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Abstract

We investigate whether joining the European Monetary Union and losing the ability to set monetary policy affected the economic growth of 12 Eurozone countries. We use the synthetic control approach to create a counterfactual scenario for how each Eurozone country would have evolved without adopting the Euro. We let this matching algorithm determine which combination of other developed economies best resembles the pre-Euro path of twelve Eurozone economies. Our estimates suggest that there were some mild losers (France, Germany, Italy, and Portugal) and a clear winner (Ireland). Nevertheless, a GDP decomposition analysis suggests that the drivers of the economic gains and losses are heterogeneous.

JEL classification: E30, E60, C32, E02, E52, E65

Keywords: Monetary union, Eurozone, Macroeconomic performance, Synthetic control method, GDP decomposition

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"For all the seven long years since the signing of the Maastricht treaty started Europe on the road to that unified currency, critics have warned that the plan was an invitation to disaster." (Krugman, 1998)

1 Introduction

In January 1999, the exchange rates between member countries' national currencies and the euro were fixed irrevocably and the European Central Bank officially took over the responsibility of conducting the unified monetary policy. Twenty years have passed since the Euro was launched and the member states gave up the ability to set their own monetary policy. In this work, we evaluate whether joining the Euro had any macroeconomic effect for twelve of the Eurozone countries.

To address this question, we develop a counterfactual scenario which represents how each Eurozone country would have evolved without adopting the Euro as their currency. For this analysis, we employ what is arguably the most important innovation in the policy evaluation literature in the last fifteen years - the synthetic control method (Athey and Imbens (2017)). We let this matching algorithm determine which combination of other OECD advanced economies best resembles the pre-Euro path of each Eurozone member. We then compare the post-Euro macroeconomic performance of each economy to its synthetic doppelganger. In particular, by decomposing the countries' and the doppelgangers' GDP into their components, we identify the main drivers of the accession gains and losses.

In the context of the Eurozone, it was expected that adopting a common currency would reduce the exchange rate volatility, the transaction costs, and any price discrimination (De Grauwe, 2020). Most likely, it could also spur trade and investment within the Euro area (Frankel and Rose (1998)). Notwithstanding, since its announcement many have been calling into question the success of the Euro (Wyplosz, 2006). They believed that the Eurozone did not satisfy the requirements of an Optimum Currency Area, especially due to the lack of labor mobility (Jonung, Drea, et al., 2009). Additionally, the Euro area countries could no longer set monetary policy independently thus becoming more exposed to external (asymmetric) shocks. Still nowadays, the rising strength of nationalism movements in Europe has intensified the doubts about the advantages of the Eurozone (Fligstein et al. (2012), Guiso et al. (2019)). Some of the arguments put forward are the loss of sovereignty and the suitability of the monetary policy undertaken to all member states.

Our contribution is two folded. First, we evaluate the macroeconomic impact of adopting the Euro measured by the real GDP. Theoretical predictions about this effect are ambiguous and depend on whether the costs outweigh the benefits of joining the Eurozone.¹ Indeed, we find that there are some mild losers (France, Germany, Italy, and Portugal) and a clear winner (Ireland). Such heterogeneous findings are in line with De Grauwe (2020) who warns us not to be overoptimistic on economic growth stemming from an inharmonious monetary union.

Next, we investigate which were the channels driving the output gains and losses and if they differed from country to country. For Ireland, the private consumption and investment notably explain almost 80% of the total output gain from joining the Euro. While for France and Portugal, the private consumption and the net exports accounted for a large share of the economic loss, in the case of Germany and Italy, the private consumption and the investment explain the negative impact of the Euro. For most countries, the trade volume was significantly higher than if they had not joined the Eurozone. Nonetheless, the common currency had a positive impact on the trade balance solely for Germany and Ireland.

Literature Review: This paper is related to two strands of the literature. The first one is directly related to the methodology used to construct the counterfactuals. To employ the Synthetic Control Method (SCM), we follow the original work by Abadie and Gardeazabal (2003), Abadie et al. (2010) and (2015) who developed the methodology.² Furthermore, we follow more recent work by Campos et al. (2018) who evaluate the impact of the European Union accession, Born et al. (2019) who assess the macroeconomic impact of the election of Donald Trump as the President of the USA, and finally Breinlich et al. (2020) and Born et al. (2020) who study the costs of economic nationalism by looking at the Brexit vote impact on the transactions and GDP, respectively.

In particular, this paper closely relates to the recent literature about the Euro adoption using the synthetic control approach (Fernández and Garcia-Perea (2015), Verstegen et al. (2017), and Gasparotti and Kullas (2019) among others). We build directly on the work of Puzzello and Gomis-Porqueras (2018) by extending their analysis from six to twelve member states which joined the Euro until 2007. Furthermore, we investigate the channels that drove the economic gains and losses of the accession.

This paper also contributes to the literature that studies the macroeconomic impact of joining a currency union. Starting from the groundbreaking contribution of Mundell (1961) on the theory of optimal currency areas, many economists have been studying the key characteristics that allow a group of countries to benefit from having the same currency. (McKinnon, 1963) and Kenen (1969) have added seminal contributions to this theory by

¹We refer the reader to Lane (2006) and Beetsma and Giuliodori (2010) who provide a more recent account of the real effects of the EMU by surveying the literature on its macroeconomic costs and benefits.

²A good overview of the literature using this methodology can be found in Abadie (2019).

exploring the role of international trade and diversified output structures in determining the costs and benefits of joining a monetary union.

More recent work from Alesina and Barro (2002) explains that forgoing monetary policy, on the one hand, implies losing a stabilization device to deal with domestic shocks but, on the other hand, can boost credibility and price stability. Alesina and Barro (2002) show that if there is a reduction in trading costs, the adoption of a common currency has a direct positive effect on trade, output, and consumption.³

There is also a broad literature which tests empirically these theoretical links. At about the time the euro was launched, Rose (2000) famously estimated that a currency union could boost up to three times bilateral trade. The relevance of these results to the euro case was immediately doubted since the sample used for the analysis was based on unions of small, poor, and remote countries. Micco et al. (2003) developed the first comprehensive study for the impact of the EMU on trade, concluding that the Euro had a positive impact not only on trade between member states but also with third parties.

Some papers have also highlighted the impact currency unions on investment. Among others, Barr, Breedon, and Miles (2003) suggest that inward investment in the countries outside the union would have been greater if they had joined the EMU. Furthermore, De Sousa and Lochard (2011) estimate that, in the Eurozone countries, investment increased with the single currency adoption.

The remainder of this paper is organized as follows. Section 2 describes the construction of the doppelganger, its implementation and the data used. Section 3 presents the results and performs robustness exercises. Section 4 explores the potential channels through which the Euro adoption affected the GDP. We briefly conclude in Section 5.

³Alesina, Barro, and Tenreyro (2002) and Barro and Tenreyro (2007) empirically investigate and present evidence in accordance with these theoretical predictions.

2 Constructing the doppelganger

2.1 The synthetic control method

To measure the impact of the European Monetary Union accession on the macroeconomic performance of the Eurozone countries, we construct a doppelganger for each Eurozone country based on the synthetic control methodology (SCM) developed by Abadie and Gardeazabal (2003), Abadie et al. (2010) and (2015).⁴ Ideally, these doppelgangers behave just like the Eurozone economies, except that they did not adopt the Euro.

The goal is to compute the treatment effect of a policy intervention:

$$\tau_{i,t} \equiv Y_{i,t}^I - Y_{i,t}^C$$

where $Y_{i,t}^I$ represents the realized outcome of country i in period t and $Y_{i,t}^C$ stands for the non-observable outcome country i would have had in period t absent from the policy intervention. Abadie and Gardeazabal (2003) proposed the SCM to estimate $Y_{i,t}^C$ by constructing a doppelganger as a weighted average of the outcomes of non-treated units. We refer to these units as "donor countries" and to the set of these countries as "donor pool" throughout the paper. Suppose that we have N+1 countries and country i=1 is exposed to the intervention of interest. Then, an unbiased estimate of the treatment effect, which we refer to as doppelganger gap throughout this paper, is:

$$\hat{\tau}_{1,t} = Y_{1,t}^I - \sum_{i=2}^{N+1} w_i Y_{i,t} \tag{1}$$

where w_i is the estimated weight assigned to donor country i used to construct the doppelganger.

The weights are chosen to minimize the difference between each treated unit and its doppelganger's pre-intervention outcome variable and predictors. The outcome variable studied is real GDP and the set of predictors used is based on Abadie and Gardeazabal (2003) and Born et al. (2019). These predictors are the average GDP shares of private consumption, government consumption, investment, exports, imports, the employed share of population, the labor productivity growth, the real GDP and its lags.⁵

Formally, we let $\mathbf{x_1}$ denote the (37×1) vector of 30 observations for real GDP and 7 covariates' averages in each Eurozone country and $\mathbf{X_0}$ denote a (37×14) matrix with observations from the donor countries. Finally, we let \mathbf{w} denote a (14×1) vector of weights

⁴A detailed exposition of the method can be find in Abadie (2019).

⁵We avoid the so-called cherry picking problem in Ferman et al. (2020) by choosing a standard set of predictors based on previous empirical literature.

 w_i , i = 2, ..., 15. Then, the optimal weighting scheme is defined by \mathbf{w}^* which minimizes the following mean squared error:

$$(\mathbf{x_1} - \mathbf{X_0}\mathbf{w})'\mathbf{V}(\mathbf{x_1} - \mathbf{X_0}\mathbf{w}) \tag{2}$$

subject to:

$$w_i > 0 \text{ for } i = 2, ..., 15$$
 (3)

$$\sum_{i=2}^{15} w_i = 1 \tag{4}$$

where V is a (14×14) symmetric and positive semidefinite weighting matrix assigning different relevance to the characteristics in \mathbf{x}_1 and \mathbf{X}_0 . Following Abadie and Gardeazabal (2003) and Abadie et al. (2010), we choose a diagonal V matrix such that the mean squared prediction error of the outcome variable (and the covariates) is minimized for the pre-treatment period.⁶

2.1.1 Implementation

The SCM offers several advantages to study the question at hands. This method is transparent regarding the construction of the counterfactual and the fit of the control unit to the treated unit. It provides the exact weigh of each donor country for the construction of the doppelganger. The fit of the counterfactual can be inspected by comparing the outcome variable and other characteristics of the treated unit with the estimated data. It is also important to highlight that this method allows the design decisions like choice of donor pool and predictors to be made regardless of the post-treatment considerations and without knowing the implication for the results and conclusions. Moreover, the SCM precludes extrapolation since the estimated weights are non-negative and sum to one.⁷

To successfully implement the SCM several contextual and data requirements should be satisfied.⁸ Especially for estimating causal effects, the credibility of the results severely depends on whether these requirements are met in the empirical application at hand. Therefore, we now present these requirements and how we address them.

First, treated units and the donor countries should be comparable. The counterfactual

⁶Including the covariates in the optimization differs from Kaul et al. (2018) who have raised concerns about including all pre-intervention outcomes together with covariates when using the SCM. The covariates used are relevant for the computation of the doppelgangers and its choice hinged on theoretical grounds.

⁷See King and Zeng (2006) for more information about the dangers of relying on extrapolation to estimate counterfactuals

⁸See Abadie (2019) for more detail on these requirements.

should be identical to the treated unit in all dimensions except for the treatment assignment. When the treated unit is a country, an "ideal" control unit rarely exists in observed data because countries differ widely across demographic, legislative and economic characteristics (Born et al., 2020). Yet, the donor pool selection should try to accommodate this need.

Unlike Puzzello and Gomis-Porqueras (2018) and Gasparotti and Kullas (2019), we ensure that the donor countries can resemble the level of economic and social development of the treated units by using only OECD economies in our baseline estimates. It is important to "restrict the donor pool to units with outcomes that are thought to be driven by the structural process as for the unit representing the case of interest" (Abadie et al., 2015).

When using developing countries with structurally higher growth rates to create a doppelganger for an advanced economy with structurally more modest growth rates, results are condemned to be biased. Using a smaller donor pool that guarantees more similarities with the treated unit should be preferred, albeit the expected poorer fit (Abadie & Gardeazabal, 2003).

Secondly, since the counterfactual weights are constructed according to the pre-intervention characteristics, we have to assure that there are no (external) differentiated shocks during the study period in the donor pool countries (Abadie (2019)). To account for this issue, contrarily to Fernández and Garcia-Perea (2015) and Verstegen et al. (2017), we only consider observations until 2007. From 2008 onward, the Great Recession affected countries in very different ways and arguably provoked structural changes in the affected economies.

It is also important to exclude any country that was treated from the donor pool. In this context, this is addressed by using only donor countries which never adopted the Euro.

Furthermore, policy interventions frequently have spillover effects to non-treated units. When employing the SCM, it is important to ensure that the *counterfactuals are not affected* by the treatment. In the context of our analysis, this is equivalent to ruling out the possibility that the Euro adoption by an individual country affected the income of the donor countries. This assumption is tested by performing in-space placebo testes in section 3.2.2.

Fourth, the intervention has no effect on the outcome before the implementation period. In section 3.2.1, potential anticipation effects are tested by changing the treatment date used in the analysis.

The SCM requires as well a sizable number of pre- and post-intervention periods. In the literature, previous SCM applications with yearly data use between 20 (Abadie & Gardeazabal, 2003) and 30 pre-treatment periods (Abadie et al., 2015). The reason is that the credibility of a synthetic control depends upon how well it tracks the treated unit's characteristics and outcomes over an extended period of time prior to the treatment. The post-treatment period should be long enough to account for delayed or dissipated effects of the intervention.

These requirements are satisfied with the data used in analysis as discussed next in section.

Finally, it is important to guarantee that there are no extreme values in the variable of interest for the treated units. The SCM is based on the idea that a combination of unaffected units can approximate the pre-intervention characteristics of the affected unit. However, if the treated unit exhibits "extreme" values for the outcome variables this is not possible. We address this issue by normalizing real GDP to unity in 1970.

2.2 Data and Sample

We use annual data since 1970 until 2007 from the Penn World Tables, version 9.1 (PWT 9.1 - Feenstra, Inklaar, and Timmer (2015)) and the World Bank. We focus on the real GDP as our main outcome variable and conduct our analysis on twelve Eurozone countries, namely Austria, Belgium, Finland, France, Germany, Greece, Ireland, Italy, Luxembourg, the Netherlands, Portugal, and Spain.⁹

We assume the treatment date takes place in 1999 for all countries except for Greece, which joined the Eurozone later in 2001. In our baseline estimate, we have at least 29 pre-intervention periods, from 1970 to 1998, which is sufficiently large to apply the SCM.

Doppelgangers are constructed on the basis of a donor pool of 14 countries selected as follows. First, only OECD countries are used to ensure that doppelgangers are sufficiently similar to the treated countries. Then, all countries that joined the European Union or the Eurozone during the post-treatment period are excluded. This guarantees that the donor countries are neither affected by the treatment nor suffer a differentiated external shock during the post-treatment period.

For our baseline estimates, we do not restrict the donor pool further except for countries for which the necessary data is not available. The pool is composed of Australia, Canada, Chile, Denmark, Iceland, Israel, Korea, Mexico, New Zealand, Norway, Sweden, Switzerland, the United Kingdom, and the United States.

We believe this donor pool of countries is just narrow enough to guarantee that these donor countries are comparable to the treated units but do not compromise the application of the SCM and estimation of the counterfactuals. Possible flukes to this belief are assessed in section 3.3.1 where we perform robustness checks by excluding individual and groups of countries from the donor pool.

⁹The outcome variable is normalized to unity in 1970 in each country. Consult Table 2 for further details on the data. Focusing on the normalized *per capita* real GDP instead does not change the results.

3 Empirical Results

This section starts by presenting the baseline results for the impact of the euro accession. Next, taking into account the assumptions discussed in Section 2.1.1, we discuss the statistically significance and causality of these results by performing two types of placebo exercises. First, we apply in-space placebo tests in Section 3.2.2 which assign the treatment to all countries in the donor pool. Then, in Section 3.2.1, we perform in-time placebo tests in which placebo treatment dates are assigned to the treated countries.

The main findings in this section corroborate the results of Puzzello and Gomis-Porqueras (2018) and adds new insights by concluding that the results for France, Germany, Ireland, Italy, and Portugal are statistically significant.

3.1 Baseline results: Assessing Euro's macroeconomic impact

It is expected that the SCM yields an imperfect pre-treatment match for some countries given that our procedure determines 14 parameters (country weights) to match 37 observations. Notwithstanding, this methodology can provide substantial improvement relative to alternative methods as differences-in-differences (Ferman & Pinto, 2019) and thus, we are confident that this data-driven approach is the best to study the problem at hands.

Table 4 displays the donor country weights that constitute each doppelganger. For instance, the synthetic Spain is composed by all countries in the donor pool yet being significantly constructed using data from the United States (46%), Mexico (19%), Switzerland (18%), and Australia (16%). We are overall confident on the plausibility and credibility of the methodology weighting scheme.¹⁰

Table 3 shows that doppelgangers are very similar to the actual countries when comparing their predictors means despite using the same specification for all countries.¹¹ Furthermore, in section A.7, we show that the doppelgangers are successful in recovering the time path of all GDP components for most of the analyzed countries.

Figure 1 displays the real GDP for each country (full black line) and doppelganger (dashed blue line) presented as the deviation from the first year of the sample in percent. The shaded area represents two standard deviations of the pre-treatment difference between the actual and the counterfactual series. When the doppelganger series deviates from the realized path in such a way that exceeds these bounds, it indicates that such deviation is non-standard compared to the pre-Euro period.

¹⁰Potential concerns regarding the use of countries which belong to the Exchange Rate Mechanism (Denmark, Sweden and United Kingdom) are addressed in Section 3.3.1.

¹¹Matching only the key variable might suffice but having further similarities in related variables is also important and ensures the robustness of the findings (Botosaru & Ferman, 2019).

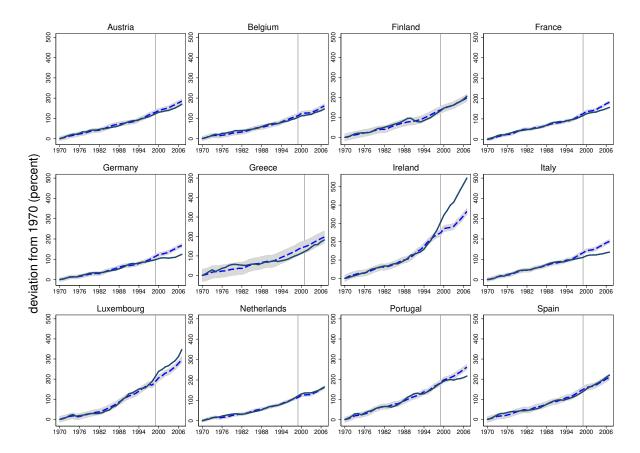


Figure 1: The impact of the Eurozone accession

Notes: In each graph, the dashed line represents the normalized real GDP for the synthetic country and the continuous line represents the series for the actual country. The vertical line represents the treatment period - 1999 for all countries except for Greece which is 2001. For each country the analysis starts in 1970 and ends in 2007.

A number of observations stand out. The pre-treatment paths for most countries and their doppelgangers are overlapping. In fact, Figure 1 shows some series embarking on a different growth trajectory relative to their counterfactuals only around the Eurozone creation.

Table 1 presents the exact doppelganger gaps measured in Euro *per capita*. Ireland benefited the most from the Euro adoption. Its GDP *per capita* was 10,781 Euro higher due to the common currency adoption. However, France, Germany, Italy and Portugal would be better off by not participating in this currency union. Yet, Germany and Italy lost the most - 5,788 and 6,089 Euro *per capita* respectively.

Table 1: Doppelganger Gap

	AUT	BEL	FIN	FRA	DEU	GRC	IRL	ITA	LUX	NLD	PRT	ESP
Gap (Euros per capita)	-1,712	-1,668	777	-2,632	-5,788	-1,098	10,781	-6,089	4,697	168	-2,558	900

Notes: This table presents the doppelganger output gap *per capita* in 2007. This measure is obtained by adjusting the real GDP gap for the population size and converting 2011 US dollars into 2011 Euro. We use the conversion rate available from the PWT 9.1 for this year (≈ 0.73).

3.2 Causality

A key assumption to study the impact of a policy intervention is that there is no reverse causality. In our context, this means that countries must not have adopted the Euro due to economic considerations. This assumption is plausible because the Eurozone accession was driven mainly by political rather than economic factors (Eichengreen and Frieden (1993), Feldstein (1997)). In fact, by not satisfying the requirements of an Optimum Currency Area, many economists believed that countries adopting the Euro would face economic losses (Jonung et al., 2009). This argument holds even for the Greek case which had decided to join the Euro before the single currency was a reality.¹²

Sections 2.1.1 and 2.2 discussed the conditions under which the SCM provides suitable estimates of causal effects and this paper address some of these requirements. To further back the notion that the doppelganger gap is indeed caused by the referendum shock, Sections 3.2.1 and 3.2.2 provides a number of placebo experiments and robustness checks. The basic idea of the placebos is very intuitive. We can be confident that the synthetic control estimator captures the causal effect of an intervention as long as similar magnitudes are not estimated in cases where the intervention did not take place (Born et al., 2020). Finally, Section discusses the statistical significance of the results 3.2.3.

3.2.1 In-time placebo test: anticipation effects

On 7 February 1992, representatives from twelve countries signed the Maastricht Treaty – Belgium, Denmark, France, Germany, Greece, Ireland, Italy, Luxembourg, Netherlands, Portugal, Spain and the United Kingdom. Upon signing the it, these member countries knew that a monetary union, with a central banking system and common currency, was to be created within the next years. It is therefore reasonable to think that countries experienced at least partly the Eurozone accession even before the Euro was launched.

¹²According to the 1998 convergence report from the European Commission, Greece did not join the single currency in 1999 because it had not fulfil any of the four convergence criteria. Notwithstanding, the decision of joining was already made.

To check for anticipation effects of the Euro adoption, we perform in-time placebo tests by inspecting different intervention periods in our analysis. We take the date the Maastricht Treaty was signed as the placebo treatment period. Figure 2 suggests that the main conclusions from Figure 1 remain unchanged.

We ran further time-placebo tests in which the placebo treatment date is set artificially to be every year since 1992 until 1998. For the sake of brevity, besides the Maastricht Treaty date 1992, we only report the tests for 1995 and 1998 in Figures 8 and 9. Reassuringly, the results remain unaltered.¹³

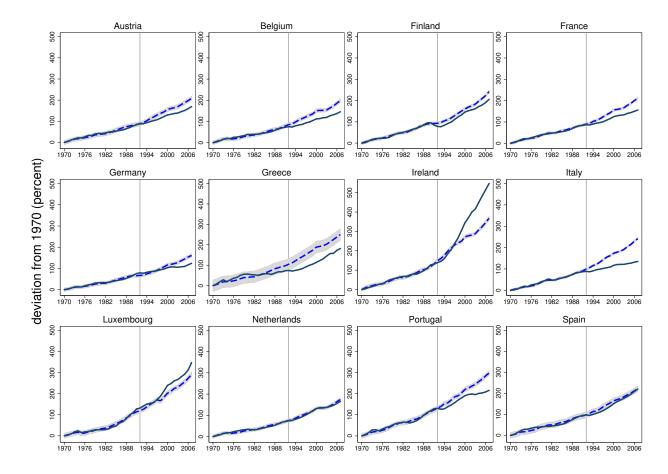


Figure 2: In-time placebo tests

Notes: In each graph, the blue dashed line represents the normalized real GDP for the synthetic country and the black full line represents the series for the actual country. The vertical line depicts the placebo treatment period - 1992 for all countries. For all countries the analysis starts in 1970 and ends in 2007.

Yet, Figure 2 presents some evidence in favour of the existence of anticipation effects for several countries. In some cases, the gap between the actual and the synthetic series becomes

¹³The remaining figures can be provided by the authors upon request.

wider than the one analyzed in Figure 1. Still, these figures should be analysed carefully. For example, the plot for Greece show some effect that starts even before the placebo treatment of 1992. This anticipated effect seems to be driven by the Greek EU accession in 1981 and not from the Euro.

Furthermore, estimates for Austria and Finland must be interpreted carefully as these countries entered the EU exactly in 1995 and the placebo treatment effect may be biased. The doppelganger gaps for Belgium, France and Italy are wider than in Figure 1 but the direction of the effect remains unchanged. Ignoring possible anticipation effects in our baseline estimate may lead to a lower bound estimate of the Euro impact for these countries.

The absence of anticipation effects for the remaining countries may be due to two things. First, the key event representing a change for most European citizens was the irrevocably the exchange rate fix in 31st December 1998 and the Euro launch in the 1st of January 1999. Second, most of these countries had already experienced trade and economic gains from joining the European Union (Campos et al., 2018). Therefore, such effects lie in our pre-treatment sample and thus, are being already considered.

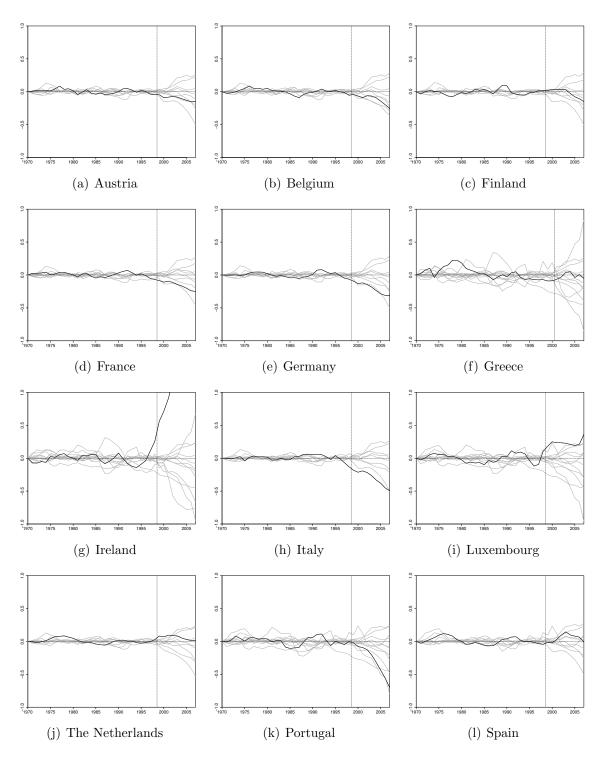
3.2.2 In-space placebo test

Following Abadie et al. (2010), Abadie and Cattaneo (2018), and Firpo and Possebom (2018), we employ the synthetic control methodology on the donor pool countries while exposing them to a placebo treatment in 1999. The idea is to sequentially "re-assign" the treatment to all units in the donor pool and, for each of them, estimate a fictitious doppelganger using the remaining donor countries and the originally treated unit. We repeat this process for every treated country.

Next, we compare the post and pre-treatment behavior of these series and inspect the differences between treated and fictionally treated units. If our benchmark estimates for each Eurozone country are picking up the causal effect of the Euro accession, these should dominate any possible impact of the fictitious event in the donor countries. On the other hand, if no difference is found, then most likely the actual intervention had no effect. Applying this idea to each country in the donor pool allows us to compare the estimated effect of the Euro accession on Eurozone countries to the distribution of placebo effects obtained for the other countries (Abadie et al., 2015).

The plots from Figure 3 depict the doppelganger gap, that is, the differences between each countries' normalized GDP and their doppelgangers' estimates. The smaller the gap for the pre-treatment period, the better the fit of the synthetic series to the outcome variable. Countries with a bad pre-intervention fit are excluded from the in-space placebo test because

Figure 3: In-space placebo tests



Notes: The plotted lines represent the prediction error for the treated country (black) and donor countries (grey) for which we impose a fictitious Euro accession. Following the literature, we plot the donor countries whose pre-treatment MSPE was four times larger than the one of the treated country.

they are not suited to inform about the post-treatment effect.¹⁴

Visually, Figure 3 reinforces the findings in Figure 1. When comparing the full black lines from each Euro-adopter country to the grey lines of fictitious treated units, it is clear that, for some countries, the post-treatment gap is unusually bigger. Specifically, it suggests a positive impact of the Euro accession on Ireland and Luxembourg and a negative impact, if any, on France, Germany, Italy, and Portugal.

3.2.3 Statistical Significance

To evaluate the statistical significance of our estimates and following Abadie et al. (2010) and Abadie and Cattaneo (2018), we use a test based on the classic framework for permutation inference which builds on the computations presented in the previous section.

Given our estimates of all fictional treatment effects in the previous section, we can evaluate the statistical significance by computing a p-value associated with the treatment. First, we compute the ratio of mean squared prediction errors in the post-intervention period relative to the pre-intervention period for treated and fictitiously treated units as follows:

$$\chi \equiv \frac{RMSPE_{post}}{RMSPE_{pre}} \equiv \frac{\sqrt{\frac{1}{T-T_0+1} \sum_{t=T_0}^{T} (x_{1,t} - \mathbf{x_{0,t}w})^2}}{\sqrt{\frac{1}{T_0-1} \sum_{t=1}^{T_0-1} (x_{1,t} - \mathbf{x_{0,t}w})^2}}$$
(5)

where $x_{1,t}$ denotes the GDP of the treated country at period t, $\mathbf{x_{0,t}}$ denotes a vector of observations of GDP for the donor countries in period t, \mathbf{w} denotes a vector of weights for the donor countries, T denotes the total number of periods, and T_0 denotes the treatment date.

This statistic already allows a quantitative analysis of the treatment effect taking into account the quality of the match produced by the SCM. A small pre-treatment RMSPE implies a good fit of the synthetic series to the actual series and a large post-treatment RMSPE suggests, for the treated units, a large intervention impact. Therefore, obtaining a larger ratio for the treated unit than for the placebo treated units would entail a significant treatment effect.¹⁵

Figure 4 depicts this relative measure for the Eurozone countries and its donors. Ireland

¹⁴We define a good pre-intervention fit following Firpo and Possebom (2018) when the pre-intervention MSPE of a donor country is at most four times greater than the Eurozone country's pre-intervention MSPE being analyzed.

¹⁵A large post-intervention RMSPE *per se* is not indicative of a large effect of the intervention. It depends on whether the synthetic control can reproduce closely the outcome of interest prior to the intervention (Abadie, 2019).

clearly stands out as the country with the highest RMSPE ratio with a post-intervention gap about 16 times larger than its pre-intervention gap.

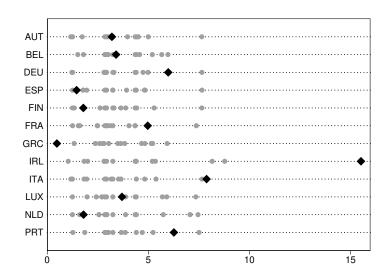


Figure 4: Ratio between the post- and pre-treatment RMSPE

Notes: The plots show the ratio between the post- and the pre-intervention RMSPE for the treated units (in black) and all donor countries (in grey).

We deem the effect of the Euro adoption significant if the estimated effect for the treated units is unusually large relative to the distribution of the placebo effects. To test this in practise, we compute a p-value which compares the value of the RMSPE for the treated country to that of all other units as follows:

$$\rho_1 = \frac{\sum_{i=1}^{N+1} I(\chi_i \ge \chi_1)}{N+1} \tag{6}$$

where I(.) denotes the indicator function, N the number of donor countries, χ_1 the RMSPE ratio for the treated unit and χ_i is the RMSPE ratio for country i which can be a donor or the treated country.

Table 5 presents the RMSPE ratios, (χ) , for all countries in the baseline analysis and correspondent p-values for the treated units. If one were to pick a country at random from the Irish sample, the chances of obtaining a ratio as high as the Irish would be 1/15 = 0.067 which we consider to be statistically significant. A closer look at Table 5 and Figure 4 shows that Ireland, Italy, Portugal, Germany, and France experienced a significant effect from adopting the Euro.

3.3 Further robustness checks

3.3.1 Changes to donor pool

This section addresses two concerns. On the one hand, some countries in the control group may have opted out of the treatment. This would suggest a reverse causality problem and raise doubts about the credibility of the results presented. As discussed in the beginning of Section 3.2, countries used in the analysis must not have opted in or out due to economic considerations.

In fact, the UK, Sweden and Denmark belonged to the European Union at the time but did not adopt the common currency. Even though, they did opted out due to political reasons, we still address this issue by excluding these countries altogether from the donor pool. We redo our analysis with this new pool and the results are presented in Figure 5. The main conclusions remain unchanged for all the Eurozone countries. With special attention for the ones whose doppelgangers' construction highly relied on this trio - Austria, Belgium, France, Germany, Italy, and Portugal.

On the other hand, there might have been spillover effects of the treatment onto the donor countries. We address this issue by iteratively re-estimating our baseline estimate for each Eurozone country excluding in each iteration one of the countries with positive weight.

We display this robustness check for countries that contribute with, at least, 10% for the construction of, at least, 2 countries' doppelgangers in Section A.6. This exercise shows that no particular donor country is driving the main conclusions. ¹⁶ So, it is unlikely that spillover effects nor one specific country in the donor pool are driving the results.

3.3.2 Changes in the sample period

The credibility of the SCM results severely depends on whether the requirements specified on Section 2.1.1 are satisfied. One of these requirements was that countries should not experience differentiated shocks during the sample period. Several analysed countries joined the European Union during the pre-treatment period which may concern the most attentive reader.

For countries which joined the EU at least 10 years before the Eurozone creation, we re-do the estimates using their accession data as the start of our sample, i.e. Ireland (1973), Greece (1981), Portugal and Spain (1986). From Figure 6, it is possible to conclude that the conclusions from the baseline analysis are robust.

Unfortunately, it is not possible to re-estimate the results for Austria and Finland because

¹⁶For the sake of brevity, the figures for all the remaining countries are not reported, but they can be provided by the authors upon request. The results remain qualitatively unchanged.

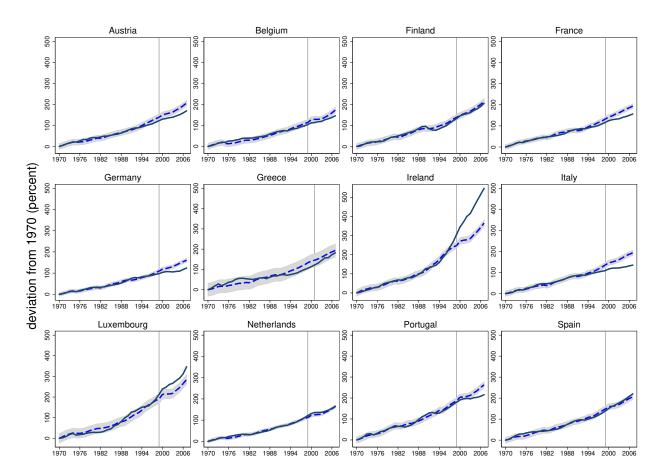


Figure 5: Impact of the Euro accession with a change in the donor pool

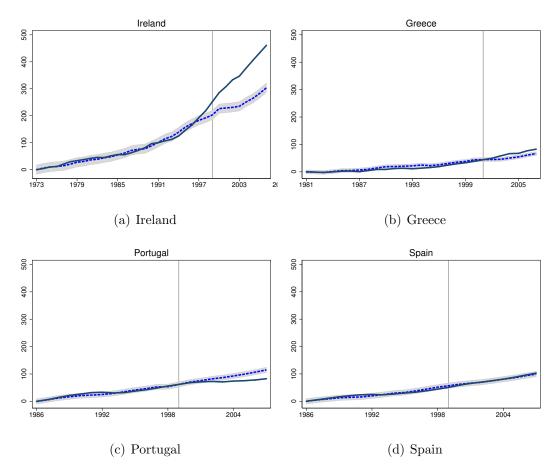
Notes: In each graph, the blue dashed line represents the normalized real GDP for the synthetic country and the black full line represents the series for the actual country. The vertical line depicts the treatment period - 2001 for Greece and 1999 for the remaining countries. For all countries the analysis starts in 1970 and ends in 2007. Relative to the baseline analysis, the donor pool now excludes Denmark, Sweden and the United Kingdom. Table 6 show the weights used to construct these results.

they joined the EU in 1995. Yet, according Campos et al. (2018), Austria and Finland were not significantly affected by the EU accession and thus we believe that this does not pose a problem in our analysis.

4 What drives the doppelganger gap?

In this section, we take one step further and investigate what drives the results presented in Figure 1 by decomposing the Euro accession response of GDP into the response of its components. First, we look at the evolution of each GDP component for both countries and

Figure 6: The impact of the Euro accession with a change in the sample period



Notes: In each graph, the blue dashed line represents the normalized real GDP for the synthetic country and the black full line represents the series for the actual country. The vertical line depicts the treatment period - 1999 for Ireland, Greece, and Portugal and 2001 for Greece. The analysis starts in 1973 for Ireland, 1981 for Greece, and 1986 for Portugal and Spain. For all countries the analysis ends in 2007.

corresponding counterfactuals. Then, we try to understand what explains the output gains and losses from the accession by accounting the role of each component for the doppelganger GDP gap. We find that countries were affected by the event through different channels.

4.1 The impact of the Eurozone accession on GDP components

In Appendix A.7, we present the GDP decomposition for all countries and doppelgangers. The synthetic shares of GDP components were constructed using the weights estimated in Section 2 and the data from the donor countries. Similarly to the construction of the synthetic GDP series, we now compute the synthetic shares of each GDP component as a weighted average of the shares of GDP components for the donor countries. Namely, we

obtain GDP shares of private consumption, investment, government consumption, exports, and imports.

We aim at understanding how the components evolved over time and, in particular, after the Euro accession. To do that, we use each component share and the GDP series to compute the five GDP components series for both countries and doppelgangers. Finally, these series are normalized to unity in 1970. The results are presented in Appendix A.7.

It is important to highlight that comparing the actual and the synthetic series from Appendix A.7 also indicates whether the doppelganger can really mimic the behaviour of each country prior to the Euro accession. We must recall that the construction of the doppelganger in Section 2 does not target the time path of GDP components and thus, a good fit in this regard can not be taken for granted (Born et al., 2020). Overall, the figures from Appendix A.7 reassure us of the fit of the SCM estimates.

With respect to the clear winner, Ireland, Figure 22 shows that all its GDP components increase consistently since the beginning of the sample and a gap between the actual and synthetic series materializes around the Euro accession date. In particular, investment and government consumption deviate significantly from their counterfactuals. The Eurozone accession especially stimulated their growth.

Even though Figure 1 and Table 5 display significant negative treatment effect on the economic growth of France, Germany, Italy, and Portugal, these countries seem to have been affected through different channels. Figures 20 and 23 indicate that private consumption and investment would have grown more if these country had not adopted the Euro. On the other hand, the treatment output loss in France and Portugal seems to be driven by a smaller net exports growth after the accession, as Figures 19 and 26 suggest.

The remaining countries, not significantly affected by the treatment, do not exhibit large differences in the GDP components' growth relative to their counterfactuals. Appendix A.7 also depicts a larger acceleration in the trade volume driven by the Eurozone accession for almost all analyzed countries. This finding is in accordance with empirical works departing from Rose (2000), Glick and Rose (2002), and Frankel and Rose (2002) who argue that countries sharing the same currency trade two to three times as much as they would with different currencies. Moreover, regarding trade with third parties, Micco et al. (2003) explains that the monetary union increases trade not just with EMU countries, but also with the rest of the world. Even though the treatment seems to have affected both imports and exports, we corroborate the negative impact on the growth of net exports in most cases as documented in Hope (2016). The common currency had solely a positive impact on trade balance growth for Germany and Ireland.

4.2 What explains the cumulative doppelganger gap?

The previous section discussed the growth patterns of the GDP components for the countries and their counterfactuals. In this Section, we compute the contribution of each GDP component for the output gap generated by the treatment. In section 2.1, equation 1 defines doppelganger gap as the difference in outcome variable (here, real GDP) between the treated and the synthetic country. The cumulative treatment effect can be estimated by computing the doppelganger gap for t = 2007, the last year of our analysis.

Now, we proceed to quantify the contribution of each component for the output doppelganger gap in four steps. Analogously to Equation 1 for output, we start by computing doppelganger gaps for each component. Then, we compute the relative weight of each component z on the output doppelganger gap in the following way:

weight of
$$z_{c,t} = \frac{z_{c,t} - z_{c,t}^{dop}}{GDP_{c,t} - GDP_{c,t}^{dop}}$$
 (7)

where z is either private consumption, government consumption, investment, exports, or imports, the subscript c stands for one of the twelve treated countries, and the subscript t represents the time period. Thereafter, we calculate the **percent** doppelganger gap for GDP as follows:

percent output doppelganger gap
$$_{c,t} = \frac{GDP_{c,t} - GDP_{c,t}^{dop}}{GDP_{c,t}^{dop}}$$
 (8)

Showing the treatment effect in percent terms allows a direct interpretation of how much larger/smaller the GDP is due to the Euro accession. Finally, we multiply the relative weight of each doppelganger gap $z_{c,t}$ by the percent output doppelganger gap. This allows us to understand the direct contribution of each channel on the treatment effect.

Figure 7 depicts, for each country, the percent GDP gap in 2007 and its decomposition into private consumption, government consumption, investment and net exports. It clearly shows that countries experienced the Euro accession heterogeneously. ¹⁷

We now take a deeper look into the countries which were significantly affected as argued in Section 3.2.

For Ireland, joining the Euro area induced a 39% higher GDP compared to the counterfactual scenario of not adopting the common currency. Even though all GDP components contribute positively for this result, the private consumption and investment together can explain almost 80% of the total output gain from the treatment.

¹⁷See Table 7 for the exact values depicted in the Figure 7.

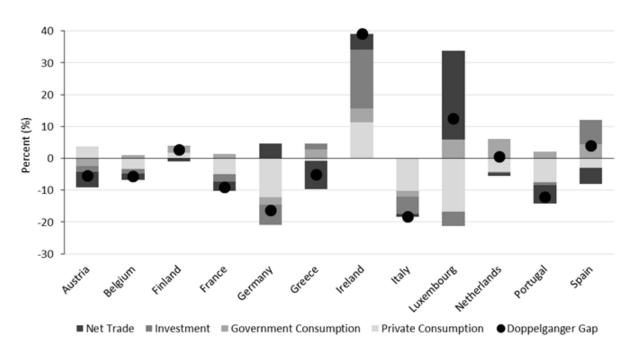


Figure 7: Doppelganger gaps and GDP components

Notes: The black dot depicts, for each country, the percent doppelganger gap of output computed as in Equation 8. The stacked bars represent the contribution of each GDP component for these gaps. For each component, we multiply the weight of the component (Equation 7) by the percent doppelganger gap of GDP (Equation 8). The sum of the values for each component sum up the percent doppelganger gap. The values represent the cumulative effect of the Euro accession since they are computed for 2007, the last year of the analysis.

Table 7 also shows that the reasons behind the economic slowdown experienced by some countries at the Euro accession differ from country to country. We find that for France and Portugal, the private consumption and the net exports accounted for a large share of the GDP gap. For Germany and Italy, it is the doppelganger difference in private consumption alongside with investment that better explain the negative economic impact of the Euro on GDP.

Before the Euro, the need to exchange local currencies implied extra transaction costs and exchange rate risk. The single currency was expected to boost cross-border trade and investment between the member states since doing business in the euro area would be more cost efficient and less risky (De Grauwe (2020)). For third parties, the Euro area would be as well an attractive place to invest. Consumers would benefit from price transparency and stability. Therefore, roughly, it would be expected an increase in investment, exports, and imports but it is not clear in which direction the trade balance would change.

Table 8 indeed reveals that, with the exception of France and Italy, all countries had

a higher trade volume than if they had not adopted the common currency. This result is in accordance to Baldwin et al. (2008) and Schmitz and Von Hagen (2011) who argue that the Euro has significantly promoted trade in the Eurozone countries. Yet, net exports changed differently across countries. Only Germany and Ireland experienced significant net trade benefits from the Euro accession. Similarly, even though the common currency was expected to attract foreign investment for the whole Euro area, Ireland stands out from the remaining member countries. Investment in Ireland increased significantly because of the Euro adoption. Therefore, country specific characteristics have significantly shaped the impact Euro across member states.

5 Conclusion

In this paper we study the impact of the European monetary union accession on the macroeconomic performance of the first twelve member states. We use the synthetic control method to construct a counterfactual of these countries' GDP. This method allows building a doppelganger which should represent the economic activity of these countries in the absence of the Euro adoption.

Our findings suggest that there are mild losers (France, Germany, Italy, and Portugal) and a clear winner (Ireland). Notwithstanding, the drivers of such estimates are heterogeneous as our GDP decomposition analysis indicates. Even though trade has increased substantially with the adoption of a common currency, only Germany and Ireland benefited in net terms. Moreover, the common currency adoption especially boosted investment in Ireland.

These evidence points out to the importance of analyzing in detail the heterogeneous responses of GDP components and their implications. For example, given the different responses of investment and government consumption across member states, it is natural to ask if the effectiveness of national fiscal policies has changed. This could be especially interesting given that, with the common currency adoption, countries forgo some important policy instruments.

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A Appendix

A.1 Variables description and source

Table 2: Variables' Description

Variable code	Description	Source
rgdpna	Real gross domestic product at constant 2011 na-	PWT 9.1
	tional prices in million 2011 US dollars normalized	
	to unity in 1970.	
emp	Total employment - number of persons engaged in	PWT 9.1
	millions.	
$\operatorname{csh_prod}$	Labor productivity growth computed by taking	PWT 9.1
	the log-difference between real gdp and total em-	
	ployment	
pop	Total population in millions.	PWT 9.1
csh_emp	Employment share - ratio between total employ-	PWT 9.1
	ment and total population	
csh_c	Private consumption expenditure (% of GDP) ob-	World Bank
	tained by subtracting general government final	
	consumption expenditure to the series of final con-	
	sumption expenditure	
csh_g	General government final consumption expendi-	World Bank
	ture ($\%$ of GDP)	
csh_i	Gross fixed capital formation (% of GDP)	World Bank
csh_x	Exports of goods and services (% of GDP)	World Bank
csh_m	Imports of goods and services (% of GDP)	World Bank

Notes: All variables collected directly from the Penn World Table are from version 9.1 (PWT 9.1) (Feenstra et al., 2015). All level variables are in real terms and at annual frequency spanning the year 1970 until 2007. GDP components were collected from the World Bank database in shares of GDP. The data were collected on the 30-10-2019.

A.2 Comparison Tables

Table 3: Predictors' means (in %) for each country during pre-treatment period

Variable Names	Country	Doppelganger			
Austria					
Share of Priv. Consumption	56.27	51.57			
Share of Gov. Consumption	17.91	20.58			
Share of Investment	26.53	26.27			
Share of Imports	32.57	32.35			
Share of Exports	31.85	33.93			
Employment Share	45.16	48.33			
Labor productivity growth	2.47	2.07			
Belgium					
Share of Priv. Consumption	54.14	54.15			
Share of Gov. Consumption	21.41	20.59			
Share of Investment	22.87	24.74			
Share of Imports	53.80	36.03			
Share of Exports	55.39	36.55			
Employment Share	38.01	49.41			
Labor productivity growth	2.33	1.57			
Finland					
Share of Priv. Consumption	52.51	52.30			
Share of Gov. Consumption	19.38	19.96			
Share of Investment	26.66	26.54			
Share of Imports	26.57	26.66			
Share of Exports	28.02	27.86			
Employment Share	47.26	47.22			
Labor productivity growth	3.08	1.87			
France					
Share of Priv. Consumption	55.16	56.10			
Share of Gov. Consumption	21.24	19.47			
Share of Investment	23.11	23.42			
Share of Imports	20.54	21.50			
Share of Exports	21.03	22.51			
Employment Share	40.49	43.80			
r J	-01-0				

Continued on next page...

... table 3 continued

Variable Names	Country	Doppelganger			
Germany					
Share of Priv. Consumption	57.15	57.29			
Share of Gov. Consumption	19.68	18.18			
Share of Investment	24.50	24.30			
Share of Imports	21.40	21.25			
Share of Exports	20.06	21.48			
Employment Share	48.36	48.23			
Labor productivity growth	2.41	1.48			
Greece					
Share of Priv. Consumption	63.06	60.84			
Share of Gov. Consumption	16.75	17.21			
Share of Investment	26.46	22.76			
Share of Imports	21.77	17.88			
Share of Exports	15.50	17.07			
Employment Share	38.77	45.51			
Labor productivity growth	1.69	1.21			
Ireland					
Share of Priv. Consumption	61.17	55.50			
Share of Gov. Consumption	18.72	24.34			
Share of Investment	21.65	24.24			
Share of Imports	52.85	34.67			
Share of Exports	51.30	30.59			
Employment Share	35.36	39.90			
Labor productivity growth	3.43	2.14			
Italy					
Share of Priv. Consumption	59.00	58.72			
Share of Gov. Consumption	17.55	17.63			
Share of Investment	22.87	23.15			
Share of Imports	19.28	19.37			
Share of Exports	19.86	19.86			
Employment Share	38.25	44.11			
Labor productivity growth	2.48	1.24			
Luxembourg					

Continued on next page...

... table 3 continued

Variable Names	Country	Doppelganger		
Share of Priv. Consumption	48.16	58.05		
Share of Gov. Consumption	15.33	10.82		
Share of Investment	22.57	29.26		
Share of Imports	81.83	39.32		
Share of Exports	95.78	41.20		
Employment Share	46.63	51.00		
Labor productivity growth	1.81	1.54		
The Netherlands				
Share of Priv. Consumption	50.90	54.78		
Share of Gov. Consumption	21.94	18.40		
Share of Investment	22.62	26.16		
Share of Imports	47.47	37.62		
Share of Exports	52.01	38.27		
Employment Share	43.78	49.70		
Labor productivity growth	1.56	1.48		
Portugal				
Share of Priv. Consumption	66.93	65.45		
Share of Gov. Consumption	14.03	13.91		
Share of Investment	26.93	21.36		
Share of Imports	31.16	21.84		
Share of Exports	23.27	21.12		
Employment Share	41.81	38.57		
Labor productivity growth	2.02	1.29		
Spain				
Share of Priv. Consumption	63.12	61.46		
Share of Gov. Consumption	14.51	14.58		
Share of Investment	23.78	23.82		
Share of Imports	18.55	17.34		
Share of Exports	17.13	17.47		
Employment Share	35.38	43.58		
Labor productivity growth	2.56	1.10		

Notes: Predictors' means for each country during the pre-treatment period. All numbers are in percent. Variables definitions can be find in Table 2.

A.3 Weights Table

Table 4: Composition of each doppel ganger: country weights (in %)

	Austria	Belgium	Finland	France	Germany	Greece	Ireland	Italy	Luxembourg	Netherlands	Portugal	Spain
Australia	< 0.1	< 0.1	26.6	< 0.1	0.1	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	15.6
Canada	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1
Chile	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	22.0	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1
Denmark	< 0.1	46.5	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	26.4	< 0.1	< 0.1
Iceland	< 0.1	< 0.1	< 0.1	< 0.1	0.2	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1
Israel	< 0.1	13.2	< 0.1	< 0.1	0.9	3.1	50.1	< 0.1	< 0.1	13.5	< 0.1	2.0
Korea	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	14.0	< 0.1	4.0	< 0.1
Mexico	< 0.1	< 0.1	< 0.1	29.2	< 0.1	< 0.1	< 0.1	22.4	< 0.1	< 0.1	43.0	18.5
New Zealand	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	42.1	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1
Norway	48.2	7.2	33.9	< 0.1	< 0.1	< 0.1	27.9	< 0.1	0.2	13.0	< 0.1	< 0.1
Sweden	33.9	< 0.1	29.0	57.8	31.8	< 0.1	< 0.1	34.9	< 0.1	< 0.1	< 0.1	< 0.1
Switzerland	7.2	33.1	4.2	< 0.1	17.4	< 0.1	< 0.1	5.6	85.8	47.1	5.1	17.6
United Kingdom	10.7	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	4.0	< 0.1	< 0.1	47.9	< 0.1
United States	< 0.1	< 0.1	6.2	13.0	49.6	54.8	< 0.1	33.1	< 0.1	< 0.1	< 0.1	46.2

A.4 RMSPE

Table 5: Relative root mean squared prediction error of the pre- and post- treatment doppelganger gaps.

·	AUS	CAN	CHL	DNK	ISL	ISR	KOR	MEX	NZL	NOR	SWE	CHE	GBR	USA	Treated	P -Value (ρ)
AUT	2.98	5.00	2.85	4.50	1.73	1.25	2.93	4.35	3.10	7.64	4.41	3.24	1.16	3.97	3.20	0.533
BEL	2.86	4.57	2.85	5.97	3.01	1.79	2.93	4.35	3.28	5.66	4.41	3.24	1.51	5.19	3.40	0.467
FIN	2.61	5.28	2.85	4.42	2.89	1.25	2.93	4.35	3.30	7.65	3.64	3.24	1.33	3.88	1.78	0.867
FRA	2.99	2.49	2.85	4.34	1.61	1.25	2.93	4.35	3.14	7.37	2.47	3.24	1.54	4.04	4.97	0.133
DEU	2.80	4.74	2.85	4.98	2.88	1.25	2.93	4.35	3.30	7.65	4.40	3.24	1.22	2.86	5.98	0.133
GRC	2.81	3.83	2.59	4.83	3.49	2.38	2.74	5.15	3.63	5.93	4.66	3.04	1.33	5.21	0.48	1.000
IRL	2.81	5.34	1.04	4.37	2.00	8.16	2.93	4.35	2.81	8.77	4.40	3.24	1.82	5.19	15.51	0.067
ITA	1.95	3.65	2.85	4.82	1.82	1.25	2.93	3.52	3.16	7.65	4.40	3.24	1.17	5.38	7.88	0.067
LUX	2.70	5.69	2.85	5.90	1.97	1.25	2.93	4.35	2.71	7.35	4.40	3.24	2.43	3.88	3.69	0.467
NLD	2.95	5.74	2.85	7.06	2.56	1.25	2.93	4.35	2.55	7.45	4.41	3.24	1.61	3.87	1.79	0.867
PRT	3.02	5.23	2.85	4.70	3.01	1.27	2.93	3.66	2.88	7.50	4.40	3.24	1.85	3.86	6.26	0.133
ESP	1.95	4.84	2.85	4.81	1.77	1.25	2.93	4.35	3.24	7.65	4.40	3.24	1.77	3.92	1.45	0.933

Notes: The column Treated displays the RMSPE ratio for each country, χ in equation 5. The column P-Value tell us the chances of obtaining a ratio as high as the treated country if one were to pick a country at random from the sample including also the treated country, ρ in equation 6. Given the small number of donor countries, we consider the results significant if there at most 2 countries with a higher RMSPE ratio. For example, for Ireland there is only 1 out of 15 countries with an RMSPE ratio of at least 15.51 yielding a p-value of 0.067.

A.5 In-time placebo test

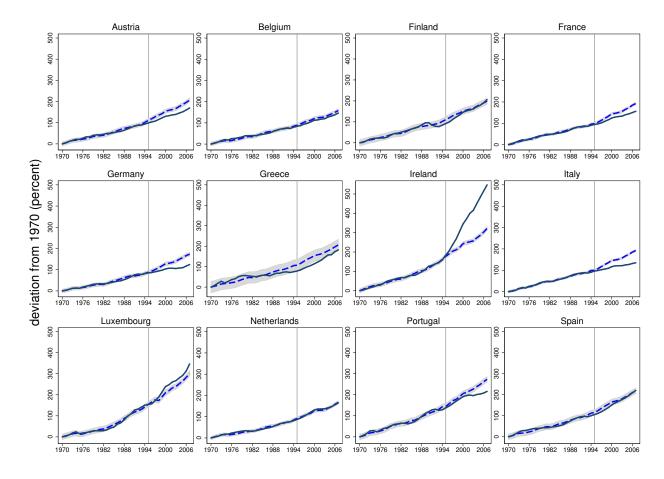


Figure 8: In-time placebo test - 1995

Notes: In each graph, the blue dashed line represents the normalized real GDP for the synthetic country and the black full line represents the series for the actual country. The vertical line depicts the placebo treatment period - 1995 for all countries. For all countries the analysis starts in 1970 and ends in 2007. This date is particularly useful to analyze because it is the one used by Puzzello and Gomis-Porqueras (2018). The reasons on why we do not use it are explained in the main text with the most compelling being that Austria and Finland are being analyzed in our work but they joined the EU in 1995.

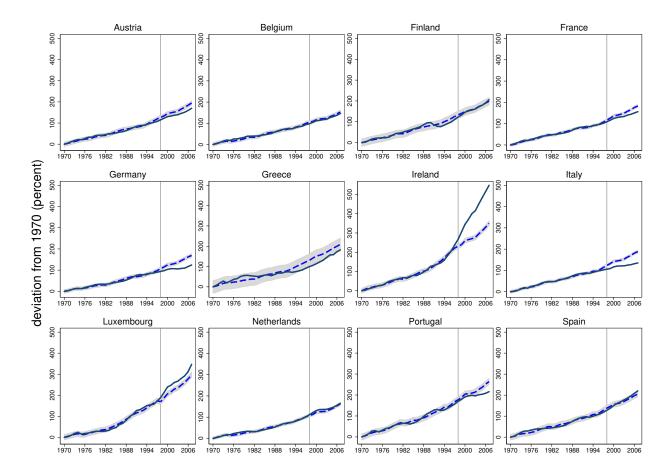


Figure 9: In-time placebo test - 1998

Notes: In each graph, the blue dashed line represents the normalized real GDP for the synthetic country and the black full line represents the series for the actual country. The vertical line depicts the placebo treatment period - 1998 for all countries. For all countries the analysis starts in 1970 and ends in 2007.

A.6 Changing donor pool

In each of the following set of graphs, the blue dashed line represents the normalized real GDP for the synthetic country and the black full line represents the series for the actual country. The vertical line depicts the treatment period. For all countries the analysis starts in 1970 and ends in 2007. We iteratively exclude different countries from the donor pool as argued in Section 3.3.1.

Figure 10: SCM without Australia

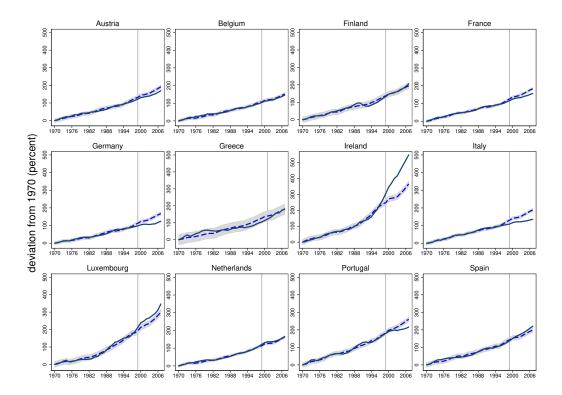


Figure 11: SCM without Israel

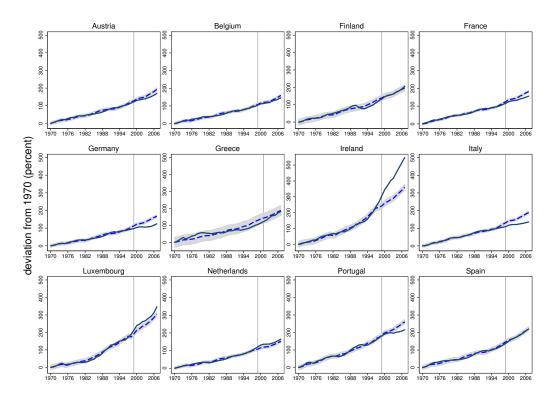


Figure 12: SCM without Mexico

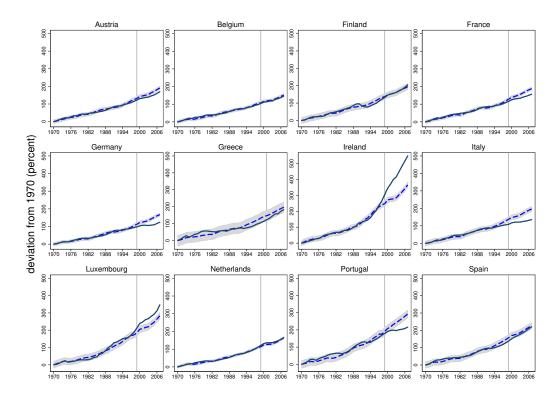


Figure 13: SCM without Norway

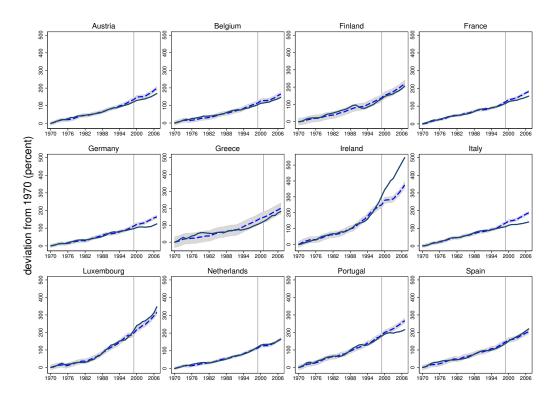


Figure 14: SCM without Switzerland

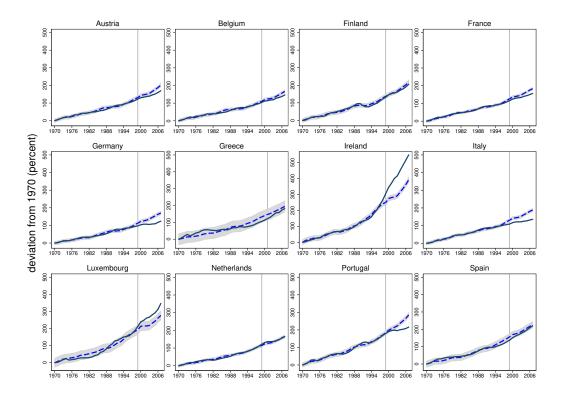


Figure 15: SCM without the United States

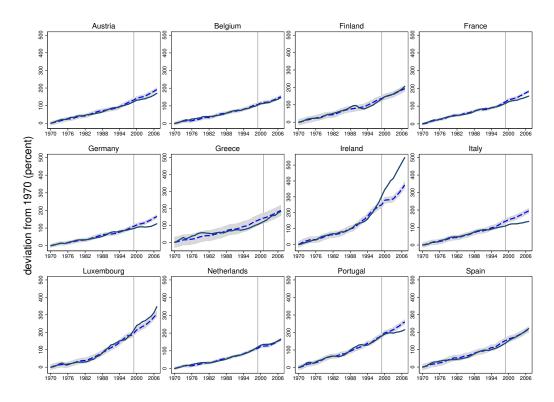
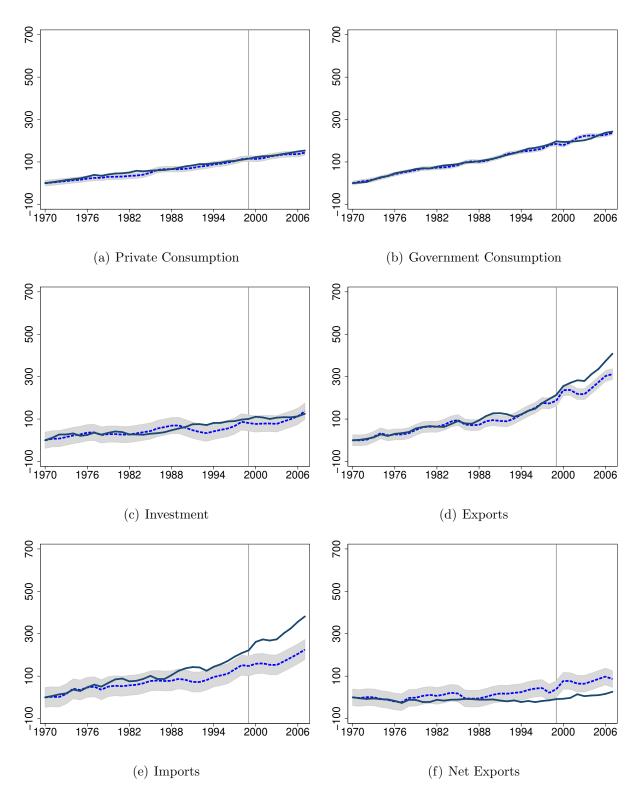


Table 6: Composition of each doppel ganger: country weights (in %)

	Austria	Belgium	Finland	France	Germany	Greece	Ireland	Italy	Luxembourg	Netherlands	Portugal	Spain
Australia	18.1	< 0.1	39.6	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	0.1
Canada	< 0.1	< 0.1	22.6	52.0	18.3	< 0.1	< 0.1	32.4	< 0.1	< 0.1	< 0.1	< 0.1
Chile	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	22.0	< 0.1	5.2	< 0.1	< 0.1	< 0.1
Iceland	< 0.1	< 0.1	17.2	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1
Israel	16.4	24.6	< 0.1	< 0.1	< 0.1	< 0.1	50.1	< 0.1	52.3	12.0	15.1	4.0
Korea	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	1.9	< 0.1
Mexico	1.8	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	< 0.1	39.8	17.5
New Zealand	18.6	< 0.1	< 0.1	32.9	10.4	30.8	< 0.1	4.1	< 0.1	< 0.1	< 0.1	< 0.1
Norway	13.3	< 0.1	< 0.1	15.1	< 0.1	< 0.1	27.9	< 0.1	< 0.1	22.6	< 0.1	< 0.1
Switzerland	31.8	75.4	20.6	< 0.1	35.6	2.6	< 0.1	12.2	42.5	65.4	26.7	21.3
United States	< 0.1	< 0.1	< 0.1	< 0.1	35.7	66.6	< 0.1	51.3	< 0.1	< 0.1	16.6	57.1

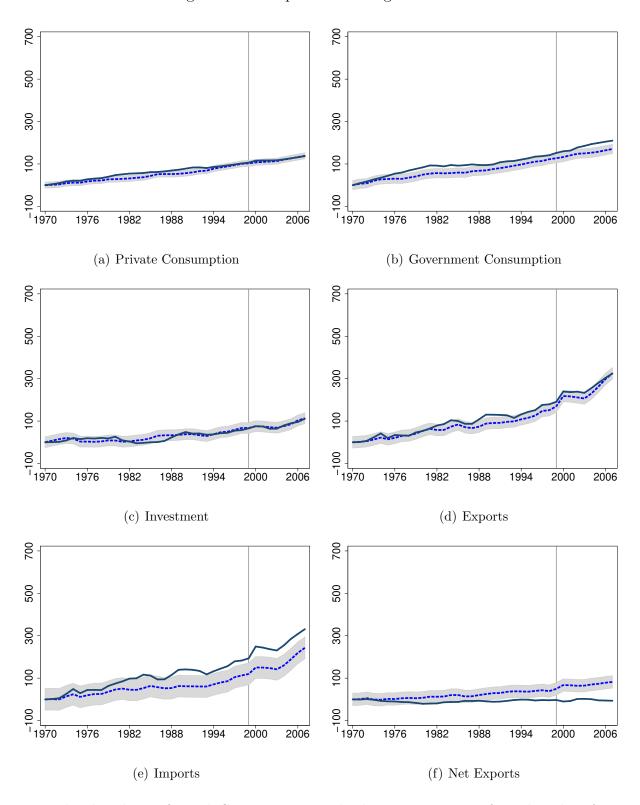
A.7 Components analysis

Figure 16: Components of Austria's GDP



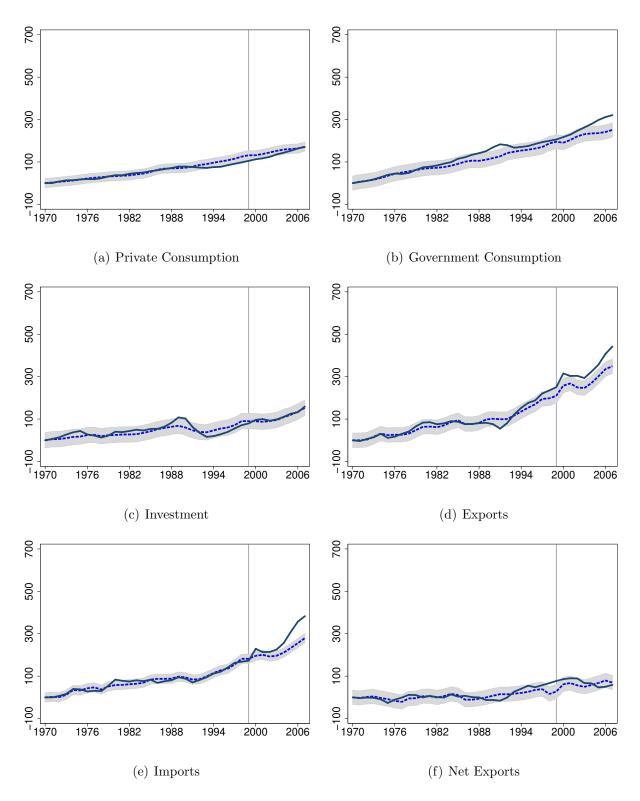
Notes: The plots depict, for each GDP component, the deviation in percent from the value of 1970. The blue dashed lines represents the synthetic Austria computed in section 2. The full black lines stand for the actual Austrian series.

Figure 17: Components of Belgium's GDP



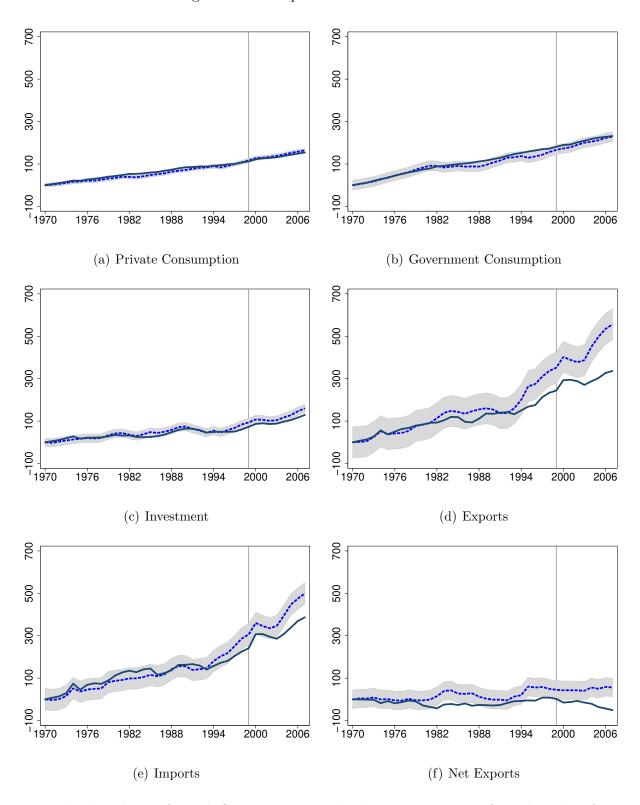
Notes: The plots depict, for each GDP component, the deviation in percent from the value of 1970. The blue dashed lines represents the synthetic Austria computed in section 2. The full black lines stand for the actual Belgian series.

Figure 18: Components of Finland's GDP



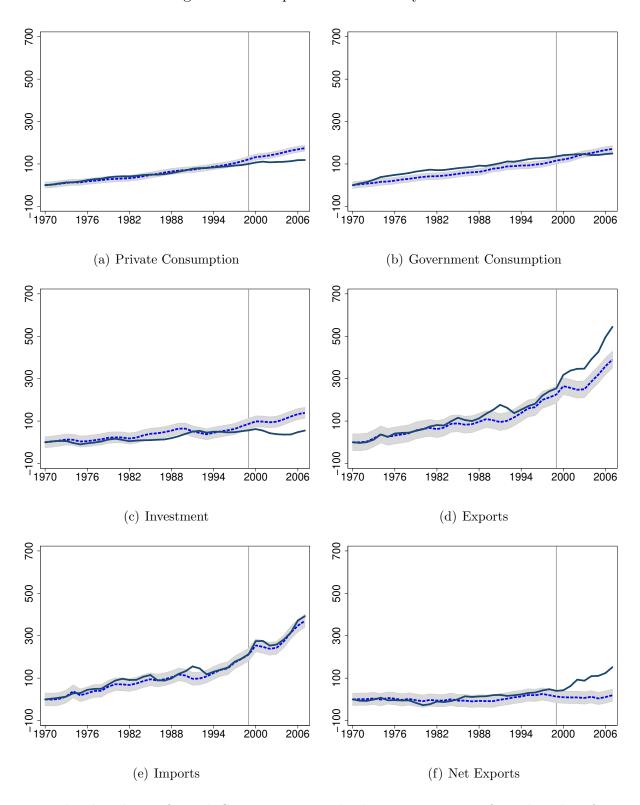
Notes: The plots depict, for each GDP component, the deviation in percent from the value of 1970. The blue dashed lines represents the synthetic Belgium computed in section 2. The full black lines stand for the actual Finnish series.

Figure 19: Components of France's GDP



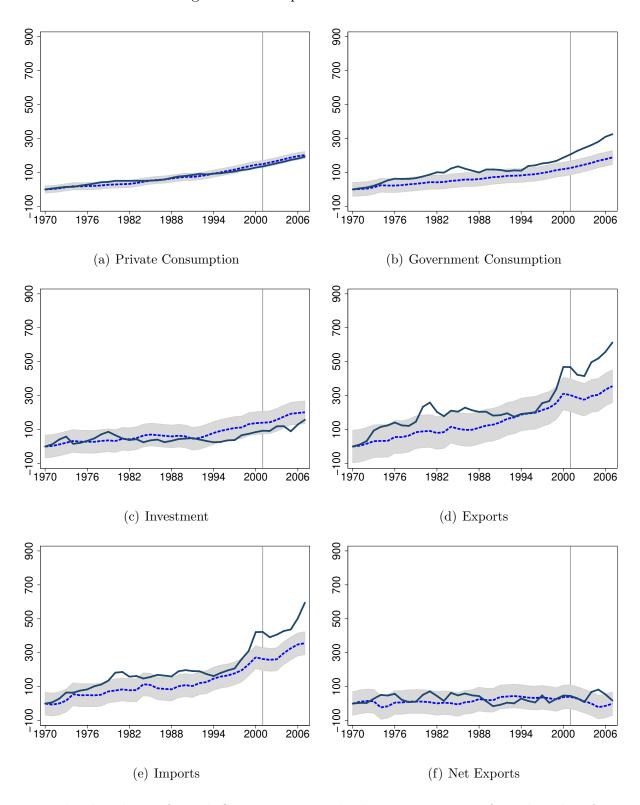
Notes: The plots depict, for each GDP component, the deviation in percent from the value of 1970. The blue dashed lines represents the synthetic France computed in section 2. The full black lines stand for the actual French series.

Figure 20: Components of Germany's GDP



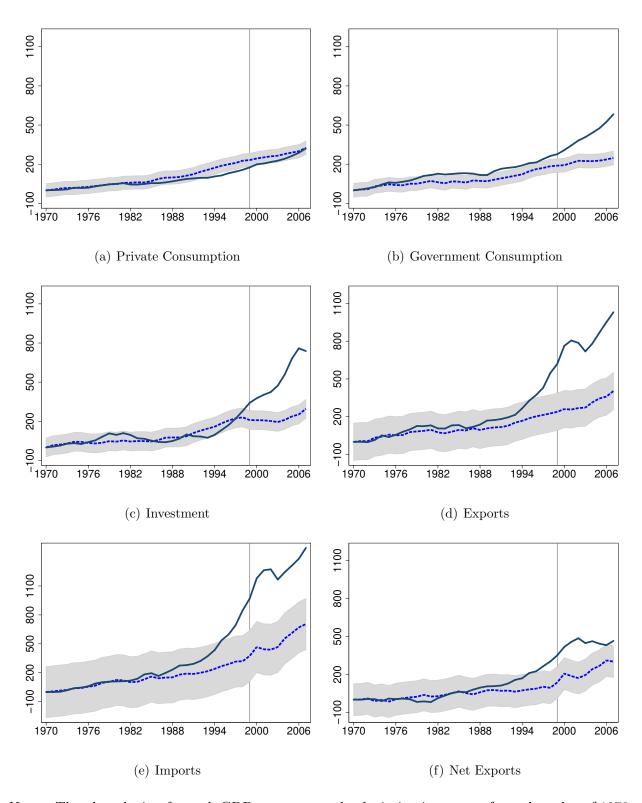
Notes: The plots depict, for each GDP component, the deviation in percent from the value of 1970. The blue dashed lines represents the synthetic Germany computed in section 2. The full black lines stand for the actual German series.

Figure 21: Components of Greece's GDP



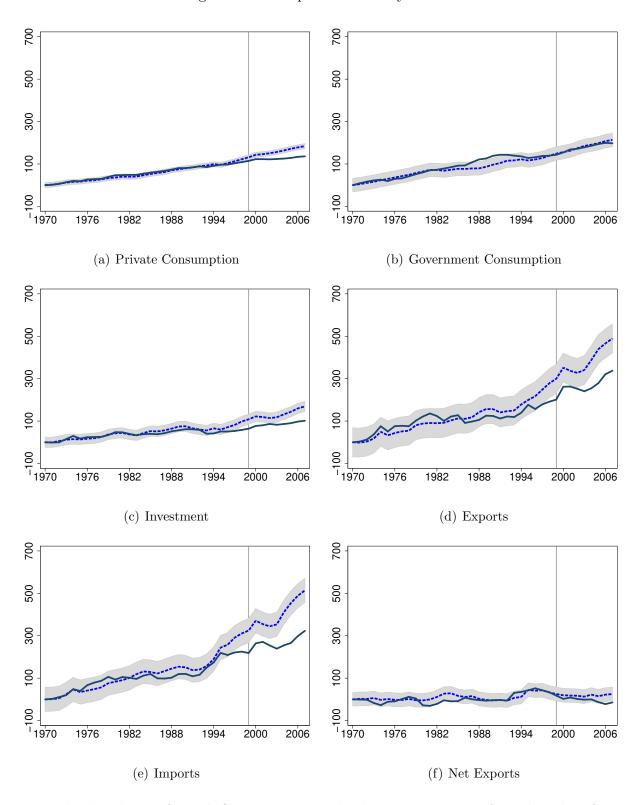
Notes: The plots depict, for each GDP component, the deviation in percent from the value of 1970. The blue dashed lines represents the synthetic Greece computed in section 2. The full black lines stand for the actual Greek series.

Figure 22: Components of Ireland's GDP



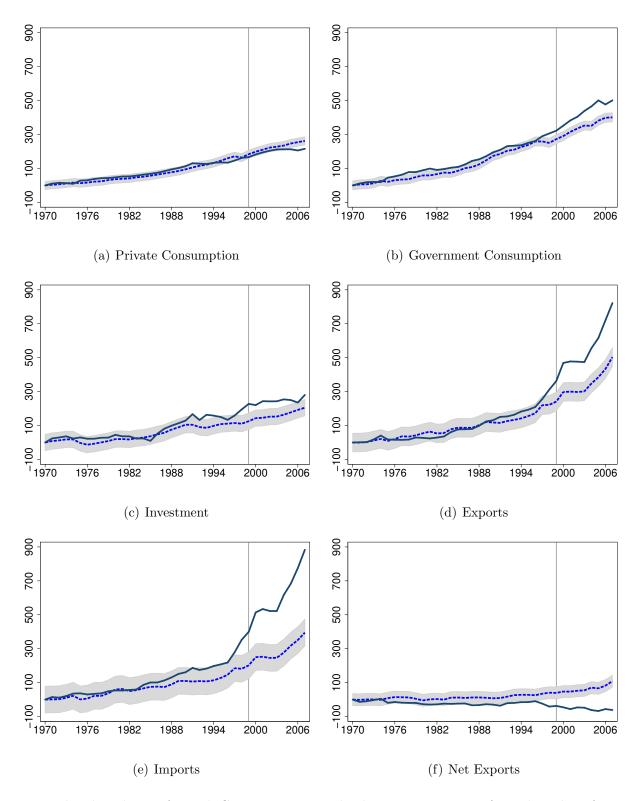
Notes: The plots depict, for each GDP component, the deviation in percent from the value of 1970. The blue dashed lines represents the synthetic Ireland computed in section 2. The full black lines stand for the actual Irish series.

Figure 23: Components of Italy's GDP



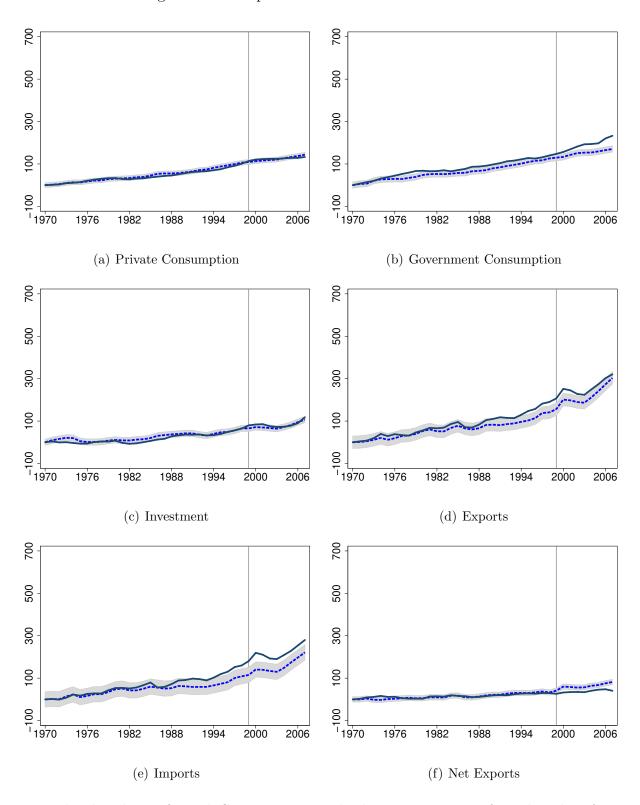
Notes: The plots depict, for each GDP component, the deviation in percent from the value of 1970. The blue dashed lines represents the synthetic Italy computed in section 2. The full black lines stand for the actual Italian series.

Figure 24: Components of Luxembourg's GDP



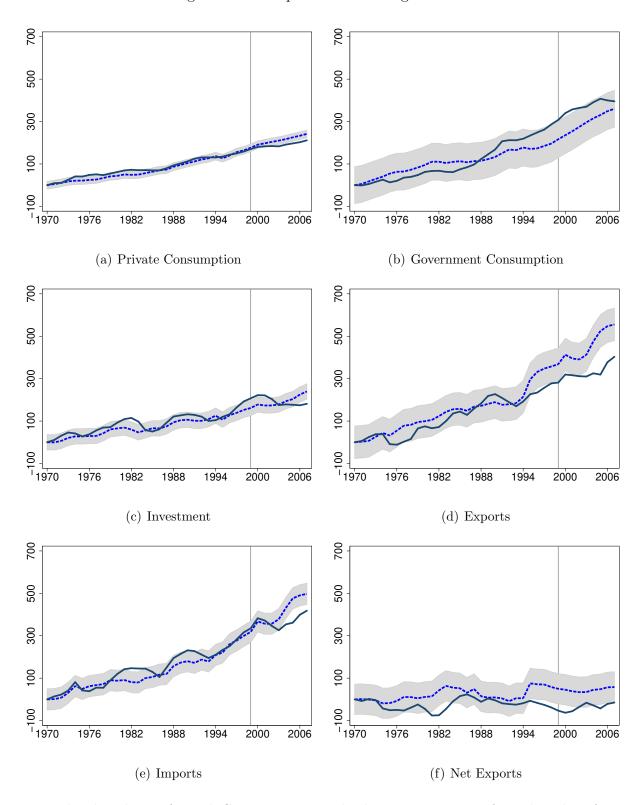
Notes: The plots depict, for each GDP component, the deviation in percent from the value of 1970. The blue dashed lines represents the synthetic Luxembourg computed in section 2. The full black lines stand for the actual Luxembourgers series.

Figure 25: Components of The Netherlands's GDP



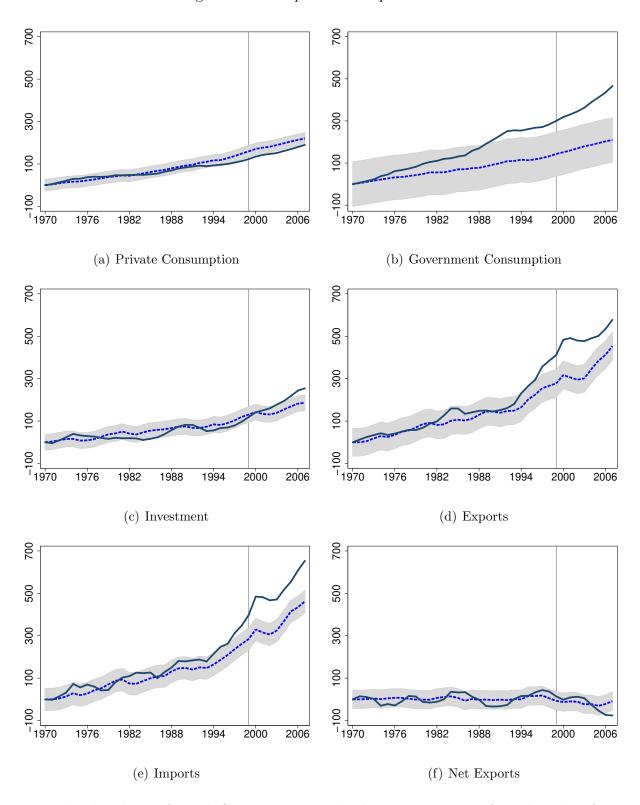
Notes: The plots depict, for each GDP component, the deviation in percent from the value of 1970. The blue dashed lines represents the synthetic Netherlands computed in section 2. The full black lines stand for the actual Dutch series.

Figure 26: Components of Portugal's GDP



Notes: The plots depict, for each GDP component, the deviation in percent from the value of 1970. The blue dashed lines represents the synthetic Portugal computed in section 2. The full black lines stand for the actual Portuguese series.

Figure 27: Components of Spain's GDP



Notes: The plots depict, for each GDP component, the deviation in percent from the value of 1970. The blue dashed lines represents the synthetic Spain computed in section 2. The full black lines stand for the actual Spanish series.

A.8 What explains the doppelganger gap? - Re do this table in Stata and add column to display the Euro per capita impact of the doppelganger gap

Table 7: What explains the cumulative doppelganger gap?

	Private Consumption	Government Consumption	Investment	Net Exports	Doppelganger Gap		
	(%)	(%)	(%)	(%)	(%)	Euro per capita	
Austria	3.77	-2.48	-1.73	-4.96	-5.40	-1,712	
Belgium	-3.35	1.08	-1.39	-2.02	-5.68	-1,668	
Finland	1.69	1.93	0.06	-0.97	2.72	777	
France	-5.02	1.31	-2.40	-2.86	-8.97	-2,632	
Germany	-12.18	-2.34	-6.43	4.70	-16.25	-5,788	
Greece	-0.83	2.83	1.87	-8.94	-5.07	-1,098	
Ireland	11.36	4.31	18.51	4.90	39.09	10,781	
Italy	-10.31	-1.79	-5.33	-0.93	-18.36	-6,089	
Luxembourg	-16.84	5.92	-4.37	27.86	12.57	4,697	
Netherlands	-4.28	6.05	-0.27	-0.98	0.52	168	
Portugal	-7.46	2.07	-0.92	-5.86	-12.17	-2,558	
Spain	-2.99	4.35	7.65	-5.11	3.90	900	

Notes: This table summarizes the cumulative doppel ganger gaps for each Euro member country and presents the channels driving the impact of the accession by decomposing GDP into its components. The doppel ganger gap represents the percentage GDP gain/loss in 2007 from adopting the common currency, i.e. for country c we define $percent\ doppel$ $ganger\ gap_{2007,c} = \frac{GDP_{2007,c}-GDP_{2007,c}^{dop}}{GDP_{2007,c}^{dop}}$. Then, the table shows the contribution of each GDP component for the GDP gain/loss. Values are constructed in a way to sum up to the doppel ganger gap. The decomposition of net exports into exports and imports is presented in Table 8.

Table 8: Net exports decomposition

	Net Exports	Exports	Imports
Austria	-4.96	5.56	10.52
Belgium	-2.02	21.17	23.19
Finland	-0.97	7.91	8.88
France	-2.86	-11.70	-8.84
Germany	4.70	4.50	-0.20
Greece	-8.94	1.62	10.56
Ireland	4.90	69.94	65.04
Italy	-0.93	-8.74	-7.81
Luxembourg	27.86	147.85	119.99
Netherlands	-0.98	15.38	16.35
Portugal	-5.86	-1.37	4.49
Spain	-5.11	1.84	6.95

Notes: This table presents the summary of the net exports decomposition into exports and imports for each treated country. It tells us how much did net exports contributed to the doppelganger gap. For example, the Austrian GDP is 4.96% smaller than of its doppelganger due to net exports. Even though exports contributed to an increase of 5.56% of GDP, imports contributed to a decrease of about 10.52% of GDP.