L), energy (E), materials (M) and service (S) inputs. The database is downloadable at <http://www.euklems.net>. See also Koszerek et al. (2007) for its extensive overview.

¹³ These adjustments were done by the EU KLEMS consortium on the basis of agreed procedures to ensure harmonisation of the data and to generate growth accounts in a consistent and uniform way. Harmonisation focused, among others, on industrial classifications, aggregation levels, reference years for volume measures, price concepts and methods for solving breaks.

Table 1 Data: definitions of variables

Variable	Definition	Source	Availability
Labour productivity (LP)	Value added per hour worked	EU KLEMS	Total sample
Relative labour productivity (RLP)	Labour productivity level relative to the euro area	EU KLEMS	Total sample
Foreign Direct Investment (FDI)	Gross FDI inflow (calculated from the change in stocks) as a share of value added	WIIW database on Foreign Direct Investment in Central, East and Southeast Europe	Available for 1995–2005 for LT, LV, PL and SI, while from 1996(7) for the other countries
Total investment (INV)	Gross fixed capital formation as a share of value added	Eurostat (NewCronos)	Not available for EE, gaps for LV and SI
Human capital (HC)	Share of high-skilled workers in total hours worked	EU KLEMS	Not available for EE, LT and LV
Import penetration (IMP)	Imports from EU-15 as a share of value added	WIIW Industrial Database Eastern Europe	Available for manufacturing only
R&D intensity (RD)	Business R&D expenditure as a share of value added	Eurostat	Wide coverage only for CZ, HU and PL
Capacity utilisation	Producers' assessment of the current level of capacity utilisation	European Commission Industry Survey	Available for manufacturing only

Relative labour productivity levels vis-à-vis the euro area for 1997 are calculated using industry-specific purchasing power parities (PPPs). Estimates for the remaining years are extrapolated using labour productivity growth rates

Estonia, Hungary, Latvia, Lithuania, Poland, Slovakia and Slovenia.¹⁴ The sectors covered are: manufacturing (13 industries: NACE DA to DN, without DC), construction (NACE F) and market services (5 industries: NACE G to K).¹⁵

We measure labour productivity (LP) as real value added per hour worked.¹⁶ Relative productivity levels (RLP) are calculated vis-à-vis the euro area using industry-specific purchasing power parities (PPPs) for 1997. Relative levels for the remaining years are extrapolated using real labour productivity growth rates. The

¹⁴ Bulgaria and Romania are not covered in the EU KLEMS database.

¹⁵ While data on mining and quarrying (NACE C), electricity, gas and water supply (NACE E) and manufacture of leather and leather products (DC) are generally available, these sections are excluded from our sample. The reason for doing so is their high regulation (C and E) or very small share in total economy's output (DC). It has to be noted that adding these industries to our sample keeps the main results qualitatively unchanged.

¹⁶ Ideally, we would want to measure productivity as total factor productivity. Unfortunately, this and related measures are not available (or are hard to estimate in a consistent way) for the group of countries we focus on, particularly at this level of disaggregation.

data are taken from the March 2007 release of the EU KLEMS database, covering the period 1995–2004.¹⁷

Our FDI variable is defined as the ratio of FDI inflow to gross value added.¹⁸ FDI inflow is calculated as a change in inward FDI stocks, taken from the May 2007 release of the WIIW database.¹⁹ The time span covered in this data set differs across countries and (to lesser extent) across industries. Most FDI data for Latvia, Lithuania and Slovenia cover the whole period of interest (1995–2005), while one or two-first years are missing for the other countries.

Human capital (HC) is measured as the share of hours worked by high-skilled persons in total hours worked. The share of high-skilled workers is available for all but the three Baltic countries. These data were taken from the EU KLEMS database.

In addition to these main variables, there are several other indicators which we use in the empirical part of the paper. The investment rate (INV) is measured as gross fixed capital formation divided by gross value added. By including the investment rate we can see whether FDI has an impact on productivity in addition to total investment. Data on gross fixed capital formation come from the Eurostat NewCronos database and are not available for Estonia, while those for Latvia and Slovenia cover only the most recent years (2003–2005 and 2000–2005, respectively).

Import penetration (IMP) is imports from EU-15 countries, scaled by value added. We treat this indicator as a proxy for competitive pressure. We use imports from EU-15 countries as this seems a better approximation to international competitive pressure than total imports. The data were taken from the WIIW Industrial Database Eastern Europe and are available only for manufacturing industries.

R&D intensity (RD) is defined as business research and development expenditures over gross value added. We include R&D expenditure as a proxy for innovation. Data on business R&D expenditure come from Eurostat and have a relatively wide industry-coverage only for the Czech Republic, Hungary and Poland.

Finally, we also use data on capacity utilisation from opinion surveys in order to capture cyclical swings in productivity. These survey data come from the European Commission's regular harmonised survey of the business sector in EU countries and are available only for total manufacturing.

6 Results

6.1 Preliminary regressions

In order to establish a benchmark for our econometric choices described in Sect. 4, we first estimate a set of simple regressions, using only the productivity gap and FDI

¹⁷ Whenever possible, data on labour productivity and nominal value added are extrapolated to 2005 using official Eurostat sources.

¹⁸ Similarly to all value-added shares defined below, this ratio was calculated by converting the relevant variables to a common currency using market exchange rates.

¹⁹ This means that our measure of FDI inflow captures not only flow of funds, but also the revaluation effect. Unfortunately, the availability of direct data on FDI inflows is very limited, so relying on them would dramatically truncate our sample.

Table 2 Preliminary regressions

$\Delta \ln LP$	(1)	(2)	(3)	(4)	(5)
$\ln RLP$	-0.057*** (0.008)	-0.518*** (0.026)	-0.148** (0.064)	-0.045*** (0.006)	-0.064*** (0.010)
FDI	0.121*** (0.026)	0.064** (0.027)	0.129** (0.056)	0.117*** (0.027)	0.098*** (0.032)
Estimation method	OLS	FE	GMM	OLS	SUR
Observations	1,075	1,075	1,075	294	294

The estimations are performed using the ordinary least squares (OLS), fixed effects with a full set of country-industry dummies (FE), the system generalized method of moments (GMM) and the seemingly unrelated regressions technique (SUR); for details on GMM and SUR estimations, see notes to Tables 3 and 4, respectively; for the time-series approach (columns 1 to 3), the sample is an unbalanced panel of yearly observations covering the period of 1996–2005; for the cross-section approach (columns 4 and 5), each regression is a system of two equations, covering two 5-year periods: 1996–2000 and 2001–2005; LP is labour productivity; RLP is labour productivity relative to that of the euro area; FDI is the foreign direct investment share in value added, numbers in parentheses are robust standard errors

*, ** and *** denote statistical significance at the level of 10, 5 and 1%, respectively

intensity as explanatory variables. In the case of the time series approach, we start off with simple ordinary least squares (OLS), then use the fixed effect estimator (FE) and finally run our preferred system GMM. The estimated econometric specification can be written as:²⁰

$$\Delta \ln LP_{ijt} = \alpha_{ij} + \alpha_t + \beta \ln RLP_{ijt-1} + \gamma FDI_{ijt-1} \quad (1)$$

where subscripts i, j and t index industry, country and year, respectively, and the variables appearing in the equation are as defined in Sect. 5.

In the cross-section approach we first run a simple OLS as a benchmark and then use our preferred SUR technique, using the following specification:²¹

$$\Delta \ln LP_{ijt} = \alpha_{it} + \alpha_{jt} + \beta \ln RLP_{ijt} + \gamma FDI_{ijt} \quad (2)$$

where subscript t denotes one of the two 5-year subperiods. As discussed in Sect. 4, each SUR regression is a system of two equations, covering two 5-year periods: 1996–2000 and 2001–2005.

The results of the preliminary regressions are reported in Table 2. As can be seen from comparing the estimates from columns 1 to 3, using OLS or FE in a dynamic panel data setup results in well-known biases of the autoregressive term. The direction of the bias is exactly as expected (Bond 2002): OLS clearly overpredicts the inertia in the dependent variable (and thus underpredicts the speed of convergence), while the opposite holds for the FE estimator. Comparing the results reported in columns 4 and 5 shows that adding country and industry dummies in a cross-section setup slightly changes the estimates. In particular, the coefficient of the gap term is now closer to that obtained using the time series approach.

²⁰ In the OLS estimation all yearly observations are pooled without imposing any cross-section structure. This implies that the first intercept is identical across all observations in the OLS specification.

²¹ In the OLS version there is only one intercept, common across all observations of a given 5-year subperiod.

Finally, we note that although all regressions yield significant estimates of the parameters of interest, our preferred techniques yield somewhat more conservative results in statistical terms, which is reassuring as regards the robustness of the conclusions we draw.

6.2 FDI and absorptive capacity

We start presenting our main results with a discussion of the regressions estimated using the system GMM method. The main results are reported in Table 3.

In column 1, we repeat for convenience the estimates of the simple specification considered in the previous subsection. Both the relative productivity level and the FDI share enter in a statistically significant way, pointing to a strong convergence effect and an important role of foreign capital in accounting for productivity growth in Central and Eastern Europe. Compared to other studies for developing countries, our estimate of the speed of convergence looks relatively high, which should not be surprising given the close integration of the central and eastern European region with the euro area. As regards the size of the coefficient on the FDI variable, those studies finding it significant usually report higher values. On the other hand, there are a

Table 3 System GMM estimation results

$\Delta \ln LP$	(1)	(2)	(3)	(4)	(5)	(6)
$\ln RLP$	-0.148** (0.064)	-0.138*** (0.044)	-0.071** (0.032)	-0.098* (0.056)	-0.099* (0.053)	-0.152** (0.076)
FDI	0.129** (0.056)	0.127*** (0.042)	-0.360 (0.282)	0.133*** (0.049)	0.050 (0.059)	0.152*** (0.055)
INV		-0.006 (0.039)				
$FDI \times \ln RLP$			0.144* (0.079)			
HC				-0.027 (0.125)	-0.081 (0.141)	-2.695 (1.788)
$HC \times \ln RLP$						0.654 (0.448)
$HC \times FDI$					0.479* (0.279)	
Observations	1,075	804	1,075	617	617	617
Serial correlation test (p -value)	0.32	0.59	0.33	0.12	0.15	0.13
Hansen test (p -value)	0.33	0.46	0.25	0.32	0.38	0.45

The estimation is done using the system GMM technique; the sample is an unbalanced panel of yearly observations covering the period of 1996–2005; all regressions include a full set of time dummies; LP is labour productivity; RLP is labour productivity relative to that of the euro area; FDI is the foreign direct investment share in value added; INV is the gross fixed capital formation share in value added; HC is the share of high-skilled workers in employment; all regressors are lagged 1 year; numbers in parentheses are robust standard errors; the null hypothesis of the serial correlation test is that errors in the first-differenced regression exhibit no second order correlation; the null hypothesis of the Hansen test is that the instruments are exogenous

*, ** and *** denote statistical significance at the level of 10, 5 and 1%, respectively

number of papers that do not find any robust relationship between FDI and growth. Our estimate is both statistically and economically significant, since it implies a long-run semi-elasticity of the productivity level with respect to the FDI share of about 0.9.²²

In column 2, we check whether FDI has effects over and above those of total investment by including the investment rate as an additional explanatory variable. This may also be justified by the fact that our measure of productivity is labour productivity rather than total factor productivity. Hence, there may be some role of capital deepening in accounting for productivity developments. However, the results hardly change compared to those reported in column 1. Since including the investment rate leads to a sizable decrease in the number of observations effectively used (see Sect. 5), we run the remaining regressions without this control.

We next examine the effect of adding an interaction between FDI and the relative productivity level. As can be seen from column 3, we find a positive and significant coefficient, which is consistent with the absorptive capacity argument of Glass and Saggi (1998). Together with the coefficient on FDI without interaction becoming insignificant, this may suggest that the inflow of foreign capital positively affects productivity only if the distance to the technological frontier is not too large.

In column 4, we augment the specification from column 1 by including the proxy for human capital. It turns out insignificant and does not change the remaining estimates in qualitative terms, while implying a slightly lower speed of convergence and a somewhat larger long-run effect of FDI.

The results reported in column 5 are obtained by adding the interaction between human capital and FDI. It turns out positive and significant, which reinforces the role of absorptive capacity in determining the impact of FDI on productivity growth and is consistent with the findings of Borensztein et al. (1998) obtained on a larger sample of developing countries.

In column 6 we replace the interaction of human capital and FDI with that of human capital and relative productivity. It does not enter in a significant way and does not lead to sizable changes in the coefficients on FDI or the relative productivity level compared to the regression reported in column 1.

We repeat the six baseline steps described above using the cross-section approach and employing the SUR estimation technique. The results are shown in Table 4.

Except for the last specification, we get a qualitatively similar picture to that of the time series approach. In particular, we find a strong convergence effect and impact of FDI inflow, the latter exhibiting patterns suggesting an important role of absorptive capacity.

An important difference emerges from comparing column 6 of Tables 3 and 4. Using the cross-section approach we find a negative and statistically significant coefficient on the interaction between human capital and the relative productivity level, while the relative productivity term becomes insignificant. This result can be interpreted as evidence for the critical role of human capital in bridging the productivity gap in Central and Eastern Europe, in line with the idea advocated by

²² This becomes apparent once one realises that our specification can be viewed as a special case of an error-correction model. By definition, $\ln RLP = \ln LP - \ln LP^*$, where an asterisk indexes the euro area. By re-arranging the terms in Eq. 1 we obtain the long-run semi-elasticity of LP with respect to FDI equal to $-\gamma/\beta$.

Table 4 SUR estimation results

$\Delta \ln LP$	(1)	(2)	(3)	(4)	(5)	(6)
$\ln RLP$	-0.064*** (0.010)	-0.056*** (0.012)	-0.068*** (0.010)	-0.045*** (0.012)	-0.050*** (0.014)	-0.001 (0.020)
FDI	0.098*** (0.032)	0.159*** (0.058)	-0.093 (0.117)	0.190*** (0.055)	-0.004 (0.098)	0.105* (0.059)
INV		-0.020 (0.034)				
FDI \times $\ln RLP$			0.057* (0.034)			
HC				-0.306 (0.314)	-0.590* (0.350)	1.149** (0.531)
HC \times $\ln RLP$						-0.372*** (0.108)
HC \times FDI					1.133** (0.471)	
Observations	294	232	294	172	172	172
R^2 for individual periods	0.61, 0.33	0.67, 0.38	0.62, 0.32	0.66, 0.45	0.66, 0.36	0.67, 0.43

The estimation is done using the SUR technique; each regression is a system of two equations, covering two 5-year periods: 1996–2000 and 2001–2005; the estimation allows for different error variances in each equation and for correlation of these errors across equations; all regressions include a full set of country and industry dummies, the coefficients on which are allowed to vary across periods; other coefficients are constrained to be the same for both periods; LP is labour productivity; RLP is labour productivity relative to that of the euro area; FDI is the foreign direct investment share in value added; INV is the gross fixed capital formation share in value added; HC is the share of high-skilled workers in employment; all variables are expressed as 5-year averages, except for RLP, which is measured for the year preceding the beginning of the relevant period; numbers in parentheses are robust standard errors

*, ** and *** denote statistical significance at the level of 10, 5 and 1%, respectively

Nelson and Phelps (1966). However, lack of support for this finding from the time series approach suggests caution in interpreting the results in this way.

6.3 The role of openness and innovation

We also estimate a set of regressions using import penetration (IMP) and R&D intensity (RD), following the same strategy as with human capital, i.e. including them alone and in interaction terms. Generally, the results are inconclusive, so we summarize them only briefly below.²³

In the system GMM regressions, import penetration alone does not enter in a statistically significant way, while its two interactions do: the one with FDI is significantly positive, while that with relative productivity is strongly negative. This might suggest that stronger competition from abroad is conducive to larger gains from foreign capital inflows and speeds up the pace of convergence at early stages of the catching-up process. However, these findings are not confirmed using the SUR technique: the interaction of import penetration with FDI has the negative sign while that with the relative productivity

²³ Detailed results are available from the authors upon request.

Table 5 System GMM estimation results: manufacturing vs. services

$\Delta \ln LP$	(1m)	(1s)	(3m)	(3s)	(5m)	(5s)
$\ln RLP$	-0.283** (0.113)	-0.066 (0.050)	-0.112** (0.045)	-0.051 (0.045)	-0.181* (0.100)	-0.028 (0.058)
FDI	0.123 (0.092)	0.129*** (0.039)	-0.543** (0.269)	0.263 (0.519)	0.214 (0.239)	0.060 (0.114)
$FDI \times \ln RLP$			0.223*** (0.086)	-0.035 (0.134)		
HC					-0.222 (0.411)	-0.085 (0.108)
$HC \times FDI$					-1.681 (2.711)	1.464*** (0.386)
Observations	685	390	685	390	386	231

See notes to Table 3. The column numbers correspond to the relevant regressions from Table 3, with 'm' denoting manufacturing (NACE DA to DN, without DC), while 's' stands for services (including construction, NACE F to K)

level is not significant. If included in the regression without interactions, import penetration turns out highly positive. All in all, although there is some evidence for the positive role of high competition in accounting for productivity growth in the countries considered in this paper, its particular channels seem rather unclear.

All regressions including R&D intensity yield insignificant coefficients on this variable and its interaction, both in the time series and the cross-section approach. It has to be emphasised, however, that the coverage of the data we have on R&D is far from satisfactory.

Including R&D intensity leaves other coefficients of interest qualitatively unchanged. As we do not have data on imports of services, adding import penetration implies that the sample is restricted to manufacturing industries. The results are therefore very similar to those obtained on the subsample of manufacturing industries (see Table 5, described in the next subsection).

6.4 Cross-section heterogeneity

The size of our sample, although quite impressive given well known problems with data availability and comparability across transition economies, does not allow us to examine cross-section heterogeneity of parameters of interest using too detailed breakdowns. Still, it is feasible and potentially interesting to check how our results differ across sufficiently broadly defined groups of sectors or countries.

We do this type of exercise for several specifications, which we consider as the key ones for the conclusions we have drawn so far, using only the system GMM approach. This is motivated by the fact that running SUR regressions even on two subsamples of equal size in a comparable setup to that used for the total sample, i.e. with a full set of country and industry dummies, makes the number of estimated parameters too large given the standard rule-of-thumb used in applied econometric works. However, to be on the safe side, we restrict our attention only to those regressions, for which the full sample GMM and SUR results coincide in qualitative terms.

Table 6 System GMM estimation results: CEE versus Baltic countries

$\Delta \ln LP$	(1CE)	(1BL)	(3CE)	(3BL)
$\ln RLP$	-0.131* (0.068)	-0.226*** (0.077)	-0.082 (0.072)	-0.151** (0.061)
FDI	0.127*** (0.041)	0.186* (0.107)	1.186** (0.516)	-0.559*** (0.193)
FDI \times $\ln RLP$			-0.288** (0.137)	0.227*** (0.072)
Observations	662	413	662	413

See notes to Table 3; the column numbers correspond to the relevant regressions from Table 3, with the following acronyms used for the two regions: ‘CE’

(Central Europe: the Czech Republic, Hungary, Poland, Slovakia, Slovenia), ‘BL’

(Baltic countries: Estonia, Latvia, Lithuania)

First, we examine heterogeneity in the key parameters between two groups of industries: manufacturing and services (including construction). We focus on three preferred specifications, corresponding to regressions 1, 3 and 5 in Table 3. The results are reported in Table 5.

Three important features stand out. One is that convergence towards euro area levels is much more pronounced in manufacturing than in services, which corroborates observations made in Sect. 2. Second, it is manufacturing where absorptive capacity measured as the relative productivity level is important for the positive effect of FDI inflow to materialise. Third and symmetrically, the beneficial role of FDI in services highly depends on a sufficient level of human capital.

In the second breakdown, we split our sample into two regions: Central Europe (the Czech Republic, Hungary, Poland, Slovakia and Slovenia) and the Baltic countries (Estonia, Latvia and Lithuania). Also, as documented in Sect. 2, the Baltic countries had relatively low initial productivity levels and embarked on the transition process towards the market economy later than the five central European countries. Since we do not have data on our proxy for human capital for any of the Baltic countries, we restrict our attention to regressions 1 and 3 from Table 3. The results are reported in Table 6.

It is apparent that the speed of convergence is substantially faster in the Baltic region than in the central European countries. Interestingly, in the former group, the extent of the benefits from FDI seems to depend positively on the absorptive capacity, measured as the relative productivity level vis-à-vis the euro area, while the opposite holds true for the latter countries. Putting it differently, a smaller distance to the technological frontier is accompanied by a larger positive effect from FDI in the Baltic countries, whereas in central Europe inflows of foreign capital led to particularly strong productivity gains at relatively early stages of the catching-up process. A possible explanation of this finding is that the Baltic countries had too low productivity levels in the first years of our sample to extract benefits from FDI inflows and they developed this capability only gradually.²⁴

²⁴ This hypothesis seems to be confirmed by the unrestricted variant of our SUR estimations: if we allow the coefficients in regression 3 from Table 4 to vary across the two subperiods, we get a positive and significant estimate of the interaction term only in the first equation, covering the period 1995–2000 (see Table 7).

6.5 Two periods of convergence

Given our findings, indicating a significant role for absorptive capacity in the convergence process, it may be interesting to examine how the relative importance of productivity determinants evolved over time. Such an exercise may be particularly useful for assessing future convergence prospects in the central and eastern European EU Member States.

Given our sample size, splitting it into two equal subperiods makes the system GMM method rather inefficient. Therefore, this time we rely on the SUR technique and account for parameter heterogeneity over time by relaxing the restrictions on parameter equality across the equations run for the two sub-periods (1995–2000 and 2001–2005). As before, we restrict our attention to three key specifications. The results are reported in Table 7.

The specification including the interaction between FDI and the relative productivity level shows the most striking differences across the two subperiods. This term is highly positive and significant in the equation estimated over the period 1995–2000, while insignificant in the second half of our sample. This suggests that productivity gains of foreign capital inflows were limited by a large technological gap at the early stages of convergence, while over time this constraint ceased to play a significant role. An important implication of these results, confirmed by those obtained from a simple specification excluding the interaction, is that FDI inflows were a main driver of productivity gains in the more advanced stages of the convergence process in the central and eastern European Member States (i.e. during the second half of our sample).

In contrast, the interaction between FDI and human capital turns out to be significant in both subperiods. This confirms that human capital is an important factor shaping the future path of convergence in the region.

Table 7 Unrestricted SUR estimation results

$\Delta \ln LP$	(1_95-00)	(1_01-05)	(3_95-00)	(3_01-05)	(5_95-00)	(5_01-05)
$\ln RLP$	-0.069*** (0.013)	-0.046*** (0.013)	-0.071*** (0.013)	-0.042*** (0.013)	-0.050** (0.100)	-0.035** (0.015)
FDI	0.063 (0.044)	0.114*** (0.038)	-0.410** (0.173)	0.430** (0.215)	-0.019 (0.140)	-0.036 (0.123)
$FDI \times \ln RLP$			0.155*** (0.056)	-0.080 (0.056)		
HC					-0.431 (0.438)	-0.669* (0.368)
$HC \times FDI$					1.614* (0.937)	1.155** (0.544)
Observations	147	147	147	147	86	86

The difference compared to the regressions reported in Table 4 is that the estimation does not restrict any of the parameters to be equal across the two subperiods; otherwise, see notes to Table 4; the column numbers correspond to the relevant regressions from Table 4, with ‘_95-00’ denoting the first 5-year period (1995–2000), while ‘_01-05’ standing for the second period (2001–2005)

6.6 Some robustness checks²⁵

As already mentioned before, one of the weaknesses of the time series approach is that its results may be driven by cyclical rather than medium- or long-term movements. There are certainly grounds to assume that the observed procyclicality of labour productivity is to some extent due to imperfect measurement of changes in utilisation of factor inputs (Basu and Kimball 1997). In our case, this problem should not be very serious, since we measure labour productivity as output per hour worked rather than per person employed, hence changes in working time are explicitly taken into account. Still, it is plausible that effort per hour is not constant over the business cycle, which means that our measure of productivity may exhibit some cyclical patterns related to imperfect measurement of effective labour input.

To deal with this issue we re-estimate all regressions from Table 3 with log changes in capacity utilisation in manufacturing as a control variable.²⁶ This does not affect any of the main results obtained from the baseline specification. It has to be noted, however, that this robustness check can be treated only as a very rough one, since we do not have industry-specific measures of capacity utilisation at the level of detail in this study.

Finally, we check whether our main results are not driven by any single industry that is insignificant for the total economy. Looking at the value added composition across industries in the central and eastern European Member States, the share of hotels and restaurants (NACE H) is relatively small in all countries, while the coke and refinery industry (NACE DF) is virtually nonexistent in the Czech Republic, Estonia, Latvia, and Slovenia. Excluding these industries from our sample does not change the main results in a qualitatively significant way, however. This is the case for both the time series and the cross-section approach.

7 Conclusions

The central and eastern European EU Member States have recorded impressive productivity gains over the past 15 years. At the broad sectoral level, manufacturing has been the main driver of productivity convergence, whereas gains in services have been less pronounced. Productivity catching-up has been accompanied by substantial inflows of FDI, particularly to financial and business-related services and, to a lesser extent, to industry. These general trends, however, mask important differences at the country and industry level.

The empirical results in this paper point to three main conclusions, which seem to be robust to a variety of tests. First, there is a strong convergence effect in productivity both at the country and at the industry level, i.e. productivity growth depends positively on its gap vis-à-vis the euro area. At the country level, this effect is highly pronounced in the Baltic region. At the industry level, the convergence

²⁵ The estimation results described in this section are available from the authors upon request.

²⁶ This is the approach pursued by Cameron et al. (2005) in a similar setup covering UK manufacturing industries.

effect is particularly strong in the manufacturing sector. Second, foreign capital, in the form of FDI inflows, plays an important role in accounting for productivity growth in the central and eastern European region. Third, the impact of FDI on productivity critically depends on the absorptive capacity. More specifically, the effect of FDI on productivity seems to be increasing with a declining productivity differential vis-à-vis the euro area. There is also evidence that the level of human capital is positively associated with a larger impact of FDI. The former type of interaction between absorptive capacity and the beneficial impact from FDI seems to be important in manufacturing, whereas the latter is more significant in services.

Acknowledgments This paper was written while Marcin Kolasa was working in DG Economics of the ECB. The authors would like to thank the participants to the internal ECB seminar and the INFER Workshop in Cluj-Napoca for useful comments. Special thanks are owed to: Hans-Joachim Klöckers, Reiner Martin, Monica Pop-Silaghi and three anonymous referees.

References

- Arellano, M., & Bond, S. R. (1991). Some tests of specification for panel data: Monte Carlo evidence and an application to employment equations. *Review of Economic Studies*, 58, 277–297.
- Arellano, M., & Bover, O. (1995). Another look at the instrumental-variable estimation of error-components models. *Journal of Econometrics*, 68, 29–51.
- Arratibel, O., Heinz, F. F., Martin, R., Przybyła, M., Rawdanowicz, L., Serafini, R., et al. (2007). *Determinants of growth in the central and eastern European EU member states—A production function approach*. (ECB occasional paper no. 61). European Central Bank, Frankfurt a.M.
- Baas, T., Brücker, H., Hauptmann, A., & Jahn, E. J. (2009). The macroeconomic consequences of labour mobility. In *Labour mobility within the EU in the context of enlargement and the functioning of the transitional arrangements*. Available at <http://www.wiiv.ac.at/>.
- Balasubramanyam, V. N., Salisu, M., & Dapsoford, D. (1999). Foreign direct investment as an engine of growth. *Journal of International Trade and Economic Development*, 8(1), 27–40.
- Barrell, R., & Holland, D. (2000). Foreign direct investment and enterprise restructuring in Central Europe. *Economics of Transition*, 8(2), 477–504.
- Barro, R. J. (1991). Economic growth in a cross section of countries. *Quarterly Journal of Economics*, 106, 407–443.
- Barro, R. J., & Sala-i-Martin, X. (1997). Technological diffusion, convergence and growth. *Journal of Economic Growth*, 1, 1–26.
- Basu, S., & Kimball, M. S. (1997). *Cyclical productivity with unobserved input variation*. (NBER working paper no. 5915). National Bureau of Economic Research, Washington, DC.
- Benhabib, J., & Spiegel, M. M. (2005). Human capital and technology diffusion. In P. Aghion, & S. Durlauf (Eds.), *Handbook of economic growth*. Amsterdam: Elsevier.
- Blanchard, O. (1997). *The economics of post-communist transition*. Clarendon.
- Blundell, R. W., & Bond, S. R. (1998). Initial conditions and moment restrictions in dynamic panel data models. *Journal of Econometrics*, 87, 115–143.
- Bond, S. R. (2002). *Dynamic panel data models: A guide to micro data methods and practice*. *Portuguese Economic Journal*, 1(2), 141–162.
- Borensztein, E., De Gregorio, J., & Lee, J. (1998). How does foreign direct investment affect economic growth? *Journal of International Economics*, 45, 115–135.
- Bound, J. D., Jaeger, D. A., & Baker, R. M. (1995). Problems with instrumental variables estimation when the correlation between the instruments and endogenous variables is weak. *Journal of the American Statistical Association*, 90, 443–450.
- Cameron, G., Proudman, J., & Redding, S. (2005). Technological convergence, R&D, trade and productivity growth. *European Economic Review*, 49, 775–807.
- Campos, N. F., & Coricelli, F. (2002). Growth in transition: What we know, what we don't, and what we should. *Journal of Economic Literature*, 40(3), 793–836.

- Carkovic, M., & Levine, R. (2005). Does foreign direct investment accelerate economic growth? In T. H. Moran, E. M. Graham, & M. Blomstrom (Eds.), *Does foreign direct investment promote development?* Peterson Institute for International Economics, Washington, DC.
- Cohen, W. M., & Levinthal, D. A. (1989). Innovation and learning: The two faces of R&D. *Economic Journal*, 99, 569–596.
- Damijan, J. P., & Rojec, M. (2007). Foreign direct investment and the catching-up process in new EU member states: Is there a flying geese pattern? *Applied Economics Quarterly*, 53(2), 91–118.
- Damijan, J. P., Knell, M., Majcen, B., & Rojec, M. (2003). The role of FDI, R&D accumulation and trade in transferring technology to transition countries: Evidence from firm panel data for eight transition countries. *Economic Systems*, 27(2), 189–204.
- Djankov, S., & Hoekman, B. M. (2000). Foreign investment and productivity growth in Czech enterprises. *World Bank Economic Review*, 14(1), 49–64.
- Doyle, P., Kuijs, L., & Jiang, G. (2001). *Real convergence to EU income levels: Central Europe from 1990 to the Long Term*. (IMF Working Paper No. 01/146). International Monetary Fund, Washington, DC.
- Duczynski, P. (2003). Convergence in a model with technological diffusion and capital mobility. *Economic Modelling*, 20, 729–740.
- European Bank for Reconstruction and Development. (1994). *Transition Report*. London.
- European Commission. (2004). *The EU Economy: 2004 Review*. Brussels.
- Fillat Castejón, C., & Wörz, J. (2006). *Good or bad? The influence of FDI on output growth: An industry-level analysis*. (WIIW working paper no. 38). The Vienna Institute for International Economic Studies, Vienna.
- Findlay, R. (1978). Relative backwardness, direct foreign investment, and the transfer of technology: A simple dynamic model. *Quarterly Journal of Economics*, 92, 1–16.
- Gersl, A., Rubene, I., & Zumer, T. (2007). *Foreign direct investment and productivity spillovers: Updated evidence from Central and Eastern Europe*. (Working paper no. 8). Czech National Bank, Praha.
- Glass, A., & Saggi, K. (1998). International technology transfer and the technology gap. *Journal of Development Economics*, 55(2), 369–398.
- Görg, H., & Greenaway, D. (2004). Much ado about nothing? Do domestic firms really benefit from foreign direct investment? *World Bank Research Observer*, 19(2), 171–197.
- Gorodnichenko, Y., Svejnar, J., & Terrell, K. (2007). *When does FDI have positive spillovers? Evidence from 17 emerging market economies*. (CEPR discussion paper no. 6546). Centre for Economic Policy Research, London.
- Griffith, R., Redding, S., & Van Reenen, J. (2004). Mapping the two faces of R&D: Productivity growth in a panel of OECD industries. *Review of Economics and Statistics*, 86, 883–895.
- Grossman, G. M., & Helpman, E. (1991). *Innovation and growth in the global economy*. Cambridge, MA: MIT Press.
- Holland, D., & Pain, N. (1998). *The diffusion of innovations in Central and Eastern Europe: A study of the determinants and impact of foreign direct investment*. (NIESR discussion paper no. 137). National Institute of Economic and Social Research, London.
- Hunya, G. (1997). Large privatisation, restructuring and foreign direct investment. In S. Zecchini (Ed.), *Lessons from the transition: Central and Eastern Europe in the 1990s*. Dordrecht: Kluwer.
- Hunya, G., & Schwarzhappel, M. (2007). WIIW database on foreign direct investment in Central, East and Southeast Europe: Shift to the East. FDI05/2007, WIIW.
- Islam, N. (1995). Growth empirics: A panel data approach. *Quarterly Journal of Economics*, 110, 1127–1170.
- Kokko, A. (1994). Technology, market characteristics, and spillovers. *Journal of Development Economics*, 43, 279–293.
- Kolasa, M. (2008). How does FDI inflow affect productivity of domestic firms? The role of horizontal and vertical spillovers, absorptive capacity and competition. *Journal of International Trade and Economic Development*, 17(1), 155–173.
- Koszerek, D., Havik, K., Mc Morrow, K., Röger, W., & Schönborn, F. (2007). An overview of the EU KLEMS growth and productivity accounts. (Economic paper no. 290). European Commission, Brussels.
- Lenain, P., & Rawdanowicz, L. (2004). *Enhancing income convergence in Central Europe after EU accession*. (OECD Economics Department working paper no. 392). Organisation for Economic Co-operation and Development, Paris.
- Mankiw, N. G., Romer, D., & Weil, D. (1992). A Contribution to the Empirics of Economic Growth. *Quarterly Journal of Economics*, 107, 407–437.

- Mencinger, J. (2003). Does foreign direct investment always enhance economic growth? *Kyklos*, 56(4), 491–508.
- Nelson, R. R., & Phelps, E. S. (1966). Investment in humans, technological diffusion, and economic growth. *American Economic Review*, 56, 69–75.
- Nelson, C. R., & Startz, R. (1990). The distribution of the instrumental variables estimator and its t-ratio when the instrument is a poor one. *Journal of Business*, 63, 125–140.
- Nickell, S. (1981). Biases in dynamic models with fixed effects. *Econometrica*, 49, 1417–1426.
- Persyn, D. (2008). *Trade as a wage disciplining device*. (LICOS discussion paper no. 210). Catholic University, Leuven.
- Polgar, E. K., & Wörz, J. (2009). Trade and wages: Winning and losing sectors in the enlarged European Union. *Focus on European Economic Integration*, Q1(09), 6–35.
- Romer, P. M. (1990). Endogenous technological change. *Journal of Political Economy*, 98, 71–102.
- Romer, P. (1993). Idea gaps and object gaps in economic development. *Journal of Monetary Economics*, 32(3), 543–573.
- Roodman, D. (2006). *How to do xtabond2: An introduction to “Difference” and “System” GMM in Stata*. (Working paper no. 103), Center for Global Development, Washington.
- Schadler, S., Ashoka, M., Abiad, A., & Leigh, D. (2006). Growth in the central and eastern European Countries of the European Union. (IMF occasional paper no. 252). International Monetary Fund, Washington, DC.
- Timmer, M., O’Mahony, M., & van Ark, B. (2007). The EU KLEMS growth and productivity accounts: An overview. University of Groningen & University of Birmingham.
- Wang, J.-Y. (1990). Growth, technology transfer, and the long-run theory of international capital movements. *Journal of International Economics*, 29, 255–271.
- World Bank. (2008). *Global economic prospects—Technology diffusion in the developing world*. Washington, DC.