Does One Size Fit All in the Euro Area? Some Counterfactual Evidence.

by

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Abstract
This paper examines whether Euro Area countries would have faced a more favorable inflation output variability tradeoff without the Euro. We provide evidence that this claim is true for the periods of the Great Recession and the European Sovereign Debt Crisis. For the Euro Area as a whole, the deterioration of the tradeoff becomes insignificant with Draghi’s ‘whatever it takes’ announcement onwards. However, a more detailed analysis shows that the detrimental effect of the Euro is more severe and long-lasting for peripheral countries, pointing to structural differences among Euro Area countries as a key element of the detrimental effect of the Euro. We base our results on a novel empirical strategy that, consistently with monetary theory, models the joint determination of the variability of inflation and output conditional on structural supply shocks. Moreover, our findings are robust to potential endogeneity concerns related to adoption of the Euro.

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1. Introduction

The economic crises that occurred in Europe since 2007 have left member countries of the Economic and Monetary Union (EMU) in very heterogeneous economic conditions. While some members of the EMU by now experience modest growth and high employment, others are still in the process of recovery and suffer from unprecedented levels of unemployment. Clearly this experience is at odds with the goals of the EMU in general and the Euro Area in particular.\(^1\)

A popular view on the economic events in the EMU during the last two decades is that, by construction of the EMU, structural heterogeneity, limited scope for fiscal policy and a union-wide monetary policy have amplified the effects of adverse shocks and lead to sub-optimal macroeconomic performance. An EMU country that adopted the Euro has chosen a monetary regime where monetary policy is delegated to the ECB (see, e.g., Ball, 2010). As a direct consequence, a Euro Area country can no longer offset country-specific shocks by a country-specific monetary policy. Moreover, the transmission of area-wide shocks may be heterogeneous due to structural differences among member countries. In consequence, ECB’s monetary policy is believed to be sub-optimal: ‘one size must fit all’ rather than ‘one size fits all’ (Issing, 2001).

Consider the suggestive evidence in Figure 1 below.\(^2\) Panels 1a and 1b compare the unconditional variances of inflation deviations from an estimated target and the output gap for Euro Area countries and non-Euro OECD countries over three periods: beginning of the Great Moderation until inception of the Euro, start of the Euro until the beginning of the Great Recession, and, the crisis period since then. The panels suggest that, according to these key indicators of macroeconomic performance, non-Euro OECD countries have been

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\(^1\)According to Papademos (2009), proponents of the Euro Area have seen its adoption as a means to promoting trade and capital flows within the Euro Area with a subsequent increase in competition, efficiency of resource allocation, and economic growth. A detailed description of the rationale behind the creation of the Euro Area is given by, for example, De Graauwe (2006).

\(^2\)These figures are based on our own calculations, which we detail in Section 3 below.
more successful in reducing output variability after the start of the Euro. Moreover, they have been more successful in stabilizing both inflation and output variability during the most recent period. However, this kind of evidence leaves many questions unanswered. Do these results depend on different shocks hitting the two country groups? Are they uniform across Euro countries, time or policy changes? In this paper, we address these questions through an empirical set-up drawing from both counterfactual analysis and the analysis of productive processes.

Figure 1: Comparison of the unconditional variances of inflation deviations from an estimated target and of the output gap between Euro Area as well as non-Euro OECD countries.

There is already a literature that tries to establish empirically if there exists a ‘one size fits all’ monetary policy for Euro Area countries. The focus of that literature is on the transmission of monetary policy shocks before and after the introduction of the Euro in 1999. On the one hand, studies such as Peersman and Smets (2003), and Cecioni and Neri (2011) at the Euro Area level and Peersman (2004) in a cross-country set up do not find asymmetric effects due to monetary policy across Euro Area countries. On the other hand, Barigozzi et al. (2014), Georgiadis (2015) and Burriel and Galesi (2018), in a cross-country empirical framework, show that the common monetary framework produces asymmetric effects driven by structural differences among Euro Area countries.\(^3\) Moreover, Ball (2010)

\(^3\) The most recent data considered by this literature, except for Burriel and Galesi (2018), is from 2009.
finds that the Euro adoption had no significant effects on indicators of macroeconomic performance such as the level or variability of inflation or GDP. Nevertheless, the focus of Ball (2010) is on the effects of adopting inflation targeting (IT). In fact, the bulk of the empirical literature that quantifies the effect of a change in the monetary regime on macroeconomic performance focuses on IT. Two key themes in this literature stand out: first, this literature quantifies the effect of a change in the monetary regime on the moment of a single variable, e.g., the variability of inflation or GDP; second, a key challenge in this literature is endogeneity, as it is unanimously recognized that the choice of IT is affected by initial conditions.

Conceptually, the focus on a single variable does not seem fully appropriate. Measurement and comparison of macroeconomic performance in a theoretical IT framework is routinely based on loss functions that involve inflation and output variability. Independently of whether one assumes optimal monetary policy or a simple Taylor (1993) rule, a central bank faces a long-run tradeoff between inflation and output variability. Moreover, the variability in these endogenous variables is jointly determined by structural supply shocks that move inflation and output in opposite directions. Hence from the standpoint of such a framework, evidence based on the variability of inflation or output in isolation appears problematic. In case such research finds lower inflation variability for Euro Area countries compared to other countries, this might simply imply that the Euro Area countries are located on a different position of the inflation output variability tradeoff, but do not face an improved tradeoff due to the Euro.

Against this background, this paper seeks to examine the claim of whether the Euro Area countries would have faced a more favorable inflation output variability tradeoff
without the Euro. To this end we propose a novel empirical research design that is coherent with the bulk of theoretical IT frameworks and tackles the endogeneity issue that comes along with a monetary regime change such as the adoption of the Euro. Furthermore, we take advantage of a more extensive dataset than that available to previous work to explore possible heterogeneities in the responses to monetary regimes across countries and time.

Our research design involves several steps. First, we build a panel data set with observations on the unconditional variance of inflation deviations from target and of the output gap for twenty OECD countries over the sample period 1985 to 2016. We also estimate the variance of the structural supply shocks for each country by the help of a Structural Vector Autoregression (SVAR). Second, as a clear novelty compared to the existing empirical literature on IT, we jointly condition the tradeoff between inflation and output gap variability on the variability of structural supply shocks using a set-up taken from the quantitative analysis of production processes (see, e.g., Kumbhakar, 2012). In brief, we interpret the variability of inflation and output gap of each country as jointly determined inputs and the variability of structural supply shocks of each country as an exogenous output, or, more generally, as a shifter. Third, in order to establish whether Euro adopters, on average, have been worse off by adopting this monetary regime, we first utilize a difference-in-differences (DiD) approach. However, as discussed in the IT literature (see, e.g., Ball, 2010), the choice of adopting the Euro might have been affected by initial conditions and therefore be subject to endogeneity. This can be interpreted as a violation of the parallel trends assumption between the treated (Euro Area countries after the adoption of the Euro) and the control group (the countries taken to construct the counterfactual), which is required by the DiD approach. In consequence, the estimates obtained via the DiD approach may be biased. Therefore, we also allow for country-specific trends and consider lagged dependent variable (LDV) model (for a detailed discussion see Angrist and Pischke, 2009). The latter requires less stringent identification assumptions and controls for potential endogeneity of
the Euro adoption.

We find that adopting the Euro worsened the macroeconomic performance of Euro countries on average. More precisely, when we account for the possibility that the effects of the Euro may vary over time, we find that the adoption of the Euro on average worsened macroeconomic performance only in the periods of the Financial Crisis and the European Sovereign Debt Crisis. Furthermore, for the Euro as a whole the detrimental effect of the Euro ceases after 2012. This timing corresponds to Mario Draghi’s announcement about ‘whatever it takes to preserve the Euro’ and the ECB’s enactment of more intense and additional unconventional policies such as the outright monetary transactions (OMTs), the targeted longer-term refinancing operations (TLTROs) and the expanded asset purchase programme (EAPP). Therefore we interpret our findings as evidence that these measures have been effective in reducing inflation and output variability for the Euro Area as a whole. These measures may have credibly signalled that the ECB was going to act as ‘buyer of last resort’ (Acharya et al., 2017), i.e., what De Grauwe (2012) describes as a lender of last resort in the government bond markets.

Disaggregating the analysis across country groups shows that the detrimental effect of the Euro is more severe in peripheral countries. In addition, while this effect of the Euro becomes insignificant for the core of the Euro Area after the above mentioned policy interventions, it remains significant for the peripheral countries. These findings suggest that structural differences among Euro Area countries are a key element of the detrimental effect of the Euro and that monetary policy in the Euro Area is best characterized as a ‘one size must fit all’ policy. They are consistent with Burriel and Galesi (2018) who find that the effects of ECB’s unconventional monetary policy measures on Euro Area countries are heterogeneous and related to Barigozzi et al. (2014) and Georgiadis (2015) who provide evidence for asymmetric effects at the country level to common monetary policy shocks in the Euro Area.
Our study is also related to the literature that uses a Taylor (1979) curve to evaluate macroeconomic performance. Cecchetti et al. (2006) evaluate macroeconomic performance for single countries, based on a comparison between two different subsamples of the radial distance of actual unconditional variances from the optimal variances implied by the Taylor (1979) curve. Mishkin and Schmidt-Hebbel (2007) extend the approach used by Cecchetti et al. (2006) to a multi-country level, utilizing a dynamic panel with fixed effects estimated through GMM in order to infer on the macroeconomic implications of IT. However, as illustrated by Angrist and Pischke (2009), identification in a panel with lagged variables and fixed effects is problematic when the policy is endogenous to initial conditions. Olson and Enders (2012) have also made use of a Taylor (1979) curve framework, but use a different metric to measure the distance between observed and optimal variances compared to Cecchetti et al. (2006). Furthermore their analysis is conducted exclusively for the US.

Unlike this literature, our research design does not require explicit assumptions on whether monetary policy in the examined countries is best described by optimal monetary policy or by a simple Taylor (1993) rule. Rather the opposite, our framework encompasses both the inflation output variability tradeoffs implied by optimal monetary policies and by simple Taylor (1993) rules. Besides, in our empirical analysis we explicitly link the jointly determined variability of inflation and output gap to an exogenous supply shock.

The remainder of the paper is organized as follows. In Section 2 we outline the theoretical framework on which we base our empirical strategy. Section 3 describes the empirical implementation and the data in use. Section 4 presents the main results based on the DiD, while Section 5 contains our extensive robustness analyses. Section 7 concludes.

2. Theoretical Framework

We start out by briefly elaborating the theoretical inflation output variability tradeoff in the context of the New Keynesian model. We take the latter as a benchmark for
measuring and comparing macroeconomic performance and argue that the inflation output variability tradeoff exists for optimal discretionary monetary policy as well as monetary policy described by a Taylor (1993) rule. Then, we develop a theory-based empirical framework to estimate the inflation output variability tradeoffs in economies independent of any assumption about the type of monetary policy.

Frameworks for measuring and comparing macroeconomic performance in theory are routinely based on loss functions. A popular approach is to consider ad hoc period loss functions such as

$$L = \pi_t^2 + \omega x_t^2,$$

where $\pi_t$ denotes the deviation of inflation from an inflation target and $x_t$ denotes the deviation of the output gap from steady state. Parameter $\omega$ captures the central bank’s preference for output gap relative to inflation stabilization. Moreover, assume that the aggregate economy is best approximated by a standard New Keynesian model under the rational expectations hypothesis. Under optimal monetary policy under discretion (as elaborated in Clarida et al., 1999). The central bank minimizes (1) subject to aggregate economy in each period. One can show that the minimum state variable solution of this model then implies the following long-run relationships in unconditional variances

$$\sigma_{\pi,s}^2 = a_{\pi,1}^2 \sigma_e^2,$$

$$\sigma_{x,s}^2 = a_{x,2}^2 \sigma_e^2,$$

where $e_t$ is an exogenous supply disturbance assumed to be $e_t \sim \text{iid}(0, \sigma_e^2).$

In short, (2) to (3) show that both the optimal variances of inflation and output gap

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6 See Galí (2015) or Woodford (2003) for more details on this model.

7 See Appendix A for the details.
depend on the supply shock variance. Moreover, one can verify that the larger the central banks’ preference for output gap stabilization, $\omega_x$, the lower $\sigma^2_{x,*}$ and the larger $\sigma^2_{\pi,*}$.

Figure 2a depicts this concept for the case of the US. The variation of $\omega_x$ allows one to depict the Taylor (1979) curve, $F_{USA,*}$, which can be thought of as an efficient frontier. The idea is that country-specific supply shocks hit an economy and, given the structure of the economy, create a domestic tradeoff between inflation and output variability. In theory, a domestic central bank, e.g., the Federal Reserve (Fed), can conduct optimal policy and locate the economy on the tradeoff, Fed$_{optimal}$ in Figure 2a.

When it comes to measuring the macroeconomic performance, in practice, the efficient frontier can be estimated, for instance, via a parsimonious reduced form VAR with a supply and demand equation, including reduced form shocks. Such a frontier states an approximation of optimal monetary policy. Actual observed variability in inflation and the output gap will routinely indicate that the economy is to the right of an estimated efficient inflation output variability tradeoff, Fed$_{actual}$ in Figure 2a. Therefore a central bank’s monetary policy can be classified as sub-optimal.

However, what if monetary policy in a country is not appropriately described by optimal monetary policy, but may be better approximated by a simple Taylor (1993) rule? The latter is a flexible way of describing monetary policies in theory. For instance, it can also involve terms for observed monetary policy inertia, feedback to real economic activity or exchange rates. Therefore, such rules may be a more suitable description of monetary policy for many countries.

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8'Three or four estimated equations are crucial for the Taylor economic model but the economy as a whole is determined by millions of equations. At most, we could hope to get a rough picture of it.'(Friedman, 2010, p.116).

9A frequently applied approach to measuring macroeconomic performance is based on the distance of actual variability in inflation and output gap, $\sigma^2_{\pi,USA,t}$ and $\sigma^2_{x,USA,t}$ from the model-implied optimal tradeoff, $\sigma^2_{\pi,USA,*}$ and $\sigma^2_{x,USA,*}$, at certain points in time (see, e.g., Cecchetti et al., 2006). One can then repeat such an exercise for a panel of countries and compare measures of macroeconomic performance for different countries at different points in time.
As we discuss next, under such a Taylor (1993) rule, there is still an inflation output variability tradeoff. However, this tradeoff is neither optimal, nor is it captured by the approach pursued in Cecchetti et al. (2006) and related studies, which explicitly assume optimal monetary policy. For instance, consider the simple interest rate rule

\[ i_t = \phi_\pi \pi_t, \quad \phi_\pi > 0. \]  

(4)

The minimum state variable solution under policy (4) then implies the following long-run relationships in unconditional variances

\[ \sigma_\pi^2 = b_\pi^2 \sigma_e^2 \]  

(5)

\[ \sigma_x^2 = b_x^2 \sigma_e^2. \]  

(6)

Thus, similar to (2) to (3), (5) to (6) show that both the variances of inflation and output

\[ ^{10} \text{Notice that the same arguments holds, if we would consider a rule that also involves feedback to the output gap, i.e., } i_t = \phi_\pi \pi_t + \phi_x x_t, \phi_x > 0. \]
gap depend on the variance of the supply shock. In addition, the smaller the central bank’s 
coefficient on inflation, $\phi_{\pi} \in (1, \infty]$, the lower $\sigma_{x}^{2}$ and the larger $\sigma_{\pi}^{2}$.\textsuperscript{11} Thus, there exists an inflation output variability tradeoff, although the latter is based on the simple interest rate 
rule (4). The challenge is then to develop an empirical framework that is flexible enough 
to encompass both the tradeoffs implied by optimal and simple monetary policy.

In this paper, we propose an empirical framework to tackle this challenge. We assume 
that a inflation output variability tradeoff exists independently of the specific monetary 
policy in a certain country. Coming back to the example of the Fed in Figure 2, actual 
variances observed for the USA may be the result of optimal or sub-optimal monetary 
policy, but a tradeoff exists at any rate, see Figure 2b. We solely assume that, consistently 
with the above theory, an exogenous supply shock shifts output and inflation in opposite 
directions and determines both the variability of inflation and output. Moreover, a stronger 
central bank preference for inflation stabilization, i.e., lower $\omega_{x}$, or, a higher coefficient on 
inflation in the interest rate rule, $\phi_{\pi}$, implies a higher variability of output and a lower 
variability of inflation. Using observations for more countries at different points in time, 
our framework allows us to fit a convex curve as depicted in Figure 3, which is shifted by 
changes in the variance of supply shocks.

Our empirical strategy builds on tools developed in the quantitative production analy-
ysis. We use a specification based on a translog transformation function (TTF). Kumbhakar 
(2012, 2013) shows that an input-oriented TTF can be used to model the determination of 
one or more endogenous production inputs, for exogenous production outputs, and technol-
ogy. Here we use the input-oriented TTF to model the joint determination of the endoge-
nous variances of inflation and output gap (i.e., the two inputs in the TTF framework), for 
a given variance of an exogenous structural supply shock (i.e., a single output or shifter 
in the TTF framework) and a given monetary policy. In this way, the macroeconomic

\textsuperscript{11}It is well known that this model lacks a determinate rational expectations equilibrium for $\phi_{\pi} \leq 1$.\textsuperscript{11}
performance of different countries can be gauged controlling for country-specific supply shocks. As we show in Appendix A, we can obtain the following empirical specification

\[-\ln(\sigma^2_x) = \alpha_0 + \alpha_e \ln(\sigma^2_e) + (1/2)\alpha_{ee} \ln(\sigma^2_e)^2 + \beta_2 \ln(\sigma^2_\pi/\sigma^2_x) + (1/2)\beta_{2,1} \left[\ln(\sigma^2_\pi/\sigma^2_x)\right]^2 + \gamma_{2,e} \ln(\sigma^2_e) \times \ln(\sigma^2_\pi/\sigma^2_x) + v.\]  

(7)

The above function is normalized with respect to \(\sigma^2_x\), but exactly the same econometric results would be obtained by normalizing on \(\sigma^2_\pi\). Furthermore, provided that shifter \(\sigma^2_e\) is exogenous, the presence of \(\sigma^2_x\) (in the \(\sigma^2_\pi/\sigma^2_x\)-ratio) among the regressors does not make OLS estimates inconsistent (see Kumbhakar, 2012, in particular, for a formal treatment of this issue for a single shifter). Estimating (7) allows one to test whether a tradeoff between the variability of output and inflation actually exists in the data. Note that conditionally on the existence of this tradeoff, estimation of (7) uses the statistical information on macroeconomic performance more efficiently than the usual estimates based on either inflation or output gap variability alone. The reason is that in this set-up one can use the variability of the output gap (respectively inflation) to model the variability of inflation (respectively output gap).
3. Empirical Implementation

Our goal is to estimate the inflation output variability tradeoff for a number of countries \( i = 1, ..., N \) over time \( t = 1, ..., T \) based on (7). However, the empirical implementation of (7) is not obvious. In principle, as said above, one can consistently estimate the inflation output variability tradeoff by a two-way fixed-effect model, i.e.,

\[
-\ln(\sigma_{x,i,t}^2) = \alpha_e \ln(\sigma_{e,i,t}^2) + (1/2)\alpha_{ee} \ln(\sigma_{e,i,t}^2)^2 + \beta_2 \ln(\sigma_{\pi,i,t}/\sigma_{x,i,t}^2)
+ (1/2)\beta_{2,1} \left[ \ln(\sigma_{\pi,i,t}/\sigma_{x,i,t}^2) \right] + \gamma_{2,e} \ln(\sigma_{e,i,t}^2) \times \ln(\sigma_{\pi,i,t}/\sigma_{x,i,t}^2) + \alpha_i + \delta_t + \epsilon_{i,t},
\]

where we assume \( v_{i,t} = \alpha_i + \delta_t + \epsilon_{i,t} \). \( \epsilon_{i,t} \) is a stochastic error term, \( \alpha_i \) a fixed effect aimed at capturing unobserved time invariant country factors and \( \delta_t \) can be thought of as a flexible (nonlinear) time trend, i.e., a common unobserved factor (shock) affecting all countries by the same amount (for further details see Smith and Fuertes, 2016).

Some major challenges emerge with regard to the estimation of (8). First, consistently with theoretical inflation output variability tradeoff, we require observations of the variances of inflation deviation from target, of the output gap, and of the structural supply shock. This in turn implies some de-trending of the inflation and output data and the derivation of structural supply shocks for each country. We estimate the latter by SVARs. Moreover, we want to examine the effect of the Euro monetary policy on macroeconomic performance of Euro Area countries after they have adopted the Euro relative to a comparable set of countries without the Euro. Hence, we need to develop an identification strategy for the effect of the Euro on macroeconomic performance. We address these issues below.\(^{12}\)

\(^{12}\)We are aware that in principle this approach (see, e.g., Peersman, 2004; Olson and Enders, 2012) is subject to the Lucas critique. The empirical significance of this critique is however an unsettled issue. Based on Favero and Hendry (1992), Ericsson and Irons (1995), Hendry (2002), Estrella and Fuhrer (2003), for a set-up comparable to ours, Olson and Enders (2012, p.1290) argue that the Lucas critique is likely to have little effect on the estimation of SVARs. We address the Lucas critique and related robustness concerns further below.
3.1. Data and Estimation of Structural Supply Shocks

Our dataset includes quarterly observations of the consumer price index and real GDP for $N = 20$ member countries of the Organization for Economic Cooperation and Development (OECD) over the period 1984Q1-2016Q4. The source is the OECD database. The countries are Australia, Austria, Belgium, Canada, Denmark, Finland, France, Germany, Ireland, Italy, Japan, Netherlands, New Zealand, Norway, Portugal, Spain, Sweden, Switzerland, UK and US. As in Ball (2010), we have chosen all countries with population above one million that were members of the OECD in 1985 (beginning of the Great Moderation). Inflation is calculated as the year-to-year percentage difference of the consumer price index (all items).

In the estimation of the structural supply shocks, we require a measure of monetary policy. For the nominal interest rate, we rely on the following indicators. For the USA we use the Shadow Rate developed by Wu and Xia (2016) since the beginning of the sample. While in normal times it shadows the effective federal funds rate, at the zero interest-rate lower bound it is aimed to represent unconventional policies adopted by the central bank. Next, for the UK, we use the treasury bill rate up to 2004Q3, then using the corresponding shadow rate developed by Wu and Xia (2016). Finally, for the Euro Area countries, we use the money market rate until 1998Q4. Then, up to 2004Q3, we use the common ECB refinancing rate, and thereafter we use the Euro Shadow Rate developed by Wu and Xia (2016). For the non-Euro OECD countries other than the UK or the US, we use the money market rate up to 2004Q3, and then, since afterwards no shadow rate is available, we use a quarterly measure of the overnight bank rate.\footnote{Ciccarelli et al. (2013) make a similar choice and use the EONIA for Euro Area countries in recent years, since before the crisis it was indistinguishable from the main refinancing operations (MRO) rate, while after the adoption of unconventional policies it dropped below the MRO rate, being more sensitive to the unconventional policies.}

In order to compute the variances of output gap and inflation deviation from target, we
use a filter recently proposed by Hamilton (2017), which avoids the spurious persistence in the cyclical component implied by the traditional Hodrick and Prescott (1997) (HP) filter. Therefore, our measure for the output gap is the difference in the log of real gross domestic product from its trend value computed through the Hamilton (2017) filter, while inflation is calculated as the year-to-year percentage difference of the consumer price index (all items) minus its trend value computed through the Hamilton (2017) filter. Here we assume that the filter-measured trend is able to capture the explicit or implicit inflation target of the countries considered. This choice is motivated by the fact that we do not observe an explicit target in all countries of the sample. Moreover, in the short run, for instance, during the recent crisis, central banks may deliberately tolerate a deviation from the explicit target, which is a long-run concept by definition. Put differently, there may be an implicit short-run target different from the explicit long-run target and the implicit target represented by the filter-measured trend may provide a better representation of it. This approach is common in the literature (see, for instance, Olson and Enders (2012), where a HP filter is adopted). Appendix B contains further details.

Next, we base the derivation of the structural supply shocks on a SVAR model. We identify the shocks via sign restrictions with the median target (MT) method proposed by Fry and Pagan (2011). We impose that for the first four quarters, there is a positive inflation and a negative output response based on 50,000 accepted draws.\footnote{We have also derived structural supply shocks on a restricted sample ending in 2007Q4. These shocks and our baseline shocks show a correlation for the common period (1985Q1 to 2007Q4) equal to 0.83. Thus, potential structural breaks following the Financial Crisis should not be particularly worrying.} Further details on the specification of the VAR and the shock identification assumptions can be found in Appendix C. Finally, it must be pointed out that structural supply shocks and their variance are by construction orthogonal to the information set available to policy decision makers. It follows that the supply shock variance is an exogenous shifter (regressor) in the sense assumed by Kumbhakar (2012). Accordingly equation (8) and its variants to be...
considered below can be consistently estimated through OLS.

3.2. Identification of the Effect of the Euro on Macroeconomic Performance

It will be recalled that our basic observations are the variances of inflation deviations from target, output gaps, and structural supply shocks. In order to compute these variances, we have divided the sample in the nine periods highlighted in Table 1. The intention is to compute variances over a sufficiently long time window. Our chosen sub-periods also seem to satisfy the need to single out interesting economic episodes. We end up with a panel dataset of dimension \( N \times T = 20 \times 9 = 180 \).

Our baseline specification is a DiD approach, where our aim is to infer whether the adoption of the Euro and its new monetary framework has on average improved the macroeconomic performance for the Euro Area countries. We augment the two-way fixed effect model (8) by a dummy \( E_{i,t} \), which is equal to zero for all countries and one for Euro countries when the policy is implemented

\[
- \ln(\sigma_{x,i,t}^2) = \beta E_{i,t} + \alpha_{e} \ln(\sigma_{e,i,t}^2) + (1/2)\alpha_{ee} \ln(\sigma_{e,i,t}^2)^2 + \beta_2 \ln(\sigma_{\pi,i,t}^2/\sigma_{x,i,t}^2) + (1/2)\beta_{2,1} \left[ \ln(\sigma_{\pi,i,t}^2/\sigma_{x,i,t}^2) \right]^2 + \gamma_{2,e} \ln(\sigma_{e,i,t}^2) \times \ln(\sigma_{\pi,i,t}^2/\sigma_{x,i,t}^2) + \alpha_i + \delta_t + \varepsilon_{i,t},
\]

where \( \beta E \) represents our estimate of interest as it captures the impact of the Euro adoption on macroeconomic performance. Equation (9) can be written more compactly as

\[
Y_{i,t} = \alpha_i + \delta_t + \beta E_{i,t} + \Omega X_{i,t} + \varepsilon_{i,t},
\]

where \( Y_{i,t} \) is the dependent variable, \( X_{i,t} \) contains all the right hand side variables shown in equation (9) and \( \Omega \) is the corresponding vector of parameters to be estimated. The key assumption here is that of parallel trends, i.e., the average outcome for treated and control group would have followed the same trend without treatment. Under this assumption and
with non-random policy assignment, the following conditional independence assumption (CIA) can be written:

\[ E(Y_{0,i,t} | \alpha_i, \delta_t, X_{i,t}, \varepsilon_{i,t}) = E(Y_{0,i,t} | \alpha_i, \delta_t, X_{i,t}) \]. \tag{10} \]

Equation (10) states that unobserved confounders are fully captured by the two way fixed effect model. In other words, it means that the assignment is determined only by the country plus time fixed effects and control variables \( X_{i,t} \), where the latter cannot be influenced by the policy. \( \beta_E \) represents the average treatment on the treated (ATT).\(^{15}\)

When there is a treatment regarding a multiplicity of periods, one way to assess the appropriateness of the parallel trends assumption within the DiD is to allow for leads and lags of the treatment, which can be written as

\[ Y_{i,t} = \alpha_i + \delta_t + \sum_{j=-m}^{q} \beta_{E,j} E_{i,t=T_0+j} + \Omega X_{i,t} + \varepsilon_{i,t}, \tag{11} \]

where \( T_0 \) is the implementation date of the Euro, i.e., 1999Q1. Thus, instead of estimating a single post-treatment effect of the policy, we estimate \( m \) leads (pre-treatments) and \( q \) lags (post-treatments) of the policy effect. If \( k \) coincides with the date of the Euro adoption, \( m \) to \( k - 1 \) would coincide with the pre-treatment period (i.e., the leads). Proposed for the first time by Autor (2003), this is defined by the literature as a placebo experiment, where one pretends that the implementation of the policy took place earlier than in reality. The test proposed by Autor (2003) would then be \( \beta_{E,j} = 0 \forall j < 0 \). However, this cannot be considered a proper (over)identification test. Since it is based only on the pre-treatment period (i.e., there is no guarantee that trends continue to be parallel after the treatment), the null can be rejected because of two not mutually exclusive reasons: first, the policy

\(^{15}\)The treatment effect identified by a non-random DiD is the ATT (see, e.g., Athey and Imbens, 2006).
effect might have been anticipated by the economy, and thus cannot be safely ascribed to
the policy itself; second, the parallel trends assumption is not a satisfactory basis for the
identification of policy effects.

Note on the other hand that if the ATT is not constant over time after policy imple-
mentation, the modelling of $\beta_{\varepsilon, j, j} \geq 0$, allows us to have estimates of the time-varying
impact of the policy regime change. This specification is of great policy interest in our
empirical application as it allows one to assess whether the impact of the monetary policy
regime changes after the inception of the Great Recession.

A simple solution adopted in the literature to check the sensitivity of the ATT esti-
mates is to augment equation (11) with unit-specific (in this case, country-specific) trends.
However, more generally, supposing that the effects of unobserved confounders is fully
controlled by the two way fixed effect model is often restrictive. Ashenfelter (1978) and
Ashenfelter and Card (1985) have for example found, in a labour market context, that par-
ticipants to a government-sponsored training programme have earning histories that have
a pre-program-dip. Indeed, the literature aimed at evaluating the change in macroeco-
nomic performance due to IT almost unanimously consider its adoption as endogenous. In
particular, policy assignment is seen as dictated by previous economic conditions (see Ball,
2010, for an interesting survey of the literature). Since the adoption of the Euro may be
subject to such considerations, we also resort to an alternative specification to corroborate
our results.

We therefore consider the LDV model, where it is possible to avoid reliance on the
parallel trends assumption, and, at the same time, to control for past outcomes. It can
generally be specified as follows

$$Y_{i,t} = \alpha_i + \delta_t + \beta_{\varepsilon} \mathcal{E}_{i,t} + \Omega X_{i,t} + \theta Y_{i,t-h} + \varepsilon_{i,t},$$

(12)
where $X_{i,t}$ contains all the right hand side variables from equation (9) and $\Omega$ is the corresponding vector of parameters to be estimated. In this case the sample starts at the date of the Euro adoption denoted as $T_0$. Moreover, note that this is not a dynamic model, since we are conditioning on a fixed vector of pre-treatment responses $Y_{i,t-h}$, where $t-h$ spans the period from $t-1$ to the earliest available observation. In this case, the less stringent conditional independence assumption is

$$E(Y_{0,i,t} \mid \delta_t, Y_{i,t-h}, X_{i,t}, \varepsilon_{i,t}) = E(Y_{0,i,t} \mid \delta_t, Y_{i,t-h}, X_{i,t}),$$

where we assume that conditional on past outcomes and time fixed effects, the potential outcomes are independent of the treatment status. Moreover, given that past outcomes are influenced by observed and unobserved components, with a long pre-treatment period, as in our case, the pre-treatment variables (i.e., the fixed vector of pre-treatment responses $Y_{i,t-h}$) represent a proxy for controlling for unobserved time-varying heterogeneity.

Also in the LDV case, adopting the Autor (2003) multiple-effect framework makes it possible to analyze whether the Euro monetary policy has had heterogeneous effects through time. Equation (13) provides a companion to (11) for the LDV model,

$$Y_{i,t} = \alpha_i + \delta_t + \sum_{j=-m}^{q} \beta E_j E_{i,t=T_0+j} + \Omega X_{i,t} + \theta Y_{i,t-h} + \varepsilon_{i,t}. \tag{13}$$

The DiD and the LDV model are not nested. So, we cannot take one of the two as a special case of the other. But if they give broadly similar results, we might be more confident about evidence obtained on our estimate of interest.\footnote{An apparently ideal strategy, where for simplicity we do not consider time fixed effects, would be to condition on both LDV and unobserved time invariant effects (i.e., fixed effects), to obtain an even weaker CIA: $E(Y_{0,i,t} \mid \alpha_i, Y_{i,t-h}, X_{i,t}, \varepsilon_{i,t}) = E(Y_{0,i,t} \mid \alpha_i, Y_{i,t-h}, X_{i,t}).$ However, as discussed in Angrist and Pischke (2009), this combined approach requires very stringent econometric conditions for identification. In this empirical study, we will therefore utilize the DiD and LDV approaches and compare the results obtained through each of them.}
Finally, it is often argued that there are two rather distinct country groups within the Euro Area. The periphery is believed to be structurally different from the core in many aspects (see, e.g., De Grauwe and Ji, 2013). This could be a potential explanation of why monetary policies by the ECB face a problem of ‘one size must fit all’ rather than being characterized by ‘one size fits all’ (Issing, 2001). In order to shed light upon this important matter, we conduct a subsample analysis for the core and periphery countries, and present results for these groups as well as for the Euro Area as a whole. In our analysis, the core consists of Austria, Belgium, Finland, France, Germany, and the Netherlands. The periphery is comprised by Italy, Ireland, Portugal, and Spain, which are the countries in our sample that experienced a sovereign debt crisis.

4. Main Results

The first three columns of Table 2 present the baseline results for the DiD specification (9). First, notice that the coefficient for the ratio of the variability of inflation to output gap, $\hat{\beta}_2$, is highly significant in all three columns, in line with the considerations developed at the end of Subsection 3.2. Next, the coefficients for the non-linear terms $\ln(\sigma^2_{e,t,i})^2$ and $\ln(\sigma^2_{x,t,i}/\sigma^2_{z,t,i})^2$ are not significantly different from zero. This is evidence in favor of a convex Cobb-Douglas inflation output variability tradeoff for the Euro Area as a whole, as well as for the core and the periphery. Second, the coefficient $\hat{\alpha}_e$ shows that the variance of the supply shock has a highly significant impact on the location of the tradeoff in all the Euro Area, the core and the periphery. It has a negative sign, $\hat{\alpha}_e < 0$, therefore, the larger the variance of the supply shock, the larger the variance of inflation and the output gap.\footnote{Appendix D contains details on the interpretation of the coefficient estimates.}

Finally, there is some evidence against the separability in the transformation function, as $\hat{\gamma}_{2,e} < 0$ is significant for the periphery and (weakly) for the Euro Area as a whole.\footnote{The separability hypothesis, which traditionally implies that the marginal rates of substitution between}
sum, the significant coefficient estimates are evidence for the existence of an inflation output variability tradeoff for the Euro Area consistent with the theoretical tradeoff discussed in Section 2.

Most importantly, the coefficient for the dummy on Euro adoption, $\hat{\beta}_E$, is highly significant for all the Euro Area, the core and the periphery. A negative sign for $\hat{\beta}_E$ means that, on average, adopters of the Euro have been worse off due to adopting this monetary regime.

We can provide some further interpretation for the coefficient estimates. For instance, for $\hat{\beta}_E = -0.827$ follows that $\exp(\hat{\beta}_E) \approx 0.44$. The latter can be interpreted as the ratio of the Euro Area (post-Euro introduction) and control group transformation functions. Thus, the inverse of this ratio is $\approx 2.29$ and means that, conditionally on the supply shock variances, the Euro Area has a joint variance of inflation and output, which is around 129% larger than that of the control group. Likewise for the core and the periphery this variance was respectively 112% and 155% larger.

Finally, notice that the high p-values for the Ramsey (1969) Reset test suggest that the null hypothesis of the test, i.e., omitted variables being orthogonal with respect to the included variables, cannot be rejected for all the Euro Area, its core and periphery. Therefore, we cannot find evidence that estimates of equation (9) suffer from misspecification.

In the next three columns of Table 2, we present results for the specification (11), where we include leads and lags of the treatment, along the lines of Autor (2003). This exercise allows us to provide a more articulated economic interpretation of the basic finding of a detrimental effect of the Euro.

Consider the fourth column in Table 2, which relates to the Euro Area as a whole. Compared to the previous results, three observations stand out. First, there is again
significant and well-specified evidence in favor of an inflation output variability tradeoff consistent with the theoretical tradeoff developed in Section 2. Second, the coefficients on the leads, $\hat{\beta}_{\varepsilon,-3}$, $\hat{\beta}_{\varepsilon,-2}$, and $\hat{\beta}_{\varepsilon,-1}$, are insignificant. We also tested the null hypothesis that these coefficients are all zero and cannot reject it. Thus, anticipation effects and/or divergent trends between treated and control group appear to play no role in explaining our basic findings. Third, coefficients $\hat{\beta}_{\varepsilon,0}$ and, $\hat{\beta}_{\varepsilon,2}$ and $\hat{\beta}_{\varepsilon,3}$, relating to periods including the Financial Crisis and the European Sovereign Debt Crisis, are significant, while $\hat{\beta}_{\varepsilon,4}$, relating to the time after Mario Draghi’s ‘whatever it takes’ and OMTs announcements and the EAPP announcement and implementation, is insignificant. We interpret these findings as evidence that the impact of the Euro adoption has changed over time. The inflation output variability tradeoff for the Euro Area countries seems to be significantly worse after the inception of the Euro (relative to the control group) until the ECB’s 2012 policy turnaround, but not thereafter. The coefficient estimates can be interpreted as follows: $\exp(\hat{\beta}_{\varepsilon,0}) \approx 0.49$, $\exp(\hat{\beta}_{\varepsilon,2}) \approx 0.33$, and $\exp(\hat{\beta}_{\varepsilon,3}) \approx 0.24$ imply that the Euro Area had a joint variance of inflation and output, which is more than 105% (205%, 323%) larger during 1999Q1 to 2002Q2 (2006Q1 to 2009Q2, 2009Q3 to 2012Q4) compared to the control group.

The fifth and sixth column in Table 2 relate to the subsample analysis on the core and periphery. Their main features can be summed up by the following four remarks. First, throughout all specifications the estimated coefficients regarding the inflation output variability tradeoff are similar to the previous results. Second, again the Reset test does not give rise to concerns about omitted variables or specification problems. Third, the effect of the Euro on the core is in line with our findings for the Euro Area as a whole. There is again a detrimental effect of the Euro, and once we allow for heterogeneous effects of the treatment over time, the effect is observed in periods 7 and 8, but not in period 9. However, and fourth, the coefficient estimates for the periphery reveal differences vis-
à-vis the former results. From 2006Q1 onwards the detrimental effect of the Euro is more severe in the periphery as the respective coefficients are larger in absolute value. This is consistent with the asymmetric effects of shocks in the Euro Area as found in Barigozzi et al. (2014), Georgiadis (2015) and Burriel and Galesi (2018). More crucially, the detrimental effect of the Euro does not cease in period 9. Our findings can be interpreted as evidence that both the core and periphery experienced a worsening of the tradeoff during the period of the financial crisis and the sovereign debt crisis. However, only the core appears to have benefited significantly from the monetary policy turnaround occurring in 2012.

5. Robustness

The purpose of this section is threefold. First, we assess the robustness of the above findings by augmenting the DiD model with country-specific linear trends. Second, we consider the LDV approach, which is an alternative identification strategy. Third, we present an additional robustness analysis relating to a different periodization of inflation, output and supply shock variances.\(^\text{19}\)

5.1. Accounting for Country-specific Trends

According to Angrist and Pischke (2009), a simple way to assess the validity of DiD identification strategy, especially with relationship to the parallel trend assumption, is to augment the baseline DiD equation with unit-specific trends (in our case, country-specific trends). In Table 3 we provide estimates for this robustness exercise. While the estimates directly related to the inflation-output variability trade off are broadly consistent with the previous evidence, we now remarkably find that the Euro dummy, \(\hat{\beta}_E\), is insignificantly different from zero. This is not an uncommon result in the literature. In two well-known

\(^{19}\)Furthermore, in order to address the Lucas critique, we ran several Chow tests on (9) and the other specifications and found no evidence of parameter instability. The potential breakpoints were taken at the inception of the Euro, at the beginning of the Financial Crisis, and also at a previous stage when inflation targeting had gained preeminence among central bankers. These tests are available upon request.
studies, Besley and Burgess (2004) and Waldinger (2010) find that treatment effects are erased or strongly reduced by the inclusion of unit-specific trends. As argued by Waldinger (2010, p.822), this robustness check is very demanding, as unit-specific trends are bound to impinge on any true treatment effect. However, we observe from the second column that the ATT’s remain significant, when we model the treatment dynamically as in Autor (2003). Much as before, relative to the control group, the trade off faced by countries within the Euro zone worsened during the height of the recession, and then improved at the end of the period under analysis. The usual calculation, \( \exp(\hat{\beta}_{E,2}) \approx 0.44 \) \( \exp(\hat{\beta}_{E,3}) \approx 0.36 \), implies a joint variance of inflation and output in the Euro Area, which is more than 2.28 (2.77) times larger during 2006Q1 to 2009Q2 (2009Q3 to 2012Q4) compared to the one in the control group.

Also in line with the baseline DiD estimates an insignificant coefficient estimate \( \hat{\beta}_{E,4} \) for the Euro Area as a whole and for the core countries suggests again that the 2012 monetary policy change had a beneficial effect on macroeconomic performance. Once more, however, this beneficial effect does not show up in the periphery. Summing up, and using Angrist and Pischke (2009) own vocabulary, it is heartening to find that our most articulated evidence is left unchanged by the inclusion of country-specific trends.

5.2. Lagged Dependent Variable Approach

As already explained in Section 3, the LDV approach is one way to account for the possibility that policy choices such as the Euro adoption depend on past economic performance. This is indeed a plausible scenario and would imply an endogeneity bias for the DiD approach. Further advantages of the LDV approach are that it does not require the parallel trends assumption and that it controls for unobserved time-varying heterogeneity.

Inspection of the first, second and third column in Table 4, relating to the estimation of equation (12), reveals that all our key findings are qualitatively unchanged vis-à-vis the previous ones. The estimates directly related to the inflation output variability tradeoff
are consistent with the previous findings. Moreover, we find that the Euro dummy, \( \hat{\beta}_E \), is significantly different from zero for the Euro Area as a whole and the periphery. Clearly, this is further evidence that, relative to the control group, countries with the Euro faced a worse tradeoff, and that this result is driven by the detrimental effect of the Euro on the periphery. The usual calculation, \( \exp(\hat{\beta}_E) \approx 0.58 \), implies a joint variance of inflation and output for all the Euro Area, which is more than 73\% larger compared to the one in the control group. For the periphery this variance is 262\% larger. Thus, the quantitative implications of the treatment are in the same ballpark of those obtained under the parallel trends assumption. In sum, our previous findings are robust to relaxing the parallel trends assumption and following an alternative empirical strategy that also controls for potential endogeneity of policy choices and unobserved time-varying heterogeneity.

Consider now the fourth, fifth and sixth column of Table 4, which provide the results from estimation of equation (13). Notice also that in this specification we cannot have ATT leads, hence \( m = 0 \). As found above, the adoption of the Euro worsens the tradeoff at the occurrence of the Financial Crisis and the European Sovereign Debt Crisis for the Euro Area as a whole, but this detrimental effect becomes insignificant in correspondence of the period which starts with the Draghi announcement and includes the OMTs announcement and the EAPP announcement and implementation.

However, the estimates \( \hat{\beta}_{E,0}, \hat{\beta}_{E,1}, \hat{\beta}_{E,2}, \hat{\beta}_{E,3}, \hat{\beta}_{E,4} \) for the core are insignificant, while \( \hat{\beta}_{E,0}, \hat{\beta}_{E,2}, \hat{\beta}_{E,3}, \hat{\beta}_{E,4} \) are significantly negative for the periphery. Once more, these results suggest that the detrimental effect found for the Euro Area as a whole is mainly associated with the effect of the Euro on the periphery.

In sum, Table 4 shows that all our main findings are qualitatively unchanged vis-à-vis the ones from Table 2, regardless of whether the ATT is modelled through a single variable or through a string of lagged variables. This conclusion is also vouched by the insignificance of the Reset test throughout the table.
5.3. Different Number of Periods in the Panel Estimation

So far, we have split our quarterly data into 9 periods as described in Subsection 3.2. While we believe that this choice is plausible, one may argue that it is arbitrary and that a different choice may potentially yield different results. In order to address this concern, we provide an alternative choice of periods splitting the sample into 7 instead of 9 periods. This new periodization is shown in the lower panel of Table 1. As a consequence, both the financial crisis and the European sovereign debt crisis are now in period 6 and the ECB’s policy turnaround is in period 7. We then re-estimate the DiD (with and without country-specific trends) and LDV models for the new periodization.

Table 5 displays only the estimates for the leads and lags of the ATT as we believe that they best characterize the previously obtained evidence. The coefficient estimates related to the inflation output variability tradeoff and the Reset tests are in line with our previous estimates.

Focusing on the treatment effects, it appears that the DiD model without country-specific trends is the one that sports the most significant discrepancies vis-à-vis its nine-period counterpart. There is no significant ATT now, except for the periphery in the last period. However the ATT is almost significant at the 10%-level for the Euro Area as a whole and in the core in period 6, and always has the right (negative) sign. Hence the different results may be at least partially explained by the lower sample size and estimation efficiency of the seven-period model.

The DiD model with country-specific trends is quite similar to the nine-period counterpart. In this case the treatment effect is not significant for period 6 in the periphery, but it should be noted that in the nine-period estimates too, the Euro’s detrimental performance begins to be significant in the periphery one period later than in the core. In line with all the other estimates, the ATT is not significant in the core and significant in the periphery for the last period.
The LDV estimates are the most similar to their nine-period counterparts. ATT’s are negative and significant throughout all country samples for period 6, while retaining significance in period 7 only for the periphery.

All in all, we conclude that our previous finding of a last-period ATT significant in the periphery and insignificant in the core is left untouched by the different choice of periods. The situation is more nuanced for period 6, while on the other hand the former evidence (of virtually no significance) is fully supported for the previous periods.

6. Discussion

Summing up, and maintaining for convenience the baseline periodization, we consistently find a detrimental effect of the Euro on macroeconomic performance in periods 7 (2006Q1 to 2009Q2) and period 8 (2009Q3 to 2012Q4). In keeping with most of the previous literature, we surmise an interpretation of these results in terms of a monetary policy for the Euro Area during these periods of crises that can be denoted a ‘one size must fit all’ rather than a ‘one size fits all’ monetary policy. Moreover, we think that our finding that the detrimental effect of the Euro ceases in period 9 has two implications.

First, it suggests that the detrimental effect during periods 7 and 8 is directly related to periods of crises and so is our ‘one size must fit all’ judgement. This is consistent with De Grauwe’s (2012) Eurozone fragility hypothesis. The ECB did not immediately react to solvency concerns regarding some peripheral Euro Area countries by signalling its willingness to act as ‘buyer of last resort’ on the market for bonds of European governments, although this would have been a natural policy for independent national central banks in these peripheral Euro Area countries.

Second, we interpret our finding that the detrimental effect of the Euro ceases in period 9 meaning that the ECB acted in such a way to make clear that it was willing to act as ‘buyer of last resort’. In turn, European sovereign debt markets calmed down, leading to
the disappearance of the detrimental effect of the Euro on macroeconomic performance for the Euro Area on average. This narrative is also supported by empirical work on the effects of these announcements on European sovereign debt markets (see, e.g., De Grauwe and Ji, 2013; Saka et al., 2015).

A more detailed analysis of the relative effectiveness of such announcements and unconventional monetary policy measures is beyond the scope of this paper. Nevertheless, one indicator to gauge the effectiveness of the monetary policies during periods 7, 8 and 9 is the evolution of the shadow rate developed in Wu and Xia (2016). We depict this measure for the USA and the UK as examples of the control group on the one side and the Euro Area on the other side in Figure 4 below. We observe that the shadow rates in the control group countries moved below zero much earlier than in the Euro Area. On the other hand, the shadow rate for the Euro Area becomes consistently negative only since mid 2013, which corresponds to our period 9. This stylized fact is broadly consistent with the interpretation we gave above for our empirical evidence.

Our subsample analysis for the effects of the Euro on the core and the periphery of the

![Figure 4: Shadow Rates](image-url)
Euro Area allows for further insights. We find that the detrimental effect of the Euro on macroeconomic performance is more severe in peripheral countries and does disappear in period 9 in the core, but not in the periphery.

One plausible explanation for this finding is the observation of zombie lending or forbearance, i.e., measures that keep non-performing loans performing. This is associated with a misallocation of bank credit and may prevent economic recovery. Acharya et al. (2019) present evidence that such zombie lending in the Euro Area was particularly severe in the peripheral countries. However, there is also evidence for zombie lending in the control group, e.g., Japan (see, Caballero et al., 2008). Moreover, Schivardi et al. (2017) provide evidence for Italy that questions the importance of zombie lending in the Euro Area. Thus, while zombie lending appears to be a plausible candidate explanation, the empirical literature has not yet reached a consensus on the quantitative importance of this candidate explanation.

Finally, could bad luck be the explanation of the detrimental effect of the Euro on macroeconomic performance in periods 7 and period 8 and the disappearance of the detrimental effect in period 9? Put differently, did the Euro Area countries experienced a more severe sequence of shocks than the economies in the control group during periods 7 and 8, while this was not the case in period 9? In principle, we are confident that this is not a reasonable explanation, as we have controlled for any (supply) shock that generates an inflation and output variability tradeoff. In consequence, we are confident that our results are not driven by such shocks.

However, it may be argued that other shocks that create variability of inflation and output, but do not necessarily shift the inflation and output variability tradeoff, can explain the detrimental effect of the Euro during periods 7 and 8. For instance, in the context of our theoretical New Keynesian model above, discount factor shocks, or, shocks to the financial market conditions may create variability of inflation and output at the zero interest-rate
lower bound as they might no longer be offset by monetary policy. In consequence, the detrimental effect of the Euro in periods 7 and 8 could be due to an idiosyncratic sequence of shocks, for instance, to financial market conditions. Yet this is not a likely explanation for the following reasons.

First, the Reset test is insignificant throughout all estimates. It has long been known (Thursby, 1981, 1982) that in this case we can safely expect that any omitted variable is orthogonal to the included regressors and hence does not bias results. Second, we have obtained similar results with the pure DiD and the LDV approaches, where the latter also controls for unobserved time-varying heterogeneity. Therefore, we conclude that bad luck does not seem to be a very plausible explanation for the detrimental effect of the Euro in periods 7 and 8. The very same considerations suggest that good luck is not a very plausible explanation for the disappearance of the detrimental effect of the Euro in period 9.

7. Concluding Remarks

This paper conducts a counterfactual analysis providing evidence that Euro Area countries would have experienced a more favorable inflation output variability tradeoff without the Euro. The deterioration of the tradeoff for Euro Area countries is related to the period of the Great Recession and the European Sovereign Debt Crisis and ceases with the Draghi and OMTs announcements as well as the EAPP announcement and implementation. Our findings are based on a novel empirical strategy that is consistent with a theory-based inflation output variability tradeoff whose position is influenced by structural supply shocks. We develop a panel data set for twenty OECD countries which consists of variances for output gap and inflation deviations from target as well as variances for the structural supply shocks. The shock variances are estimated via a structural model that uses sign restrictions to identify shocks. In the estimation of the tradeoff, we model the joint deter-
mination of the variability of inflation and output by the structural supply shock through a transformation function taken from the quantitative analysis of production processes. The counterfactual evidence relating to ATT’s is robust throughout various empirical specifications.

We interpret the higher inflation and output variability in the Euro Area during the periods of crisis as evidence that the ECB measures during these periods have not been effective to reduce inflation and output variability to levels comparable with other economies. The disappearance of this detrimental effect cannot be found after the Draghi announcement onwards. This suggests that the policy moves subsequent to Draghi’s ‘whatever it takes’ announcement have been effective in reducing inflation and output variability in the Euro Area on average. We argue that this is the case, because these moves credibly signalled that the ECB was going to act as ‘buyer of last resort’.

Most importantly, our more detailed analysis shows that the detrimental effect of the Euro is more severe for peripheral countries of the Euro Area. Moreover, while the Draghi and OMTs announcements as well as the EAPP announcement and implementation had a beneficial effect on the macroeconomic performance of the core, this was not the case for the periphery. Hence, structural differences among Euro Area countries may be the underlying reason for the detrimental effect of the Euro.

References


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Tables

Table 1: Overview on the periods considered in the panel regressions

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<td>Draghi announcement (July 26th, 2012)</td>
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<td>Outright Monetary Transactions (OMTs) announcement</td>
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Table 2: Estimated parameters for all countries, core and periphery, for DiD model

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<th>Estimates</th>
<th></th>
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<tbody>
<tr>
<td></td>
<td></td>
<td>(9)</td>
<td>All</td>
<td>Core</td>
<td>Periphery</td>
</tr>
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<td></td>
<td></td>
<td>All</td>
<td>Core</td>
<td>Periphery</td>
</tr>
<tr>
<td>$\varepsilon_{i,t}$</td>
<td>$\beta_{\varepsilon}$</td>
<td>-0.827**</td>
<td>-0.753**</td>
<td>-0.937**</td>
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<tr>
<td></td>
<td></td>
<td>(0.310)</td>
<td>(0.340)</td>
<td>(0.377)</td>
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<tr>
<td>$\varepsilon_{i,T_{-3}}$</td>
<td>$\beta_{\varepsilon_{-3}}$</td>
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<tr>
<td>$\varepsilon_{i,T_{-2}}$</td>
<td>$\beta_{\varepsilon_{-2}}$</td>
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<tr>
<td></td>
<td></td>
<td>-0.054</td>
<td>-0.044</td>
<td>-0.086</td>
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<td></td>
<td></td>
<td>(0.389)</td>
<td>(0.524)</td>
<td>(0.452)</td>
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</tr>
<tr>
<td>$\varepsilon_{i,T_{-1}}$</td>
<td>$\beta_{\varepsilon_{-1}}$</td>
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<td></td>
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</tr>
<tr>
<td></td>
<td></td>
<td>-0.419</td>
<td>-0.192</td>
<td>-0.702</td>
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<tr>
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<td>(0.389)</td>
<td>(0.287)</td>
<td>(0.610)</td>
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<td>$\varepsilon_{i,T_0}$</td>
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<tr>
<td></td>
<td></td>
<td>-0.718*</td>
<td>-0.664*</td>
<td>-0.781</td>
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<td>(0.363)</td>
<td>(0.363)</td>
<td>(0.588)</td>
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</tr>
<tr>
<td>$\varepsilon_{i,T_1}$</td>
<td>$\beta_{\varepsilon_1}$</td>
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<tr>
<td></td>
<td></td>
<td>-0.584</td>
<td>-0.656</td>
<td>-0.382</td>
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</tr>
<tr>
<td></td>
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<td>(0.413)</td>
<td>(0.473)</td>
<td>(0.434)</td>
<td></td>
</tr>
<tr>
<td>$\varepsilon_{i,T_2}$</td>
<td>$\beta_{\varepsilon_2}$</td>
<td></td>
<td></td>
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</tr>
<tr>
<td></td>
<td></td>
<td>-1.115**</td>
<td>-0.968*</td>
<td>-1.233</td>
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</tr>
<tr>
<td></td>
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<td>(0.490)</td>
<td>(0.511)</td>
<td>(0.722)</td>
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<td>$\varepsilon_{i,T_3}$</td>
<td>$\beta_{\varepsilon_3}$</td>
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<tr>
<td></td>
<td></td>
<td>-1.442***</td>
<td>-1.234**</td>
<td>-1.807***</td>
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<tr>
<td></td>
<td></td>
<td>(0.484)</td>
<td>(0.537)</td>
<td>(0.549)</td>
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<td>$\varepsilon_{i,T_4}$</td>
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<td>-0.813</td>
<td>-0.262</td>
<td>-1.597***</td>
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<td></td>
<td></td>
<td>(0.494)</td>
<td>(0.560)</td>
<td>(0.432)</td>
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<tr>
<td>$\ln(\sigma^2_{\varepsilon_{i,t}})$</td>
<td>$\alpha_{\varepsilon}$</td>
<td>-0.508***</td>
<td>-0.495***</td>
<td>-0.559***</td>
<td>-0.540*** -0.517***</td>
</tr>
<tr>
<td></td>
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<td></td>
<td>-0.581***</td>
</tr>
<tr>
<td></td>
<td></td>
<td>(0.089)</td>
<td>(0.115)</td>
<td>(0.095)</td>
<td>(0.095)</td>
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<td>(0.125)</td>
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<td></td>
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<td>(0.105)</td>
</tr>
<tr>
<td>$\ln(\sigma^2_{\varepsilon_{i,t}})^2$</td>
<td>$\alpha_{\varepsilon \varepsilon}$</td>
<td>-0.191</td>
<td>-0.351</td>
<td>-0.180</td>
<td>-0.233</td>
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<td></td>
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<td></td>
<td>-0.351</td>
</tr>
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<td></td>
<td></td>
<td>(0.269)</td>
<td>(0.256)</td>
<td>(0.272)</td>
<td>(0.228)</td>
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<td></td>
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<td>(0.242)</td>
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<td></td>
<td></td>
<td></td>
<td>(0.189)</td>
</tr>
<tr>
<td>$\ln(\sigma^2_{\varepsilon_{i,t}}/\sigma^2_{x_{i,t}})$</td>
<td>$\beta_2$</td>
<td>0.530***</td>
<td>0.546***</td>
<td>0.563***</td>
<td>0.559*** 0.579***</td>
</tr>
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<td></td>
<td></td>
<td>0.581***</td>
</tr>
<tr>
<td></td>
<td></td>
<td>(0.048)</td>
<td>(0.055)</td>
<td>(0.059)</td>
<td>(0.055) 0.068***</td>
</tr>
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<td></td>
<td>0.068***</td>
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<tr>
<td></td>
<td></td>
<td></td>
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<td></td>
<td>(0.057)</td>
</tr>
<tr>
<td>$\ln(\sigma^2_{\varepsilon_{i,t}}/\sigma^2_{x_{i,t}})^2$</td>
<td>$\beta_{21}$</td>
<td>-0.021</td>
<td>0.027</td>
<td>0.004</td>
<td>-0.016</td>
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<td>0.034</td>
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<td>(0.029)</td>
<td>(0.046)</td>
<td>(0.031)</td>
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<td>(0.050)</td>
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<td></td>
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<td></td>
<td></td>
<td></td>
<td>(0.029)</td>
</tr>
<tr>
<td>$\ln(\sigma^2_{\varepsilon_{i,t}}) \times \ln(\sigma^2_{\varepsilon_{i,t}}/\sigma^2_{x_{i,t}})$</td>
<td>$\gamma_{2\varepsilon}$</td>
<td>-0.206*</td>
<td>-0.198</td>
<td>-0.278**</td>
<td>-0.221*</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>-0.221*</td>
</tr>
<tr>
<td></td>
<td></td>
<td>(0.114)</td>
<td>(0.159)</td>
<td>(0.116)</td>
<td>(0.110)</td>
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<td>(0.166)</td>
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<tr>
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<td></td>
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<td>(0.115)</td>
</tr>
</tbody>
</table>

Country fixed effect: yes, yes, yes, yes, yes, yes
Time fixed effect: yes, yes, yes, yes, yes, yes

N 20 20 20 20 20 20
Number of observations 180 144 126 180 144 126
$R^2$ 0.808 0.793 0.807 0.820 0.805 0.834

Specification tests:
- Ramsey (1969) RESET 0.606 0.309 0.833 0.730 0.622 0.974
- $\beta_{\varepsilon_{-3}} = \beta_{\varepsilon_{-2}} = \beta_{\varepsilon_{-1}} = 0$

*p-values are reported for all tests.

---

*a* The dependent variable is the variance of the output gap, i.e., $-\ln(\sigma^2_{\varepsilon_{i,t}})$.

*b* ***p<0.01; **p<0.05; *p<0.10; Standard errors are in parentheses (cluster-robust standard errors, robust to serial correlation and heteroskedasticity).
Table 3: Estimated parameters for all countries, core and periphery, for DiD model and country-specific trends

<table>
<thead>
<tr>
<th>Variables</th>
<th>Coefficient Estimates</th>
<th>( \text{All} )</th>
<th>( \text{Core} )</th>
<th>( \text{Periphery} )</th>
<th>( \text{All} )</th>
<th>( \text{Core} )</th>
<th>( \text{Periphery} )</th>
</tr>
</thead>
<tbody>
<tr>
<td>( \varepsilon_{i,t} )</td>
<td>( \beta_E )</td>
<td>(-0.215)</td>
<td>(-0.630)</td>
<td>(0.409)</td>
<td>(0.145)</td>
<td>(0.319)</td>
<td>(0.067)</td>
</tr>
<tr>
<td>( \varepsilon_{i,T,-3} )</td>
<td>( \beta_{E,-3} )</td>
<td>(0.145)</td>
<td>(0.319)</td>
<td>(0.067)</td>
<td>(0.145)</td>
<td>(0.319)</td>
<td>(0.067)</td>
</tr>
<tr>
<td>( \varepsilon_{i,T,-2} )</td>
<td>( \beta_{E,-2} )</td>
<td>(0.058)</td>
<td>(0.078)</td>
<td>(0.183)</td>
<td>(0.058)</td>
<td>(0.078)</td>
<td>(0.183)</td>
</tr>
<tr>
<td>( \varepsilon_{i,T,-1} )</td>
<td>( \beta_{E,-1} )</td>
<td>(-0.220)</td>
<td>(-0.068)</td>
<td>(-0.302)</td>
<td>(-0.220)</td>
<td>(-0.068)</td>
<td>(-0.302)</td>
</tr>
<tr>
<td>( \varepsilon_i )</td>
<td>( \beta_{E,0} )</td>
<td>(-0.460)</td>
<td>(-0.498)</td>
<td>(-0.229)</td>
<td>(-0.460)</td>
<td>(-0.498)</td>
<td>(-0.229)</td>
</tr>
<tr>
<td>( \varepsilon_{i,T+1} )</td>
<td>( \beta_{E,1} )</td>
<td>(-0.239)</td>
<td>(-0.437)</td>
<td>(0.283)</td>
<td>(-0.239)</td>
<td>(-0.437)</td>
<td>(0.283)</td>
</tr>
<tr>
<td>( \varepsilon_{i,T+2} )</td>
<td>( \beta_{E,2} )</td>
<td>(-0.824**)</td>
<td>(-0.818*)</td>
<td>(-0.559)</td>
<td>(-0.824**)</td>
<td>(-0.818*)</td>
<td>(-0.559)</td>
</tr>
<tr>
<td>( \varepsilon_{i,T+3} )</td>
<td>( \beta_{E,3} )</td>
<td>(-1.017***)</td>
<td>(-0.969*)</td>
<td>(-0.847**)</td>
<td>(-1.017***)</td>
<td>(-0.969*)</td>
<td>(-0.847**)</td>
</tr>
<tr>
<td>( \varepsilon_{i,T+4} )</td>
<td>( \beta_{E,4} )</td>
<td>(-0.480)</td>
<td>(-0.097)</td>
<td>(-0.877**)</td>
<td>(-0.480)</td>
<td>(-0.097)</td>
<td>(-0.877**)</td>
</tr>
<tr>
<td>( \ln(\sigma^2_{\varepsilon,i,t}) )</td>
<td>( \alpha_\varepsilon )</td>
<td>(-0.548***)</td>
<td>(-0.531***)</td>
<td>(-0.590***)</td>
<td>(-0.590***)</td>
<td>(-0.601***)</td>
<td>(-0.601***)</td>
</tr>
<tr>
<td>( \ln(\sigma^2_{\varepsilon,i,t})^2 )</td>
<td>( \alpha_{\varepsilon\varepsilon} )</td>
<td>(-0.364*)</td>
<td>(-0.540**)</td>
<td>(-0.430**)</td>
<td>(-0.361*)</td>
<td>(-0.527**)</td>
<td>(-0.420**)</td>
</tr>
<tr>
<td>( \ln(\sigma^2_{\varepsilon,i,t}/\sigma^2_{\pi,i,t}) )</td>
<td>( \beta_2 )</td>
<td>(0.507***)</td>
<td>(0.543***)</td>
<td>(0.528***)</td>
<td>(0.549***)</td>
<td>(0.586***)</td>
<td>(0.543***)</td>
</tr>
<tr>
<td>( \ln(\sigma^2_{\varepsilon,i,t}/\sigma^2_{\pi,i,t})^2 )</td>
<td>( \beta_{2,1} )</td>
<td>(0.034)</td>
<td>(0.034)</td>
<td>(0.071)</td>
<td>(0.042)</td>
<td>(0.045)</td>
<td>(0.082)</td>
</tr>
<tr>
<td>( \ln(\sigma^2_{\varepsilon,i,t}) \times \ln(\sigma^2_{\pi,i,t}/\sigma^2_{\pi,i,t}) )</td>
<td>( \gamma_{2,e} )</td>
<td>(-0.082)</td>
<td>(0.008)</td>
<td>(-0.116)</td>
<td>(-0.090)</td>
<td>(-0.008)</td>
<td>(-0.104)</td>
</tr>
</tbody>
</table>

Country fixed effect: yes, Time fixed effect: yes

N | 20 | 20 | 20 | 20 | 20 | 20
Number of observations | 180 | 144 | 126 | 180 | 144 | 126
\( R^2 \) | 0.875 | 0.871 | 0.897 | 0.882 | 0.882 | 0.906

Specification tests:
- Ramsey (1969) Reset: 0.133 | 0.368 | 0.177 | 0.198 | 0.736 | 0.095
- Test of significance of country-specific trends: 0.000 | 0.000 | 0.000 | 0.000 | 0.000 | 0.000

* The dependent variable is the variance of the output gap, i.e., \( -\ln(\sigma^2_{\varepsilon,i,t}) \).

** p<0.01; *p<0.05; *p<0.10; Standard errors are in parentheses (cluster-robust standard errors, robust to serial correlation and heteroskedasticity).

\( \gamma_{2,e} \) p-values are reported for all tests.
Table 4: Estimated parameters for all countries, core and periphery, for lagged dependent variable model

<table>
<thead>
<tr>
<th>Variables</th>
<th>Coefficient</th>
<th>Estimates&lt;sup&gt;b&lt;/sup&gt;</th>
<th>(12)</th>
<th>(13)</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td>All</td>
<td>Core</td>
<td>Periphery</td>
</tr>
<tr>
<td>E&lt;sub&gt;i,t&lt;/sub&gt;</td>
<td>β&lt;sub&gt;E&lt;/sub&gt;</td>
<td>-0.550*</td>
<td>0.378</td>
<td>-1.286***</td>
</tr>
<tr>
<td></td>
<td></td>
<td>(0.302)</td>
<td>(0.368)</td>
<td>(0.276)</td>
</tr>
<tr>
<td>E&lt;sub&gt;i,T&lt;sub&gt;0–3&lt;/sub&gt;&lt;/sub&gt;</td>
<td>β&lt;sub&gt;E&lt;/sub&gt;</td>
<td>-0.336</td>
<td>0.533</td>
<td>-0.930**</td>
</tr>
<tr>
<td>E&lt;sub&gt;i,T&lt;sub&gt;0–2&lt;/sub&gt;&lt;/sub&gt;</td>
<td>β&lt;sub&gt;E&lt;/sub&gt;</td>
<td>-0.120</td>
<td>0.618</td>
<td>-0.460</td>
</tr>
<tr>
<td>E&lt;sub&gt;i,T&lt;sub&gt;0–1&lt;/sub&gt;&lt;/sub&gt;</td>
<td>β&lt;sub&gt;E&lt;/sub&gt;</td>
<td>-0.805**</td>
<td>0.153</td>
<td>-1.415***</td>
</tr>
<tr>
<td>E&lt;sub&gt;i,T&lt;sub&gt;0&lt;/sub&gt;&lt;/sub&gt;</td>
<td>β&lt;sub&gt;E&lt;/sub&gt;</td>
<td>-0.995**</td>
<td>0.044</td>
<td>-1.846***</td>
</tr>
<tr>
<td>E&lt;sub&gt;i,T&lt;sub&gt;0&lt;/sub&gt;&lt;/sub&gt;</td>
<td>β&lt;sub&gt;E&lt;/sub&gt;</td>
<td>-0.421</td>
<td>0.939</td>
<td>-1.744***</td>
</tr>
<tr>
<td>ln(σ&lt;sup&gt;2&lt;/sup&gt;&lt;sub&gt;e,i,t&lt;/sub&gt;)</td>
<td>α&lt;sub&gt;e&lt;/sub&gt;</td>
<td>-0.537**</td>
<td>-0.457</td>
<td>-0.635**</td>
</tr>
<tr>
<td>ln(σ&lt;sup&gt;2&lt;/sup&gt;&lt;sub&gt;e,i,t&lt;/sub&gt;)</td>
<td>α&lt;sub&gt;ee&lt;/sub&gt;</td>
<td>-0.195</td>
<td>0.242</td>
<td>0.052</td>
</tr>
<tr>
<td>ln(σ&lt;sup&gt;2&lt;/sup&gt;&lt;sub&gt;π,i,t&lt;/sub&gt;/σ&lt;sup&gt;2&lt;/sup&gt;&lt;sub&gt;x,i,t&lt;/sub&gt;)</td>
<td>β&lt;sub&gt;2&lt;/sub&gt;</td>
<td>0.563***</td>
<td>0.534***</td>
<td>0.567***</td>
</tr>
<tr>
<td>ln(σ&lt;sup&gt;2&lt;/sup&gt;&lt;sub&gt;π,i,t&lt;/sub&gt;/σ&lt;sup&gt;2&lt;/sup&gt;&lt;sub&gt;x,i,t&lt;/sub&gt;)</td>
<td>β&lt;sub&gt;2,1&lt;/sub&gt;</td>
<td>0.014</td>
<td>0.068</td>
<td>0.027</td>
</tr>
<tr>
<td>ln(σ&lt;sup&gt;2&lt;/sup&gt;&lt;sub&gt;π,i,t&lt;/sub&gt;/σ&lt;sup&gt;2&lt;/sup&gt;&lt;sub&gt;x,i,t&lt;/sub&gt;) × ln(σ&lt;sup&gt;2&lt;/sup&gt;&lt;sub&gt;e,i,t&lt;/sub&gt;/σ&lt;sup&gt;2&lt;/sup&gt;&lt;sub&gt;x,i,t&lt;/sub&gt;)</td>
<td>γ&lt;sub&gt;2,2&lt;/sub&gt;</td>
<td>-0.020</td>
<td>-0.138</td>
<td>-0.192</td>
</tr>
</tbody>
</table>
Country fixed effect no no no no no no | Time fixed effect yes yes yes yes yes yes | N | 20 | 20 | 20 | 20 | 20 | 20 |
| Number of observations | 100 | 80 | 70 | 100 | 80 | 70 |
| R<sup>2</sup> | 0.818 | 0.851 | 0.870 | 0.833 | 0.868 | 0.901 |

<sup>a</sup>The dependent variable is the variance of the output gap, i.e., −ln(σ<sup>2</sup><sub>x,i,t</sub>).

<sup>b</sup>***p<0.01; **p<0.05; *p<0.10; Standard errors are in parentheses (cluster-robust standard errors, robust to serial correlation and heteroskedasticity).

<sup>c</sup>p-values are reported for all tests.
Table 5: Estimated parameters all countries, core and periphery, for DiD model as well as the former augmented with either country-specific trends or lagged dependent variables with seven periods in the panel.

<table>
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<tr>
<th>Variables</th>
<th>Coefficient</th>
<th>Estimates$^b$</th>
<th>Estimates$^b$</th>
<th>Estimates$^b$</th>
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<td>(11)</td>
<td>(11)</td>
<td>(13)</td>
</tr>
<tr>
<td>All</td>
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<td>Core</td>
</tr>
<tr>
<td>$\beta_E$</td>
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<td></td>
</tr>
<tr>
<td>$\beta_{E,1}$</td>
<td>0.192 0.142</td>
<td>0.243</td>
<td>0.118</td>
<td>0.169</td>
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<td></td>
<td>(0.415)</td>
<td>(0.509)</td>
<td>(0.519)</td>
<td>(0.392)</td>
</tr>
<tr>
<td>$\beta_{E,2}$</td>
<td>-0.037 0.262</td>
<td>-0.454</td>
<td>-0.199</td>
<td>0.211</td>
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<td></td>
<td>(0.336)</td>
<td>(0.407)</td>
<td>(0.461)</td>
<td>(0.375)</td>
</tr>
<tr>
<td>$\beta_{E,3}$</td>
<td>-0.294 -0.400</td>
<td>-0.211</td>
<td>-0.498</td>
<td>-0.399</td>
</tr>
<tr>
<td></td>
<td>(0.404)</td>
<td>(0.429)</td>
<td>(0.577)</td>
<td>(0.374)</td>
</tr>
<tr>
<td>$\beta_{E,4}$</td>
<td>-0.345 -0.438</td>
<td>-0.199</td>
<td>-0.703*</td>
<td>-0.530</td>
</tr>
<tr>
<td></td>
<td>(0.472)</td>
<td>(0.492)</td>
<td>(0.684)</td>
<td>(0.409)</td>
</tr>
<tr>
<td>$\beta_{E,5}$</td>
<td>-0.786 -0.709</td>
<td>-0.853</td>
<td>-1.172***</td>
<td>-0.751*</td>
</tr>
<tr>
<td></td>
<td>(0.473)</td>
<td>(0.448)</td>
<td>(0.606)</td>
<td>(0.359)</td>
</tr>
<tr>
<td>$\beta_{E,6}$</td>
<td>-0.339 0.019</td>
<td>1.327**</td>
<td>-0.841***</td>
<td>-0.065</td>
</tr>
<tr>
<td></td>
<td>(0.405)</td>
<td>(0.444)</td>
<td>(0.515)</td>
<td>(0.284)</td>
</tr>
<tr>
<td>$\ln(\sigma^2_{e,i,t})$</td>
<td>$\alpha_e$</td>
<td>-0.807*** -0.783*** -0.802***</td>
<td>-0.902*** -0.922*** -0.902***</td>
<td>-0.841*** -0.667*** -0.941***</td>
</tr>
<tr>
<td></td>
<td>(0.194)</td>
<td>(0.137)</td>
<td>(0.135)</td>
<td>(0.136)</td>
</tr>
<tr>
<td>$\ln(\sigma^2_{x,i,t})$</td>
<td>$\alpha_e$</td>
<td>-0.412 -0.886 -0.540</td>
<td>-0.494 -1.134** -0.626</td>
<td>-0.335 -0.519 -0.353</td>
</tr>
<tr>
<td></td>
<td>(0.544)</td>
<td>(0.996)</td>
<td>(0.565)</td>
<td>(0.534)</td>
</tr>
<tr>
<td>$\ln(\sigma^2_{E,i,t}/\sigma^2_{x,i,t})$</td>
<td>$\beta_2$</td>
<td>0.584*** 0.628*** 0.490***</td>
<td>0.575*** 0.620*** 0.484***</td>
<td>0.647*** 0.558*** 0.572***</td>
</tr>
<tr>
<td></td>
<td>(0.067)</td>
<td>(0.064)</td>
<td>(0.075)</td>
<td>(0.056)</td>
</tr>
<tr>
<td>$\ln(\sigma^2_{E,i,t}/\sigma^2_{x,i,t})^2$</td>
<td>$\beta_3$</td>
<td>-0.004 0.012 0.051</td>
<td>0.048 -0.030 0.115**</td>
<td>-0.013 -0.151 0.028</td>
</tr>
<tr>
<td></td>
<td>(0.006)</td>
<td>(0.107)</td>
<td>(0.044)</td>
<td>(0.061)</td>
</tr>
<tr>
<td>$\ln(\sigma^2_{E,i,t}) \times \ln(\sigma^2_{E,i,t}/\sigma^2_{x,i,t})$</td>
<td>$\gamma_{2e}$</td>
<td>-0.087 -0.114 -0.140</td>
<td>-0.039 0.050 -0.060</td>
<td>-0.105 -0.154 -0.070</td>
</tr>
<tr>
<td></td>
<td>(0.181)</td>
<td>(0.186)</td>
<td>(0.108)</td>
<td>(0.117)</td>
</tr>
</tbody>
</table>

Country fixed effect: yes yes yes no no no yes yes yes
Time fixed effect: yes yes yes yes yes yes yes yes yes
Number of observations: 20 20 20 20 20 20 20 20 20

Specification tests:
- Ramsey (1969) Reset: 0.634 0.969 0.813
- $\sum_{i=1}^{n} (\hat{y}_{i,t} - \hat{y}_{i,t-1})^2$: 0.789 0.615 0.964
- Test of significance of country-specific trends: 0.000 0.001 0.000

$^a$ The dependent variable is the variance of the output gap, i.e., $-\ln(\sigma^2_{x,i,t})$.

$^b$ ***p<0.01; **p<0.05; *p<0.10; Standard errors are in parentheses (cluster-robust standard errors, robust to serial correlation and heteroskedasticity).

$^c$ p-values are reported for all tests.
A. More on the Theoretical Framework

**Optimal Monetary Policy.** Assume that the aggregate economy is best approximated by a standard New Keynesian model under the rational expectations hypothesis, i.e.,

\[
x_t = E_t x_{t+1} - \sigma^{-1} (i_t - E_t \pi_{t+1}) + g_t
\]

(A.1)

\[
\pi_t = \beta E_t \pi_{t+1} + \lambda x_t + e_t.
\]

(A.2)

In this model, \(i_t\) denotes the nominal interest rate controlled by the central bank. \(\beta\) and \(\sigma\) are structural parameters, \(\lambda\) is a composite term comprising several structural parameters. \(g_t\) denotes an exogenous demand disturbance and \(e_t\) denotes an exogenous supply disturbance. In addition, the shocks are assumed to be \(g_t \sim \text{iid}(0, \sigma_g^2)\), and \(e_t \sim \text{iid}(0, \sigma_e^2)\).\(^{20}\)

Consider optimal monetary policy under discretion (as elaborated in Clarida et al., 1999). The central bank minimizes (1) subject to (A.2) in each period. The first-order necessary condition is \(\pi_t = -(\omega_x/\lambda)x_t\) and one can show that, under this policy, the model implies an inflation output variability tradeoff as first developed in Taylor (1979). In particular, solving the model for given parameters implies a minimum state variable solution \(\pi_t = a_\pi e_t\) and \(x_t = a_x e_t\), where \(a_\pi \equiv \omega_x/(\omega_x + \lambda^2)\) and \(a_x \equiv -\lambda/(\omega_x + \lambda^2)\). This implies the long-run relationships in unconditional variances (2) and (3).

**Taylor (1993) Rule.** Next, consider (4) and, for simplicity, assume \(g_t = 0\) for all \(t\) (i.e., we abstract from demand shocks).\(^{21}\) Then, it can be easily verified that the model (A.1) and (A.2) under policy rule (4) has the solution \(\pi_t = b_\pi e_t\) and \(x_t = b_x e_t\), where \(b_\pi \equiv (1 + \sigma^{-1} \phi_\pi \lambda)^{-1}\) and \(b_x \equiv -\sigma^{-1} \phi_\pi /(1 + \sigma^{-1} \phi_\pi \lambda).\) This solution implies the long-run

\(^{20}\)In the theoretical literature \(e_t\) is usually denoted a cost-push shock. Notice that allowing for autocorrelation in the exogenous shocks would not alter any conclusion.

\(^{21}\)This is solely for ease of exposition. If \(g_t \neq 0\), then (5) to (6) both depend on the variance of the demand shock. We discuss the potential relevance of demand shocks for our analysis further below.

39
relationships in unconditional variances given by (5) and (6).

**Input-oriented TTF.** Next we show that the input-oriented TTF provides us with a functional form that captures the basic characteristics of an inflation output variability tradeoff. Formally, one assumes that the relationship between the supply shock (as the single output, $y$) and the variances of inflation and output gap (as the $K = 2$ inputs, $z$) can be described by $A f(y, z) = 1$, where we have one output $y$ and $K$ inputs $z$. Moreover, $A$ captures factors that affect the TTF neutrally. We will be more specific about the assumptions further below. Next, we assume a *translog* functional form, i.e.,

\[
\ln(f(y, z)) = \alpha_y \ln(y) + \frac{1}{2} \alpha_{yy} \ln(y)^2 + \sum_k \beta_k \ln(z_k) + \frac{1}{2} \sum_k \sum_l \beta_{k,l} \ln(z_k) \times \ln(z_l) + \sum_k \gamma_{k,y} \ln(y) \times \ln(z_k),
\]

(A.3)

where the following symmetry is imposed: $\beta_{k,l} = \beta_{l,k}$. Equation (A.3) requires $K + 2$ additional identification, or, normalization restrictions. As discussed in Kumbhakar (2012), it is possible to impose the restrictions such that a single equation framework emerges that allows for simultaneous estimation of more than one endogenous input (e.g., input-oriented) or output (e.g., output-oriented).

Since, in the case of the inflation output variability tradeoff, we have simultaneous endogeneity of $\sigma_n^2$ and $\sigma_x^2$, while $\sigma_e^2$ is exogenous, we can consider the former two variances as inputs, while the latter variance is the output. Therefore, we adopt a normalization with respect to an input. This gives rise to an input-oriented TTF. Following Kumbhakar (2012), we rewrite (A.3) as

\[
\ln(f(y, z)) = \alpha_y \ln(y) + \frac{1}{2} \alpha_{yy} \ln(y)^2 + \sum_k \beta_k \ln(z_k / z_1) + \frac{1}{2} \sum_k \sum_l \beta_{k,l} \ln(z_k / z_1) \times \ln(z_l / z_1) + \sum_k \gamma_{k,y} \ln(y) \times \ln(z_k / z_1) + \Upsilon,
\]

where the following symmetry is imposed: $\beta_{k,l} = \beta_{l,k}$. Equation (A.3) requires $K + 2$ additional identification, or, normalization restrictions. As discussed in Kumbhakar (2012), it is possible to impose the restrictions such that a single equation framework emerges that allows for simultaneous estimation of more than one endogenous input (e.g., input-oriented) or output (e.g., output-oriented).
where each input $k$ has to be combined with the remaining inputs $l$ as described in this equation. $\Upsilon$ is a composite term that follows from writing the second and third line in expression (A.3) in ratios (see, e.g., Kumbhakar, 2012, for the details).

Next we impose the normalization restrictions, $\sum_k \beta_k = -1$, $\sum_k \beta_{k,l} = 0 \forall k$, and, $\sum_k \gamma_{k,y} = 0$. As a consequence, the composite term $\Upsilon$ is eliminated and we obtain the input-oriented TTF that we use as our empirical specification

$$-\ln(z_1) = \alpha_0 + \alpha_y \ln(y) + \frac{1}{2} \alpha_{yy} \ln(y)^2 + \sum_{k=2} \beta_k \ln(z_k/z_1)$$

$$+ \frac{1}{2} \sum_{k=2} \sum_{l=2} \beta_{k,l} \ln(z_k/z_1) \times \ln(z_l/z_1) + \sum_{k=2} \gamma_{k,y} \ln(y) \times \ln(z_k/z_1) + v,$$

where $\ln(A) = \alpha_0 + v$. In this case we normalize our function on $z_1$. We would get exactly the same econometric results by normalizing the function on $z_k$. In the particular case of the inflation output variability tradeoff, we have $y = \sigma_e^2$, $z_1 = \sigma_x^2$ and $z_2 = \sigma_\pi^2$.

### B. Computing Variances of Output Gap and Inflation Deviation from Target

Our choice of the Hamilton (2017) filter with respect to the more traditional HP filter is motivated by the considerations made by Hamilton (2017), who shows that the persistence present in the cyclical part of the HP filter is not related to the underlying data generating process. Starting from the closed form solution of the HP minimization problem, with quarterly data and $t$ more than 15 years from the start or end of a sample the cyclical component $c_t = \hat{y}_t - \hat{y}_t^*$ can be approximated by

$$c_t = \frac{\tilde{\lambda}(1-L)^4}{F(L)} \hat{y}_{t+2}, \quad \text{(B.1)}$$

---

22The normalization restrictions imply homogeneity, symmetry and monotonicity properties of the TTF.
with \( F(L) = 1 + \tilde{\lambda}(1 - L)^2(1 - L)^2 \), which shows that the HP filter might be expected to produce a stationary series as long as the fourth differences of the original series is stationary, since it takes the fourth difference of \( \hat{y}_{t+2} \) and applies the operator \([F(L)]^{-1}\). However, De Jong and Sakarya (2016) have shown that there might still be some non stationarity coming from the beginning or end of the sample, while Phillips and Jin (2015) show that even with an I(1) series the HP filter might not be able to remove the trend. Cogley and Nason (1995) show that for a random walk \( \hat{y}_t = \hat{y}_{t-1} + \varepsilon_t \) (where first differences are completely unpredictable) equation (B.1) near the middle of the sample can be approximated as \( c_t = \tilde{\lambda}(1 - L)^3/F(L)\varepsilon_{t+2} \) with \( \tilde{\lambda} = 1600 \) (as is usual for quarterly data) the HP filter will produce an \( \varepsilon_t \), which is random, and a cycle which is predictable (as a function of past and future observations). The persistence of the cycle is due to the fact that the coefficients in \([F(L)]^{-1}\) depend on the value chosen for \( \tilde{\lambda} \) and may not reflect any features of the data generating process (see Hamilton, 2017, for further details). Even when the properties of the data are such to make the HP filter optimal, Hamilton (2017) shows that the estimated value of \( \lambda \) is always much lower than the value of 1600 customarily used for quarterly data.

Hamilton (2017) proposes a new filter hinging on the forecast made two years in advance of \( \hat{y}_{t+h} \) on the basis of \( p \) current and past values. The suggested cycle (at time \( h \)), \( \tilde{v}_{t+h} \), derived from the population linear projection, in case of quarterly data would be

\[
\hat{y}_{t+h} = \zeta_0 + \zeta_1 \hat{y}_t + \zeta_2 \hat{y}_{t-1} + \zeta_3 \hat{y}_{t-2} + \zeta_4 \hat{y}_{t-3} + \tilde{v}_{t+h}.
\]

(B.2)

For quarterly data, we would have \( h = 8 \) and Hamilton (2017) suggests to take \( p = 4 \) in this case, interpreting the resulting forecast error \( \tilde{v}_{t+h} \) as the cycle at time \( t + h \). Den Haan (2000) has shown that such a forecast error would be stationary for a large class of nonstationary processes. Most importantly, Hamilton (2017) shows that the primary
reason of wrongly predicting most of the macroeconomic and financial variables two years
in advance would be due to cyclical factors.

As discussed by Hamilton (2017), it is not necessary to know the nature of the nonsta-
tionarity and to have the correct model for forecasting the time series. Even in the case
of an I(2) variable, where the series should be differenced twice \((d = 2)\) in order to have
a stationary process, Hamilton (2017) shows that with \(p > d\), equation \((B.2)\) will use two
coefficients to ensure that the residual are stationary and the remaining coefficients will
be determined by the parameters that characterize the stationary variable \(\tilde{v}_{t+h}\). This filter
can deal with the same situation considered for the HP filter (in case we need to take the
fourth difference of the time series to obtain stationarity), but without any of its previously
discussed drawbacks, in particular without the introduction of spurious persistence in the
cyclical component.

C. Estimation of the Structural Supply Shock

In order to derive our structural supply shock we consider a vector autoregression with
exogenous variables (VARX), whose reduced form of order \((p, q)\) can be represented by

\[
Y_t = C + \sum_{i=1}^{p} A_i Y_{t-i} + \sum_{k=1}^{q} \Phi_k X_{t-k} + u_t, \tag{C.1}
\]

where \(C\) is a \(3 \times 1\) vector of constant terms, \(Y_t\) is a \(3 \times 1\) vector of endogenous variables
including a measure of the output gap, inflation (as difference from target) and nominal
interest rate, \(X_t\) is a \(3 \times 1\) vector of exogenous variables including USA output gap, in-
flation (as difference from target) and a nominal interest rate, aimed at capturing world
macroeconomic stance and \(u_t\) is a vector of reduced-form disturbances with \(E[u_t] = 0\) and
\(E[u_t u'_t] = \Sigma_u\). We consider all countries with the exception of USA as open economies (for
a more detailed discussion of this choice see Favero and Giavazzi, 2008). Identification of
the VAR in equation (C.1), requires to impose enough restrictions to decompose $u_t$ in order to obtain economically meaningful structural innovations. A matrix $A$ is required such that $Ae_t = u_t$, where $e_t$ represent the vector of structural shocks. At least $n \times (n - 1)/2$ restrictions on $A$ are required to obtain identification. Usually identification is obtained via Cholesky factorization of $\Sigma_u$.

We adopt a sign restrictions identification strategy. Uhlig (2005) amongst others, shows how to obtain identification of the above VAR imposing sign restrictions on a (sub)set of the variables responses to shocks. This approach is highly germane to our model, where supply shocks are defined by their acting upon inflation and output gap in opposite directions. A further advantage of this procedure is that only a minimum amount of economically meaningful sign restrictions are required in order to identify the structural shocks. In case of a single shock, Uhlig shows that any impulse vector $a$ can be recovered if there is an $n$-dimensional vector $q$ of unit length such that $a = Aq$, where $\hat{A}$ is the Cholesky factor of $\Sigma_u$.

More precisely, starting with estimation of the above reduced form model using OLS, identification of a single shock by sign restrictions (as in our case) can be obtained as follows:

1. derive the impulse-responses for the $n$ variables corresponding to a given impulse vector $a_j$ up to period $f$ on which sign restrictions are intended to be imposed;
2. draw an $n$-dimensional $q$ vector of independent $N(0,1)$ and divide it by its norm, obtaining a candidate draw $q$ from which an impulse vector $a_j = Aq$ can be derived for then calculating the corresponding impulse responses;
3. if the resulting impulse responses meet the sign restrictions imposed accept the draw, otherwise discard it;
4. repeat 2 and 3 until a desired number of accepted draws is obtained.

Generally, the median of accepted draws is considered as the central estimate of interest.
for impulse responses. However, as shown by Fry and Pagan (2011), the median responses may combine information from several identification schemes (i.e., different $q$’s). In order to overcome this problem, Fry and Pagan (2011) suggest the median target (MT) method, taking the responses to a shock, which are overall as close as possible to the median responses, while imposing that the responses are generated from a single identifying vector $q$.\textsuperscript{23}

D. Details on the Interpretation of Coefficient Estimates

$\hat{\beta}_2$: One can see this by ignoring all other terms in (9) except for the one involving $\hat{\beta}_2$, thus

\[
-\ln(\sigma_{x,t}^2) = \hat{\beta}_2 \ln(\sigma_{\pi,t}^2/\sigma_{x,t}^2) = \hat{\beta}_2 \left[ \ln(\sigma_{\pi,t}^2) - \ln(\sigma_{x,t}^2) \right] \\
\Leftrightarrow \ln(\sigma_{x,t}^2) = \left[ \hat{\beta}_2/(1 - \hat{\beta}_2) \right] \ln(\sigma_{\pi,t}^2),
\]

where $\hat{\beta}_2 \in [0,1)$ implies that $[\hat{\beta}_2/(1 - \hat{\beta}_2)] < 0$, i.e., a higher inflation variability implies a lower output variability. An estimate of $\hat{\beta}_2 \in [0,1)$ significantly different from zero already implies a non-linear inflation output variability tradeoff. The equation is linear in the natural logarithms of variances. However, if we apply $\exp(\cdot)$ on both sides, one can see that the relationship between the variances in inflation and output is non-linear and convex as suggested by economic theory.

\textsuperscript{23}As implied by the identification scheme, the variance of shocks is normalized to unity. This however relates to the whole sample. As single observations do not uniformly differ from the mean, different subsamples (time windows) may have different shock variances, which can be related to shifts in the inflation output variability tradeoff.
\( \hat{\alpha}_e \): Ignore all other terms in (9) apart from the ones involving \( \hat{\alpha}_e \) and \( \hat{\beta}_2 \), thus

\[
- \ln (\sigma_{x,i,t}^2) = \hat{\alpha}_e \ln (\sigma_{x,i,t}^2) + \hat{\beta}_2 \ln (\sigma_{\pi,i,t}^2 / \sigma_{x,i,t}^2)
\]

\[\Leftrightarrow \ln (\sigma_{x,i,t}^2) = [\hat{\alpha}_e / (-1 + \hat{\beta}_2)] \ln (\sigma_{\pi,i,t}^2) + [\hat{\beta}_2 / (-1 + \hat{\beta}_2)] \ln (\sigma_{x,i,t}^2)
\]

\[\Leftrightarrow \ln (\sigma_{\pi,i,t}^2) = -[\hat{\alpha}_e / \hat{\beta}_2] \ln (\sigma_{\pi,i,t}^2) + \left[(-1 + \hat{\beta}_2)/\hat{\beta}_2\right] \ln (\sigma_{x,i,t}^2)\]

and, as \( \hat{\alpha}_e < 0 \) and \( \hat{\beta}_2 \in [0, 1) \), it follows that \( [\hat{\alpha}_e / (-1 + \hat{\beta}_2)], -[\hat{\alpha}_e / \hat{\beta}_2] > 0 \). We conclude from the equations above that the relationship between the variance of the supply shock and the variances of the output gap and inflation is positive, which is consistent with the economic theory discussed above.

\( \hat{\beta}_\varepsilon \): Similar arguments as above, yield

\[
\ln (\sigma_{x,i,t}^2) = [\hat{\alpha}_e / (-1 + \hat{\beta}_2)] \ln (\sigma_{x,i,t}^2) + [\hat{\beta}_2 / (-1 + \hat{\beta}_2)] \ln (\sigma_{\pi,i,t}^2) + [\hat{\beta}_\varepsilon / (-1 + \hat{\beta}_2)] \ln (\sigma_{\varepsilon,i,t}^2)
\]

\[
\ln (\sigma_{\pi,i,t}^2) = -[\hat{\alpha}_e / \hat{\beta}_2] \ln (\sigma_{\pi,i,t}^2) + \left[(-1 + \hat{\beta}_2)/\hat{\beta}_2\right] \ln (\sigma_{x,i,t}^2) - [\hat{\beta}_\varepsilon / \hat{\beta}_2] \ln (\sigma_{\varepsilon,i,t}^2)
\]

and, as \( \hat{\beta}_\varepsilon < 0 \) and \( \hat{\beta}_2 \in [0, 1) \), it follows that \( [\hat{\beta}_\varepsilon / (-1 + \hat{\beta}_2)], -[\hat{\beta}_\varepsilon / \hat{\beta}_2] > 0 \).
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