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Economic Integration and Growth at the Margin: A Space-Time Incremental Impact Analysis

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### Timo Mitze and Philipp Breidenbach<sup>1</sup>

# Economic Integration and Growth at the Margin: A Space-Time Incremental Impact Analysis

#### **Abstract**

We use the case of EU enlargement in 2004 to investigate the impact of economic integration on regional income growth. Being particularly interested in studying the effects 'at the margin', we track the relative performance of regions adjacent to both sides of the integration border vis-à-vis non-border regions. We use a space-time incremental difference-in-difference (IDID) analysis to account for spatial spillovers, early anticipation and adjustment dynamics over time. Our findings indicate that EU-15 regions up to a distance of 100 km from the integration border experience positive integration effects, but we do not observe additional income growth effects for NMS-10 border regions compared to non-border regions. The results are found to be robust for alternative regression specifications including doubly robust estimation, varying sample settings and placebo tests. Country-specific estimates for the EU-15 finally indicate that in particular East German regions have benefited from EU enlargement potentially reflecting their proximity to Poland as largest NMS market, their favorable investment conditions, i.e. modern infrastructure, and preferential historical ties to the NMS-10.

JEL Classification: C23, F15, O47, R11

Keywords: Economic integration; regional income growth; EU enlargement; spatial spillovers; space-time incremental difference-in-difference estimation

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#### 1. Introduction

A central research question in economic geography is how exogenous shocks affect the geographic structure of an economic system and how the resulting spatial patterns of development can be explained. In this paper, we use the fifth and hitherto largest enlargement of the European Union in 2004 as an exogenous institutional shock that has heterogeneously affected the income growth path of EU regions. Our focus is set on identifying the relative performance of regions located at the internal border between 'old' (EU-15) and 'new' EU-member states (NMS-10) vis-à-vis non-border regions, i.e. we focus on regions 'at the margin' of the pre-accession territorial boundaries. We motivate our focus on border regions by that fact that these regions – although not being the prime target of economic integration – can be expected to be significantly affected in terms of market access, factor allocation/mobility and institution building among other factors (Hansen 2005, Redding and Sturm 2008).

Surprisingly, the spatial implications of economic integration are still far from being well understood. While there is a consensus among international economists that economic integration is net welfare increasing for the integrating countries (Baldwin and Venables 1995), it is not so clear how the gains from economic integration are distributed within countries (Niebuhr and Stiller 2002). With regard to the economic effects of EU integration for border regions, it can be argued that, on the one hand, the improved cross-border market access increases the regions' potential for economic growth and development (European Commission 2001, Brülhart et al. 2004, Brülhart 2011 and Hanson 2005). In this logic, increased proximity to markets in the NMS-10 together with specific territorial assets in EU-15 border regions would have allowed these regions to grow faster than more distant non-border regions.

On the other hand, granting EU-wide market access to the relatively 'poor' regions from the NMS-10 may also induces short-run costs to EU-15 border regions, particularly related to an increased cost competition from their spatially proximate neighboring regions (Niebuhr 2005).<sup>2</sup> Additionally, Behrens et al. (2007) have shown in a two-country,

<sup>&</sup>lt;sup>2</sup> In a comprehensive review of the theoretical literature on the spatial effects of economic integration Niebuhr and Stiller (2002) point out that it is a priori difficult to make predictions whether border regions are more affected by the integration process than non-border regions. Similarly, Rodríguez-Pose (2006)

two-region new economic geography (NEG) model that a country's internal economic geography constitutes a significant conditioning factor for the regional economic effects of international economic integration. In this logic, border regions, which are typically lagging behind in terms of their economic development compared to more central regions within a country due to their distorted market area, will only catch up with central regions if interregional trade costs are sufficiently high. Otherwise, it can be expected that economic integration is likely to increase regional inequalities within the integrating country in favor of central regions (e.g. Monfort and Nicolini 2000).

Given the ambiguity of theoretical predictions with regard to the direction and magnitude of economic integration effects on border regions, we shift the identification to the empirical level. As Monastiriotis et al. (2017) have recently pointed out, the few existing studies have mostly examined the link between EU association and economic growth at the national level (e.g. Henrekson et al. 1997, Badinger 2005) and virtually no study exists that provides direct evidence for the link between regional growth outcomes and the EU association process.

Bridging the gap between national and missing regional-level evidence, Monastiriotis et al. (2017) provide an extensive analysis of the spatial effects of EU integration for the subset of Central and Eastern European (CEE) regions. Using an event-analysis approach, the authors find evidence for heterogeneous regional income growth effects in the progression of EU accession between 1991 and 2008. Their results further show that the EU accession process has particularly strengthened agglomeration forces in CEE countries favoring regions with a high market potential, industry concentrations and regional specializations in increasing returns sectors. While the study does not provide direct empirical evidence on the relative performance of border regions vis-à-vis non-border in the CEE countries, it shows that central regions with higher population density, larger market access and particular industry structures benefit most from the EU accession process. Moreover, the results also indicate that increasing geographical distances to the respective national capital and to the old EU member states (proxied by the distance to Brussels)

argues that trade integration can significantly affect relative regional advantages by creating new productivity and agglomeration advantages that are unequally distributed across space. Thereby it is unclear whether border regions gain or lose in this process.

reduce regional income growth. Some of these effects are thereby found to vary over time from the early transition shock to post-accession after 2004.

A more direct focus on border regional development effects in the course of economic integration is placed in the study by Brakman et al. (2012). Analysing population trends for 1,457 regions and 2,410 cities in the EU since 1973, the authors find evidence for a positive population growth effect of EU economic integration. This effect is found to be significant at the regional and urban level for regions/municipalities located within a 70km radius from national borders. It holds for both sides of the integration border and amounts to roughly 0.15%. Niebuhr (2008) adds to this long-run view by studying the income effects of EU enlargement in 2004 using a three-region economic geography model calibrated with pre-accession data for 1995-2000. The simulation results indicate that border regions realize higher integration benefits than non-border regions with the strongest effects found for CEE regions along the former external EU-15 border. Finally, a related strand of the literature also focusses on the employment and wage effects of EU enlargement in border regions (see, e.g., Braakmann and Vogel 2011, Marin 2011). Compared to the positive regional population and income growth effects, these studies – predominately conducted at the firm level - mainly find small to moderate negative effects of EU enlargement in old EU member states such as Austria and Germany.

Here, we add to the earlier empirical evidence on regional income growth effects of EU enlargement. While our approach is conceptually related to the empirical setup used in Monastiriotis et al. (2017), we offer a range of novelties: First, our focus is set on assessing the relative income growth performance of border regions located on both sides of the integration border relative to non-border regions in the EU. For this reason, we build a sample of 1,228 EU-25 NUTS-3 regions over the period 1998 to 2008 and track the regions' growth performance in an empirical convergence model setup. Second, we account for potential neighborhood effects and quantify the evolution of spatial integration effects when gradually moving away from the integration border between EU-15 and NMS-10 countries; and third, as for non-selective spatial integration effects, we also account for potential lead- and lag-effects of EU integration in 2004 over time.

In order to properly measure these mutual effects by means of an econometric model, we propose the concept of a space-time incremental difference-in-difference (IDiD) analysis to track the relative growth performance of EU (border) regions over space and time. By using this two-dimensional grid search approach, we hope to provide further insights into the relationship between market access and the spatial distribution of economic activity. While our results do not show significant excess income growth rates for border regions from the NMS10 countries in the post-enlargement period (relative to a suitable comparison group of NMS-10 non-border regions), we observe a positive effect for border regions located in the EU15 (particularly in East Germany) relative to EU15 non-border regions. This effect is found to be statistically significant for regions directly adjacent to the border and their first-order geographical neighbors up to 100 km off the border line, which thus supports the existence of a distance decay effect to cross-border market access (Brakman et al. 2012).

When we finally apply the full space-time IDiD estimator, the resulting grid of estimated effects further indicate that the income growth effect of EU enlargement partly phases-in in the year prior to enlargement and builds-up over time in the post-accession period. While our results thus hint at some early anticipation effects possibly reflecting the continuous process of EU accession, they also point to the fact that the development potential of economic integration can only be fully reaped by border regions in the mid run. Decomposing the space-time effects by affected EU-15 countries, the results show that mainly German border regions experience sustained integration effects, while we only observe a positive short-run effect for Austrian border regions in 2004 and even negative effects for Italian regions. We argue that this effect is likely to be driven by several factors: First, East German regions are located in close proximity to the large Polish market with a size of approximately 10% of EU population. Second historical ties between East Germany and the NMS support cross-border interactions and third, during the period of EU enlargement, East Germany has offered relatively attractive investment conditions, i.e. modern infrastructure (Gauselmann et al. 2011).

The remainder of this paper is organized as follows: Section 2 outlines the study design and introduces the space-time IDiD estimation setup. Section 3 describes the data and some stylized facts. The empirical results for alternative empirical specifications and

a series of robustness checks are presented in Section 4. Section 5 discusses the policy implications of the empirical results and finally concludes the paper.

#### 2. Study design

Quantifying the income growth effects of economic integration offers many challenges. One central issue is related to the potential problem of endogeneity as the event of EU accession cannot be seen as a source of exogenous variation to the national performance, including economic growth. This two-way link between national development and EU accession mainly stems from the fact that a good economic performance partly reflects a successful transition policy and the adoption of certain institutions linked to democratic governance and a functioning market economy, which in turn are a prerequisite for signing accession agreements (i.e. the so-called Copenhagen criteria for EU accession). In response to this conceptual challenge, we follow the argumentation in Brakman et al. (2012) pointing to the fact that EU integration did not primarily target the economic development in border regions and that hence the macroeconomic enlargement 'shock' can be seen as an exogenous source of variation for these 'marginal' regions. In addition, we lean on the empirical identification approach used in Monastiriotis et al. (2017) and specify a regional growth model that incorporates a set of regional-level variables, capturing time-varying regional structures and geographical characteristics that are assumed to influence regional economic growth besides a pure enlargement effect. Thus, by embedding our empirical identification approach in the well-established related literature on the regional effects of economic integration, we can ensure to properly measure the growth effects of EU enlargement in 2004 'at the margin'.

A second challenge relates to the issues of the appropriate i) spatial definition of border regions and ii) the timing of economic effects stemming from the EU enlargement 'shock'. Here, we estimate the income growth effects of EU enlargement by means of a space-time incremental difference-in-difference (IDiD) estimation approach. This allows us to account for spatial neighborhood effects and time leads/lags in the transmission process from EU enlargement to regional economic effects. In fact, although the fifth and hitherto biggest enlargement came into force on 1 May 2004, negotiations between old and new member states had already started in 1997 with the ratification process of the treaty being finalized on 16 April 2003 (European Union 2007). Hence, on the one hand,

it therefore cannot be ruled out that the accession of the NMS-10 was already anticipated by market actors prior to 2004. At the same time, on equal grounds it cannot be ruled out that economic adjustments such as cross-border investment and migration processes took only place with a time delay. The latter point was further fueled by the so-called '2+3+2' rule allowing old member states to protect their national labor market during a transition period of maximum seven years (Koikkalainen 2011).

In the econometric literature such early anticipation and potentially sluggish adjustment processes are linked to common estimation problems such as 'Ashenfelter's dip' indicating that treated regions are systematically different from non-treated regions in the period prior to treatment (Ashenfelter, 1978). Our proposed incremental difference-indifferences (IDiD) estimation approach is able to account for such potential lead- and lageffects in the distribution of regional economic integration effects resulting from EU enlargement. Besides slicing the potential integration effects of enlargement over time, in similar veins, we also slice the potential effects over space.

As benchmark specification, we start estimating a standard difference-in-difference (DiD) specification within the framework of an empirical growth model traditionally used to study income convergence (e.g. Tondl 2001, Acemoglu 2009, Gennaioli et al., 2014). In our adaption of this type of growth models, the (log-transformed) regional income level  $(y_{it})$  in region i at time t is specified as a function of lagged income levels  $(y_{it-1})$ , (log-transformed) regional-level covariates  $(\mathbf{x}_{it})$ , and a DiD term  $(d_i^{border} \times d_t^{post04})$ , where we interact a treatment group dummy for border regions  $(d_i^{border})$  with a time dummy splitting the sample into a pre- and post-enlargement period  $(d_t^{post04})$  as

$$y_{it} = \beta_0 + \beta_1 y_{it-1} + \beta_2' \mathbf{x}_{it} + \tau d_t^{post04} + \lambda \left( d_i^{border} \times d_t^{post04} \right) + \mu_i + \varepsilon_{it}, \tag{1}$$

where  $\beta_0, \beta_1, \beta_2, \tau, \lambda$  are coefficients (vectors) to be estimated,  $\mu_i$  is a vector of region-specific fixed effects and  $\varepsilon_{it}$  is a residual term. Please note that due to the inclusion of region-specific fixed effects the time-invariant treatment group dummy  $(d_i^{border})$  cancels

out from eq.(1). Following Acemoglu (2009), we estimate eq.(1) as a function of income levels rather than income growth rates  $(\Delta y_{it} = y_{it} - y_{it-1})$ .

The main coefficient of interest is  $\lambda$ , which measures the difference in growth trajectories of border and non-border regions in the post-treatment period from 2004 onwards. If  $\lambda$  is found to be statistically significant and positive, this indicates that – controlling for other regional-specific determinants of regional growth – border regions along the integration border between old and now member states have grown faster than other EU regions after the 2004 enlargement. In the econometric literature, the coefficient  $\lambda$  is therefore also referred to as the average treatment effect of the treated (ATT). We refer to the standard estimation approach of eq.(1) including a set of regional confounding factors as regression adjusted DiD estimation.

The basic regression equation can then be extended in various dimensions. First, we test for heterogeneous growth effects among border regions in the old and new member states as

$$y_{it} = \beta_0 + \beta_1 y_{it-1} + \beta_2' \mathbf{x}_{it} + \tau d_t^{post04} + \lambda_1 \left( d_i^{EU15-border} \times d_t^{post04} \right)$$
$$+ \lambda_2 \left( d_i^{NMS10-border} \times d_t^{post04} \right) + \mu_i + \varepsilon_{it},$$
 (2)

where  $d_i^{EU15-border}$  and  $d_i^{NMS10-border}$  are separate treatment group dummies for border regions in the old and new EU member states, respectively. As before, these time-invariant group dummies enter the DiD terms but otherwise cancel out from eq.(2) due to the inclusion of region-specific fixed effects. We can then test for differences in the growth effects for these two treatment groups when comparing the coefficients for the DiD terms  $\lambda_1$  and  $\lambda_2$ . Moreover, besides estimating eq.(2) with two DiD terms based on the overall sample of EU-25 regions and just one common comparison group of non-border regions, we also estimate the income growth effects of EU enlargement for border regions using separate sub-samples for the EU-15 and the NMS-10. The main reason for conducting

<sup>&</sup>lt;sup>3</sup> See Appendix B for further details on the equivalence of growth model specifications in first-differences and levels of the dependent variable.

sub-sample estimations is that this likely increases the homogeneity of treatment (border) and comparison (non-border) regions in the light of structural differences between the EU-15 and NMS-10 countries after the fall of the iron curtain in the early 1990s and thus may support the validity of the crucial common trend assumption of DiD models implying that – in the absence of treatment – the difference between the treatment and comparison group is constant over time (Lechner 2011).

Together with other assumptions needed to ensure the consistency of the DiD estimator, namely i) exogeneity of the included control variables ( $\mathbf{x}_{it}$ ) with regard the treatment and outcome, ii) common support implying that no other systematic factors are varying across geography and over time as well as iii) absence of relevant interactions between the members of the population, the validity of the common trend assumption implies that we can interpret the estimation outcome of the DiD approach as the causal impact of economic integration on income growth in border regions along the 2004 EU enlargement border. To reduce potential estimation biases stemming from a violation of these assumptions, we apply several modifications and extensions to the standard regression approach shown in eq.(1). First, to account for the fact that the chosen regression model does not properly control for all confounding factors and thus violates the exogeneity and common support assumptions, we also estimate eq.(1) as doubly robust regression specification, which combines the method of inverse probability weighting by a propensity score with standard regression adjustment in DiD estimation (Funke et al., 2011).

Second, we account for the fact that the effects of economic integration – although subject to geographical distance decay – may not only have an impact on 'direct' border regions but also their geographical neighbors. Clarke (2017) has recently pointed out that the SUTVA may be too strong, when dealing with regional data and using territorial boundaries to estimate treatment effects. The reason is that territorial borders are porous and may give rise to spatial spillovers. In order to estimate unbiased treatment effects in the presence of spillovers Clarke (2017) proposes the use of a weaker condition than

<sup>&</sup>lt;sup>4</sup> The assumption of the absence of relevant interactions between the members of the population is also referred to as the stable unit treatment value assumption (SUTVA, see Rubin 1977).

<sup>&</sup>lt;sup>5</sup> See Appendix C for further details on the doubly robust estimation of the DiD model.

SUTVA, relying on the assumption that there existing at least some subset of units which are not affected by the treatment status of others. As it can be assumed that those economic actors living in comparison areas 'close to' treatment areas are able to either partially or fully access treatment, the subset of regions unaffected by the treatment can be determined by their (geographic, economic etc.) distance to treated units. Hence, starting from eq.(1), we capture this latter gradual distance decay effect by including a set of treatment group dummies and associated interaction terms for direct border regions, first, second- and higher-order neighbors as

$$y_{it} = \beta_0 + \beta_1 y_{it-1} + \beta_2' \mathbf{x}_{it} + \tau d_t^{post04} + \sum_{k=1}^K \lambda_k (d_i^k \times d_t^{post04}) + \mu_i + \varepsilon_{it}, \quad (3)$$

where the index k=1,...,K counts the total number of treatment group dummies  $(d_i^k)$  included in the empirical specification. Each treatment group thereby represents a 'slice of space' defined by a distance belt relative to the 2004 EU enlargement border. This process can be seen as testing for incremental changes in economic integration effects over space, where the hypothesis from standard inter-country and inter-regional trade models is that a potential growth effect of economic integration decreases with further distance to the border due to increasing transportation costs to the newly accessed market area (Niebuhr and Stiller 2002). Through the inclusion of multiple DiD terms, eq.(3) can hence be seen as a spatial IDiD estimator. Details about the empirical operationalization of the K spatial treatment groups will be given in Section 3.

Third, we apply the IDiD estimator to slice treatment effects over time as originally proposed in Dolton et al. (2010). The specification of a temporal IDiD estimator shall thereby account for potential lead and lag structures in the distribution of the economic integration effect on regional income growth over time. The underlying assumption is that that economic integration effects captured by the coefficient of the DiD term ( $\lambda$ ) in eq.(1) are not uniformly distributed over time but may gradually build-up or phase-out due to the fact that economic costs and benefits of economic integration are not equally distributed over time (Niebuhr 2008). Starting from the benchmark specification in eq.(1), we capture lead- and lag-effects before/after 2004 as

$$y_{it} = \beta_0 + \beta_1 y_{it-1} + \beta_2' \mathbf{x}_{it} + \sum_{s=2002}^{2008} \tau_s d_t^s + \sum_{s=2002}^{2008} \lambda_s (d_i^{border} \times d_t^s) + \mu_i + \varepsilon_{it}, \quad (4)$$

where eq.(4) tests for the presence of year-specific effects ( $\lambda_s$ ) between 2002 and 2008 relative to the 'off' sample period 1998 to 2001. As Dolton et al. (2010) point out, the advantage of the temporal IDiD estimation procedure is that it facilitates the estimation of year-to-year incremental effects of economic integration and allows to test for the presence of Ashenfelter's dip. Finally, one may also think of combinations of eq.(3) and eq.(4) to identify incremental income growth effects of economic integration over space and time as

$$y_{it} = \beta_0 + \beta_1 y_{it-1} + \beta_2' \mathbf{x}_{it} + \sum_{s=2002}^{2008} \tau_s d_t^s + \sum_{s=2002}^{2008} \sum_{k=1}^K \lambda_{k,s} (d_i^k \times d_t^s) + \mu_i + \varepsilon_{it}.$$
 (5)

Eq.(5) hence represents a full space-time IDiD estimator and allows us to conduct a 'grid search' for significant IDiD coefficients over slices of time and space in order to provide a comprehensive assessment of the regional income growth effects of EU enlargement in 2004. As default specification, all models are estimated as fixed effects model (FEM). A distinct advantage of including region-specific fixed effects is that we are able to account for any latent regional heterogeneity which is roughly time constant such as rural-urban settlement structures, population density, regional institutions and trust or further regional amenities. This, in fact, enables us to further reduce the likelihood of introducing an estimation bias due to omitted variables when quantifying the regional growth effects of economic integration at the integration border.

However, given the inclusion of the lagged dependent variable as a right-hand-side regressor in eq.(1) to eq.(5), we also estimate the growth models by means of dynamic panel data estimators in order to avoid the risk of including the so-called Nickell bias (Nickell 1981). Specifically, here we use a (bootstrap) bias-corrected fixed effects model (FEMc) as suggested by Kiviet (1995) and Everaert and Pozzi (2007). Once the model

properly controls for serial correlation through the inclusion of  $y_{it-1}$ , the models' error terms  $(\varepsilon_{it})$  are assumed to be distributed as  $\varepsilon_{it} \sim N(0, \sigma^2)$ .

#### 3. Data and stylized facts

In order to estimate eq.(1) to eq.(4), we collect data from Eurostat for NUTS3 regions covering the period 1998-2008. While the selection of small-scale regional entities at the NUTS3 level is a natural choice to properly isolate immanent border effects, the selection of the sample length is mainly driven by two facts: First, regional historical data from Eurostat prior to 1998 is subject to a high degree of uncertainty and missing information (especially for regions in the NMS-10 countries). Second, in order to avoid biases in the estimation of structural growth regressions arising from strong business cycle movements in the course of the global economic crisis after 2008, we limit the sample period to 2008. This sample limitation is typically applied in the related literature (see, for instance, Monastiriotis et al. 2017, Breidenbach et al. 2018). The model's dependent variable is the regional per capita GDP level. Regional-level control variables include – besides lagged per capita GDP levels – the labor participation rate, the employment share in the service sector, the unemployment rate and population growth. All variables are used in logarithmic transformations. Variable definitions and summary statistics are given in Table A.1 in the online supplementary data and research materials.

Geo information on the EU's internal territorial borders is extracted from a shapefile on administrative units in the EU obtained from Eurostat. Direct border regions in the EU-15 are defined as those regions whose administrative boundaries intersect with a region from the NMS-10 countries and vice versa as shown in Figure A.1 in the online supplementary data and research materials (i.e. border regions are defined via land borders here). To measure the degree of spatial neighborhood effects we also define higher-

<sup>&</sup>lt;sup>6</sup> For FEMc estimation, we use the xtbcfe command for STATA documented in De Vos et al. (2015).

<sup>&</sup>lt;sup>7</sup> NUTS stands for 'Nomenclature of Statistical Territorial Units'. NUTS3 regions are defined by minimum and maximum population thresholds of 150 to 800 thousand inhabitants and are thus much smaller compares to NUTS2 regions as the next higher regional aggregate with a population range of 800 thousand to 3 million inhabitants.

<sup>&</sup>lt;sup>8</sup> See Capello et al. (2018a) for a discussion of alternative methods to define border regions.

order neighbors with regard to their geographical distance from the border. For this purpose, we calculate for all regions which are not classified as direct border regions the geographical distance from the region's centroid to the closest location at the border. Using 50km threshold distances k with k={50km, 100km, 150km,..., 300km}, we then build additional treatment group dummies for regions within these 50km distance belts from the border and test for spatially distributed integration effects as shown in eq.(3) and – in combination with temporally distributed integration effects – in eq.(5).

A visual inspection of the average annual income growth rates for border and non-border regions in the EU-15 and the NMS-10 is given in Figure A.2 in the online supplementary data and research materials. The figure shows that particularly EU-15 border and non-border regions follow almost identical pre-treatment trends prior to 2004, which supports the validity of the common trend assumption of DiD estimation as outline above. In 2004, border regions in the EU-15 start to grow faster compared to non-border regions, potentially hinting at the presence of positive integration effects of EU enlargement. A reverse tendency can, however, be observed for NMS-10 regions. Here, the average growth rate of non-border regions persistently outdates the respective growth rate of border regions. The gap in the relative growth performance of these two groups particularly increases from 2006 onwards. More thorough statistical tests on the differences in growth rates of treated and non-treated regions over space and time will be performed next.

#### 4. Empirical results

Table 1 shows the estimation results for the benchmark DiD specifications in eq.(1) and eq.(2). Additionally, separate sub-sample estimates for the EU-15 and NMS-10 are reported as well. The latter sub-sample estimates narrow down the group of comparison regions used in the DiD approach (i.e. only non-border regions in the EU-15 are used as comparison group when estimating the regional income effects for EU-15 border regions and likewise for the case of NMS-10 border regions). This limits the degree of regional heterogeneity across macro regions in Europe that may result in an unequal comparison of treated and comparison units. As the results show, we find a positive and significant

<sup>&</sup>lt;sup>9</sup> The validity of the common trend assumption is also tested with a placebo test (i.e. including a placebo treatment for the pre-integration year 2001) as shown in Table 2.

coefficient for the overall DiD term in column (I), which covers both EU-15 and NMS-10 border regions in a joint dummy variable. The additional marginal growth effect of EU integration amounts to 0.5% on top of the overall increase in the growth performance of EU-25 regions after 2004 of 3.4% (measured by means of the included time dummy taking a value of one from 2004 onwards). If we separate the two treatment groups for EU-15 and NMS-10 border regions, column (II) shows that only the income growth trend of NMS-10 border regions is significantly larger compared to EU-25 non-border regions after 2004.

#### << Table 1 about here >>

However, the problem associated with this overall growth specification becomes apparent when we additionally run separate sub-sample estimates for the EU-15 and NMS-10 in columns (III) and (IV), respectively. It can be seen that both macro-regions follow entirely different common growth trends in the post enlargement period. While average income growth accelerated by 3.3% in EU-15 regions, it is almost doubled for NMS-10 regions (6.5%). This implies that any comparison of NMS-10 border regions with non-border regions in the EU-25 (as done in column (II)) will lead to an upward bias in the estimated border region effect of NMS-10 regions. This is exactly what can be observed in column (III). The NMS-10 border regions did not benefit over and above the rest of the NMS-10 regions. Once, the DiD estimation setup accounts for the much stronger overall growth trend in NMS-10 regions after 2004, the coefficient for the DiD term does not report any additional growth impact in border regions compared to NMS-10 non-border regions. This result is in line with earlier studies such as Monastiriotis et al. (2017) arguing that particularly densely populated capital regions benefited from EU integration and not necessarily those proximate to the integration border.

Quite differently, however, column (IV) in Table 1 indicates that border regions in the EU-15 have experienced an additional growth stimulus compared to EU-15 non-border regions. While being only marginally significant at the 10% critical level, this latter result becomes more well-defined if we apply the doubly robust DiD estimation approach by weighting comparison regions with regard to their similarity in regional characteristics  $\mathbf{x}_{it}$  in the pre-enlargement period 1998 to 2001. As shown in column (V), we find an

additional growth effect of 0.8%, which is statistically significant at the 5% significance level. In comparison, applying doubly robust DiD estimation to the NMS-10 subsample, the integration effect on border regions actually becomes significantly negative providing further support to the hypothesis that geography is not a key factor in explaining post-EU enlargement growth in the new member states. Carefully speaking, we thus get first empirical evidence for a positive causal effect of economic integration on border regions in the EU-15.

This overall result is also supported by several robustness checks shown in Table 2. First, column (I) indicates that the significance of the estimated effects for EU-15 border regions remains intact when evaluated for the subsample of EEC-6 countries<sup>10</sup>, which can be seen as a more homogeneous comparison group compared to the EU-15 aggregate. Second, we decompose the estimated treatment effects for the EU-15 at the national level. Column (II) shows that the positive overall effect can mainly be attributed to excess income growth in German regions. In comparison, we find statistically insignificant effects for Austrian regions and even negative effects for Italian regions.

We offer a range of placebo tests to critically inspect the credibility and robustness of the estimated treatment effects. First, we introduce an artificial treatment in 2001 which is used to test for significant growth difference between border and non-border region of the EU-15 in a subsample ranging from 1998 to 2001. The subsample range is chosen in order to exclude later 'real treatment' effects. The non-significant placebo treatment effect of this estimation (column III) supports the common trend assumption since a significantly different trend between border and non-border-regions would show up in this estimation.<sup>12</sup>

<sup>&</sup>lt;sup>10</sup> The EEC-6 aggregate comprises the six founding nations of the European Economic Community in 1957: Belgium, France, Germany, Italy, Luxembourg and the Netherlands. While comparison regions from Austria are thus excluded from the sample, the treatment group remains unaffected.

<sup>&</sup>lt;sup>11</sup> Country-specific DiD estimates should, however, only be interpreted carefully as the number of border regions varies significantly across countries and is particularly small for Italy (Germany: 27 regions, Austria: 12 regions, Italy: 3 regions).

<sup>&</sup>lt;sup>12</sup> Due to the rigorous sample restriction, the number of observations is strongly reduced. However, note that all other coefficients and significances are in line with the previous estimations. A non-significant

Second, we also search for placebo effects in space and select certain groups other than border regions as being the treatment group in the light of EU enlargement in 2004 (systematically or randomly). In terms of their physical geography, coastal regions share several characteristics with border regions. That is, due to their sea border they often have less neighbors and thus a potentially limited market access However, different from the group of border regions, coastal regions are not expected to benefit over-proportionally from EU integration as their restriction to market access in the immediate geographical vicinity has not been lifted. The DiD estimation results support this hypothesis and shows no significant positive (or even negative) effects when coastal regions are taken as treatment group (column IV and V). Finally, besides this systematically selected placebo treatment group, the results also show that randomly assigned placebo treatments among EU-15 regions do not provide statistically significant results which underline the significant findings for EU-15 border regions as shown in Table 1.

#### << Table 2 about here >>

In order to further investigate the nature of the economic integration effect in EU-15 (border) regions, we then move on to the application of the incremental IDiD estimations. First, we check for the spatial extent of integration effects for the aggregate treatment period from 2004 onwards by adding further treatment groups to the regression setup, which are defined as first, second and higher order neighbors of border regions. The results from the extended estimations are summarized graphically in Figure 1 and are further reported in Table A.2 in the online supplementary data and research materials.<sup>14</sup> As

<sup>(</sup>placebo) treatment effect is also found if the subsample is extended to 2002 with a placebo treatment for 2002 (coefficient: 0.012; standard error: 0.009; *p*-value: 0.14). Detailed estimation results can be obtained from the authors upon request.

<sup>&</sup>lt;sup>13</sup> Actually treated EU-15 border regions have been excluded from this estimation setup in order not to bias the economic development in comparison regions.

<sup>&</sup>lt;sup>14</sup> Results are based on regression adjusted IDiD estimation. Using a doubly robust estimator gave us very similar coefficients and significance values. These results can be obtained from the authors upon request. Additionally, the results for the NMS-10 subsample, as shown in Table A.2, did mostly support the impression of insignificant or negative effects for border and neighboring regions. We will not further focus on these results in the following.

the Figure shows, we indeed find positive neighborhood effects for a maximum distance of 100km off the border.<sup>15</sup> This additional integration effect for first-order neighboring regions, which is in line with findings in Brakman et al. (2012) for population growth, potentially reflects the internal geography of border regions with larger urban areas not being located directly at the border up instead within some moderate distance (exceptions are, for instance, Frankfurt/Oder and Görlitz in Germany, Trieste in Italy).

#### << Figure 1 about here >>

We then also check for the presence of incremental enlargement effects over time and apply the full space-time IDiD estimator for the subset of EU-15 regions. Again, we provide a graphical summary of the empirical results in Figure 2, while detailed regression outputs are reported in Table A.3 in the appendix. Panel A highlights the temporal evolution of integration effects in EU-15 border regions accounting for potential leads in 2002 and 2003 prior to EU enlargement. Building upon the insights from the above spatial IDiD estimation, the border region group has been defined as direct border regions plus neighboring regions within 100km from the border. As the estimation results show, we do not find any significant anticipation effect in 2002 compared to the 'off'-period 1998 to 2001. However, the significant point estimate for the year 2003 indicates that some early anticipation effects are in order – though of smaller magnitude compared to the short-run post-EU accession effect in 2004. Afterwards, the effect remains roughly stable over time with a minor slack in 2005 (as already indicated by the stylized growth trends for EU-15 border and non-border regions in Figure A.1).

<< Figure 2 about here >>

<sup>&</sup>lt;sup>15</sup> Note that we have generally used 50km steps to define different groups of neighboring regions. However, given that there were very few regions within a distance of up to 50km to the border which were no direct border regions, we have integrated these regions in the set of 100km neighboring regions.

<sup>&</sup>lt;sup>16</sup> Results for NMS-10 regions, which are in line with the insignificant findings at the aggregate level, can be obtained from the authors upon request.

On top, Panel B reports the full grid of results for slices of treatments over space and time. The results thereby confirm our earlier estimates pointing to significant additional income growth effects in EU-15 border regions after 2004 with some evidence for early anticipation effects in 2003. Moreover, the effects are shown to be the strongest for first-order regions within a distance up to 100km from the border. And, finally, the grid in Panel B shows to have two peaks over time – one immediately around the timing of EU enlargement and a second peak in 2007/2008. This latter result may provide some preliminary evidence that the benefits of economic integration only gradually build up over time while they appear to be stable over space (at least for moderate distances from the EU enlargement border).

As a final exercise, we decompose the IDiD estimation results by EU-15 countries with border regions. The year-specific effects shown in Figure 3 thereby support the aggregate estimates from Table 2 indicating that German regions benefited most from the EU enlargement. As Panel A of Figure 3 shows, the estimated effects for German border and first-order neighboring regions build up gradually over time with no indication of early anticipation effects. They remain statistically significant over the entire treatment period (except for direct border regions in 2006). For Austria, Panel B shows a significantly positive one-period growth effect for border regions in 2004 pointing at short-run gains from EU enlargement. Finally, for Italian border regions we get persistent negative effects, which, however, are already in place prior to 2004. Given the very small number of Italian border (and first-order neighboring) regions, the latter results should only be interpreted very carefully as the underlying assumptions for the chosen DiD estimation may not be met (see, for instance, Donald and Lang 2007, Conley and Taber 2011 for alternative DiD setups with very few treatment groups).

<< Figure 3 about here >>

#### 5. Discussion and conclusion

We have analysed the spatial implications of economic integration. Specifically, using EU enlargement in 2004 as an exogenous institutional shock, we have studied the income growth effects 'at the margin', i.e. for regions in the geographical vicinity to the integration border. Our emphasis on border regional effects goes along with the distinct focus of

EU policy makers to closely monitor the performance of internal land border regions in the EU (European Commission 2017). This intensive monitoring reflects worries that border regions are typically expected to have a lower development potential compared to non-border regions in the light of their remoteness related to limited market access, public service provision etc. Accordingly, it is of key interest for policy makers to gain insights on factors that improve their economic development potential and, hence, studying the regional effects of economic integration may add to the latter.

Although trade models, traditional and modern location theories generally predict that the lifting of borders translates into positive welfare effects, these theoretical models do not provide clear-cut predictions whether border regions may particularly benefit from economic integration or not and –if so– over which time dimension these benefits can be reaped. Shifting identification to the empirical level, we have shown that EU-15 NUTS3 border regions have indeed experience a positive development stimulus once trade and mobility restrictions of national borders have been removed in the process of EU enlargement in 2004. Our proposed incremental difference-in-difference estimation approach thereby further allowed us to disentangle these effects over space and time. As such, we find that this positive effect is bound to regions with a maximum distance of 100km from the integration border. With regard to the temporal pattern, we find that the enlargement effect is roughly stable over time characterized by two moderate peaks around the immediate timing of the EU accession in 2004 and later on in 2007/08.

While we get some evidence for early anticipation effects in 2003, generally we do not find evidence that the common trend assumption is systematically violated, i.e. border and non-border regions follow similar pre-treatment trends (this particularly holds for the EU-15). Various robustness tests underline our key finding that –at least– some significant positive integration effects are in order for EU-15 border regions. Country-specific estimates for the EU-15 further indicate that German regions have benefited most from EU integration. We argue that this effect is likely to reflect several underlying factors: First, it can be explained with the specific geographic location of East German regions being in close proximity to the large Polish market with roughly 10% of EU population. Gravity model applications to bilateral trade flows between the EU-15 and the NMS-10

in the run-up to EU enlargement have largely predicted a first mover advantage for Germany (e.g. Buch and Piazolo 2000, Kunze and Schumacher 2003).

Second, besides their privileged geographic location, East German regions have also been reported to be relatively well equipped with factors that may attract national and foreign investments. For instance, Gauselmann et al. (2011) find for the period 2000-2008 that East German regions have possessed a distinct location advantage over Poland and the Czech Republic in terms of their advanced transition path with efficient institutions, their proximity to West European markets and their modern infrastructure. In fact, Behrens (2011) identifies infrastructure as an important prerequisite for a balanced spatial development path in the progress of economic integration. The relative success of East German states in attracting foreign direct investments (FDI) is also documented in Mitze et al. (2010) when estimating gravity models for trade and FDI flows between German federal states and the EU-27 in 1993-2005.

Third, long-run historical ties between East Germany and the NMS-10 may have given East German regions a relative advantage in cross-border interactions over other EU-15 border regions in Austria and Italy. For instance, German was the only "Western" language that could be learned and practiced freely in Eastern Europe before 1989 (Piazolo 1997). Finally, our empirical findings are also in line with studies looking more carefully into the specific types of transnational interactions taking place in borderlands. For instance, Schäffler et al. (2017) find that German foreign direct investment in the Czech republic are mainly made by firms in border regions that would otherwise not be able to invest abroad if higher transaction costs would have to be borne.

In the light of our empirical findings, the proposed actions by the European Commission in September 2017 in order to boost growth and cohesion in EU border regions can be seen as an appropriate means to account for the prevailing structural differences in border regions compared to non-border regions. Beside the specific support of firms in border regions to access larger markets and transnational networks (found by Schäffler et al. 2017) or the adoption of properly functioning institutions (found, e.g., by Pinkovskiy 2017), our results suggest that ongoing integration and a consequent facilitation of crossborder trade and mobility helps to accelerate economic development in border regions. Existing literature (see, e.g., Bosker et al. 2010, Kashiha et al. 2017, Capello et al. 2018b)

shows that national borders still have strong impacts on trade and economic prosperity within the European Union, thus leaving ample space for future integration efforts aiming to support the economic growth of border regions.

#### **Conflict of Interest**

The authors declare that they have no conflict of interest.

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Table 1: Baseline DiD estimation results for EU-25 NUTS3 regions in 1998–2008

Dep. Var.: ln(GDPpc <sub>i,t</sub> )	(I)	(II)	(III)	(IV)	(V)	(VI)
Sample	EU-25	EU-25	EU-15	NMS-10	EU-15	NMS-10
Estimation Method	RA	RA	RA	RA	DR	DR
$ln(GDPpc_{i,t-1})$	0.908***	0.907***	0.909***	0.803***	0.909***	0.803***
	(0.0183)	(0.0187)	(0.0188)	(0.1071)	(0.0188)	(0.1071)
ln(Service <sub>i,t</sub> )	0.106***	0.107***	0.045**	0.194***	0.077***	0.249***
	(0.0144)	(0.0143)	(0.0206)	(0.0576)	(0.0294)	(0.0709)
$ln(Emp_{i,t})$	-0.042	-0.043	-0.032**	0.017	-0.018	0.072
	(0.0374)	(0.0372)	(0.0154)	(0.0996)	(0.0249)	(0.1247)
$ln(Unemp_{i,t})$	-0.021***	-0.021***	-0.010***	-0.037***	-0.021***	-0.061***
	(0.0022)	(0.0022)	(0.0020)	(0.0139)	(0.0025)	(0.0121)
$ln(\Delta Pop_{i,i})$	-0.650***	-0.652***	-0.751***	0.337	-0.879***	0.244
	(0.0785)	(0.0792)	(0.1092)	(0.3986)	(0.0905)	(0.4153)
Time Dummy <sub>2004</sub>	0.034***	0.034***	0.033***	0.065***	0.043***	0.081***
	(0.0020)	(0.0021)	(0.0016)	(0.0110)	(0.0014)	(0.0166)
Diff-in-Diff <sup>2004</sup>	0.005**					
	(0.0026)					
Diff-in-Diff <sup>2004</sup>		-0.001	0.004*		0.008**	
		(0.0032)	(0.0025)		(0.0039)	
Diff-in-Diff <sup>2004</sup>		0.015**		-0.01		-0.014**
		(0.0060)		(0.0089)		(0.0068)
Number of region-year obs.	9720	9720	8381	1339	8381	1339
Region-fixed effects	YES	YES	YES	YES	YES	YES

Note: \*\*\*, \*\*, \*= denote significance at the 1%, 5% and 10% critical level. Robust standard errors are calculated on the basis of nonparametric bootstrap where the number of bootstrap samples is set of 50. RA = regression adjusted estimation; DR = doubly robust estimation.

Table 2: Robustness tests for alternative econometric specifications, sample settings and placebo treatments

Dep. Var.: ln(ΔGDPpc)	(I)	(II)	(III)	(IV)	(V)	(VI)	(VII)
Sample:	EEC-6	EU-15	EU-15	EU-15	EEC-6	EU-15	EU-15
Time:	1998-2008	1998-2008	1998-2001	1998-2008	1998-2008	1998-2008	1998-2008
Treatment:	k = EU-15 border	k = Ctry border	k = EU-15 border	k = Coastal	k = Coastal	k = Random  5%	k = Random 10%
$ln(GDPpc_{i,t-1})$	0.898***	0.909***	0.962***	0.908***	0.894***	0.908***	0.907***
	(0.0270)	(0.0197)	(0.0560)	(0.0244)	(0.0276)	(0.0209)	(0.0248)
ln(Service <sub>it</sub> )	0.064***	0.045**	0.411***	0.046**	0.056**	0.046**	0.046**
	(0.0228)	(0.0193)	(0.1011)	(0.0222)	(0.0276)	(0.0189)	(0.0222)
$ln(Emp_{i,t})$	-0.066***	-0.032	-0.020	-0.025	-0.055	-0.026	-0.027
	(0.0242)	(0.0194)	(0.1438)	(0.0252)	(0.0550)	(0.0164)	(0.0239)
ln(Unemp <sub>it</sub> )	-0.001	-0.010***	-0.035***	-0.011***	0.001	-0.010***	-0.010***
	(0.0038)	(0.0019)	(0.0113)	(0.0029)	(0.0033)	(0.0023)	(0.0030)
$ln(\Delta Pop_{it})$	-0.879***	-0.749***	-0.852**	-0.726***	-0.908***	-0.742***	-0.741***
	(0.1371)	(0.1085)	(0.3709)	(0.0682)	(0.1581)	(0.0690)	(0.0673)
Time Dummy2004	0.030***	0.033***		0.034***	0.030***	0.033***	0.033***
	(0.0016)	(0.0016)		(0.0014)	(0.0014)	(0.0012)	(0.0014)
Diff-in-Diff <sup>2004</sup>	0.008**			-0.004***	0.001	-0.001	-0.002
₩ W K	(0.0034)			(0.0012)	(0.0018)	(0.0035)	(0.0021)
$Diff$ -in- $Diff_{k=GER}^{2004}$	,	0.009**		,	(	(,	(**** )
		(0.0039)					
$Diff$ -in- $Diff_{k=AUT}^{2004}$		0.001					
⇒ K=AUI		(0.0031)					
$Diff$ -in- $Diff_{k=1TA}^{2004}$		-0.014***					
₩ W K≡II A		(0.0030)					
Time Dummy <sub>2001</sub>		(010000)	-0.024***				
• •••			(0.0050)				
Diff-in-Diff <sup>2001</sup>			-0.025				
к			(0.0153)				
Number of region-year obs.	5755	8381	1715	8036	5530	8036	8036
Region-fixed effects	YES	YES	YES	YES	YES	YES	YES

Note: \*\*\*, \*\*= denote significance at the 1%, 5% and 10% critical level. Robust standard errors for RA estimates are calculated on the basis of nonparametric bootstrap where the number of bootstrap samples is set of 50. EEC-6 = European Economic Community including France, Italy, Germany, Netherlands, Belgium and Luxembourg (founding member states of European Union). Random treatment assignment has been done on the basis of the randtreat command for STATA (Carril 2015). Different values for the share of treated regions in all sample regions have been used (5% and 10%) in order to closely proxy the actual share of border regions in NUTS3 regions (see summary statistics in Table A.1 in the online supplementary data and research materials).

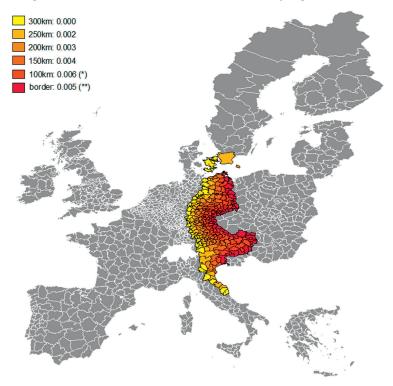
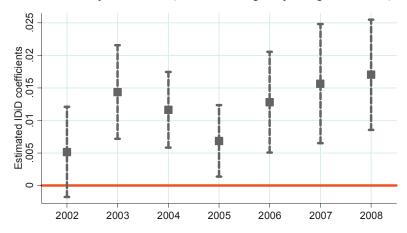


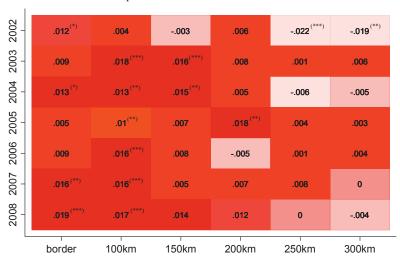
Figure 1: Spatial IDiD coefficients for alternative treatment groups in EU-15 sub-sample

*Notes*: (\*\*), (\*) = statistically significant results at the 5% and 10% critical level, respectively. Underlying Estimation results are shown in Table A.2.

Figure 2: Space-time incremental IDiD estimation results for regions in EU-15 Panel A: Year-specific effects (direct border regions plus regions <100km)

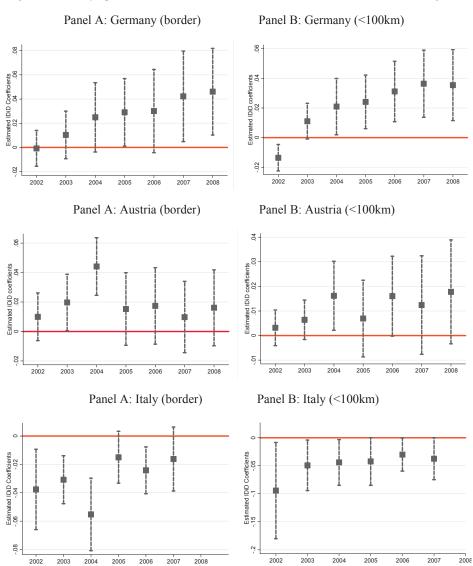


Panel B: Space-time distribution IDiD coefficients



*Notes:* Panel A reports the estimated coefficients plus 95% confidence intervals for year-specific IDiD terms measuring the spatially cumulative integration effect for EU-15 regions within 100 km from the border over time. Panel B shows the grid of space-time IDiD coefficients for alternative combinations of years and treatment groups; underlying estimation outputs can be found in Table A.3; (\*\*\*), (\*\*) = statistically significant results at the 1%, 5% and 10% critical level, respectively.

Figure 3: Country-specific DiD estimation results for German, Austrian and Italian regions



Notes: Underlying estimation outputs can be found in Table A.4.

## Appendix A. Summary statistics, additional figures and estimation results

Table A.1: Variable definitions and summary statistics

Variable	Description	Mean	S.D.	Min.	Max.
$GDPpc_{i,t}$	Per capita GDP (in €)	20940.73	9130.94	3200	152700
Service <sub>i,t</sub>	Share of service sector employment in total employment (in %)	64.949	11.281	20.371	92.472
$Emp_{i,t}$	Share of employees in total workforce (in %, excl. agriculture)	86.661	7.735	23.644	1
$Unemp_{i,t}$	Unemployment rate (in %)	8.475	5.158	0	34.5
$\Delta Pop_{i,t}$	Population growth rate (in %)	0.194	0.984	-29.092	13.242
EU-25 Border <sub>i</sub>	Dummy for EU-25 border regions	0.052	0.222	0	1
EU-15 Border <sub>i</sub>	Dummy for EU-15 border regions	0.034	0.182	0	1
NMS-10 Border <sub>i</sub>	Dummy for NMS-10 border regions	0.018	0.133	0	1
$Coastal_i$	Dummy for coastal regions	0.329	0.469	0	1

Note: All data for NUTS3 regions are gathered from the Eurostat general and regional statistics database.

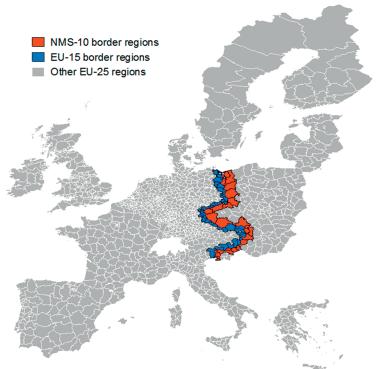
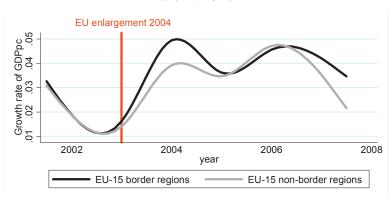


Figure A.1: EU enlargement in 2004 and border regions in the EU-15 and NMS-10

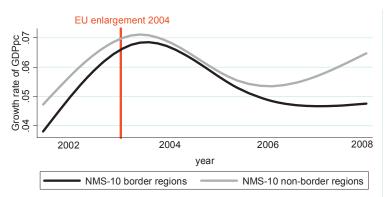
Note: Border regions are calculated for land borders only. See main text for further details.

Figure A.2. Average GDP growth rates of border and non-border regions in EU-15 and NMS-10.

Panel A: EU-15



Panel B: NMS-10



*Notes:* For definition of border regions see main text and Figure A.1. To highlight structural trends of GDP growth before and after EU enlargement, the growth rates shown in Figure A.1 have been smoothened by a cubic spline to better capture structural patterns in the data.

Table A.2: Incremental DiD estimates for spatial distribution of regional income growth effects of EU enlargement in 2004

Sample	Dep. Var.: ln(GDPpci,t)	(I)	(II)	(III)	(IV)	(V)
EU-15	Time Dummy2004	0.032***	0.032***	0.032***	0.032***	0.032***
	•	(0.0016)	(0.0015)	(0.0015)	(0.0014)	(0.0014)
EU-15	Diff-in-Diff <sup>2004</sup>	0.005*	0.005*	0.005**	0.005**	0.005**
		(0.0025)	(0.0026)	(0.0025)	(0.0025)	(0.0025)
EU-15	Diff-in-Diff <sup>2004</sup>	0.006*	0.006*	0.006*	0.006*	0.006*
		(0.0032)	(0.0032)	(0.0033)	(0.0033)	(0.0033)
EU-15	Diff-in-Diff <sup>2004</sup>		0.004	0.004	0.004	0.004
			(0.0040)	(0.0040)	(0.0040)	(0.0040)
EU-15	Diff-in-Diff <sup>2004</sup>			0.002	0.003	0.003
				(0.0038)	(0.0039)	(0.0039)
EU-15	Diff-in-Diff <sup>2004</sup> <sub>N250km</sub>				0.003	0.003
					(0.0026)	(0.0027)
EU-15	Diff-in-Diff <sup>2004</sup>					0.001
						(0.0025)
NMS-10	Time Dummy <sub>2004</sub>	0.065***	0.065***	0.063***	0.062***	0.063***
		(0.0137)	(0.0150)	(0.0094)	(0.0119)	(0.0118)
NMS-10	Diff-in-Diff <sup>2004</sup>	-0.010	-0.010	-0.008	-0.008	-0.008
		(0.0070)	(0.0092)	(0.0082)	(0.0069)	(0.0069)
NMS-10	$Diff$ -in- $Diff_{NIOOkm}^{2004}$	-0.004	-0.004	-0.002	-0.002	-0.002
		(0.0067)	(0.0047)	(0.0057)	(0.0073)	(0.0049)
NMS-10	$Diff$ -in- $Diff_{NISOkm}^{2004}$		0.001	0.004	0.004	0.004
			(0.0162)	(0.0120)	(0.0122)	(0.0122)
NMS-10	$Diff$ -in- $Diff_{N200km}^{2004}$			0.019	0.019	0.019*
				(0.0140)	(0.0125)	(0.0102)
NMS-10	$Diff$ -in- $Diff_{NZSOkm}^{2004}$				0.004	0.003
	2004				(0.0119)	(0.0126)
NMS-10	$Diff$ -in- $Diff_{N300km}^{2004}$					-0.01
						(0.0085)
Number of region-year obs. [EU-15/NMS-10]		8381/1339	8381/1339	8381/1339	8381/1339	8381/1339
Covariates included		YES	YES	YES	YES	YES
Region-fixed effects		YES	YES	YES	YES	YES

Note: \*\*\*, \*\*, edenote significance at the 1%, 5% and 10% critical level. Robust standard errors for RA estimates are calculated on the basis of nonparametric bootstrap where the number of bootstrap samples is set of 50.

Table A.3: Incremental DiD estimates for space-time distribution of regional income growth effects for EU-15 border regions

Dep. Var.: In (ΔGDPpc)	(I)	(II)	(III)	(IV)	(V)	(VI)
Treatment:	k = border	k = <100 km	k = <150 km	k = <200 km	k = <250 km	k = <300 km
$Diff$ -in- $Diff_k^{2002}$	0.012*	0.004	-0.003	0.006	-0.022***	-0.019**
	(0.0065)	(0.0065)	(0.0074)	(0.009)	(0.0065)	(0.0087)
$Diff$ -in- $Diff_k^{2003}$	0.009	0.018***	0.016***	0.008	0.001	0.006
	(0.0065)	(0.0046)	(0.0063)	(0.0069)	(0.0051)	(0.0057)
$Diff$ -in- $Diff_k^{2004}$	0.013*	0.013**	0.015**	0.005	-0.006	-0.005
	(0.0077)	(0.0060)	(0.0068)	(0.0064)	(0.0068)	(0.0103)
$Diff$ -in- $Diff_k^{2005}$	0.005	0.010**	0.007	0.018**	0.004	0.003
	(0.0059)	(0.0047)	(0.0059)	(0.0092)	(0.0062)	(0.0073)
$Diff$ -in- $Diff_k^{2006}$	0.009	0.016***	0.008	-0.005	0.001	0.004
	(0.0073)	(0.0052)	(0.0067)	(0.0058)	(0.0050)	(0.0062)
$Diff$ -in- $Diff_k^{2007}$	0.016**	0.016***	0.005	0.007	0.008	0
	(0.0072)	(0.0045)	(0.0064)	(0.0059)	(0.0052)	(0.0071)
$Diff$ -in- $Diff_k^{2008}$	0.019***	0.017***	0.014	0.012	0	-0.004
•	(0.0067)	(0.0060)	(0.0096)	(0.0091)	(0.0061)	(0.0062)
Number of region-year obs.	8381	8381	8381	8381	8381	8381
Covariates included	YES	YES	YES	YES	YES	YES
Region-fixed effects	YES	YES	YES	YES	YES	YES

Note: \*\*\*, \*= denote significance at the 1%, 5% and 10% critical level. Robust standard errors for RA estimates are calculated on the basis of nonparametric bootstrap where the number of bootstrap samples is set of 50.

Table A.4: Country-specific incremental DiD estimates for temporal distribution of regional income growth effects

Dep. Var.: In (ΔGDPpc)	(I)	(II)	(III)	(IV)	(V)	(VI)
Country:	GER	AUT	ITA	GER	AUT	ITA
Treatment:	k = border	k = border	k = border	k = < 100 km	k = <100 km	k = <100 km
$Diff$ -in- $Diff_k^{2002}$	-0.001	0.009	-0.038***	-0.016***	0.003	-0.094**
	(0.0075)	(0.0098)	(0.0144)	(0.0046)	(0.0044)	(0.0439)
$Diff$ -in- $Diff_k^{2003}$	0.010	0.019*	-0.031***	0.011*	0.006	-0.049**
	(0.0099)	(0.0116)	(0.0086)	(0.0061)	(0.0049)	(0.0229)
$Diff$ -in- $Diff_k^{2004}$	0.025*	0.044***	-0.055***	0.021**	0.016**	-0.044**
	(0.0146)	(0.0119)	(0.0131)	(0.0097)	(0.0081)	(0.0207)
$Diff$ -in- $Diff_k^{2005}$	0.029**	0.015	-0.015	0.024***	0.007	-0.042**
	(0.0142)	(0.0149)	(0.009)	(0.0092)	(0.0095)	(0.0217)
$Diff$ -in- $Diff_k^{2006}$	0.030*	0.017	-0.024***	0.031***	0.016	-0.030**
	(0.0175)	(0.0157)	(0.0084)	(0.0104)	(0.0099)	(0.0150)
$Diff$ -in- $Diff_k^{2007}$	0.042**	0.009	-0.016	0.036***	0.012	-0.037*
	(0.0191)	(0.0147)	(0.0115)	(0.0116)	(0.0122)	(0.0191)
$Diff$ -in- $Diff_k^{2008}$	0.046**	0.016	_	0.035***	0.018	_
	(0.0183)	(0.0156)		(0.0122)	(0.0128)	
Number of region-year obs.	8381	8381	8381	8381	8381	8381
Covariates included	YES	YES	YES	YES	YES	YES
Region-fixed effects	YES	YES	YES	YES	YES	YES

Note: \*\*\*, \*\*, \*= denote significance at the 1%, 5% and 10% critical level. Robust standard errors for RA estimates are calculated on the basis of nonparametric bootstrap where the number of bootstrap samples is set of 50. Due to missing observations year-specific IDiD coefficients for Italy could not be estimated in 2008.

#### Appendix B. Equivalence of empirical growth model specifications

This appendix shows the equivalence of estimating an empirical growth model in first differences or levels of an outcome variable  $(Y_{i,t})$  for region (or country) i at time t. The focus is set on interpreting the vector of coefficients  $\beta_2$  for the set of included regressors  $(\mathbf{X}_{i,t})$ . A fairly standard workhorse specification in the spirit of Mankiw et al. (1992) can be formulated as

$$\Delta y_{it} = \alpha_i + \beta_1 y_{it-1} + \beta_2' \mathbf{x}_{it} + \lambda_t + u_{i,t}, \tag{A.1}$$

where lower case letters indicate log transformed variables and the dependent variables is specified as  $\Delta y_{it} = (y_{it} - y_{it-1})$  with  $\Delta$  being the first-difference operator. Besides fixed effects for regions  $(\alpha_i)$  and years  $(\lambda_i)$ , eq.(A.1) typically includes variables proxying input factors of production in region/country I, which may enter the regression equation either contemporaneously or with a time lag (see, e.g., Mohl and Hagen 2010). In the empirical growth literature, the coefficient  $\beta_1$  can be used to measure the region's rate of convergence towards steady state income and to test for the validity of different theoretical growth model predictions. The coefficients  $\beta_2$  measure the marginal effects of changes in  $\mathbf{x}_{it}$  on  $\Delta y_{it}$ . Following Acemoglu (2009), we can rearrange the growth model into a regression equation in levels of the outcome variables, which allows estimating equation (A.1) as a dynamic panel data model of the following form:

$$y_{it} = \alpha_i + \delta y_{it-1} + \beta_2' \mathbf{x}_{it} + \lambda_t + u_{i,t}, \tag{A.2}$$

It is thereby important to note that rearranging the estimation equation only affects the coefficient measuring the region's rate of convergence toward steady state income with  $\delta = \beta_1 + 1$ , while leaving the coefficients for the set of regional regressors ( $\mathbf{X}_{i,t}$ ) unchanged (Tondl 2001). This also implies that the coefficient of the DiD term introduced in eq.(1) in the main text can still be interpreted as the marginal effect of economic integration on regional income growth.

#### Appendix C. Doubly robust difference-in-difference estimation

Doubly robust estimators combine an outcome regression with a model for the selection into treatment (i.e., the propensity score) in order to estimate unbiased treatment effects (see, for instance, Funke et al. 2011 for an overview). We implement the doubly robust DiD estimation as a two-stage approach: In the first stage, we first estimate the propensity score  $(\pi_{ij})$  as the conditional probability of a region to receive a particular treatment j (= being subject to economic integration as a border region). To do so, we thereby define a multivalued treatment variable (border<sub>i</sub>) as

$$border_i = \begin{cases} 0 & \text{region } i \text{ is a non-border region,} \\ 1 & \text{region } i \text{ is an EU-15 border region,} \\ 2 & \text{region } i \text{ is a NMS-10 border region.} \end{cases}$$

The variable is then taken as regressand in a multinomial logit model as (McCaffrey et al., 2013)

$$\pi_{i,j} = \mathbf{P}(border_i = j | X_i) = \frac{e^{\beta_j' x_i}}{1 + \sum_{j=1}^2 e^{\beta_j' x_i}} \quad \text{for } j = 1,2 \text{ and } t < 2002,$$

$$\mathbf{P}(border_i = 0 | X_i) = \frac{1}{1 + \sum_{j=1}^2 e^{\beta_j' x_i}} \quad \text{for } t < 2002,$$
(B.1)

where j indicates the treatment status with j=0,1,2 as indicated above,  $\mathbf{x}_i$  is the set of included regional characteristics to control for selection into the treatment status and  $\beta_j'$  are unknown coefficients to be estimated from the data. The category j=0 (non-border regions) serves as base outcome in the multinomial logit regression model. We restrict the first-stage estimation to the sample period t < 2002 in order to ensure that the selection into the treatment (economic integration of a border region) is specified as a function of pre-integration regional characteristics.

The obtained  $\pi_{ij}$  are then used to construct weights for the estimation of pairwise ATTs. Particularly, we are interested in estimating ATT<sub>1</sub> (EU-15 border regions versus non-border regions) and ATT<sub>2</sub> (NMS-10 versus non-border regions) with associated sample weights

$$w_{i,1} = \frac{\mathbf{P}(border_i = 1|\mathbf{x}_i)}{\mathbf{P}(border_i = 0|\mathbf{x}_i)}$$

and

$$w_{i,2} = \frac{P(border_i = 2|\mathbf{x}_i)}{P(border_i = 0|\mathbf{x}_i)}.$$

Intuitively speaking,  $w_{i,1}$  and  $w_{i,2}$  weight each individual region by the reciprocal of their probability of receiving the treatment that they received (with non-border regions being the base group) relative to the probability of being a EU-15 or NMS-10 border region. This ensures that regions with covariate values that are much more common in their own treatment group than in the target group(s) get small weights. In a second stage, we then apply DiD estimation using weighted least square (WLS) regression. Given that the use of propensity score weights for estimating multiple treatment effects is based on several conditions which may only be partially fulfilled in a sample of regions with a limited set of covariates (McCaffrey et al., 2013),<sup>17</sup> we only use the doubly robust DiD as a particular robustness check for the standard regression adjusted DiD estimation.

For EU-15 regions, mainly regional unemployment and population growth rates prior to 2001 are identified as significant factors characterizing border regions vis-à-vis non-border regions. For NMS-10 regions, we additionally find that higher shares of service sector activity are negatively correlated

<sup>&</sup>lt;sup>17</sup> First, each region has a non-zero probability of receiving each treatment and second, the set of observed pre-treatment covariates is sufficiently rich to include all variables directly influencing the treatment and outcome variable.

with the probability of being a border region. First-stage regression outputs can be obtained from the authors upon request.